## Applied Mathematics and Statistical Mechanics and their Applications

Lead Guest Editor: El-Sayed Abo-Dahab Guest Editors: A. M. Abd-Alla, Fateh Mebarek-Oudina, and E.M. Khalil


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## Contents

Computing Simplicial Depth by Using Importance Sampling Algorithm and Its Application Fanyu Meng, Wei Shao (D), and Yuxia Su
Research Article (11 pages), Article ID 6663641, Volume 2021 (2021)
Heat Transport Improvement and Three-Dimensional Rotating Cone Flow of Hybrid-Based Nanofluid
Azad Hussain, Qusain Haider (D), Aysha Rehman (DD, M. Y. Malik (ID, Sohail Nadeem (D), and Shafiq Hussain
Research Article (11 pages), Article ID 6633468, Volume 2021 (2021)

A New Heat Dissipation Model and Convective Two-Phase Nanofluid in Brittle Medium Flow over a Cone
Azad Hussain, Qusain Haider (D), Aysha Rehman (D), Aishah Abdussattar, and M. Y Malik (D) Research Article (11 pages), Article ID 6688747, Volume 2021 (2021)

A Computational Model for the Radiated Kinetic Molecular Postulate of Fluid-Originated Nanomaterial Liquid Flow in the Induced Magnetic Flux Regime
Azad Hussain, Aysha Rehman (iD, Sohail Nadeem, M. Riaz Khan, and Alibek Issakhov
Research Article (17 pages), Article ID 6690366, Volume 2021 (2021)

The Mixture of the Marshall-Olkin Extended Weibull Distribution under Type-II Censoring and Different Loss Functions
Refah Alotaibi, Mervat Khalifa, Lamya A. Baharith, Sanku Dey, and H. Rezk (iD
Research Article (15 pages), Article ID 6654101, Volume 2021 (2021)

Stationary Wavelet with Double Generalised Rayleigh Distribution
Hassan M. Aljohani (D)
Research Article (14 pages), Article ID 6646287, Volume 2021 (2021)

Combined Effects of Heat and Mass Transfer on MHD Free Convective Flow of Maxwell Fluid with Variable Temperature and Concentration
Muhammad Bilal Riaz (D), Maryam Asgir, A. A. Zafar, and Shaowen Yao (D)
Research Article (36 pages), Article ID 6641835, Volume 2021 (2021)

## Generalized Type-I Hybrid Censoring Scheme in Estimation Competing Risks Chen Lifetime

 PopulationsNeveen Sayed-Ahmed (ID, Taghreed M. Jawa, Tahani A. Aloafi, F. S. Bayones, Azhari A. Elhag, J. Bouslimi, and G. A. Abd-Elmougod
Research Article (12 pages), Article ID 6693243, Volume 2021 (2021)
Kinematics Analysis of 6-DoF Articulated Robot with Spherical Wrist
Seemal Asif (i) and Philip Webb
Research Article (11 pages), Article ID 6647035, Volume 2021 (2021)

```
A Dynamic Intersecting Arrangement Model Based on Isolated Draw Zones for Stope Structure Optimization during Sublevel Caving Mining
Guanghui Li (D), Fengyu Ren (D), Hangxing Ding (D), Huan Liu (D), Mingzhi Sun (D), and Guang Li (D)
Research Article (13 pages), Article ID 6669558, Volume 2021 (2021)
Managing a Dual-Channel Supply Chain with Fairness and Channel Preference
Xianjin Du (D) and Weijie Zhao
Research Article (10 pages), Article ID 6614692, Volume 2021 (2021)
Thermoelastic Characteristics of Functionally Graded Circular Disk Models under the Loading of Contact Forces
Jaegwi Go
Research Article (12 pages), Article ID 8845028, Volume 2021 (2021)
```


## New Vertically Planed Pendulum Motion

A. I. Ismail (D)

Research Article (6 pages), Article ID 8861738, Volume 2020 (2020)

Nonlinear Dynamics and Suppressing Chaos in Magnetic Bearing System
Shun-Chang Chang (D)
Research Article (10 pages), Article ID 6669941, Volume 2020 (2020)

On a System of $\boldsymbol{k}$-Difference Equations of Order Three
Ibrahim Yalçınkaya, Hijaz Ahmad (D), Durhasan Turgut Tollu (iD, and Yong-Min Li (iD
Research Article (11 pages), Article ID 6638700, Volume 2020 (2020)

Second-Order Differential Equation: Oscillation Theorems and Applications
Shyam S. Santra, Omar Bazighifan (DD, Hijaz Ahmad (D), and Yu-Ming Chu (D)
Research Article (6 pages), Article ID 8820066, Volume 2020 (2020)

The Stability Conditions for a Heavy Solid Motion
A. I. Ismail

Research Article (4 pages), Article ID 8829439, Volume 2020 (2020)

Existence and Uniqueness of Uncertain Fractional Backward Difference Equations of Riemann-Liouville Type
Pshtiwan Othman Mohammed (D), Thabet Abdeljawad (iD, Fahd Jarad (D), and Yu-Ming Chu (D)
Research Article (8 pages), Article ID 6598682, Volume 2020 (2020)

# Computing Simplicial Depth by Using Importance Sampling Algorithm and Its Application 

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Simplicial depth (SD) plays an important role in discriminant analysis, hypothesis testing, machine learning, and engineering computations. However, the computation of simplicial depth is hugely challenging because the exact algorithm is an NP problem with dimension $d$ and sample size $n$ as input arguments. The approximate algorithm for simplicial depth computation has extremely low efficiency, especially in high-dimensional cases. In this study, we design an importance sampling algorithm for the computation of simplicial depth. As an advanced Monte Carlo method, the proposed algorithm outperforms other approximate and exact algorithms in accuracy and efficiency, as shown by simulated and real data experiments. Furthermore, we illustrate the robustness of simplicial depth in regression analysis through a concrete physical data experiment.

## 1. Introduction

With the development of computer technology and multivariate statistical analysis, scientists deal with a large amount of multidimensional data in many fields, such as biogenetics and industrial engineering. The demand for multivariate data analysis tools has become increasingly urgent. As a powerful multivariate nonparametric and robust statistical tool, the statistical depth function extends the concept of one-dimensional data order statistics and provides the central-outward sorting of multivariate data [1-4]. In recent years, the interest of researchers in statistical depth has increased due to the extensive application of the statistical depth function in multivariate statistical analysis, robust estimation, discriminant analysis, hypothesis testing, machine learning, economics, and hydrological data analysis $[5,6]$.

The first statistical depth function concept, which was proposed by Tukey in 1975, is known as the halfspace depth (also known as the Tukey depth) [7-9]. The other concepts of the statistical depth function include projection depth $[3,10]$, simplicial depth (SD) $[11,12]$, and regression depth
[13, 14]. Zuo and Serfling defined a general structural property of the statistical depth function [1]. Among the many concepts of this statistical depth function, SD is a relatively attractive one not only because of its simple form and ability to achieve the maximum depth value in the center and satisfy monotonicity but also because of its important applications in sign test and centralization test [1, 12].

However, the computation of SD is complicated. The exact calculation of SD is an NP problem, which is only feasible when the dimension is no higher than three. Serfling and Wang emphasized that the computation of SD for higher-dimensional data still requires further study [12]. The computation and application of the statistical depth function are active research topics.

Similarly, Monte Carlo (MC) methods have become important statistical, computational tools that are widely used in finance, engineering computation, genetic biology, computational chemistry, and other related fields [15-18]. As a critical MC strategy, the importance sampling (IS) method concentrates most of the test samples in the important area of the objective function by introducing the transfer probability density function [15, 19]. This method
dramatically improves the computational efficiency and is an important MC acceleration algorithm. In this study, we apply an efficient IS algorithm to the approximate computation of SD and demonstrate the advantages of such an algorithm over other MC methods and exact algorithms through simulated and real data examples. Furthermore, we extend the SD to regression analysis and obtain a robust estimation of regression analysis. The results of a real physical data experiment show that the estimation based on the SD method is more robust than that based on the traditional least squares (LS) method.

The remainder of this paper is organized as follows. In Section 2, we review the preliminary concept and existing algorithms for SD. Section 3 describes the IS algorithm used for the computation of SD. The advantages of the IS algorithm are illustrated through simulated data examples in Section 4. The extension of SD to regression analysis and a real data experiment are presented in Section 5. Lastly, the conclusions are provided in Section 6.

## 2. Preliminary of SD and the State of the Art

In this section, we present the preliminary of SD and the existing algorithms for its computation.

Consider a sample set $\mathbf{X}^{n}=\left\{\mathbf{X}_{1}, \mathbf{X}_{2}, \ldots, \mathbf{X}_{n}\right\}$ in $\mathbb{R}^{d},(d \geq 2)$, where $\mathbf{X}^{n}$ is one sample of size $n$ in $\mathbb{R}^{d}$ and $\mathbf{x}$ is a given point in $\mathbb{R}^{d}$. The sample version [11] of SD of $\mathbf{x}$ with respect to the sample set $\mathbf{X}^{n}$ is expressed as

$$
\begin{equation*}
\mathrm{SD}\left(\mathbf{x}, \mathbf{X}^{n}\right)=\frac{1}{C_{n}^{d+1}} \sum_{1 \leq i_{1}<\ldots<i_{d+1} \leq n} \mathbf{1}_{\left\{\mathbf{x} \in S\left[\mathbf{x}_{i_{1}, \ldots, \mathbf{x}_{i_{d+1}}}\right]\right\}} \tag{1}
\end{equation*}
$$

where $\mathbf{1}_{\{A\}}$ denotes the indicator function of event $A$, and $S\left[\mathbf{X}_{i_{1}}, \ldots, \mathbf{X}_{i_{d+1}}\right]$ denotes the simplex determined by the $d+1$ sample points $\left\{\mathbf{X}_{i_{1}}, \ldots, \mathbf{X}_{i_{d+1}}\right\}$.

Serfling and Wang stated that no algorithms are faster than simply generating all simplices and counting the ones enclosing the given point (using $O\left(n^{d+1}\right)$ complex time) when dimension $d \geq 5$ [12]. Therefore, designing an efficient approximate algorithm for the computation of SD is necessary.

A direct MC method for the computation of SD contains two steps: (1) randomly selecting $d+1$ points from $\mathbf{X}^{n}$ and then (2) taking the average of the points that enclose the given point $\mathbf{x}$ (i.e., using $\widehat{\mathrm{SD}}\left(\mathbf{x}, \mathbf{X}^{n}\right)$ to estimate the true SD value $\left.\operatorname{SD}\left(\mathbf{x}, \mathbf{X}^{n}\right)\right)$.

$$
\begin{equation*}
\left.\widehat{\mathrm{SD}}\left(\mathbf{x}, \mathbf{X}^{n}\right)=\frac{1}{M} \sum_{i=1}^{M} \mathbf{1}_{\{\mathbf{x} \in S}\left[\mathbf{x}_{\left.i_{1}, \ldots, \mathbf{X}_{i_{d+1}}\right]}\right]\right\} \tag{2}
\end{equation*}
$$

where $\left\{\mathbf{X}_{i_{1}}, \ldots, \mathbf{X}_{i_{d+1}}\right\}$ is randomly chosen from $\mathbf{X}^{n}$ and $M$ is the trying number for the estimation.

Another approach for the computation of SD is the use of the IS algorithm, which is the proposed method in this study. The computation of SD is an expectation computation. Therefore, SD can be estimated by the IS algorithm. The simple MC method uses the randomly selected $d+1$ points to estimate the SD, whereas the IS approach selects $d+1$ points with a high probability that they contain the given point $\mathbf{x}$. Theoretically, the results of the latter will have a
smaller variance than those of the former. The simulated data examples in Section 4 illustrate the advantage of the IS algorithm over the MC method.

## 3. New Algorithm for SD in $R^{d}$

3.1. Overview of the IS Algorithm. Many engineering problems can be expressed as computations of a multidimensional integral. Using the MC method to compute the integral involves drawing samples from a uniform distribution on a regular area and using the sample mean to approximate the true integral. In higher-dimensional cases, the efficiency of the MC method is extremely low if the region where the target function is not equal to zero is extraordinarily sparse. On the contrary, the IS algorithm draws most samples in the important area. This strategy improves the efficiency of the integral computation. The IS algorithm plays an important role in the field of statistical physics, molecular simulation, and Bayesian statistics.

For example, we want to compute the integral of $h(\mathbf{x})$ on region $A$; that is,

$$
\begin{equation*}
\mu=\int_{A} h(\mathbf{x}) \mathrm{d} x \tag{3}
\end{equation*}
$$

and the integral computation (3) can be treated as an expectation calculation:

$$
\begin{equation*}
\mu=E_{\pi}\left(\frac{h(\mathbf{X})}{\pi(\mathbf{X})}\right) \tag{4}
\end{equation*}
$$

where $\mathbf{X}$ is a random variable (r.v.) with its own probability density function (p.d.f.) $\pi(\mathbf{x})$; that is, $\mathbf{X} \sim \pi(\mathbf{x})$. If $\mathbf{X}_{1}, \ldots, \mathbf{X}_{n}$ denote samples with size $n$ from $\mathbf{X}$, the MC method draws $\mathbf{X}$ from a uniform distribution on region $A$. From the Law of Large Numbers [20], the sample mean can be used to estimate the expectation in (4) as

$$
\begin{equation*}
\widehat{\mu}=\frac{1}{n} \sum_{i=1}^{n} h\left(\mathbf{X}_{i}\right) \cdot S(A) \tag{5}
\end{equation*}
$$

where $S(A)$ is the area of $A$ and $\mathbf{X}$ is the r.v. from the uniform distribution on $A(\mathbf{X} \sim U(\mathbf{x}))$.

However, the efficiency of the MC method (5) will be extremely low if region $A$ is extremely wide or sparse (especially in high-dimensional cases). By contrast, the IS method uses a special p.d.f. $g(\mathbf{x})$ instead of $\pi(\mathbf{x})$ in (4) to compute mean $\mu$ and utilizes the corresponding sample mean to estimate the expectation in (4):

$$
\begin{equation*}
\widetilde{\mu}=\frac{1}{n} \sum_{i=1}^{n} \frac{h\left(\mathbf{X}_{i}\right)}{g\left(\mathbf{X}_{i}\right)}, \tag{6}
\end{equation*}
$$

and the variance $\operatorname{Var}(\widetilde{\mu})=1 / n \operatorname{Var}\left(h\left(\mathbf{X}_{i}\right) / g\left(\mathbf{X}_{i}\right)\right)$, which means that we can choose appropriate $g(\mathbf{x})$ close to $h(\mathbf{x})$ to reduce the variance of $\tilde{\mu}$. In extreme situations, if we select $g(\mathbf{x}) \propto h(\mathbf{x})$, that is, $g(\mathbf{x})=c \cdot h(\mathbf{x})$ (where $\left.c=1 / \int_{A} h(\mathbf{x}) \mathrm{d} \mathbf{x}\right)$, the variance of $\tilde{\mu}$ will drop to zero, and $\widetilde{\mu}$ is equal to the exact value $\int_{A} h(\mathbf{x}) \mathrm{d} \mathbf{x}$. However, we cannot directly use the IS method defined in (6) during such an extreme situation because we do not know the exact value of $\int_{A} h(\mathbf{x}) \mathrm{d} \mathbf{x}$ in advance.

Nevertheless, it gives us a significant hint that the closer $g(\mathbf{x})$ and $h(\mathbf{x})$ are, the more accurate the IS method's result is. The steps of the IS method for the computation of integral (3) are listed as follows:
(1) Draw the samples $\infty$ from $g(\mathbf{x})$.
(2) Compute the importance weights $\omega_{i}=h\left(\mathbf{x}_{i}\right) / g\left(\mathbf{x}_{i}\right)$.
(3) Use the mean of the computed weights to estimate the integral in (3):

$$
\begin{equation*}
\widetilde{\mu}=\frac{1}{n} \sum_{i=1}^{n} \omega_{i} \tag{7}
\end{equation*}
$$

The following theorem shows that the IS estimator in (7) is unbiased.

Theorem 1. The IS estimator $\tilde{\mu}$ in (7) is an unbiased estimator of $\mu$.

Proof. To prove that the IS estimator is unbiased, we only need to show that the expectation of $\widetilde{\mu}$ is equal to $\mu$ :

$$
\begin{align*}
E(\widetilde{\mu}) & =E\left(\frac{1}{n} \sum_{i=1}^{n} \omega_{i}\right) \\
& =\frac{1}{n} \sum_{i=1}^{n} E\left(\omega_{i}\right)  \tag{8}\\
& =E\left(\omega_{i}\right) .
\end{align*}
$$

Because $\omega_{i}$ is a r.v. and $\omega_{i}=h\left(\mathbf{X}_{i}\right) / g\left(\mathbf{X}_{i}\right)$,

$$
\begin{align*}
E\left(\omega_{i}\right) & =E\left(\frac{h\left(\mathbf{X}_{i}\right)}{g\left(\mathbf{X}_{i}\right)}\right) \\
& =\int_{A} \frac{h(\mathbf{x})}{g(\mathbf{x})} \cdot g(\mathbf{x}) \mathrm{d} \mathbf{x}  \tag{9}\\
& =\int_{A}(\mathbf{x}) \mathrm{d} \mathbf{x} \\
& =\mu
\end{align*}
$$

We obtain the expression $E(\widetilde{\mu})=\mu$, which verifies that the IS estimator $\widetilde{\mu}$ in (7) is unbiased. So we complete the proof of this theorem.

Aside from being an unbiased estimator of the integral presented in (3), the IS estimator exhibits a more efficient and powerful integral computation than the MC estimator defined in (5), especially in higher-dimensional cases.
3.2. IS Algorithm for SD Computation. We use the previously described IS method to compute the SD. Using the definition of SD in (1) is not appropriate in computing the SD value of a data point with respect to a dataset. The MC method in (6) becomes extremely inefficient when dimension $p$ or sample size $n$ is excessively large because the number of simplices
containing the original data point decreases with the increase in $p$ or $n$.

The IS algorithm can transform the original p.d.f. into a highly efficient one that can construct the simplex containing the original data point. In the computation of SD, the MC method randomly selects $p$ data points to construct the simplex, whereas the IS method chooses the data points that are likely to let the original data point inside the simplex. Figure 1 is a 2 D example that is composed of 20 sample data points. The data point $x_{0}$ is used to compute the SD value. After sampling the two data points ( $x_{1}$ and $x_{2}$ ), only two more ( $x_{3}$ or $x_{4}$ ) are needed to construct the simplex that contains the original data point $x_{0}$. In this illustrated example, we do not need to count all the simplices after getting $x_{1}$ and $x_{2}$; only $x_{3}$ or $x_{4}$ is considered as the final vertex of the simplices containing $x_{0}$.

We list the details of the IS algorithm for the computation of SD in high-dimensional cases. Suppose that $\mathbf{X}^{n}$ is a sample with size $n$ in $\mathbb{R}^{d}$ (i.e., $\left.\mathbf{X}^{n}=\left\{\mathbf{X}_{1}, \mathbf{X}_{2}, \ldots, \mathbf{X}_{n}\right\}\right)$ and $x$ is a given point in $\mathbb{R}^{d},(d \geq 2)$. The data points are in general position (i.e., any $d$ data points can define a unique $d$-1-dimensional hyperplane in $\left.\mathbb{R}^{d}\right)$. The procedure of using the IS algorithm to compute SD (i.e., the computation of $\operatorname{SD}\left(\mathbf{x}, \mathbf{X}^{n}\right)$ ) is summarized as follows:
(1) Set the IS parameters, including the number of samples tries $N$.
(2) Let $t=1,2, \ldots, N$. Compute the importance weight $\omega_{t}$ for the $t$-th sample try.
(i) Randomly choose d sample points from $\left\{\mathbf{X}_{1}, \mathbf{X}_{2}, \ldots, \mathbf{X}_{n}\right\}$, and denote them as $\left\{\mathbf{X}_{1}^{t}, \mathbf{X}_{2}^{t}, \ldots, \mathbf{X}_{d}^{t}\right\}$.
(ii) Let $k=1, \ldots, d$, and compute the simplex data point set $\mathscr{U}_{k}^{t}$ (i.e., the datasets consist of the possible data points that can construct a simplex containing the original data point $\mathbf{x}$ ).
Replace the $k$-th data point $\mathbf{X}_{k}^{t}$ with the original data point $x$ to obtain a dataset $P_{k}^{t}$ with size $d$ (i.e., $P_{k}^{t}=\left\{X_{1}^{t}, \ldots, X_{k-1}^{t}, x, X_{k+1}^{t}, \ldots, X_{d}^{t}\right\}$ ).

Compute the unique director $\mathbf{d}_{k}^{t}$ which is perpendicular to the hyperplane determined by $P_{k}^{t}$.
Project all data points $\left\{\mathbf{X}_{1}, \mathbf{X}_{2}, \ldots, \mathbf{X}_{n}\right\}$ and $\mathbf{x}$ along $\mathbf{d}_{k}^{t}$, and compute the projected value $\left\{x_{t, k}^{l}\right\}$, wheret $=1, \ldots, N ; k=1, \ldots, d ; \quad$ and $l=0,1, \ldots, n$, and $x_{t, k}^{0}$ is the projected value of $\mathbf{x}$ along $\mathbf{d}_{k}^{t}$.
Compute the simplex data point set $\mathscr{U}_{k}^{t}=\left\{\mathbf{X}_{l} \mid\left(x_{t, k}^{l}-x_{t, k}^{0}\right) \times\left(x_{t,} \quad k^{k}-x_{t, k}^{0}\right)<0, l=\right.$ $1, \ldots, n\}$.
(iii) Let $\mathscr{U}^{t}=\cap{ }_{k=1}^{d} \mathscr{U}_{k}^{t}$, and set $\omega_{t}=\#\left\{\mathscr{U}^{t}\right\} /(n-d)$, where $t=1, \ldots, N$.
(3) The sample mean of $\omega_{t}(t=1, \ldots, N)$ can be treated as the IS estimator of $\mathrm{SD}\left(x, X_{n}\right)$; that is,


Figure 1: A 2D simulated example (sample size $=20$ ) which indicates that only two possible trilaterals (or simplices) can contain the original data point $x_{0}$ (the vertexes of the trilaterals are $\left\{x_{1}, x_{2}, x_{3}\right\}$ and $\left\{x_{1}, x_{2}, x_{4}\right\}$ ).

$$
\begin{equation*}
\widetilde{\mathrm{SD}}\left(\mathbf{x}, \mathbf{X}_{n}\right)=\frac{1}{N} \sum_{t=1}^{N} \omega_{t} . \tag{10}
\end{equation*}
$$

Theorem 2. The computational complexity of using the IS algorithm to calculate $S D$ is

$$
\begin{equation*}
O\left(N d^{5} n\right) \tag{11}
\end{equation*}
$$

where $N$ is the number of samples tries of the IS algorithm, $d$ is the dimension of the sample data, and $n$ is the sample size.

Proof. According to the steps for computing SD using the IS algorithm, we need to compute every $\omega_{i}$ for $i=1, \ldots, N$. For every $\omega_{i}$, every selected sample data point $\mathbf{X}_{k}^{t}$ for $k=1, \ldots$, d must be replaced. The computational complexity of finding the unique director perpendicular to the hyperplane is $O\left(d^{3}\right)$, whereas that of projecting all data points to the unique director is $O(d n)$. The total computational complexity is $O\left(N d^{5} n\right)$. Then we complete the proof of this theorem.

Theorem 2 shows that the computational complexity of the IS algorithm for the computation of SD is a polynomial with dimension $d$ and sample size $n$ as its input arguments. While all other exact algorithms for the computation of SD are NP problems, especially when the dimension $d \geq 5$, there is no algorithm that can run faster than simply generating all simplices and computing the exact SD value (i.e., using $O\left(n^{d+1}\right)$ time) [12]. According to the definition of the IS algorithm in (7) and Theorem 1, the IS estimator defined in (10) is an unbiased estimator of $\operatorname{SD}\left(\mathbf{x}, \mathbf{X}_{n}\right)$.

## 4. Performance Comparison

This section presents simulated and real data examples of SD computation. All results are obtained using $R$ (version X64 3.6.2) and MATLAB (R2017a) on a Lenovo K42-80 laptop
computer (Intel(R) Core(TM) i7-6500U CPU@2.5GHz, RAM 16.00 GB , Windows 10 ). The $R$ and MATLAB codes for the results in this section are available upon request from the corresponding author.
4.1. 2D Simulated Data Example. In the simulated data experiment, we compare the computed SD results of the IS, exact, and approximate algorithms, including the MC method. The simulated dataset is sampled from a 2 D multivariate normal distribution (i.e., $N\left(\overrightarrow{\mathbf{0}}_{2}, \mathbf{E}_{2}\right)$, where $\overrightarrow{\mathbf{0}}_{2}$ is 2 D zeros vector and $\mathrm{E}_{2}$ is a 2 D unit matrix), and the sample size is 100 .

We used the exact algorithm [21], the MC method, and the IS algorithm to compute the SD. The selected points $\mathbf{x}$ are $(0,0),(0.5,0.5),(1,1)$, and $(2,2)$. We used the exact and approximate algorithms to compute the SD of $\mathbf{x}$ with respect to the dataset. The number of random simplices was set to 100 for the MC and IS algorithms. All computations were repeated 50 times. The computed results (mean, standard deviation (sd), and total CPU time (s)) are summarized in Table 1 and Figure 2.

Since there is an exact algorithm for the SD computation in the 2D case, we can evaluate the accuracy of the IS and MC methods through their mean values and sd values. Moreover, the total CPU time consumed by every algorithm can reflect its efficiency. So, in this experiment, we use these three indicators (mean, sd, and total CPU time) to compare the performances of these algorithms (exact, MC, and IS methods) for the computation of SD.

The results reveal that (1) the exact algorithm consumes less CPU time (approximately 0.1 s ), (2) the approximate algorithms (MC and IS) can achieve accurate results because their means are extremely close to the exact value, (3) IS performs better than MC as indicated by the smaller sd of the results of the former compared with those of the latter under the same CPU time, (4) all computed SD results from exact and approximate algorithms are zeros at point $(2,2)$, which means that $(2,2)$ is outside the data cloud, and (5), with the

Table 1: Computed results (mean, sd, and total CPU time in s) from different algorithms (exact, MC, and IS) in 2D experiments.

|  |  | Exact | MC | IS |
| :---: | :---: | :---: | :---: | :---: |
| $(0,0)$ | Mean | 0.2407 | 0.2380 | 0.2437 |
|  | sd | 0 | 0.0420 | 0.0150 |
|  | Time | 0.099 | 1.545 | 1.229 |
| $(0.5,0.5)$ | Mean | 0.1607 | 0.1528 | 0.1608 |
|  | sd | 0 | 0.0326 | 0.0177 |
|  | Time | 0.149 | 1.51 | 1.364 |
| $(1,1)$ | Mean | 0.0545 | 0.0594 | 0.0559 |
|  | sd | 0 | 0.0245 | 0.0127 |
|  | Time | 0.125 | 1.587 | 1.332 |
| $(2,2)$ | Mean | 0 | 0 | 0 |
|  | sd | 0 | 0 | 0 |
|  | Time | 0.223 | 1.594 | 1.492 |



Figure 2: Boxplots of the results from different algorithms (exact, MC, and IS) in 2D experiments. (a-d) The results from the computed SD of $(0,0),(0.5,0.5),(1,1)$, and $(2,2)$, respectively.
exact algorithm, the simulated example also indicates that the IS algorithm can obtain highly accurate results.
4.2. Higher-Dimensional Simulated Data Example. In this subsection, we compute the SD of different data points by
using the MC and IS algorithms in 3D and five-dimensional simulated dataset. We did not use the exact algorithm [21] because it cannot obtain any result within three hours.

In the 3D case, the dataset was sampled from $N\left(\overrightarrow{\mathbf{0}}_{3}, \mathbf{E}_{3}\right)$, and the sample size was 1000 . We used MC and IS methods to compute the SD of points $(0,0,0),(0.5,0.5,0.5)$, and

TABle 2: Results (mean, sd, and total CPU time in s) were obtained by MC and IS methods with respect to different data points in 3D experiments.

|  | $(0,0,0)$ | $(0.5,0.5,0.5)$ | $(1,1,1)$ |
| :--- | :---: | :---: | :---: |
| MC | $0.1254(0.0330)$ | $0.0460(0.0242)$ | $0.0054(0.0073)$ |
| IS | $0.1280(0.0121)$ | $0.0449(0.0078)$ | $0.0050(0.0033)$ |

The sd values are listed in the parentheses behind the mean values.



Figure 3: Boxplots of the results obtained by MC and IS methods in 3D experiments.
$(1,1,1)$. We set the number of random simplices to 100 and repeated the computation 50 times. The computed results are summarized in Table 2 and Figure 3.

Because the exact algorithm cannot get any computed SD results within three hours when dimension $d \geq 3$, we can only use MC and IS methods for the computation of SD in this subsection. Three indicators (mean, sd, and total CPU time) are summarized for the evaluation of the approximate methods. The mean values can be seen as the final computed SD results and the sd reflects the accuracy of the method (the smaller, the more accurate). The total CPU time reflects the efficiency of the method because it is more efficient if the method consumes less CPU time in the same computation of SD.

Table 2 and Figure 3 indicate that (1) the computed SD results decrease when the data points are changed from $(0,0,0)$ to $(1,1,1)$; the data point $(0,0,0)$ is deeper than the data point $(1,1,1)$ with respect to the dataset; (2) the two methods have similar computational efficiencies because they consume almost the same total CPU time; (3) the sd obtained by the IS method is smaller than that calculated by the MC method, which means that the former is more accurate than the latter in this case.

In the five-dimensional case, the dataset was sampled from $N\left(\overrightarrow{\mathbf{0}}_{5}, \mathbf{E}_{5}\right)$, and the sample size was 1000 . We used

MC and IS methods to compute the SD of points $(0,0,0,0,0),(0.5,0.5,0.5,0.5,0.5)$, and $(1,1,1,1,1)$. The number of random simplices was 100 , and the computations were repeated 50 times. The computed results (mean, sd, and total CPU time in s) are presented in Table 3 and Figure 4.

Table 3 and Figure 4 show that (1) the computed SD values decrease when the data points are changed from ( $0,0,0,0,0$ ) to ( $1,1,1,1,1$ ), thereby suggesting that the former is deeper than the latter; (2) the SD values in the fivedimensional examples are slightly smaller than those in 3D examples because the sparsity of the data points increases when the dimension is increased from three to five; (3) the IS algorithm performs better than the MC approach as indicated by the smaller sd of the results of the former compared with those of the latter; (4) the two approximate algorithms consume almost the same CPU time; (5) even after using 100 random simplices, the MC algorithm cannot find any simplex containing point, whereas the IS algorithm can identify many simplices. In conclusion, the IS method outperforms the MC method in terms of accuracy in these simulated examples.

We also evaluated the MC and IS methods with other numbers of random samples tries in different datasets. The findings show that the result's accuracy increases with the

Table 3: Results (mean, sd, and total CPU time in s) were obtained by MC and IS methods in five-dimensional experiments.

|  | $(0,0,0,0,0)$ | $(0.5,0.5,0.5,0.5,0.5)$ | $(1,1,1,1,1)$ | Total CPU time |
| :--- | :---: | :---: | :---: | :---: |
| MC | $0.0360(0.0160)$ | $0.0036(0.0066)$ | $0(0)$ | 7.67 |
| IS | $0.0315(0.0076)$ | $0.0028(0.0023)$ | $2.8 \times 10^{-6}\left(7.3 \times 10^{-6}\right)$ | 8.42 |

The sd values are listed in the parentheses behind the mean values.



Figure 4: Boxplots of the results obtained by MC and IS methods in five-dimensional experiments.
increase in the number of random samples tries. In addition, the number of random samples tries can be used by IS method in lots of datasets. It is found in our experiments that, if we set the number of random samples tries $N=1000$, the IS method can obtain the computed SD results within one second when dimension $d \leq 10$ and sample size $n \leq 10000$.

## 5. Application to Regression and Real Data Example

One of the most important extensions of SD is the robust estimation of regression based on SD. To demonstrate the relevant concept, we consider the linear regression model as follows:

$$
\begin{equation*}
Y=\alpha+\beta X+\varepsilon \tag{12}
\end{equation*}
$$

where random variables $X$ and $Y$ are in $R^{1}, \varepsilon \sim N\left(0, \sigma^{2}\right)$, and $\alpha, \beta$, and $\sigma^{2}$ are unknown parameters.

Considering that $\operatorname{SD}\left(\mathbf{x}, \mathbf{X}^{n}\right)$ can measure the depth of $\mathbf{x}$ with respect to $\mathbf{X}^{n}$, we extend the definition of SD to regression (12) and determine the simplicial regression depth:

$$
\begin{equation*}
\operatorname{SD}\left(\theta, \mathbf{W}^{n}\right)=\binom{n}{3}^{-1} \sum_{i<j<k} A\left(r_{i}(\theta), r_{j}(\theta), r_{k}(\theta)\right) \tag{13}
\end{equation*}
$$

where $\theta=(\alpha, \beta)$ are the parameters, $\mathbf{W}^{n}=\left(Y^{n}, X^{n}\right)$ are the samples of the model defined in (12), and $r_{i}(\theta)=Y_{i}-\alpha-$ $\beta X_{i}$ is the residual based on the $i$-th sample and

$$
\begin{align*}
& A\left(r_{i}(\theta), r_{j}(\theta), r_{k}(\theta)\right) \\
& = \begin{cases}1, & r_{i}(\theta), r_{j}(\theta), r_{k}(\theta) \text { have alternating signs, } \\
0, & \text { otherwise }\end{cases} \tag{14}
\end{align*}
$$

The SD based estimator of (12) can be defined as the maximum of $\operatorname{SD}\left(\theta, \mathbf{W}^{n}\right)$; that is,

$$
\begin{equation*}
\hat{\theta}_{\mathrm{SD}}=\arg \max _{\theta} \operatorname{SD}\left(\theta, \mathbf{W}^{\eta}\right) \tag{15}
\end{equation*}
$$

We consider the physical experiment data concerning the relationship between the atmospheric pressure and boiling point of water, which was discussed by a Scottish physicist named James D. Forbes [22]. In the mid-nineteenth century, this experiment can illustrate whether the simple measurement of the boiling point of water can substitute for the direct reading of the barometric pressure. The dataset was collected in the Alps in Scotland (Table 4 and Figure 5).

The linear regression model in (12) was used to fit the Forbes dataset. We used LS and SD methods to estimate the parameters of the model in (12). The function " lm " in $R$ Stats package ("stat") can be used to determine the LS estimator of

Table 4: Forbes data collected in Alps and Scotland. The temperature is measured using the boiling point $\left({ }^{\circ} \mathrm{F}\right)$ and the pressure is obtained using barometric pressure ( mmHg ).

| No. | Temperature | Pressure |
| :--- | :---: | :---: |
| 1 | 194.5 | 20.79 |
| 2 | 194.3 | 20.79 |
| 3 | 197.9 | 22.40 |
| 4 | 198.4 | 22.67 |
| 5 | 199.4 | 23.15 |
| 6 | 199.9 | 23.35 |
| 7 | 200.9 | 23.89 |
| 8 | 201.1 | 23.99 |
| 9 | 201.4 | 24.02 |
| 10 | 201.3 | 24.01 |
| 11 | 203.6 | 25.14 |
| 12 | 204.6 | 26.57 |
| 13 | 209.5 | 28.49 |
| 14 | 208.6 | 27.76 |
| 15 | 210.7 | 29.04 |
| 16 | 211.9 | 29.88 |
| 17 | 212.2 | 30.06 |

the model in (12). For the SD based method, we combined quasi-Newton [23] and IS methods to find the maximum point of (15). Moreover, we performed three statistical tests (i.e., the $R$ square value, normality test, and the test of goodness of fit [24]) for every fitted regression model to get a more insightful analysis. The $R$ square (or adjusted $R$ square) value from the significance test gives the percentage that the dependent variable $(Y)$ can be explained by the fitted model $(\alpha+\beta X)$ (see (12)). The normality test is used to test whether the residuals of the fitted model obey normal distribution which is the basis of other statistical tests. For example, under the assumption of normality, the F statistic value in the test of goodness of fit can be used to determine whether the fitted regression model makes sense.

We first used the LS method and SD approaches to compute the linear regression model with the original Forbes dataset (Table 4, denoted as original data in this section). The computed regression results are summarized in Table 5 and Figure 5(a); their corresponding statistical tests are summarized in Table 6 and Figure 6.

Table 5 and Figure 5(a) show that the LS and SD estimators obtained the very similar intercept parameter and slope parameter. This finding suggests that the SD method can capture the same accurate regression results compared with LS method.

The statistical test results have also confirmed the finding since the results from LS and SD methods were also very similar. They have very high $R$ square values which indicate more than variance of the dependent variable that can be explained by the fitted model. Under significance level 0.01 , we accept the assumption of normality and they pass the goodness of fit test (i.e., the $p$ value of F statistic is almost zero). In addition, if one needs a higher level of significance (such as 0.05 ) in this example, then some statistical techniques (e.g., Box-Cox transformation or strong influence points detection) can be used to improve the regression model (see more details in [22]). However, this is another


Figure 5: Regression analysis results from original (a) and contaminated (b) datasets.
research topic and there is a lack of sample points in this example; we only focus on the robustness of the regression model computed from different methods, especially when the dataset is contaminated, and that is what we do in the next experiment.

In the following experiment, we worked with a contaminated dataset from Forbes data. We intentionally changed the pressure of the 16th data point from 29.88 to 59.76. The new dataset was denoted as the contaminated data (Figure 5(b)). We compared the SD and LS methods' performances in the linear regression model with the contaminated dataset. The regression results are presented in Table 5 and Figure 5(b). Their corresponding statistical tests are summarized in Table 6 and Figure 7.

Table 5: Computed regression analysis results (original and contaminated) from Forbes data.

|  |  | Original data |  | Contaminated data |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
|  | $\alpha$ |  | $\alpha$ | $\beta$ |  |
| LS estimator | -81.0637 |  | 0.5229 | -181.527 |  |
| SD estimator | -78.2631 | 0.5086 | -78.2623 | 1.0266 |  |

Table 6: The statistical tests for regression analysis with original data and contaminated data using LS and SD methods.

|  |  | Original data |  | Contaminated data |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $R^{2}\left(\right.$ Adjusted $\left.R^{2}\right)$ | $F(p-$ value $)$ | $S-W(p-$ value $)$ | $R^{2}\left(\right.$ Adjusted $\left.R^{2}\right)$ | $F(p-$ value $)$ | $S-W(p-$ value $)$ |
| LS | $0.9944(0.9941)$ | $2677(0)$ | $0.8723(0.024)$ | $0.7650(0.7493)$ | $48.82\left(4.37 \times 10^{-6}\right)$ | $0.6473\left(3.10 \times 10^{-5}\right)$ |
| SD | $0.9922(0.9918)$ | $1928(0)$ | $0.8574(0.0140)$ | $0.9917(0.9912)$ | $1812(0)$ | $0.2861\left(3.13 \times 10^{-8}\right)$ |

Note: (Adjusted) is the adjusted $R$ square value, F and $p$ value are the F statistic value and its corresponding $p$ value, and $\mathrm{S}-\mathrm{W}$ and $p$ value are the Shapiro-Wilk test statistic value and its corresponding $p$ value.


Figure 6: Quantile-quantile (QQ) plots of the residuals from LS and SD methods with the original data. (a) The QQ plot from the LS method. (b) The QQ plot from the SD method.


Figure 7: Quantile-quantile (QQ) plots of the residuals from LS and SD methods with the contaminated data. (a) The QQ plot from the LS method. (b) The QQ plot from the SD method.

The results show that the LS estimator is greatly influenced by the contaminated data point, whereas the SD estimator can maintain satisfactory performance. The slope parameter estimated by the LS estimator changes from 0.5229 to 1.0266 , which cannot reflect the actual variation trend of the pressure-temperature curve. By contrast, the SD estimator is not affected by the contaminated data point and can still provide the actual variation trend. The estimated slope parameters obtained using SD method for two different datasets are 0.5086 and 0.5085 , respectively. The statistical test results show that, under the influence of the contaminated data point, the residuals of the fitted models from the two methods do not pass the normality test. However, the $R$ square (or Adjusted $R$ square) value from the SD method ( 0.9917 ) is much large than that of the LS method ( 0.7650 ) which means that the regression line from the SD method can explain more percentage of the variance of dependence variable compared with that of the LS method. These results imply that the SD estimator outperforms the LS estimator in the contaminated dataset experiment in terms of robustness.

## 6. Conclusions

The concept of statistical depth plays an important role in mathematical sciences, engineering, regression analysis, and life sciences. In this study, we computed the SD using the IS method and found that this new approach performs better than other exact and MC methods in terms of accuracy and efficiency. The simulated and real data examples illustrated the advantage of this new method. Finally, we tested the SD method based regression analysis through a concrete physical data example. The result indicated the excellent robustness of the proposed method compared with the LS estimation.

Given the many favorable properties of the proposed method, further research can be conducted on different angles. First, the IS parameter (i.e., number of sample tries $N$ ) plays an important role in the computation of SD, so the determination of $N$ before the performance of IS algorithm is yet to be thoroughly investigated. Second, the IS method for the SD computation can be improved by sampling the data points via other more important simplices (not the last data point in the possible simplices). Third, with the development of modern computer science, multicore high-performance computer is gaining popularity. Therefore, the IS method can be extended to a parallel computation based version. Lastly, the approximate algorithms (advanced MC methods) for other statistical depths (e.g., halfspace depth, projection depth, and regression depth) can be further explored.

## Data Availability

The experimental data used to support the findings of this study are included within the article.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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# Heat Transport Improvement and Three-Dimensional Rotating Cone Flow of Hybrid-Based Nanofluid 

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#### Abstract

The current research aims to study the mixed convection of a hybrid-based nanofluid consisting of ethylene glycol-water, copper (II) oxide $(\mathrm{CuO})$ and titanium dioxide $\left(\mathrm{TiO}_{2}\right)$ in a vertical cone. A hybrid base blend model is used to examine the nanofluid's hydrostatic and thermal behaviors over a diverse range of Reynolds numbers. The application of mixed nanoparticles rather than simple nanoparticles is one of the most imperative things in increasing the heat flow of the fluids. To test such a flow sector, for the very first time, a hybrid-based mixture model was introduced. Also, the mixture framework is a single-phase model formulation, which was used extensively for heat transfer with nanofluids. Comparison of computed values with the experimental values is presented between two models (i.e., the model of a mixture with the model of a single-phase). The natural convection within the liquid phase of phase change material is considered through the liquid fraction dependence of the thermal conductivity. The predicted results of the current model are also compared with the literature; for numerical results, the bvp4c algorithm is used to quantify the effects of nanoparticle volume fraction diffusion on the continuity, momentum, and energy equations using the viscous model for convective heat transfer in nanofluids. Expressions for velocity and temperature fields are presented. Also, the expressions for skin frictions, shear strain, and Nusselt number are obtained. The effects of involved physical parameters (e.g., Prandtl number, angular velocity ratio, buoyancy ratio, and unsteady parameter) are examined through graphs and tables.


## 1. Introduction

Nanofluid is the mixture of hard nanoparticles with the base fluid. The study of nanofluid is of huge interest for the evaluation of increasing thermal conductivity, In the engineering, cooling is important, such as the cooling of nano-electromechanical systems and semiconductors. The convection of nanofluids flow in nanowires such as microchannels and microtubes is mandatory to use nanofluids for these low-scale cooling techniques. Nanofluids are served in related works with single-phase heterogeneous fluids (whereas the nanoparticles are consistently distributed in base fluids). Free convection is
critical in thermal engineering in nanofluid within enclosures because rising heat flow is a significant problem for energy efficiency. The first attempts to improve heat transport using nanofluid. They simulated the heat transfer features of nanofluids in a two-dimensional insertion and originate that the heat transfer rate dramatically increases with postponed nanoparticles at every Grashof value. Elaziz and Marin [1] investigated one significant feature of theory, and it does not account for thermal energy dissipation. We discover a method for dealing with elastic interactions that do not take into account energy dissipation caused by heat sources and body forces. Remote as literature analysis is revised,
[2-12] scholars are doing notable nanofluid work. The analysis of heat transmission and nanofluid flow is an important unsolidified rheology issue. Experimenting on Cu -water nanofluid rheology, in which we noticed the conduct of a shear-thinning fluid obtained by Chang et al., [13] Santra et al. [14] introduced the forced conduction of Cu -water in Newtonian and non-Newtonian fluids in a channel. Das et al. [15] extended Aziz's attempt by looking at the Buongiorno fluid method for nanofluid flow. Xuan and Li [16] explored Cu -water nanofluid flow characteristics. Infrared photons are visible; sunlight or infrared and illumination are shown by the material nature produced from those radiations. Based on the way solar energy is collected and transmitted or transformed into solar power, energy sources and their technologies are mostly known as either active solar or inactive solar. Blackbody radioactivity is the radiation of electromagnetic waves from a superficial that exceeds absolute zero. In several practical applications, heat transport occurs via a porous medium flow. These inspections cover a wide range of fields of science and engineering, mainly grain storage, chemical hydrogen reactors, dampness movement by air-filled rubber protection, and much more. The efficiency of common base fluid thermal systems is relatively low. Suspending metallic nanoparticles in the sordid fluid is a recent way to improve the efficiency of those systems. Sheikholeslami et al. [17] investigated the free convection warmth transfer in a concentration halo between warm four-sided and heated curve cylinders in the non-attendance of the magnetic field. Kandelousi [18] investigated the consequence on ferrofluid flow of especially variable magnetic fields by considering the constant heat flux endpoint state. Sheikholeslami et al. [19-21] examined nanofluid flow alongside convective heat transfer through a different geometry. The fluid flows including chemical reaction has wide range in the processes of extrusion, refrigeration, and polymer industries. Under GN electromagnetic theory, Abd-Elaziz et al. [1] demonstrated the effect of Thomson and initial stress in a thermo-porous elastic solid. Vlase et al. [22] looked into the motion equation for a versatile one-dimensional element used in a multibody system's dynamical analysis. Malik et al. [23] proposed the idea of an incompressible fluid past MHD natural reaction over a heat-producing porous layer. The electrically transmitting non-Newtonian fluids can be used as a refrigeration liquid because their flow can be controlled by the outdoor magnetic field, which to some degree controls the heat transfer. Its usage of magnetic fields that impact heat preoccupation/generation system has several engineering applications, like crystallization and bottling of copper wires by dragging continuous polymers through inactive fluid [24-29].

Shirejini et al. [30] used a nanofluid and a gyratory scheme to restore the heat transfer rate after a decrease. Turkyilmazoglu [31-33] investigated the thermal broadcast of an electrically conductive fluid over a rotating infinite disk. Digital devices for stowage, rotating equipment, thermal energy generation systems, electrode material, geothermic industry, gas turbines, biological courses, and
different types of medical equipment are examples of applications of such problems. Turkyilmazoglu investigated fluid stream and heat allocation on a rotating disk that was traveling vertically. A spinning cone induces warmth transfer and enables flow in a quiescent liquid. Kumar et al. [34] used a finite element method to research the randomness production of a nanofluid containing copper and aluminum oxide nanoparticles in the spaces between two coaxial spinning disks.

The above studies indicate that no attempt has been complete to analyze the 3D hybrid nanofluid flow model around the cone as poignant or immobile under fluid control. The effect of copper oxide $(\mathrm{CuO})$ and titanium dioxide $\left(\mathrm{TiO}_{2}\right)$ nanoparticles on the thermal performance properties of ethylene glycol-water is investigated in this study, which has an extensive scientific and technological value. The second significance is to build on the principle of Refs. [35-37] which also contain the most important studies about the current model. In the case of counterrotating, create a mathematical model for rotating cones that are called moving or stationary. The flow reckonings are reduced to an ordinary scheme, and then bvp4c is used to solve them. Figures illustrate the effects of corporeal relevant variable quantity on velocity and temperature. Superficial grind force and temperature incline numerical outputs are tabulated contrary to stimulating physical objects. The uniqueness of the latest work is emphasized.
(1) The current study considers three-dimensional $\mathrm{CuO}+\mathrm{TiO}_{2} / \mathrm{C}_{2} \mathrm{H}_{6} \mathrm{O}_{2}$ hybrid nanofluid flow, while previous research [38, 39] has concentrated on viscous fluids and nanofluids.
(2) The MATLAB bvp4c algorithm has been used for the explanation of the non-linear problem.
(3) In comparison to other fluids, hybrid nanofluids have been found to increase the thermal efficiency of base fluids quickly.

## 2. Mathematical Formulations

To find another way to simplify the process of convection in fluids, the basis for this paper is a three-dimensional (3D) natural heat transfer of Newtonian two-phase nanofluid flow composed of $\mathrm{TiO}_{2} / \mathrm{CuO}$ hybrid particles/ethylene glycolwater ( 50 percent- 50 percent) combination of base fluid due to a pivoting cone. All conclusions and conditions considered for the geometry of this paper are clearly shown in Figure 1. Differential equations that model the problem according to the assumptions mentioned above and the physical terms that affect the problem are

$$
\begin{gather*}
x \frac{\partial u}{\partial z}+u+x \frac{\partial w}{\partial z}=0  \tag{1}\\
\frac{\partial u}{\partial t}+u \frac{\partial u}{\partial x}-\frac{v^{2}}{x}+w \frac{\partial u}{\partial z}=-\frac{v_{e}^{2}}{x}+\frac{v_{h n f} \partial^{2} u}{\partial z^{2}}+g \zeta \cos \alpha^{*}\left(T-T_{\infty}\right), \tag{2}
\end{gather*}
$$



Figure 1: Corporeal geometry of the problem.

$$
\begin{gather*}
\frac{\partial v}{\partial t}+u \frac{\partial v}{\partial x}+u \frac{v}{x}+w \frac{\partial v}{\partial z}=\frac{\partial v_{e}}{\partial t}+v_{h n f} \frac{\partial^{2} v}{\partial z^{2}}  \tag{3}\\
u \frac{\partial T}{\partial x}+w \frac{\partial T}{\partial z}+\frac{\partial T}{\partial t}=\alpha_{h n f} \frac{\partial^{2} T}{\partial z^{2}} \tag{4}
\end{gather*}
$$

The velocity gears in the paths of $x, y$, and $z$-axis, separately, are in the above equations ( $u, v$, and $w$ ). Also, $\rho_{\text {hnf }}$ is the concentration of nanofluid, $\mu_{h n f}$ is the fluid viscosity of nanofluid, $v_{e}$ is the free flow velocity, $(\rho \beta)_{\text {hnf }}$ is the coefficient of growth and contraction because of the temperature difference, $T$ is the dynamic temperature, $\zeta$ is the electrical conductivity of the fluid, $v_{h n f}$ is kinematic viscosity, $\left(\rho C_{P}\right)_{h n f}$ is nanofluid's heat capacity, $k_{h n f}$ is nanofluid's heat conductivity, $A$ is the Deborah number, and $\alpha_{\text {hnf }}$ is the thermal diffusivity.

The boundary conditions are

$$
\begin{align*}
u(0, x, z) & =v=w=u_{i}, v_{i} w_{i} \\
T & =T_{i} \\
u(t, 0, z) & =w=0  \tag{5}\\
v & =\frac{\Omega_{1} x \sin \alpha^{*}}{1-s t^{*}} \\
T & =T_{W}
\end{align*}
$$

The momentum, temperature, and boundary conditions for this problem are [38, 39].

The most recent method would be the utilization of hybrid nanoparticles rather than single nanoparticles to advance the process of convection in fluids. Nanofluid formed by hybrid nanoparticles has higher conduction than nanofluid generated by one single nanoparticle. Furthermore, the impact of using nanoparticles of different shapes on conductivity and reducing the amount of convection cannot be simply overlooked.

Thermophysical Properties. The following are the different thermal properties of hybrid nanofluid and water [39]:

$$
\begin{align*}
\rho_{\text {hnf }}= & {\left[\left(1-\phi_{2}\right)\left(1-\phi_{1}\right) \rho_{f}+\phi_{1} \rho_{c u o}\right]+\phi_{2} \rho_{T i o_{2}}, } \\
\left(\rho C_{p}\right)_{h n f}= & {\left[\left(1-\phi_{2}\right)\left(1-\phi_{1}\right)\left(\rho c_{p}\right)_{f}+\phi_{1}\left(\rho c_{p}\right)_{c u o}\right] } \\
& +\phi_{2}\left(\rho c_{p}\right)_{T i o_{2}},  \tag{6}\\
\frac{k_{h n f}}{k_{n f}}= & \frac{k_{c u o}+2 k_{n f}-2 \phi_{1}\left(k_{n f}-k_{c u o}\right)}{k_{c u o}+2 k_{n f}+\phi_{1}\left(k_{n f}-k_{c u o}\right)} \\
\frac{k_{n f}}{k_{f}}= & \frac{k_{T i o_{2}}+2 k_{n f}-2 \phi_{2}\left(k_{n f}-k_{T i_{2}}\right)}{k_{T i o_{2}}+2 k_{n f}+\phi_{2}\left(k_{n f}-k_{T i o_{2}}\right)} .
\end{align*}
$$

Define the following transformation:

$$
\begin{align*}
& \eta=\frac{\left(x \Omega \sin \alpha^{*}\right)^{0.5} z}{\nu\left(1-s t^{*}\right)^{0.5}}, \\
& v_{e}=\frac{x \Omega_{2} \sin \alpha^{*}}{1-s t^{*}}, \\
& \alpha=\frac{\Omega_{1}}{\Omega}, \\
& t^{*}=\Omega \sin \alpha^{*} t, \\
& w=\frac{\left(\sin \alpha^{*}\right)^{(1 / 2)}(\nu \Omega)^{(1 / 2)} f(\eta)}{\left(1-s t^{*}\right)^{(1 / 2)}}, \\
& T-T_{\infty}=\left(T_{W}-T_{\infty}\right) \theta(\eta), \\
& u(t, x, z)=-\frac{2^{-1} \sin \alpha^{*} f^{\prime} \Omega x}{1-s t^{*}},  \tag{7}\\
& v=\Omega x \sin \alpha^{*}\left(1-s t^{*}\right)^{(1 / 2)} g(\eta), \\
& T_{W}-T_{\infty}=\frac{\left(T_{0}-T_{\infty}\right) x L^{-1}}{\left(1-s t^{*}\right)^{2}}, \\
& G r_{1}=\frac{\cos \alpha^{*}\left(T_{0}-T_{\infty}\right) g \beta L^{3}}{\nu^{2}}, \\
& \gamma_{1}=\frac{G r_{1}}{\operatorname{Re}_{L}^{2}}, \\
& \operatorname{Re}_{L}=\sin \alpha^{*} \Omega L^{2} \nu^{-1}, \\
& \operatorname{Pr}=\frac{\nu}{\alpha} .
\end{align*}
$$

The functions required for the conversion of the partial differential equations (PDEs) ((2), (3), and (4)) to the ordinary differential equations (ODEs) are as follows.

In which the hybrid angular velocity is $\omega=\omega_{1}+\omega_{2}$, the angular velocities of a cone are $\omega_{1}$, the free torrent liquid is $\omega_{2}$, and the unstable parameter is $S$. Also, $\theta$ and $\zeta$ are the variable and temperature without dimensions, respectively.

After substituting equation (7) into equations (2)-(4) and modifying and converting, the usual differential equations
relating the flow and temperature, together with the boundary conditions, are as follows:

$$
\begin{align*}
& \left(\frac{f^{\prime \prime \prime}}{(1-\phi)^{2.5}\left(1-\phi_{2}\right)^{2.5}\left[\left(1-\phi_{2}\right)\left\{\left(1-\phi_{1}\right)-\phi_{1}\left(\rho_{c u o} / \rho_{f}\right)\right\}+\phi_{2}\left(\rho_{T i o_{2}} / \rho_{f}\right)\right]}-\left(f+\frac{1}{2} s \eta\right) f^{\prime}+\left(\frac{1}{2} f^{\prime}-s\right) f^{\prime}-2\left(g-\left(1-\alpha_{1}\right)^{2}\right)-2 \gamma_{1} \theta\right)  \tag{8}\\
& \left(\frac{g^{\prime \prime}}{(1-\phi)^{2.5}\left(1-\phi_{2}\right)^{2.5}\left[\left(1-\phi_{2}\right)\left\{\left(1-\phi_{1}\right)-\phi_{1}\left(\rho_{c u o} / \rho_{f}\right)\right\}+\phi_{2}\left(\rho_{T i o_{2}} / \rho_{f}\right)\right]}-\left(-f^{\prime} g+f g^{\prime}\right)+s\left(1-\alpha_{1}-g-\frac{1}{2} \eta g^{\prime}\right)\right)  \tag{9}\\
& \left(\frac{1}{\left.\operatorname{Pr}\left(\frac{\kappa_{h n f}}{\kappa_{f}}\right) \theta^{\prime \prime}-\left(f \theta^{\prime}-\frac{f^{\prime} \theta}{2}\right)-2 s \theta+\frac{1}{2} s \eta \theta^{\prime}\right)} .\right. \tag{10}
\end{align*}
$$

Now, the boundary conditions are

$$
\begin{align*}
f(0) & =0 \\
g(\infty) & =-1+\alpha_{1}, \\
g(0) & =\alpha_{1}, \\
\theta(0) & =-1,  \tag{11}\\
f^{\prime}(\infty) & =0 \\
f^{\prime}(0) & =0 \\
\theta(\infty) & =0
\end{align*}
$$

The important natural quantities impacting the flow and the transfer of heat are the coefficient of skin friction $C_{f x}, C_{f y}$ and the local Nusselt number $N u_{x}$, respectively, which are clear as follows:

$$
\begin{align*}
& C_{f x}=-\operatorname{Re}_{x}^{-(1 / 2)}\left(\frac{2 \mu \partial u}{\partial z}\right)_{z=0}  \tag{12}\\
& C_{f y}=\left(\frac{2 \mu \partial v}{\partial z}\right)_{z=0}\left(-\operatorname{Re}_{x}^{-(1 / 2)}\right) \tag{13}
\end{align*}
$$

Their dimensionless form is as follows:

$$
\begin{align*}
& C_{f x} \operatorname{Re}_{x}^{(1 / 2)}=\left(\frac{1}{\left(1-\phi_{1}\right)^{2.5}\left(1-\phi_{2}\right)^{2.5}}\right)\left(-f^{\prime \prime}\right)_{\eta=0}  \tag{14}\\
& C_{f y} \operatorname{Re}_{x}^{(1 / 2)}=\left(\frac{1}{\left(1-\phi_{1}\right)\left(1-\phi_{2}\right)}\right)\left(-g^{\prime}\right)_{\eta=0} \tag{15}
\end{align*}
$$

The factor of heat allocation in dimensionless form is given as

$$
\begin{equation*}
N u_{x} \operatorname{Re}_{x}^{(1 / 2)}=-\left(\frac{k_{h n f}}{k_{f}}\right) \theta^{\prime}(0) \tag{16}
\end{equation*}
$$

## 3. Numerical Solution

The coupled ordinary, non-linear differential equations (8)-(10) and the limit conditions set out in equation (11) are numerically solved using the bvp4c MATLAB algorithm.

$$
\begin{align*}
F & =y_{1}, \\
\rho_{f} & =b, \\
F^{\prime} & =y_{2}, \\
g & =y_{4}, \\
F^{\prime \prime} & =y_{3}, \\
\rho_{s} & =a, \\
g^{\prime} & =y_{5}, \\
g^{\prime \prime} & =y_{5}^{\prime}, \\
F^{\prime \prime} & =y_{3}^{\prime},  \tag{17}\\
\theta(0) & =y_{6}, \\
\theta^{\prime} & =y_{7}, \\
\theta^{\prime \prime} & =y_{7}^{\prime}, \\
k_{s} & =m, \\
k_{f} & =n, \\
\left(\rho c_{p}\right)_{s} & =c, \\
\left(\rho c_{p}\right)_{f} & =d .
\end{align*}
$$

Now, the new equations are

$$
\begin{align*}
y_{3}^{\prime}= & {\left[\left[(1-\phi)^{2.5}\left(1-\phi_{2}\right)^{2.5}\left(1-\phi_{2}\right)\left(1-\phi_{1}\right)-\phi_{1}\left(\frac{a}{b}\right)+\phi_{2}\left(\frac{c}{b}\right)\right]\right.}  \tag{18}\\
& \left.\cdot\left[\left(y(1) y(3)+\frac{1}{2} x s y(3)\right)-\frac{1}{2} y^{2}(2)+s y(2)\right)+2 y^{2}(4)-\left(1-\alpha_{1}\right)^{2}+2 \gamma_{1} y(6)\right], \\
y_{5}^{\prime}= & {\left[(1-\phi)^{2.5}\left(1-\phi_{2}\right)^{2.5}\left(1-\phi_{2}\right)\left(1-\phi_{1}\right)-\phi_{1}\left(\frac{a}{b}\right)+\phi_{2}\left(\frac{c}{b}\right)\right] }  \tag{19}\\
& \cdot\left[y(1) y(5)-y(4) y(2)-s+s \alpha_{1}+s y(4)+\frac{1}{2} s x y(5)\right], \\
y_{7}^{\prime}= & {\left[\frac{(m+2)(n+2) \phi(n-m)}{(m+2)(n-2) \phi(n-m)}\right]\left[\operatorname{Pr} y(7) y(1)-\frac{y(6) y(2)}{2}\left(2 s y(6)-\frac{\operatorname{Pr} 1}{2} s x y(7)\right)\right], } \tag{20}
\end{align*}
$$

along with limitation

$$
\begin{align*}
y(1) & =0, \\
y_{\infty}(2) & =0, \\
y(2) & =0, \\
y(4) & =\alpha_{1},  \tag{21}\\
y(6) & =-1, \\
y_{\infty}(6) & =0 .
\end{align*}
$$

## 4. Graphical Observations and Discussion

Non-linear standard differential equations (8)-(10) concerning boundary condition equation (11) are solved by the bvp4c method of numerical technique for evaluating the various physical parameters. Results indicate the effect on velocity $-f^{\prime}(\eta), g(\eta)$ and temperature $\theta(\eta)$ profiles of nondimensional governing parameters laterally with the skin friction constant and limited Nusselt number for recommended fence temperature (PWT) cases. We considered entirely dimensional parameter values for numerical algorithms as $s=2.0, \alpha_{1}=0.6, \phi=0.8, \gamma_{1}=1.5$, and $\operatorname{Pr}=7.0$. These parameter values are samein the entire article except for the disparities in the corresponding figures and tables. We learned that the heat transfer rate has been further increased due to hybrid nanofluid $\left(\mathrm{TiO}_{2}-\mathrm{Cuo} /\right.$ ethylene glycol-water). The rate of the heat transfer decreases when we increase the rotation parameter and the capacity fraction of nanoparticles. Figures 2-5 display the block diagram of the speed and temperature profiles for different models of the volume fractions of $\mathrm{TiO}_{2}$ and Cuo nanoparticles. The rise in the medium fraction of nanotubes augments the tangential velocity $-f^{\prime}(\eta)$ field and fluctuates the azimuthal velocity $g(\eta)$ field as well as the temperature profiles in the PWT case. As assumed, the improvement of the medium fraction of nanoparticles would enhance the colloidal interruption here amid solid particles, and due to this, the velocity fields are reduced. By contrast, the field of $\mathrm{TiO}_{2}$ nanoparticle velocity is faster by enhancing the values of the nanoparticle volume fraction. For this motive, we saw an enrichment in the field of tangential velocity. Figures 2(a) and 2(b) describe the effect of the $\alpha_{1}$ on prescribed wall temperature (PWT) case velocity $-f^{\prime}$
profiles. The velocity curve decreases when increasing the values of $\alpha_{1}$ in Cuo- TiO ${ }_{2}$ cases. In Figures 4(a) and 4(b), $\alpha_{1}$ indicates mixed solutions for both cases in fields of azimuthal velocity $g$. In both cases, velocity decreases if the value of this parameter increases. But with increasing values of $\alpha_{1}$ the tangential boundary layer of velocity $-f^{\prime}$ enables. Physically, the parameter of $\alpha_{1}$ helps to increase the velocity of the flow because for this reason the velocity fields are initially enhanced. At Figures 3(a) and 3(b) under the impact of the parameter $\gamma_{1}$ almost similar behavior is observed, but here the thickness of the boundary layer increases rapidly with the growth. Velocity behavior for s the maximum value of s sluggish down the tangential speed of the fluid as well as velocity $-f^{\prime}$ goes to zero far absent from the cone superficial for high values of $s$ velocity profile goes down rdThe effects of the $\operatorname{Pr}$ on $\theta$ in both cases are shown in Figures 5(a) and 5(b). The depth of the thermal boundary layer and $\theta(\eta)$ increases by enhancing the values of Prandtl number Pr. The effect of the Reynolds number on tangential velocity is seen in Figures 5(a) and 5(b). It is represented in figures with an increase in $R e$ because inertial forces have a direct relationship with $R e$; The curve in Figure 6(a) increases and in 6(b) decreases with the increasing values of tangential velocities near the cone wall. Figures 7(a) and 7(b) are devoted to the manifestation of the effect of $\gamma_{1}$ (buoyancy parameter) on the tangential skin friction coefficient. From these figures, it is clear that $C_{f x}$ grows as $\gamma_{1}$ increases, while it decreases as $\alpha_{1}$ increases. Figures 8(a) and 8(b) indicate the impact on the skin friction coefficients $C_{f y}$ at the cone boundary of the rotation parameter $\left(\alpha_{1}\right)$ and Reynolds number. Skin friction coefficients are shown in figure 8(b) by increasing values of function of rotation parameters and $8(a)$ is decreasing by function of $\alpha_{1}$ while showing declining behavior for the number of Reynolds. The effect of $P r$ is shown in Figure 9(a) on the Nusselt number by an improvement in the value of $\operatorname{Pr}$ the Nusselt number decrease. From Figure 9(b), it is noted that the Nusselt number is decreased when enhancing the value of Pr. Tables 1 and 2 demonstrate the impact of several parameters on skin friction and Nusselt number for Reynolds model. Amount of $\alpha_{1}$ and $\gamma_{1}$ expand the skin and Nusselt number decline for $\alpha_{1}$ and grow for $\alpha_{1}$. Table 1 demonstrated the hybrid surface fluid and nanoparticles have fluid dynamics-physical properties


Figure 2: (a) Impact of $\alpha_{1}$ on the velocity field distribution $-f^{\prime}(\eta)$ for Cuo nanoparticles. (b) Impact of $\alpha_{1}$ on the velocity distribution $-f^{\prime}(\eta)$ for $\mathrm{TiO}_{2}$ nanoparticles.


Figure 3: (a) Impact of $\gamma_{1}$ on velocity distribution $-f^{\prime}$ for Cuo nanoparticles. (b) Impact of $\gamma_{1}$ on velocity distribution $-f^{\prime}$ for $\mathrm{TiO}_{2}$ nanoparticles.


Figure 4: (a) Impact of $\alpha_{1}$ on velocity profile $g(\eta)$ for CuO nanoparticles. (b) Impact of $\alpha_{1}$ on velocity profile $g(\eta)$ for Tio ${ }_{2}$ nanoparticles.


Figure 5: (a) Deviation of $\operatorname{Pr}$ on temperature profile $\theta(\eta)$ for CuO nanoparticles. (b) Deviation of $\operatorname{Pr}$ on temperature profile $\theta(\eta)$ for $\mathrm{TiO}_{2}$ nanoparticles.


Figure 6: (a) Impact on skin friction $C_{f x}$ along $x$-direction with $\alpha_{1}$, for CuO nanoparticles. (b) Impact on skin friction $C_{f x}$ along $x$-direction with $\alpha_{1}$, for $\mathrm{TiO}_{2}$ nanoparticles.


Figure 7: (a) Impact on skin friction of $\gamma_{1}$ along $x$-direction for $\mathrm{CuO}-\mathrm{C}_{2} \mathrm{H}_{6} \mathrm{O}_{2}$. (b) Impact on skin friction of $\gamma_{1}$ along $x$-direction for $\mathrm{TiO}_{2}-\mathrm{C}_{2} \mathrm{H}_{6} \mathrm{O}_{2}$ + water.


Figure 8: (a) Influence on skin friction $C_{f y}$ along $x$-direction of $\alpha_{1}$ for $\mathrm{CuO} / C_{2} \mathrm{H}_{6} \mathrm{O}_{2}+$ water. (b) Influence on skin friction $C_{f y}$ along $y$-direction of $\alpha_{1}$ for $\mathrm{TiO}_{2}-\mathrm{C}_{2} \mathrm{H}_{6} \mathrm{O}_{2}$.


Figure 9: (a) Influence on Nusselt number of Pr for $\mathrm{CuO}-\mathrm{C}_{2} \mathrm{H}_{6} \mathrm{O}_{2}$. (b) Influence on Nusselt number of Pr for $\mathrm{TiO}_{2}-\mathrm{C}_{2} \mathrm{H}_{6} \mathrm{O}_{2}$.

Table 1: Impact of different parameters and dimensions of nanomaterials on velocity and temperature profiles in CuO-hybrid base fluid.

| $\alpha_{1}$ | $\gamma_{1}$ | S | Pr | $\mathrm{C}_{\text {fx }}$ | $\mathrm{C}_{\text {fy }}$ | $\mathrm{Nu}_{\mathrm{x}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.6 | 0.5 | 1.0 | 7.0 | 0.01573 | 0.01530 | -5.01480 |
| 0.7 |  |  |  | 0.02965 | 0.00882 | -4.83310 |
| 0.8 |  |  |  | 0.04081 | 0.00202 | -4.45480 |
| 0.9 |  |  |  | 0.05035 | 0.00710 | -4.01480 |
| 0.7 | 0.0 | 2.0 | 5.0 | 0.01171 | 0.02283 | -7.24294 |
|  | 0.5 |  |  | 0.02535 | 0.01107 | -7.20510 |
|  | 1.0 |  |  | 0.03678 | 0.08875 | -7.16484 |
| 0.8 | 1.5 | 0.5 | 3.0 | 0.04706 | 0.00132 | -7.12153 |
|  |  | 1.0 |  | 0.00848 | 0.01523 | -8.89947 |
|  |  | 1.5 |  | 0.02186 | 0.01552 | -8.88430 |
|  |  | 2.0 |  | 0.03367 | 0.00754 | -8.86861 |
| 0.9 | 0.5 | 0.5 | 1.0 | 0.04379 | 0.00375 | -8.85267 |
|  |  |  | 3.0 | 0.00517 | 0.01573 | -10.2863 |
|  |  |  | 5.0 | 0.01849 | 0.01171 | -10.2791 |
|  |  |  | 7.0 | 0.03054 | 0.00898 | -10.2119 |

Table 2: Results of skin friction and Nusselt coefficient for $\mathbf{T I O}_{2}$.

| $\alpha_{1}$ | $\gamma_{1}$ | S | Pr | $\mathrm{C}_{\mathrm{fx}}$ | $\mathrm{C}_{\mathrm{fy}}$ | $\mathrm{Nu}_{\mathrm{x}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.6 | 0.5 | 1.0 | 7.0 | 0.01150 | 0.0114280 | -8.09597 |
| 0.7 |  |  |  | 0.00487 | 0.0216298 | -8.15971 |
| 0.8 |  |  |  | -0.00617 | 0.0306057 | -8.22153 |
| 0.9 |  |  |  | -0.04201 | 0.0383555 | -8.22153 |
| 0.7 | 0.0 | 2.0 | 5.0 | 0.01440 | 0.0049780 | -11.4494 |
|  | 0.5 |  |  | 0.00666 | 0.0151825 | -11.4824 |
|  | 1.0 |  |  | 0.00190 | 0.0241606 | -11.5149 |
| 0.8 | 1.5 | 0.5 | 3.0 | -0.00204 | 0.0319124 | -11.5471 |
|  |  | 1.0 |  | 0.01146 | 0.0082032 | -14.0226 |
|  |  | 1.5 |  | 0.00759 | 0.0184064 | -14.0448 |
|  |  | 2.0 |  | 0.00352 | 0.0273834 | -14.0448 |
| 0.9 | 0.5 | 0.5 | 1.0 | -0.00042 | 0.0351342 | -14.0669 |
|  |  |  | 3.0 | 0.01109 | 0.0017524 | -16.1919 |
|  |  |  | 5.0 | 0.00799 | 0.0119582 | -16.2087 |
|  |  |  | 7.0 | 0.00454 | 0.0209374 | -16.2253 |

Table 3: The hybrid surface fluid and nanoparticles having fluid dynamic-physical properties.

| Physical Properties | $\mathbf{C u O}$ | $\mathbf{T i O}_{2}$ | $\mathbf{C}_{2} \mathbf{H}_{6} \mathbf{O}_{2}-\mathbf{H}_{2} \mathbf{O}$ |
| :--- | :---: | :---: | :---: |
| $\rho\left(\mathrm{~kg} / \mathrm{m}^{3}\right)$ | 6500 | 4250 | 1063.8 |
| $c_{p}(J / \mathrm{kgk})$ | 540 | 686.2 | 3630 |
| $K(w / m k)$ | 18 | 8.9538 | 0.387 |

for Cuo and in Table 2 also shows the properties of $\mathrm{TiO}_{2}$. The thermophysical properties of the nanofluid are discussed in Table 3.

## 5. Concluding Remarks

In the present paper, the influence of rotation and buoyancy force parameters on velocity and temperature is discussed in hybrid base nanofluid over a gyrating cone in the occurrence of gravity and film condensation and heat dissipation effect. By using the bvp4c algorithm, we solve PDEs with minimum errors and correct results. The results indicate that by increasing the value of $\alpha_{1}$, the tangential and azimuthal velocity reduces near the boundary of the cone for CuO and TiO 2 cases. Also, with the increase of $\gamma_{1}$, azimuthal velocity increases. The inclination of the Prandtl number results in an increase in the temperature profile. The skin friction factor is increasingby rotation and unstable parameters while it is decreasing with Reynolds number. The Nusselt number increases for larger $\operatorname{Pr}$ near the wall of the cone. The major outcomes of this study are given as follows:
(1) $\mathrm{TiO}_{2}$ nanofluid has a higher coefficient of friction factor as opposed to Cuo nanofluid. However, the heat transfer rate of $\mathrm{Tio}_{2}$ nanofluid is lower than that of Cuo nanofluid. Cuo nanofluid, therefore, improves the thermal transfer more than the $\mathrm{Tio}_{2}$ nanofluid.
(2) The parameter of viscous variation improves both temperature and the rate of heat transfer. Thus, we can say that viscosity dependent on temperature is helpful for processes of heat transfer modification.
(3) Hybridity reduces the velocity distribution while increasing the temperature distribution.
(4) As compared to nanofluid, hybrid nanofluid can have better heat transfer efficiency.
(5) The optimal heat transfer rate in hybrid nanofluid can be achieved by choosing distinct and sufficient nanoparticle increases.
(6) The heat source decreases the temperature field and enhances the heat transfer rate.

## Abbreviations

Pr: Prandtl number
$t, t^{*}: \quad$ Dimensional and dimensionless times, respectively
$T: \quad$ Temperature
$(x, y, z): \quad$ The distance measured along the meridian of a circular segment parallel to the cone's superficial
$C_{f x}: \quad$ Local skin friction in $x$-direction
$\alpha: \quad$ Semi-upright angle of the cone
$C_{f y}: \quad$ Skin friction in $y$-direction
$\eta: \quad$ Similarity variable
$f, g: \quad$ Dimensionless stream function and velocity component in $x$ - and $y$-direction, respectively
$\theta: \quad$ Dimensionless temperature
$K, L: \quad$ Thermal conductivity and characteristic length, respectively
$\mathrm{Km}^{-1} \mathrm{~K}^{-1} \gamma_{1}$ : Buoyancy parameter due to temperature
$\mu: \quad$ Dynamic viscosity $\left(\mathrm{Nms}^{-2}\right)$
$v: \quad$ Kinematic viscosity $\left(\mathrm{m}^{2} \mathrm{~s}^{-1}\right)$
$N u_{x}: \quad$ Local Nusselt number
$\rho: \quad$ Density $\left(\mathrm{kgm}^{-3}\right)$
$R e_{x}: \quad$ Reynold number based on $x$

| $R e_{L}:$ | Reynold number based on $L$ |
| :--- | :--- |
| $\left(\rho C_{p}\right)_{h n f}:$ | Heat capacity of hybrid nanofluid $\left(j k^{-1}\right)$ |
| $\rho_{\text {hf }}:$ | Nanofluid density $\left(\mathrm{kgm}^{-3}\right)$ |
| $\mu_{f}:$ | The viscosity of fluid $\left(\mathrm{Nms}^{-2}\right)$ |
| $(u, v, w):$ | Velocity components $\left(\mathrm{ms}^{-1}\right)$ |
| $\mu_{h n f}:$ | Hybrid nanofluid viscosity $\left(\mathrm{Nms}^{-2}\right)$ |
| $\alpha_{h n f}:$ | Hybrid nanofluid thermal diffusivity |
|  | $\left(m^{2} \mathrm{~s}^{-1}\right)$. |

## Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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Research Article

# A New Heat Dissipation Model and Convective Two-Phase Nanofluid in Brittle Medium Flow over a Cone 

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#### Abstract

A time-dependent convectional flow of a two-phase nanofluid over a rotating cone with the impact of heat and mass rates is elaborated in this article. The instability in the flow field is induced by the cone angular velocity that depends on the time. The Navier-Stokes selfsimilar solution and the energy equations are obtained numerically. Here, the achieved solution is not only for Navier-Stokes equations but also for the equations of the boundary layers. In this work, the concentration, Brownian motion, and thermal buoyancy effects have important significance. We have assumed viscous dissipation with heat-absorbing fluid. Similarity answers for spinning cones with divider temperature boundary conditions give an arrangement of nonlinear differential conditions that have been handled numerically. The MATLAB methodology BVP4C is used to resolve the reduced structure of nonlinear differential equations numerically. Observation for skin friction and Nusselt number is also taken into account. Velocity and temperature impact is depicted graphically, while the outward shear stress values and heat allocation rate are included in tables.


## 1. Introduction

The design of reliable equipment in manufacturing industries relies heavily on convective flow ended a cone through radiative heat and form transfer. Due to its relevance in modern technology and applications in geothermal engineering, as well as other hydrological and astrophysical biofluid studies, researchers have shown a strong interest in heat and mass transfer in Newtonian flows in recent years. The analysis of fluid flow in a cone encompasses a wide variety of subjects. It is used in plastic processing, elastic sheet cooling, polymer technology, polymer chemistry, and engineering, to name a few. So, because of its enormous applications, the researchers are holding a purpose in this area. Dependence of viscosity on temperature plays a vital role in the realm of fluids flow. The viscosity of water decreases as the temperature increases, while the viscosity of gases rises as the temperature rises. The increase in temperature in lubricating fluids causes internal friction that affects the fluid's viscosity and will no longer remain
constant. Because of this inadequacy, many researchers are interested to understand the effects of using different variable viscosity models.

Nanofluids remain a class of heat allocation fluids which caused suspended nanoparticles are distributed in the fluid ( $1-100 \mathrm{~nm}$ ). In base fluids, locomotive oil, polymer solutions, bio-fluids, other critical fluids, water, and organic fluids (e.g., ethylene and diethylene) are commonly used. Nanoparticles are generally made of carbon in diverse edifices (e.g., carbon nanotubes, black lead, and diamond), metals (e.g., copper, hoary, and gold), and metal oxides (e.g., titanium and zirconia), besides functionalized nanoparticles. A wide-range of possible applications has been initiated for the use of nanofluids. Choi remained the first to research updraft conductivity development in nanofluids [1-12]. Nanofluids, bio and pharmacological nanofluids, remedial nanofluids, environmental nanofluids, and other heat transfer fluids are categorized according to their applications. Many researchers have looked into how extent, concentration, form, and other
assets affect the warmth transfer rate of a fluid. For the Prandtl number of airs, Hering and Grosh [13] investigated a steady mixed convection boundary layer flow from a vertical cone in an ambient fluid. For a broad range of Prandtl numbers, Himasekhar et al. [14] solved the similarity solution of the mixed convection boundary layer flow over a vertical rotating cone in an ambient fluid. Body solutions of unstable mixed convection flow from a rotating cone in a rotating fluid were obtained by Anil Kumar and Roy [15] a few years ago. Chamkha and Mudhaf [16] examined heat generation, consumption, and unstable heat and mass transfer from a revolving vertical cone with a magnetic field. Ravindran et al. [17] suggested a new approach for investigating the effects of fluid flow (suction/ injection) on a vertical porous cone's steady natural convection boundary layer flow. The impact of heat-dependent viscosity with viscous heat generation on third-grade fluid flow in a standard pipe was examined by Nadeem and Hussain [18]. Exploitation the finite simple difference method, numerical solutions were obtained. Different fluids like water, ethylene glycol, and oil, due to their poor thermal conductivity, have low heat transfer properties. It is now realized that by retaining nano-sized metal flakes such as Al , titanium, silver, gold, Cu , or their oxides, the thermophysical characteristics of such fluids could be enhanced, ending in what is commonly known as nanofluid [19]. Several researchers have spent the last few years studying the edge layer movement of nanofluid fluids in various geometries and under various conditions. Kameswaran et al. [20] investigated the flow of hydromagnetic nanofluids due to a shrinking surface. Over a stretching field, Kameswaran et al. [21] discovered solutions for the deflation-point flow equations. Fauzi et al. [22] explored the time-independent nanofluid boundary film flowing along a perpendicular cone in a brittle medium. Boutra et al. [23] investigated unrestricted convection induction in a nanofluid-filled framework through round heaters, and Ambethkar and Kumar [24] investigated 2D noncompressible flow solutions with the transfer of heat in a powered square cavity singing stream function-vorticity model. Cheng [25] addressed natural convection flowing through a formatted cone in a brittle medium in the boundary layer. Chamkha et al. [26] examined the issue of mixed convection boundary layer flow in a continuous, turbulent flow over a rigid cone enclosed in a brittle thermal radiation medium. Nadeem and Saleem [27] examined turbulent nanofluid flow in a turning cone subjected to an induced magnetic field. In this paper, we look at a two-phase nanofluid flow along a vertically stretching cone.

The work of Heiring and Grosh [28] on natural convection over a multi-isothermal cone is one of the most recent cone-shaped surface boundary layer studies. A theoretical study of forced convection flow in relation to a rotating cone was suggested by Tien and Tsuji [29]. Koh and Price [30] have evaluated the transfer of heat past a pivoting cone. Ellahi et al. [31] have studied the analysis of simplified third-grade slide Couette fluid flow. There are some related studies about this phenomenon provided in References [32-43]. Extrusion processes, plastic product processing, polymers, and silicone slips, wire and copper-coating, glass and fiber optic production, hot spinning manufacture, metal rolling, food preparation, and a variety of other topics are frequently mentioned.

The combination of fluids has a broad range of applications, including cooling systems, heating processes, as well as biomedical and automotive science and technologies that control heat and mass transfer rates. The persistence of this article is to inspect the flow of liquid-based two-phase nanofluids (copper oxide and silver added to water) over a rotating cone. Two-phase flows include the flows that transition from pure liquid to vapor as a result of external heating, separated flows, and scattered two-phase flows in which phase is observed in a continuous carrier phase in the form of particles, droplets, or leaks. We have solved governing differential equations with the assistance of the BVP4V scheme under MATLAB. It also describes the possessions of relevant bodily parameters that affect the velocities, surface strain tensors, temperature, and convection rate with the help of graphs and tables.

## 2. Mathematical Formulations

We deliberate the flow of a compressible viscous nanofluid along an erect turning cone enclosed in a brittle medium as a two-dimensional time-dependent boundary layer. Figure 1 displays the scheme of coordinates and the corporal model. We have used a rectangular coordinate system in which the $x$-axis is determined along a meridian, the $y$-axis is determined along a round section, and the $z$-axis is determined on the cone's surface. Let $u$, v , and $w$ be velocity gears, with $x$ (tangential), $y$ (azimuthal), and $z$ (horizontal) orders (normal). The equations can be written as follows:

$$
\begin{gather*}
u+\frac{x \partial u}{\partial z}+x \frac{\partial w}{\partial z}=0  \tag{1}\\
\frac{\partial u}{\partial t}+u \frac{\partial u}{\partial x}+\frac{w \partial u}{\partial z}-\frac{v^{2}}{x}=\frac{v_{n f} \partial^{2} u}{\partial z^{2}}-\frac{v_{e}^{2}}{x}+g \zeta \cos \alpha^{*}\left(T-T_{\infty}\right), \tag{2}
\end{gather*}
$$



Figure 1: Bodily geometry of the problem.

$$
\begin{align*}
& \frac{\partial v}{\partial t}+\frac{u \partial v}{\partial x}+w \frac{\partial v}{\partial z}+u \frac{v}{x}=\frac{\partial v_{e}}{\partial t}+v_{\mathrm{nf}} \frac{\partial^{2} v}{\partial z^{2}}  \tag{3}\\
& u \frac{\partial T}{\partial x}+\frac{\partial T}{\partial t}+w \frac{\partial T}{\partial z}=\alpha_{\mathrm{nf}} \frac{\partial^{2} T}{\partial z^{2}} \tag{4}
\end{align*}
$$

The boundaries conditions are set, subject to initial terms and conditions as follows:

$$
\begin{align*}
& u(0, x, z)=v=w=u_{i} \\
& v_{i} w_{i} \\
& T=T_{i} \\
& u(t, 0, z)=w=0  \tag{5}\\
& v=\frac{\Omega_{1} x \sin \alpha^{*}}{1-s t^{*}} \\
& T=T_{W}
\end{align*}
$$

Reference [15] provides the momentum, temperature, initial conditions, and boundary conditions for this issue.

The following transformation is defined where $A$ is the Deborah number and $\gamma_{1}$ and $\gamma_{2}$ are the buoyancy parameters. The things of physical importance use in the two-phase model are as follows: $\alpha_{\mathrm{nf}}$ is the thermal diffusivity, $\nu_{\mathrm{nf}}$ is the kinematic viscosity, $\mu_{\mathrm{nf}}$ is the effective dynamic viscosity, $\rho_{\mathrm{nf}}$ is the density, $\left(\rho C_{P}\right) \mathrm{nf}$ is the heat capacity, $\kappa_{\mathrm{nf}}$ is the nanofluid thermal conductivity, and $(\rho \beta)_{\mathrm{nf}}$ is the nanofluid thermal expansion coefficient:

$$
\begin{align*}
& \eta=\frac{\left(x \Omega \sin \alpha^{*}\right)^{0.5} z}{\nu\left(1-s t^{*}\right)^{0.5}}, \\
& v_{e}=\frac{x \Omega_{2} \sin \alpha^{*}}{1-s t^{*}}, \\
& \alpha=\frac{\Omega_{1}}{\Omega}, \\
& t^{*}=\Omega \sin \alpha^{*} t, \\
& w=\frac{\left(\sin \alpha^{*}\right)^{1 / 2}(\nu \Omega)^{1 / 2} f(\eta)}{\left(1-s t^{*}\right)^{1 / 2}}, \\
& T-T_{\infty}=\left(T_{W}-T_{\infty}\right) \theta(\eta), \\
& u(t, x, z)=-\frac{2^{-1} \sin \alpha^{*} f^{\prime} \Omega x}{1-s t^{*}},  \tag{6}\\
& v=\Omega x \sin \alpha^{*} \underset{\left(1-s t^{*}\right)}{1 / 2} g(\eta), \\
& T_{W}-T_{\infty}=\frac{\left(T_{0}-T_{\infty}\right) x L^{-1}}{\left(1-s t^{*}\right)^{2}}, \\
& G r_{1}=\frac{\cos \alpha^{*}\left(T_{0}-T_{\infty}\right) g \beta L^{3}}{\nu^{2}}, \\
& \gamma_{1}=\frac{G r_{1}}{\operatorname{Re}_{L}^{2}}, \\
& \operatorname{Re}_{L}=\sin \alpha^{*} \Omega L^{2} \nu^{-1}, \\
& \operatorname{Pr}=\frac{\nu}{\alpha} . \\
& \nu_{\mathrm{nf}}=\frac{\mu_{\mathrm{nf}}}{\rho_{\mathrm{nf}}}, \\
& \alpha_{\mathrm{nf}}=\frac{k_{\mathrm{nf}}}{\left(\rho c_{p}\right)_{\mathrm{nf}}} \text {, } \\
& \mu=\mu_{f}(1-\phi)^{-2.5}, \\
& \rho_{\mathrm{nf}}=(1-\phi) * \rho_{f}+\phi \rho_{s} \text {, }  \tag{7}\\
& \left(\rho c_{p}\right)_{\mathrm{nf}} \text {, } \\
& \rho_{f}=(1-\phi) *\left(c_{p} \rho\right)_{f}+\phi\left(\rho c_{p}\right)_{f}, \\
& \frac{k_{\mathrm{nf}}}{k_{f}}=\frac{\left(\left(k_{s}+2 k_{f}\right)-2 \phi\left(k_{f}-k_{s}\right)\right)}{\left(k_{s}+2 k_{f}+\phi\left(k_{f}-k_{s}\right)\right)}, \\
& (\rho \beta)_{\mathrm{nf}}=(1-\phi)(\rho \beta)_{f}+\phi(\rho \beta)_{s} .
\end{align*}
$$

The transformation's equations (7) and (6) are replaced into (1)-(3). After that equation (1) will be fulfilled automatically, and equations (2)-(4) diminish to the form as follows:

$$
\begin{align*}
& \left(\frac{f^{\prime \prime \prime}}{(1-\phi)^{2.5}\left(1-\phi+\phi\left(\rho_{s} / \rho_{f}\right)\right.}-\left(f+\frac{1}{2} s \eta\right) f^{\prime}+\left(\frac{1}{2} f^{\prime}-s\right) f^{\prime}-2\left(g-\left(1-\alpha_{1}\right)^{2}\right)-2 \gamma_{1} \theta\right), \\
& \left(\frac{g^{\prime \prime}}{(1-\phi)^{2.5}\left(1-\phi+\phi\left(\rho_{s} / \rho_{f}\right)\right.}-\left(-f^{\prime} g+f g^{\prime}\right)+s\left(1-\alpha_{1}-g-\frac{1}{2} \eta g^{\prime}\right)\right),  \tag{8}\\
& \left(\frac{1}{\operatorname{Pr}}\left(\frac{\kappa_{\mathrm{nf}}}{\kappa_{f}}\right) \theta^{\prime \prime}-\left(f \theta^{\prime}-\frac{f^{\prime} \theta}{2}\right)-2 s \theta+\frac{1}{2} s \eta \theta^{\prime}\right) .
\end{align*}
$$

Now, the boundary conditions are

$$
\begin{align*}
f(0) & =0, \\
g(\infty) & =-1+\alpha_{1}, \\
g(0) & =\alpha_{1}, \\
\theta(0) & =-1,  \tag{9}\\
f^{\prime}(\infty) & =0, \\
f^{\prime}(0) & =0, \\
\theta(\infty) & =0 .
\end{align*}
$$

The skin friction $C_{f x}$ across the $x$-axis, $C_{f y}$ along $y$-axis, and Nusselt sum $\mathrm{Nu}_{x}$ are physical quantities of our distinct interest where

$$
\begin{align*}
& C_{f x}=-\mathrm{Re}_{x}^{-(1 / 2)}\left(\frac{2 \mu \partial u}{\partial z}\right)_{z=0}, \\
& C_{f y}=\left(\frac{2 \mu \partial v}{\partial z}\right)_{z=0}\left(-\operatorname{Re}_{x}^{-(1 / 2)}\right), \tag{10}
\end{align*}
$$

or in the form of dimensionless

$$
\begin{align*}
& C_{f x} \operatorname{Re}_{x}^{1 / 2}=\left(-f^{\prime \prime}\right)_{\eta=0}  \tag{11}\\
& C_{f y} \operatorname{Re}_{x}^{1 / 2}=(-g \prime)_{\eta=0}
\end{align*}
$$

In dimensionless form, the heat transfer coefficient is given as

$$
\begin{equation*}
\mathrm{Nu}_{x} \mathrm{Re}_{x}^{1 / 2}=-\theta \prime(0) \tag{12}
\end{equation*}
$$

Now, the Reynolds number is

## 3. Numerical Solution

The numerical technique performs the solutions of the joined linear sinusoidal differential equations. The following initial estimates and nonlinear operators are $f_{0}, g_{0}$, and $\theta_{0}$, respectively, for velocity components and temperature fields.

New variables are defined simplifying differential equations of high order in the form of first-order equation, i.e.,

$$
\begin{align*}
F & =y_{1}, \\
\rho_{f} & =b, \\
f^{\prime} & =y_{2}, \\
g & =y_{4}, \\
F^{\prime \prime} & =y_{3}, \\
\rho_{s} & =a, \\
g^{\prime} & =y_{5}, \\
g^{\prime \prime} & =y_{5}^{\prime}, \\
F^{\prime \prime \prime} & =y_{3}^{\prime},  \tag{14}\\
\theta(0) & =y_{6}, \\
\theta^{\prime} & =y_{7}, \\
\theta^{\prime \prime} & =y_{7}^{\prime}, \\
k_{s} & =m, \\
k_{f} & =n, \\
\left(\rho c_{p}\right)_{s} & =c, \\
\left(\rho c_{p}\right)_{f} & =d, \tag{13}
\end{align*}
$$

Now, the new equations are

$$
\begin{align*}
& y_{3}^{\prime}=\left[\left[(1-\phi)^{2.5}\left(1-\phi+\frac{\phi a}{b}\right)\right]\left[\left(y(1) y(3)+\frac{1}{2} x s y(3)\right)-\frac{1}{2} y^{2}(2)+s y(2)\right)+2 y^{2}(4)-\left(1-\alpha_{1}\right)^{2}+2 \gamma_{1} y(6)\right], \\
& y_{5}^{\prime}=\left[(1-\phi)^{2.5}\left(1-\phi+\frac{\phi a}{b}\right)\right]\left[y(1) y(5)-y(4) y(2)\left(-s+s \alpha_{1}+s y(4)+\frac{1}{2} s x y(5)\right],\right.  \tag{15}\\
& y_{7}^{\prime}=\left[\frac{(m+2)(n+2) \phi(n-m)}{(m+2)(n-2) \phi(n-m)}\right]\left[\operatorname{Pry}(7) y(1)-\frac{y(6) y(2)}{2}\left(2 s y(6)-\frac{\operatorname{Pr} 1}{2} s x y(7)\right)\right] .
\end{align*}
$$

Along with limitation,


Figure 2: (a) Impact of $\alpha_{1}$ on the velocity distribution $-f_{\prime}(n)$ for Cuo nanoparticles. (b) Impact of $\alpha_{1}$ on the velocity distribution $-f_{\prime}(n)$ for Ag nanoparticles.

$$
\begin{align*}
y(1) & =0, \\
y(6) & =-1, \\
y_{\infty}(2) & =0,  \tag{16}\\
y(2) & =0, \\
y(4) & =\alpha_{1}, \\
y_{\infty}(6) & =0 .
\end{align*}
$$

## 4. Graphical Observations and Discussion

This section of the learning includes the graphical and mathematical outcomes of multiple major parameters on velocities, temperature, coefficients of surface stress, and coefficient of heat transfer. Such variations are noted in figures. Figures 2(a) and 2(b) and Figures 3(a) and 3(b) are sketched to demonstrate primary velocity activity for parameter mixed convection. The positive parameter of buoyancy functions as a desirable gradient of pressure is mended to improve the property of the fluid. It is foretold from Figures 2(a) and 2(b) and Figures 3(a) and 3(b) that the thickness of the upper and lower layers will decrease with increase in $\alpha_{1}$ and $\gamma_{1}$ values; further, the primary velocity will have a higher magnitude for $\gamma_{1}$. The effect of mixed convection on buoyancy parameter $\gamma_{1}$ is to decrease the secondary velocity $g$ (see Figures 4(a) and 4(b), respectively). The secondary velocity $g$ is also seen to have the greater magnitude for $\alpha_{1}$. A vertically spinning or expanding cone was analyzed on the unstable frontier layer flow of both water-based nanofluids. The flow was contrary to viscous debauchery, the cohort of excess heat, and a natural process. The numerical technique is used to resolve the equations. We studied the belongings on the nanofluid velocity $(f \& g)$ and temperature $(\theta)$ profiles and even the skin friction $\left(C_{f x} \& C_{f y}\right)$ coefficient, energy, and mass exchange coefficients of the nanoparticle volume segment, buoyancy parameter $\gamma_{1}$, heat production, and chemical reaction. We considered nanoparticles of copper oxide (Cuo) and silver
$(\mathrm{Ag})$ with water as the basis fluid. In Figures 2(a) and 2(b) and Figures 3(a) and 3(b), see the variation of the angular velocity ratio $\alpha_{1}$, and the buoyancy coefficient $\gamma_{1}$ on tangential velocity $-f \prime$ is plotted. Tangential velocities are observed to decrease for $\alpha_{1}$ and $\gamma_{1}$ parameters. Figures 4(a) and $4(\mathrm{~b})$ display the variance of the angular velocity ratio $\alpha_{1}$ on azimuthal velocity $g$. At $g$, the action of $\alpha_{1}$ is contrary to that of tangential velocity $-f \prime$. Here, Figures 5(a) and 5(b) are shown in the temperature sector $\theta$ for specific $\operatorname{Pr}$ values. The width of the thermoelectric boundary layer is indicated to reduce for rising Pr values. This is because the higher Prandtl number fluid has more heat conductivity resulting in a softer heat boundary layer. Now, see in Cuo case Figure 5(a) temperature decrease with increases the value of Pr but contrary in Ag case Figure 5(b). Figures 6(a) and 6(b) address the variance in the ratio of the $\gamma_{1}$ buoyancy parameter on the secondary velocity skin friction coefficient. Skin values decrease in Cuo case but enhance in $\mathrm{Tio}_{2}$ case. Figures 7(a) and 7(b) show changes in the coefficient of $C_{f y}$ skin friction by rising $\alpha_{1}$. Figure 7(a) shows that $C_{f y}$ values increase in Cuo but Figure 7(b) shows decline values of $C_{f y}$. Physically, we can conclude that the surface temperature is higher than the fluid temperature close to the cone boundaries; therefore, larger $\gamma_{1}$ gives the greater values of skin friction. It is examined that the coefficient of tangential skin friction $\left(C_{f x} \& C_{f y}\right)$ decreases as $\alpha_{1}$ increases (see Figures 7(a) and 7(b)). Figures 8(a) and 8(b) and Figures 9(a) and 9(b) show that the primary skin friction coefficients increase or decrease with the rise in $\alpha_{1}$, also the same behavior for $\gamma_{1}$. In Figures 8(a) and 9(a), $C_{f y}$ values decline for Cuo, and we see that in Figures 8(b) and 9(b), the values of $C_{f x}$ are enhanced when the values of $\alpha_{1}$ and $\gamma_{1}$ are increased. Since the impact of $\operatorname{Pr}$ in the primary $-f^{\prime}$ and secondary $g$ directions on the velocity profiles is relatively small, therefore, the profiles are ignored. In Figures 10(a) and 10(b), the heat exchange rate has decreased, with increase in Pr. In Figures 10(a) and 10(b), see that the Nusselt number decreases when the values of Pr increase. Finally, figures also display the positive impact of heat dissipation on the local


Figure 3: (a) Impact of $\gamma_{1}$ on velocity distribution $-f \prime$ for Cuo nanoparticles. (b) Impact of $\gamma_{1}$ on velocity distribution $-f \prime$ for Ag nanoparticles.


FIgure 4: (a) Impact of $\alpha_{1}$ on velocity profile $g(\eta)$ for Cuo nanoparticles. (b) Impact of $\alpha_{1}$ on velocity profile $g(\eta)$ for $\operatorname{Ag}$ nanoparticles.


Figure 5: (a) Deviation of $\operatorname{Pr}$ on temperature profile $\theta(\eta)$ for Cuo nanoparticles. (b) Deviation of $\operatorname{Pr}$ on temperature profile $\theta(\eta)$ for Ag nanoparticles.


FIgure 6: (a) Impact on skin friction $C_{f y}$ along $y$-direction of $\gamma_{1}$ for Cuo nanoparticles. (b) Impact on skin friction $C_{f y}$ along $y$-direction of $\gamma_{1}$ for Ag nanoparticles.

(a)

(b)

Figure 7: (a) Impact on skin friction of $\alpha_{1}$ besides $y$-direction for Cuo - water. (b) Impact on skin friction of $\alpha_{1}$ besides $y$-direction for Ag - water.


Figure 8: (a) Influence on skin friction $C_{f x}$ along $x$-direction of $\alpha_{1}$ for Cuo - water. (b) Influence on skin friction $C_{f x}$ along $x$-direction of $\alpha_{1}$ forAg - water


Figure 9: (a) Influence on skin friction of $\gamma_{1}$ along $y$-direction for Cuo - water. (b) Influence on skin friction of $\gamma_{1}$ along $y$-direction for Ag - water.


Figure 10: (a) Impact on the Nusselt number of Pr for Cuo - water of nanoparticles. (b) Impact on the Nusselt number of Pr for water Ag of nanoparticles.

Table 1: Variations of $\alpha, \gamma, \mathbf{s}$, and $\mathbf{P r}$ on the coefficient of skin friction $\mathbf{C}_{\mathrm{fx}}$ and $\mathbf{C}_{\mathrm{fy}}$ and local Nusselt number ( $\mathbf{N u} \mathbf{u}_{\mathrm{x}}$ ) for $\mathbf{C u o}$ - water.

| $\alpha_{1}$ | $\gamma_{1}$ | $s$ | Pr | $C_{f x}$ | $C_{f y}$ | $N u_{x}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.5 | 1.5 | 2.0 | 1.0 | 0.56971 | 0.64524 | -33.7303 |
|  | 2.5 |  |  | 0.55597 | 0.61825 | -33.8036 |
|  | 3.5 |  |  | 0.54147 | 0.59124 | -33.8762 |
|  | 4.5 |  |  | 0.52733 | 0.56420 | -33.9481 |
| 0.8 |  |  |  | 0.75065 | 0.35930 | -65.5158 |
| 1.0 | 2.5 | 0.8 | 2.5 | 0.73655 | 0.55678 | -65.6143 |
| 1.2 |  |  |  | 0.72245 | 0.81300 | -65.7122 |
| 1.4 |  |  |  | 0.70835 | 0.86329 | -65.8094 |
|  |  | 0.6 |  | 0.55559 | 0.54225 | -71.2271 |
|  |  | 0.8 |  | 0.54147 | 0.52817 | -71.1483 |
|  |  | 1.0 |  | 0.52735 | 0.51408 | -71.1877 |
| 1.2 | 2.5 | 1.2 | 3.5 | 0.51323 | 0.50000 | -71.1087 |
|  |  |  | 1.0 | -21.5615 | -26.2363 | -30.2121 |
|  |  |  | 1.5 | -21.7700 | -26.3719 | -30.3089 |
| 1.2 | 3.5 | 1.0 | 2.0 | -21.9711 | -26.5047 | -30.4045 |
|  |  |  | 2.5 | -22.1854 | -26.1654 | -30.4988 |

Nusselt number. It must be noted that the shape effects in all figures are positive and growing factor in the ratio of heat flow. Table 1 and Table 2 represent the variations of $\alpha_{1}, \gamma_{1}, s$,
and $\operatorname{Pr}$ on the coefficient of skin friction $\left(C_{f x} \& C_{f y}\right)$ and local Nusselt number $\left(\mathrm{Nu}_{x}\right)$. Table 3 shows the physical properties of copper oxide and silver.

Table 2: Effects of the parameter on skin friction and Nusselt number for $\mathbf{A g}$ - water.

| $\alpha_{1}$ | $\gamma_{1}$ | $s$ | $\operatorname{Pr}$ | $C_{f x}$ | $C_{f y}$ | $N u_{x}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.5 | 1.5 | 2.0 | 1.0 | 0.160744 | 0.161692 | -33.7303 |
|  | 2.5 |  |  | 0.159905 | 0.164064 | -33.8036 |
|  | 3.5 |  |  | 0.161692 | 0.166214 | -33.8762 |
|  | 4.5 |  |  | 0.161762 | 0.161692 | -33.9481 |
| 0.8 |  |  |  | 0.161825 | 0.161568 | -65.5158 |
| 1.0 | 2.5 | 0.8 | 2.5 | 0.161692 | 0.161442 | -65.6143 |
| 1.2 |  |  |  | 0.160717 | 0.161692 | -65.7122 |
| 1.4 |  |  |  | 1.159742 | 0.162193 | -65.8094 |
|  |  | 0.6 |  | 0.161692 | 0.863293 | -71.2271 |
|  |  | 0.8 |  | 0.227089 | 0.542258 | -71.1483 |
|  |  | 1.0 |  | 0.292715 | 0.528174 | -71.1877 |
| 1.2 | 2.5 | 1.2 | 3.5 | 0.161692 | 0.514088 | -71.1087 |
|  |  |  | 1.0 | 0.093331 | 0.500001 | -30.2121 |
|  |  |  | 1.5 | 0.032939 | -26.2363 | -30.3089 |
| 1.2 | 3.5 | 1.0 | 2.0 | 0.161692 | -26.3719 | -30.4045 |
|  |  |  | 2.5 | 0.039527 | -26.5047 | -30.4988 |

Table 3: Base fluid and nanoparticles have thermophysical characteristics.

| Physical properties | $C_{P}$ | $\rho$ | $\kappa$ |
| :--- | :---: | :---: | :---: |
| $\mathrm{H}_{2} \mathrm{O}$ | 4179 | 997.1 | 0.613 |
| Cuo | 531.80 | 6320 | 76.50 |
| Ag | 235 | 10500 | 429 |

## 5. Concluding Remarks

$\backslash$ In this numerical study, we have considered an unstable twophase nanofluid flow and heat transfer attitudes over a cone littered with two diverse metal types specifically Ag and Cuo. Unsteady mixed convection flow has been investigated in a moving viscous fluid on a turning cone. Numerical solution of ordinary differential equations BVP4C has been implemented successfully. The afresh determined outcomes are recognized to be accessible in the literature in traditional agreement with the findings previously reported. Viscous dissipation was found to have the consequence of swelling the nanofluid temperature within the gravity effects area when the heat transfer rate from the layer decreases with an increase in a viscous heat generation. The analysis summary is as follows:
(1) For increasing $\gamma_{1}$ and $\alpha_{1}$, the tangential velocity field $-f^{\prime}$ declines. However, $s$ near to the boundary often causes $-f \prime$ to decrease and increases it far away from the boundary for increasing $\gamma_{1}$ and $\alpha_{1}$.
(2) On elevating $\alpha_{1}$, the azimuthal velocity field $g$ decreases and decreases with increase in $s$.
(3) For higher values of $\operatorname{Pr}$, the temperature profile $\theta(\eta)$ increases for Ag but decreases for Cuo.
(4) Growing the importance of shape effects has raised the temperature profile and also the local Nusselt number.
(5) The two-phase nanoparticle model has always a bigger impact than nanoparticles on the temperature profile.

## List of Symbols

| $\mathrm{Pr}:$ | Prandtl number |
| :--- | :--- |
| $T:$ | Temperature |
| $C_{f x}:$ | In the $x$-direction, there is local skin friction |
| $C_{f y}:$ | Skin friction in $y$-direction |
| $f, g:$ | The velocity of a dimensionless stream function <br> component in $x$ - and $y$-direction, respectively |
| $K, L:$ | Thermal conductivity and area of the indentation, |
|  | respectively, $\mathrm{Km}^{-1} \mathrm{~K}^{-1}$ |
| $\mu:$ | Dynamic viscosity $\mathrm{Nms}^{-2}$ |
| $\mathrm{Nu}_{x}:$ | Local Nusselt number |
| $\mathrm{Re}_{x}:$ | Reynold number based on $x$ |
| $\mathrm{Re}_{L}:$ | Reynold number based on $L$ |
| $\left(\rho C_{p}\right)_{n f}:$ | Heat capacity of nanofluid $j k^{-1}$ |
| $(u, v, w):$ | Velocity components $\mathrm{ms}^{-1}$ |

## Data Availability

The data used to support the findings of the study are available from the corresponding author upon request.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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# A Computational Model for the Radiated Kinetic Molecular Postulate of Fluid-Originated Nanomaterial Liquid Flow in the Induced Magnetic Flux Regime 

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#### Abstract

The performance of mass transfer rate, friction drag, and heat transfer rate is illustrated in the boundary layer flow region via induced magnetic flux. In this recent analysis, the Buongiorno model is introduced to inspect the induced magnetic flux and radiative and convective kinetic molecular theory of liquid-initiated nanoliquid flow near the stagnant point. The energy equation is modified by radiation efficacy using the application of the Rosseland approximation. Through similarity variables, the available formulated partial differential equations are promoted into the nondimensional structure. The variation of the induced magnetic field near the wall goes up, and very far away, it decays when the size of the radiation characteristic ascends. The velocity amplitude expands by enlargement in the amount of the magnetic parameter, mixed convection, thermophoresis parameter, and fluid characteristic. The nanoparticle concentration reduces if the reciprocal of the magnetic Prandtl number expands. The temperature spectrum declines by enhancing the amount of the magnetic parameter. Drag friction decreases by the increment in the values of radiation and thermophoresis parameters. Heat transport rate increases when there is an increase in the values of Brownian and magnetic parameters. Mass transfer rate increases when there is incline in the values of the magnetic Prandtl and fluid parameter.


## 1. Introduction

Improving the thermal efficiency of fluid flows under different conditions and applications has always been a famous research area. Besides, the significance of this issue because of the very wide range of applications in industries has made this area attractive to scientists and companies working in this field. To improve effectiveness, any solution proposed for this in any application can have different technical aspects that should be considered and investigated adequately [1]. For example, Ellahi et al. [2] explored the slip effect in the Newtonian fluid two-phase flow. Particles of the nanosized Hafnium are utilized in the base fluid. Two cases are discussed for fluid under
consideration, namely, (i) phase of particles and (ii) fluid phase. Three forms of geometries are investigated in both cases. Relevant studies of nanofluids are discussed in recent articles [3-19].

There are many engineering applications of the mixed convective boundary layer flow such as food processing, solidification system, and nuclear reactors. Convection also plays an important role in managing the production cycle such as medications and cosmetics. The transverse magnetic field that merged with the boundary layer-mixed convection flow towards an inclined plate with a wave is examined. The retardation inflow far from the magnetic field and leadingedge yield acceleration in the leading-edge close flow of the wavy sheet is observed by Wang and Chi-Chang [20]. Many
researchers analyzed the regime of mixed convection flows in their articles [21-26].

There are several uses of free convection in the presence of Lorentz forces, such as fire engineering and geophysics. Newly proposed nanotechnology is a new passive way to enhance heat transfer [27]. The induced magnetic field's influence on temperature curves is represented by Ghosh et al. [28]. Vanita and Kumar [29] have examined transient flow towards a cone with the inclined magnetic field. Akbar et al. [30] investigated nanoparticle interaction for peristaltic flow in an asymmetric channel towards the magnetic field induced. Nadeem and Ijaz [31] explored the impulse of nanoparticles as a drug carrier to investigate theoretically stenosis arteries in the existence of the induced magnetic effect. Hayat et al. [32] observed second-grade nanoliquid flow with the induced magnetic field towards a stretched convectively heated surface. Rashid et al. [33] inspected the induced magnetic field effects of the Williamson peristaltic fluid flow in a curved channel.

The radiation may be sunlight, infrared, or visible and the nature of the material emitted by such radiation depends on its exposure. Depending on how solar heat is collected and distributed or converted into solar electricity, a heat source and its systems are also categorized as either passive solar or active solar. Thermal radiations are defined as electromagnetic emissions from a sheet with a temperature greater than zero [34]. Viskant and Grosh [35] noted that when considering the power plants, hypersonic flights, cooling systems, and combustion chambers, radiations became an important part. Using Rosseland's approximations, they addressed thermal radiation impacts for the flow of Falkner-Skan. Hayat et al. [36] portrayed the $\mathrm{Ag}-\mathrm{CuO} / \mathrm{H}_{2} \mathrm{O}$ rotating hybrid nanofluid flow in the existence of partial slip radiation impacts. Hussain et al. [37] noted the nonNewtonian fluid flow with radiation efficacy and timedependent viscosity. Li et al. [38] examined the radiation impacts in the heat storage system by adding nanoparticles. Zeeshan et al. [39] investigated, due to entropy generation, the impacts of electromagnetohydrodynamics radiative diminishing internal energy of the pressure-driven dioxide-water titanium nanofluid flow.

The non-Newtonian fluid is more naturalistic to consider because of the rheological characteristics of physiological and industrial fluids. There is no extensive model that can describe the moving structure of all fluids due to the complex behavior of non-Newtonian fluids. Thus, to study nonNewtonian fluid flow characteristics, numerous models have been developed. The Eyring-Powell model was obtained from a liquid molecular hypothesis. And the inclusion of additional analytical constants was further improved. It accurately reflects the essence of Newtonian for low and high shear values. For example, rubber melts, condensed liquids, toiletries, cosmetics, and vegetable products are included in these fluids [40]. Appropriate studies of Eyring-Powell fluid may be mentioned in these articles [41-45].

This report is to narrate the specifications of radiative mass and heat transfer enhancement and flow analysis of
the molecular kinetic theory of liquid-initiated boundary layer stagnation point nanofluid towards a vertical stretched surface. The non-Newtonian nanofluid model is manifested with the induced magnetic field, radiation efficacy, combined convection, Brownian, and thermophoresis diffusion. The flow field describes that, in the form of partial differential equations, the laws of conservation of momentum are considered. By reducing the number of independent variables by using the technique of similarity transformation, these coupled equations are then purified into the system of ordinary differential equations. The results are interpreted by the MATLAB bvp4c technique. Induced magnetic pattern near the wall decreases, and far away, it increases with an increase in the values of the reciprocal of the magnetic Prandtl number. The concentration curve enhances when the number of magnetic, stretching, and Prandtl characteristics incline. This study of nanofluid is mainly applied in heat transfer devices such as electrical cooling systems, radiators, and heat exchangers.

## 2. Mathematical Formulations

Consider the incompressible, steady, two-dimensional (2D) stagnant point flow under the assumption of the induced magnetic field in the molecular kinetic theory of liquidinitiated nanofluid and heat transport enhancement in the existence of combined convection and radiation towards a vertical stretched sheet as shown in Figure 1. The surface stretching with velocity $u_{w}(x)=\mathrm{d} x$ and ambient velocity is $u_{\infty}(x)=b x$ while the origin is fixed at $O$; see Figure 1.

Taken the Cartesian coordinate structure, the velocity of liquid flow will change through $x$ - and $y$-axes in a way that $x$-axis is assumed vertically and $y$-axis is supposed horizontally. The fundamental form of the kinetic molecular postulate of liquid-originated nanofluid is [45]

$$
\begin{equation*}
\tau=\left[\mu+\frac{1}{\delta \dot{\gamma}} \sinh ^{-1}\left(\frac{1}{c_{1}} \dot{\gamma}\right)\right] A_{1} \tag{1}
\end{equation*}
$$

where

$$
\begin{equation*}
\dot{\gamma}=\sqrt{\operatorname{tr}\left(A_{1}^{2}\right) \frac{1}{2}} \tag{2}
\end{equation*}
$$

where $\delta, c_{1}, \mu, A_{1}$, and $\operatorname{tr}$ are fluid parameters, dynamic viscosity, first Rivlin-Ericksen tensor, and trace, respectively. Here, $\tau$ is the extra stress tensor and $A_{1}=\left[(\text { gradv })^{t}+\right.$ gradv $]$. We consider the second-order approximation for $\sinh ^{-1}$ function as

$$
\begin{equation*}
\sinh ^{-1}\left(\frac{1}{c_{1}} \dot{\gamma}\right) \cong \frac{\dot{\gamma}}{c_{1}}-\frac{\dot{\gamma}^{3}}{6 c_{1}^{3}}, \quad \text { where }\left|\frac{1}{c_{1}} \dot{\gamma}\right| \ll 1 . \tag{3}
\end{equation*}
$$

Then, equation (1) becomes

$$
\begin{equation*}
\tau=\left[\mu+\frac{1}{\delta c_{1}}-\frac{\dot{\gamma}^{2}}{6 \delta c_{1}^{3}}\right] A_{1} . \tag{4}
\end{equation*}
$$

Under these premises, the governing equations of this particular investigation are as follows:


Figure 1: Geometry of the problem.

$$
\begin{align*}
& \frac{\partial u}{\partial x}+\frac{\partial v}{\partial y}=0,  \tag{5}\\
& \frac{\partial H_{1}}{\partial x}+\frac{\partial H_{2}}{\partial y}=0,  \tag{6}\\
& u \frac{\partial u}{\partial x}+v \frac{\partial u}{\partial y}-\frac{\mu_{\infty}}{4 \pi \rho_{f}}\left(H_{1} \frac{\partial H_{1}}{\partial x}+H_{2} \frac{\partial H_{1}}{\partial y}\right)=\left(u_{\infty} \frac{\mathrm{d} u_{\infty}}{\mathrm{d} x}-\frac{\mu_{\infty} H_{\infty}}{4 \pi \rho_{f}} \frac{\mathrm{~d} H_{\infty}}{\mathrm{d} x}\right)+\left[v+\frac{1}{\delta c_{1} \rho_{f}}\right] \frac{\partial^{2} u}{\partial y^{2}} \\
&  \tag{7}\\
& \quad-\left[\frac{1}{2 \delta c_{1}^{3} \rho_{f}}\left(\frac{\partial u}{\partial y}\right)^{2} \frac{\partial^{2} u}{\partial y^{2}}\right]+g \frac{(1-C) \beta \rho_{\mathrm{fm}}}{\rho_{f}}\left(T-T_{\infty}\right) \\
&  \tag{8}\\
& \quad-g\left(\frac{\rho_{p}-\rho_{\mathrm{fm}}}{\rho_{f}}\right)\left(C-C_{\infty}\right),  \tag{9}\\
& u \frac{\partial H_{1}}{\partial x}+v \frac{\partial H_{1}}{\partial y}-H_{1} \frac{\partial u}{\partial x}-H_{2} \frac{\partial u}{\partial y}=\alpha_{1}^{*} \frac{\partial^{2} H_{1}}{\partial y^{2}},  \tag{10}\\
& u \frac{\partial T}{\partial x}+v \frac{\partial T}{\partial y}=\alpha \frac{\partial^{2} T}{\partial y^{2}}+\frac{\left(\rho c_{p}\right)_{p}}{\left(\rho c_{p}\right)_{f}}\left[D_{B} \frac{\partial C}{\partial y} \frac{\partial T}{\partial y}+\left(\frac{D_{T}}{T_{\infty}}\right)\left(\frac{\partial T}{\partial y}\right)^{2}\right]-\frac{1}{\left(\rho c_{p}\right)_{f}}\left(\frac{\partial q_{r}}{\partial y}\right) \\
& \frac{\partial C}{u x}+v \frac{\partial C}{\partial y}=D_{B} \frac{\partial^{2} C}{\partial y^{2}}+\left(\frac{D_{T}}{T_{\infty}}\right) \frac{\partial^{2} T}{\partial y^{2}} .
\end{align*}
$$

The above equations narrate the viscosity coefficient $\nu$, where $(u, v)$ and $\left(H_{1}, H_{2}\right)$ describe the velocity and
magnetic field components along the $x$ and $y$ directions, respectively, whereas $u_{w}(x)=\mathrm{d} x$ and $H_{\infty}(x)=x H_{0}$ are the
$x$ velocity and $y$ magnetic field at the edge of the boundary layer and $H_{0}$ is the uniform value of the vertical magnetic field at infinity.

The radiation heat flux is given by using Rosseland approximation:

$$
\begin{equation*}
q_{r}=-\frac{4 \sigma^{*}}{3 k^{*}}\left(\frac{\partial T^{4}}{\partial y}\right) \tag{11}
\end{equation*}
$$

where $\sigma^{*}, k^{*}$ are the Stefan-Boltzmann and the mean absorption coefficient, respectively, whereas via extending $T^{4}$ about $T_{\infty}$ in Taylor's series and ignoring the larger terms,

$$
\begin{equation*}
T^{4} \cong 4 T_{\infty}^{3} T-3 T_{\infty}^{4} \tag{12}
\end{equation*}
$$

Substituting equations (11) and (12) into (9), the heat equation takes the form

$$
\begin{align*}
& u \frac{\partial T}{\partial x}+v \frac{\partial T}{\partial y}=\frac{k}{\left(\rho c_{p}\right)_{f}} \frac{\partial^{2} T}{\partial y^{2}} \\
& +\frac{\left(\rho c_{p}\right)_{p}}{\left(\rho c_{p}\right)_{f}}\left[D_{B} \frac{\partial C}{\partial y} \frac{\partial T}{\partial y}+\left(\frac{D_{T}}{T_{\infty}}\right)\left(\frac{\partial T}{\partial y}\right)^{2}+\frac{16 \sigma^{*} T_{\infty}^{3}}{3 \rho c_{p} k^{*}}\left(\frac{\partial^{2} T}{\partial y^{2}}\right)\right] \tag{13}
\end{align*}
$$

The relevant boundary conditions are

$$
\begin{aligned}
u & =u_{w}(x)=\mathrm{d} x, \\
v & =0, \frac{\partial H_{1}}{\partial y}=0, \\
H_{2} & =0, \quad T \longrightarrow T_{w}, \text { at } y \longrightarrow 0, \\
u & =u_{\infty}(x)=b x, \\
H_{1} & =H_{\infty}, \\
T & \longrightarrow T_{\infty}, \quad C \longrightarrow C_{w}, \text { at } y \longrightarrow \infty .
\end{aligned}
$$

Invoking similarity transformations are defined by

$$
\begin{align*}
\psi & =\sqrt{\mathrm{d} v} x F(\gamma), \\
\gamma & =\sqrt{\frac{\mathrm{d}}{v}} y, \\
u & =\frac{\partial \psi}{\partial y}=\mathrm{d} x F(\gamma), \\
v & =-\frac{\partial \psi}{\partial x}=-\sqrt{\mathrm{d} v} F(\gamma), \\
H_{1} & =\left(\frac{H_{0} x}{L}\right) h^{\prime}(\gamma),  \tag{16}\\
H_{2} & =-\sqrt{\left(\frac{\nu}{d}\right)}\left(\frac{H_{0}}{L}\right) h(\gamma), \\
u & =u_{w}(x)=\mathrm{d} x, \\
H_{\infty} & =H_{0} \frac{x}{L}, \\
\theta(\gamma) & =\frac{T-T_{\infty}}{T_{w}-T_{\infty}}, \\
\varphi(\gamma) & =\frac{C-C_{\infty}}{C_{w}-C_{\infty}} .
\end{align*}
$$

The magnetized pressure is described as

$$
\begin{equation*}
p=p+\frac{\mu|H|^{2}}{8 \pi} \tag{17}
\end{equation*}
$$

Equations (5) and (6) are satisfied identically. Equations (7), (8), (10), and (13)-(15) reduce to

$$
\begin{array}{r}
{\left[1+\varepsilon-\varepsilon m_{1}\right] F^{\prime \prime \prime}+F F^{\prime \prime}-\left(F^{\prime}\right)^{2}+\eta^{2}+\beta\left[\left(h^{\prime}\right)^{2}-h h^{\prime \prime}-1\right]+\lambda_{1} \theta-N_{r} \varphi=0} \\
\alpha_{1} h^{\prime \prime \prime}+F h^{\prime \prime}-h F^{\prime \prime}=0, \\
\frac{1}{\operatorname{Pr}}\left(1+\frac{4}{3} R_{d}\right) \theta^{\prime \prime}+\theta^{\prime} F+\operatorname{Nb} \varphi^{\prime} \theta^{\prime}+\mathrm{Nt}\left(\theta^{\prime}\right)^{2}=0 \\
\varphi^{\prime \prime}+L e \operatorname{Pr} F \varphi^{\prime}+\frac{N_{t}}{N_{b}} \theta^{\prime \prime}=0 \tag{21}
\end{array}
$$

with boundary conditions

$$
\begin{align*}
F & =0, \\
F^{\prime} & =1, \\
\theta & =1,  \tag{22}\\
\varphi & =0, \\
h & =0, \\
h^{\prime \prime} & =0, \quad \text { at } \gamma \longrightarrow 0, \\
F^{\prime} & =\eta, \\
\theta & =0,  \tag{23}\\
\varphi & =1, \\
h^{\prime} & =1, \quad \text { at } \gamma \longrightarrow \infty .
\end{align*}
$$

Here, prime denotes derivative for $\gamma$ and other dimensionless parameters are described as

$$
\begin{align*}
\lambda_{1} & =\frac{(1-C) \beta \rho_{\mathrm{fm}}\left(T_{w}-T_{\infty}\right) g}{\mathrm{~d}^{2} x \rho_{f}}, \\
N_{r} & =\frac{\left(\rho_{p}-\rho_{\mathrm{fm}}\right)\left(C_{w}-C_{\infty}\right) g}{\mathrm{~d}^{2} x \rho_{f}}, \\
\eta & =\frac{b}{d}, R_{d}=\frac{4 T_{\infty}^{3} \sigma^{*}}{k^{*} k_{f}}, N_{t}=\frac{\left(\rho c_{p}\right)_{p} D_{T}\left(T_{w}-T_{\infty}\right)}{\left(\rho c_{p}\right)_{f} \nu T_{\infty}}, \operatorname{Pr}=\frac{v}{\alpha}, \\
N_{b} & =\frac{\left(\rho c_{p}\right)_{p} D_{B}\left(C_{w}-C_{\infty}\right)}{\left(\rho c_{p}\right)_{f} \nu}, \quad \mathrm{Le}=\frac{\alpha}{D_{B}}, \quad \alpha=\frac{K}{\left(\rho c_{p}\right)_{f}}, \\
\varepsilon & =\frac{1}{\delta c_{1} \mu}, m_{1}=\frac{\mathrm{d}^{3} x^{2}}{2 c_{1}^{2} v}, \nu=\frac{\mu}{\rho_{f}}, \quad \alpha_{1}=\frac{\alpha_{1}^{*}}{v}, \quad \mathrm{Re}_{x}^{(1 / 2)}=\sqrt{\frac{u_{w} x}{v}}, \\
\beta & =\frac{H_{0}^{2} \mu_{\infty}}{4 d^{2} \pi \rho_{f}} . \tag{24}
\end{align*}
$$

Physical quantities are very valuable from an engineering point of view. These quantities reported the flow behavior which is defined by local Nusselt number $\mathrm{Nu}_{x}$, skin friction $C_{f}$, and local Sherwood number $\mathrm{Sh}_{x}$ definitions as follows:

$$
\begin{align*}
C_{f} & =\frac{\tau_{w}}{\rho u_{w}^{2}} \\
\mathrm{Nu}_{x} & =\frac{x q_{w}}{k\left(T_{w}-T_{\infty}\right)},  \tag{25}\\
\mathrm{Sh}_{x} & =\frac{x q_{m}}{D_{B}\left(C_{w}-C_{\infty}\right)},
\end{align*}
$$

where $\tau_{w}$ represents the surface shear stress, $q_{w}$ denotes the surface heat flux, and $q_{m}$ presents the surface mass flux for fluid:

$$
\begin{align*}
& \tau_{w}=\left[\left(\mu+\frac{1}{\delta c_{1}}-\frac{1}{6 \delta c_{1}^{3}}\left(\frac{\partial u}{\partial y}\right)^{2}\right) \frac{\partial u}{\partial y}\right]_{y=0}, \\
& q_{w}=\left[-k \frac{\partial T}{\partial y}-\frac{16 \sigma^{*} T_{\infty}^{3}}{3 k^{*}}\left(\frac{\partial T}{\partial y}\right)\right]_{y=0}  \tag{26}\\
& q_{m}=-D_{B}\left[\frac{\partial C}{\partial y}\right]_{y=0}
\end{align*}
$$

Using invoking transformation equation (16), the dimensionless local Nusselt number, skin friction, and the local Sherwood number become

$$
\begin{align*}
C_{f} \operatorname{Re}_{x}^{(1 / 2)} & =\left[(1+\varepsilon) f^{\prime \prime}(0)-\frac{\varepsilon m_{1}}{3}\left(f^{\prime \prime}(0)\right)^{3}\right] \\
\mathrm{Nu}_{x} \mathrm{Re}_{x}^{-(1 / 2)} & =-\left[1+\frac{4}{3} R_{d}\right] \theta^{\prime}(0),  \tag{27}\\
\mathrm{Sh}_{x} \mathrm{Re}_{x}^{-(1 / 2)} & =-\varphi^{\prime}(0) .
\end{align*}
$$

## 3. Results and Discussion

Coupled nonlinear differential equations (18)-(21) and their boundary conditions (22) and (23) are numerically worked out by employing the MATLAB scheme. This portion illustrates the impact of nondimensional sundry characteristics on induced magnetic, temperature, velocity, and concentration flow characteristics numerically and graphically. Figure 2 portrays the impact of $\eta$ versus $h^{\prime}(\gamma)$. It is noticeable that the induced magnetic spectrum falls when increasing in $\eta$. Figure 3 designates the effects of $\alpha_{1}$ on the induced magnetic pattern. Dual behavior has been seen for $\alpha_{1}$, near the wall, it is getting down, and far away, it moves upward. The field of $h^{\prime}(\gamma)$ rises by enhancing the amount of $\beta$ in Figure 4. With larger values of mixed convection characteristic, the flow of induced magnetic expands as shown in Figure 5. Figures 6 and 7 demonstrate the effect of Brownian and thermophoresis diffusion on the induced magnetic field, respectively. When there is increase in the amount of Brownian diffusion, the induced magnetic field decreases, and the induced magnetic field increases by growing quantity of thermophoresis. Figure 8 shows the consequence of the Prandtl number on the induced magnetic curve, and profile falls by a bigger amount of Prandtl. The impacts of thermal radiation flux on the magnetic field are described in Figure 9. The variation of $h^{\prime}(\gamma)$ near the wall moves upward and very far away it moves downward by increasing the amount of thermal radiation. Figure 10 exhibits the outcome of the fluid parameter on $h^{\prime}(\gamma)$ expands in the values of $\varepsilon$ which cause rise in $h^{\prime}(\gamma)$. Figure 11 exhibits the deviation of $\beta$ on velocity curve. When there is expansion in the amount of $\beta$, then there is increase in


Figure 2: Outcomes of $\eta$ on $h^{\prime}(\gamma)$.


Figure 3: Impacts of $\alpha_{1}$ on $h^{\prime}(\gamma)$.
velocity flow. Figure 12 portrays the effect of $\lambda_{1}$ on the flow of velocity. The diagram shows that the fluctuation shoots up by increment in the mixed convection parameter. Nanofluid behaviour is shown by Brownian and thermophoretic characteristics in in Figures 13 and 14. It is depicted that the contrary attitude showed variation in velocity decline for Nb and incline for Nt by enhancing the quantity of these parameters. Figure 15 investigates that the velocity graph expands if the fluid parameter $\varepsilon$ rises. Figure 16 demonstrates that the field of velocity shrinks by rising the amount of the reciprocal of the magnetic Prandtl number. Figure 17 shows that when increasing stretching parameter, the field of $F^{\prime}(\gamma)$ enhances. Figures 18 and 19 show opposite behavior against the parameters of $\alpha_{1}$ and $\beta$, respectively. The temperature profile increases and decreases against the characteristics of $\alpha_{1}$ and $\beta$. Figure 20 explores the impact of Lewis number on temperature distribution, $\theta(\gamma)$ rises
near the wall, and far away, it declines compared to Le. The temperature reduction against the Brownian motion diffusion is shown in Figure 21. Figure 22 describes that temperature sketch grows by inclining the quantity of thermophoresis diffusion. It is easily noticed that enlarging the amount of Prandtl number causes the figure of temperature decline in Figure 23. Figure 24 depicts the influence of $\eta$ on the temperature field. The amplitude of temperature diminishes when the size of the stretching parameter boosts up. Figure 25 explores the influence of radiation characteristics on temperature portraits; field enhances by rising the amount of radiation parameter. Figure 26shows the influence of buoyancy ratio parameter on temperature distribution field increases by increasing the amount of buoyancy ratio parameter. The nanoparticle concentration field decreases if the reciprocal of the magnetic Prandtl number rises in Figure 27. Figure 28 exhibits the outcome of $\beta$ on the concentration field. The figure of concentration inclines when the number of $\beta$ grows. Figure 29 portrays the result of buoyancy ratio characteristics on nanoparticle concentration. It is easily observed that the field of concentration reduces if the number of $N_{r}$ increases. Figures 30 and 31 define the matching outcomes on the concentration profile. When the size of $\operatorname{Pr}$ and $\eta$ expands, the nanoparticle concentration grows. Streamlines graphs against the distinct amount of $\alpha_{1}$ and $\varepsilon$ are shown in Figures 32-35. Table 1 shows the impact of different characteristics on drag friction $C_{f} \operatorname{Re}_{x}^{(1 / 2)}$. The Nusselt number for noticeable amounts of $R_{d}, \mathrm{Nt}, \mathrm{Nb}, \varepsilon, \lambda_{1}$, $m_{1}, \beta$, and $\alpha_{1}$ is analyzed and characterized in Table 2. Table 3 portrays the deviation of different amounts of parameters on the local Sherwood number.

## 4. Concluding Remarks

We studied the molecular theory of liquid-originated non-Newtonian nanofluids which are commonly used in heat transfer devices such as heat exchangers, engine oils, electrical cooling systems (such as flat plates), nuclear


Figure 4: Upshot of $\beta$ on $h^{\prime}(\gamma)$.


Figure 5: Results of $\lambda_{1}$ on $h^{\prime}(\gamma)$.


Figure 6: Effects of Nb on $h^{\prime}(\gamma)$.


Figure 7: Consequences of Nt on $h^{\prime}(\gamma)$.


Figure 8: Upshot of $\operatorname{Pr}$ on $h^{\prime}(\gamma)$.


Figure 9: Consequence of $R_{d}$ on $h^{\prime}(\gamma)$.


Figure 10: Upshot of $\varepsilon$ on $h^{\prime}(\gamma)$.


Figure 11: Result of $\beta$ on $F^{\prime}(\gamma)$.


Figure 12: Outcome of $\lambda_{1}$ on $F^{\prime}(\gamma)$.


Figure 13: Result of Nb on $F^{\prime}(\gamma)$.


Figure 14: Impact of Nt on $F^{\prime}(\gamma)$.


Figure 15: Upshot of $\varepsilon$ on $F^{\prime}(\gamma)$.


Figure 16: Influence of $\alpha_{1}$ on $F^{\prime}(\gamma)$.


Figure 17: Outcome of $\eta$ on $F^{\prime}(\gamma)$.


Figure 18: Upshot of $\alpha_{1}$ on $\theta(\gamma)$.


Figure 19: Result of $\beta$ on $\theta(\gamma)$.



Figure 20: Outcome of Le on $\theta(\gamma)$.



Figure 21: Influence of Nb on $\theta(\gamma)$.


Figure 22: Deviation of Nt on $\theta(\gamma)$.


Figure 23: Consequence of $\operatorname{Pr}$ on $\theta(\gamma)$.


Figure 24: Influence of $\eta$ on $\theta(\gamma)$.


Figure 25: Conclusion of $R_{d}$ on $\theta(\gamma)$.


$$
\begin{array}{ll}
-N_{r}=0 & -N_{r}=2 \\
N_{r}=1 & -N_{r}=3
\end{array}
$$

Figure 26: Result of $N_{r}$ on $\theta(\gamma)$.


$$
\begin{array}{ll}
-\alpha_{1}=0.3 \\
\alpha_{1}=0.4 & -\alpha_{1}=0.6 \\
\alpha_{1}=100
\end{array}
$$

Figure 27: Outcome of $\alpha_{1}$ on $\theta(\gamma)$.


Figure 28: Upshot of $\beta$ on $\theta(\gamma)$.


Figure 29: Effect of $N_{r}$ on $\phi(\gamma)$.


Figure 30: Impact of $\operatorname{Pr}$ on $\phi(\gamma)$.



Figure 32: Flowlines for $\alpha_{1}=0.5$.


Figure 33: Flowlines for $\alpha_{1}=1.5$.


Figure 34: Flowlines for $\varepsilon=0.1$.


Figure 35: Flowlines for $\varepsilon=1$.

Table 1: Variation of $C_{f}\left(\operatorname{Re}_{x}\right)^{(1 / 2)}$ for distinct amounts of nondimensional parameters.

| $R_{d}$ | Nt | Nb | $\varepsilon$ | Pr | $\lambda_{1}$ | $m_{1}$ | $\eta$ | $N_{r}$ | Le | $\beta$ | $\alpha_{1}$ | $C_{f}\left(\mathrm{Re}_{x}\right)^{(1 / 2)}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.1 | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | $\begin{aligned} & -0.85974 \\ & -0.87093 \\ & -0.88148 \end{aligned}$ |
| 0.3 |  |  |  |  |  |  |  |  |  |  |  |  |
| 0.5 |  |  |  |  |  |  |  |  |  |  |  |  |
|  | 0.2 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -0.93303 |
| 0.1 | 0.3 |  |  |  |  |  |  |  |  |  |  | -1.00570 |
|  | 0.5 |  |  |  |  |  |  |  |  |  |  | -1.14835 |
|  | 0.1 | 0.2 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -0.82344 |
| 0.1 |  | 0.3 |  |  |  |  |  |  |  |  |  | -0.81182 |
|  |  | 0.4 |  |  |  |  |  |  |  |  |  | -0.80635 |
|  | 0.1 | 0.1 | 1.2 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -1.43201 |
| 0.1 |  |  | 1.3 |  |  |  |  |  |  |  |  | -1.40042 |
|  |  |  | 1.4 |  |  |  |  |  |  |  |  | -1.37091 |
|  | 0.1 | 0.1 | 1 | 7 | 0.2 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -0.83200 |
| 0.1 |  |  |  |  | 0.3 |  |  |  |  |  |  | -0.80432 |
|  |  |  |  |  | 0.4 |  |  |  |  |  |  | -0.77670 |

Table 1: Continued.

| $R_{d}$ | Nt | Nb | $\varepsilon$ | $\operatorname{Pr}$ | $\lambda_{1}$ | $m_{1}$ | $\eta$ | $N_{r}$ | Le | $\beta$ | $\alpha_{1}$ | $C_{f}\left(\mathrm{Re}_{x}\right)^{(1 / 2)}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.2 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -0.84285 |
| 0.1 |  |  |  |  |  | 0.4 |  |  |  |  |  | -0.84659 |
|  |  |  |  |  |  |  | 0.6 |  |  |  |  | -0.85046 |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.2 | 0.5 | -0.66689 |
| 0.1 |  |  |  |  |  |  |  |  |  | 0.4 |  | -0.09788 |
|  |  |  |  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  | 0.1 | 0.1 | 0.1 | 0.1 | 0.2 |
| 0.1 |  |  |  |  |  |  |  |  | 0.4 | -0.84019 |  |  |
|  |  |  |  |  |  |  |  |  | 0.85400 |  |  |  |

Table 2: Variation of $\mathrm{Nu}_{x}\left(\operatorname{Re}_{x}\right)^{-(1 / 2)}$ for distinct amounts of nondimensional parameters.

| $R_{d}$ | Nt | Nb | $\varepsilon$ | Pr | $\lambda_{1}$ | $m_{1}$ | $\eta$ | $N_{r}$ | Le | $\beta$ | $\alpha_{1}$ | $\mathrm{Nu}_{x}\left(\mathrm{Re}_{x}\right)^{-(1 / 2)}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.1 | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | 3.68235 |
| 0.3 |  |  |  |  |  |  |  |  |  |  |  | 3.46053 |
| 0.5 |  |  |  |  |  |  |  |  |  |  |  | 3.25753 |
|  | 0.2 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | 2.10453 |
| 0.1 | 0.3 |  |  |  |  |  |  |  |  |  |  | 2.04608 |
|  | 0.5 |  |  |  |  |  |  |  |  |  |  | 1.91548 |
|  | 0.1 | 0.2 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | 2.21832 |
| 0.1 |  | 0.3 |  |  |  |  |  |  |  |  |  | 2.27728 |
|  |  | 0.4 |  |  |  |  |  |  |  |  |  | 2.33814 |
|  | 0.1 | 0.1 | 1 | 9 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | 2.46411 |
| 0.1 |  |  |  | 12 |  |  |  |  |  |  |  | 2.86748 |
|  |  |  |  | 15 |  |  |  |  |  |  |  | 3.22622 |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.2 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | 2.15845 |
| 0.1 |  |  |  |  |  | 0.4 |  |  |  |  |  | 2.15809 |
|  |  |  |  |  |  | 0.6 |  |  |  |  |  | 2.15774 |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.2 | 0.5 | 2.18123 |
| 0.1 |  |  |  |  |  |  |  |  |  | 0.4 |  | 2.24457 |
|  |  |  |  |  |  |  |  |  |  | 0.6 |  | 2.32438 |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.2 | 2.16042 |
| 0. |  |  |  |  |  |  |  |  |  |  | 0.4 | 2.15919 |
|  |  |  |  |  |  |  |  |  |  |  | 0.6 | 2.15807 |

Table 3: Variation of $\mathrm{Sh}_{x}\left(\mathrm{Re}_{x}\right)^{-(1 / 2)}$ for distinct amounts of nondimensional parameters.

| $R_{d}$ | Nt | Nb | $\varepsilon$ | Pr | $\lambda_{1}$ | $m_{1}$ | $\eta$ | $N_{r}$ | Le | $\beta$ | $\alpha_{1}$ | $\mathrm{Sh}_{x}\left(\mathrm{Re}_{x}\right)^{-(1 / 2)}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.1 | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | $\begin{aligned} & -1.93921 \\ & -2.20711 \\ & -2.43144 \end{aligned}$ |
| 0.3 |  |  |  |  |  |  |  |  |  |  |  |  |
| 0.5 |  |  |  |  |  |  |  |  |  |  |  |  |
| 0.1 | 0.2 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -3.70463 |
|  | 0.3 |  |  |  |  |  |  |  |  |  |  | -5.34866 |
|  | 0.5 |  |  |  |  |  |  |  |  |  |  | -8.21711 |
| 0.1 | 0.1 | 0.2 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -1.03388 |
|  |  | 0.3 |  |  |  |  |  |  |  |  |  | -0.73383 |
|  |  | 0.4 |  |  |  |  |  |  |  |  |  | -0.58575 |
|  | 0.1 | 0.1 | 1 | 9 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -2.21493 |
| 0.1 |  |  |  | 12 |  |  |  |  |  |  |  | -2.57936 |
|  |  |  |  | 15 |  |  |  |  |  |  |  | -2.90386 |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.2 | 0.5 | 0.1 | 0.1 | 0.1 | 0.5 | -1.93906 |
| 0.1 |  |  |  |  |  | 0.4 |  |  |  |  |  | -1.93874 |
|  |  |  |  |  |  | 0.6 |  |  |  |  |  | -1.93842 |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.2 | 0.5 | -1.95958 |
| 0.1 |  |  |  |  |  |  |  |  |  | 0.4 |  | -2.01667 |
|  |  |  |  |  |  |  |  |  |  | 0.6 |  | -2.08858 |
|  | 0.1 | 0.1 | 1 | 7 | 0.1 | 0.1 | 0.5 | 0.1 | 0.1 | 0.1 | 0.2 | -1.94083 |
| 0. |  |  |  |  |  |  |  |  |  |  | 0.4 | -1.93972 |
|  |  |  |  |  |  |  |  |  |  |  | 0.6 | -1.93872 |

reactors, biomedicine lubricants, and radiators. The key points of observation in the recent analysis are as follows:
(i) Induced magnetic pattern near the wall declines, and far away, it inclines when $\left(\alpha_{1}\right)$ intensifies. The variation of $h^{\prime}(\gamma)$ field near the wall goes up and very far away it decays when the size $\left(R_{d}\right)$ ascends.
(ii) $h^{\prime}(\gamma)$ (nondimensionless induced magnetic function) falls, whereas $(\eta)$, Brownian diffusion ( Nb ), and Prandtl number (Pr) rise. The field $h^{\prime}(\gamma)$ expands by enhancing the amount of magnetic parameter $(\beta)$, mixed convection $\left(\lambda_{1}\right)$, thermophoresis parameter ( Nt ), and fluid parameter $(\varepsilon)$. The variation of $h^{\prime}(\gamma)$ profile near the wall moves upward and very far away it moves down when the size $\left(R_{d}\right)$ ascends.
(iii) The velocity amplitude expands by enlargement in the amount of magnetic parameter $(\beta)$, mixed convection $\left(\lambda_{1}\right)$, thermophoresis parameter $(\mathrm{Nt})$, fluid characteristic $(\varepsilon)$, and stretching parameter $(\eta) \cdot F^{\prime}(\gamma)$ collapses by Brownian motion ( Nb ) and $\left(\alpha_{1}\right)$.
(iv) The temperature spectrum increases when the values of $\left(\alpha_{1}\right)$, radiation parameter $\left(R_{d}\right)$, and buoyancy ratio $\left(N_{r}\right)$ increases and decreases by Prandtl number, magnetic parameter $(\beta)$, Brownian motion diffusion $(\mathrm{Nb})$, and stretching parameter $(\eta)$. $\theta(\gamma)$ (nondimensional temperature function) rises near the wall, and far away, it diminishes when there is increase in the values of Le.
(v) The nanoparticle concentration portrait reduces if the reciprocal of the magnetic Prandtl number $\left(\alpha_{1}\right)$ and $\left(N_{r}\right)$ rises. Concentration enlarges when the number of $(\beta),(\operatorname{Pr})$, and $(\eta)$ grows.
(vi) Drag friction decays by the inclination in the values of $\left(R_{d}\right),(N t),\left(m_{1}\right)$, and $\left(\alpha_{1}\right)$. When inclining the amount of $(\mathrm{Nb}),(\varepsilon),\left(\lambda_{1}\right)$, and $(\beta)$, drag force expands.
(vii) Heat transfer rates are increased when there is an increase in the values of $(\mathrm{Nb}),(\operatorname{Pr})$, and $(\beta)$ and decrease when there is an increase in the values of $\left(R_{d}\right),(\mathrm{Nt}),\left(m_{1}\right)$, and $\left(\alpha_{1}\right)$.
(viii) Mass transfer rates diminish, for $\left(R_{d}\right),(\mathrm{Nt}),(\operatorname{Pr})$, and $(\beta)$, but increases by $(\mathrm{Nb}),\left(m_{1}\right)$, and $\left(\alpha_{1}\right)$.

[^0]$(u, v)$ : Velocity components $\left(\mathrm{ms}^{-1}\right)$
$m_{1}$ : Fluid characteristic (-)
$\tau: \quad$ Extra stress tensor (-)
$\varepsilon: \quad$ Fluid characteristics (-)
$\mathrm{Sh}_{x}$ : Local Sherwood number (-)
$R_{d}: \quad$ Radiation parameter (-)
$\lambda_{1}$ : Mixed convection parameter (-)
$\rho_{f}$ : Density of the base fluid $\left(\mathrm{kg} \mathrm{m}^{-3}\right)$
$q_{w}$ : Surface heat flux (-)
np: $\quad$ Nanoparticle (-)
$\alpha$ : Thermal diffusivity $\left(\mathrm{m}^{2} \mathrm{~s}^{-1}\right)$
$\alpha_{1}^{*}: \quad$ Magnetic diffusivity (-)
$\theta$ : $\quad$ Dimensionless heat transfer function (-)
$\gamma$ : Similarity variable (-)
$\rho: \quad$ Density $\left(\mathrm{kg} \mathrm{m}^{-3}\right)$
$\beta$ : Magnetic parameter (-)
tr: Trace (-)
$C_{w}$ : Hot fluid concentration $\left(\mathrm{kg} \mathrm{m}^{-3}\right)$
$A_{1}$ : Rivlin-Ericksen tensor (-)
$C_{f}$ : Skin friction coefficient (-)
$D_{B}$ : Brownian diffusion coefficient $\left(\mathrm{m}^{2} \mathrm{~s}^{-1}\right)$
$D_{T}$ : Thermophoresis diffusion coefficient $\left(\mathrm{m}^{2} \mathrm{~s}^{-1}\right)$
$g: \quad$ Gravity acceleration $\left(\mathrm{ms}^{-2}\right)$
Le: Lewis number (-)
Nt : Thermophoresis parameter (-)
$\mathrm{Re}_{x}$ : Local Reynolds number (-)
$\sigma^{*}: \quad$ Stefan-Boltzmann constant $\left(\mathrm{Wm}^{-2} \mathrm{~K}^{-4}\right)$
$T_{\infty}$ : Ambient temperature (K)
$u_{w}: \quad$ Stretching sheet velocity $\left(\mathrm{ms}^{-1}\right)$
$u_{\infty}$ : External flow velocity $\left(\mathrm{ms}^{-1}\right)$
$(x, y)$ : Cartesian coordinate components ( $m$ )
$\delta: \quad$ Fluid characteristics ( - )
$c_{1}$ : Fluid parameter (-)
Pr: Prandtl number (-)
$N_{r}$ : Buoyancy ratio characteristics (-)
$\tau_{w}$ : Surface shear stress (-)
$q_{m}$ : $\quad$ Surface mass flux ( - )
$f: \quad$ Base fluid (-)
$\eta: \quad$ Stretching parameter (-)
$h$ : Dimensionless magnetic function (-)
$\varphi$ : Dimensionless concentration function (-)
$\rho_{p}: \quad$ The density of the nanoparticles $\left(\mathrm{kg} \mathrm{m}^{-3}\right)$.

## Data Availability

The data that support the findings of this study are available from the corresponding author upon request.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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# The Mixture of the Marshall-Olkin Extended Weibull Distribution under Type-II Censoring and Different Loss Functions 

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#### Abstract

To study the heterogeneous nature of lifetimes of certain mechanical or engineering processes, a mixture model of some suitable lifetime distributions may be more appropriate and appealing as compared to simple models. This paper considers a mixture of the Marshall-Olkin extended Weibull distribution for efficient modeling of failure, survival, and COVID-19 data under classical and Bayesian perspectives based on type-II censored data. We derive several properties of the new distribution such as moments, incomplete moments, mean deviation, average lifetime, mean residual lifetime, Rényi entropy, Shannon entropy, and order statistics of the proposed distribution. Maximum likelihood and Bayes procedure are used to derive both point and interval estimates of the parameters involved in the model. Bayes estimators of the unknown parameters of the model are obtained under symmetric (squared error) and asymmetric (linear exponential (LINEX)) loss functions using gamma priors for both the shape and the scale parameters. Furthermore, approximate confidence intervals and Bayes credible intervals (CIs) are also obtained. Monte Carlo simulation study is carried out to assess the performance of the maximum likelihood estimators and Bayes estimators with respect to their estimated risk. The flexibility and importance of the proposed distribution are illustrated by means of four real datasets.


## 1. Introduction

In the history of statistics, the use of finite mixture models is very old. They were particularly used to model population heterogeneity, generalize distributional assumptions, clustering and classification, etc. The concept of the finite mixture distribution was pioneered by Newcomb [1] for modeling outliers. Eight years later, Pearson [2] considered a mixture of two univariate Gaussian distributions to estimate the parameters of the model using the method of moments (MOM) to analyze a dataset containing ratios of the forehead to body lengths for 1,000 crabs. Since then, several authors have studied finite mixture models under different scenarios. Mendenhall and Hader [3] considered exponentially
distributed failure time distributions based on censored lifetime data to estimate the model parameters using the maximum likelihood method. In their study, they divided the failure population into two subpopulations, each representing a different cause or type of failure. Radhakrishna et al. [4] considered both moment and maximum likelihood estimators of the unknown parameters of a two-component mixture of generalized gamma distribution. Ahmed et al. [5] obtained approximate Bayes estimators for parameters of the mixture of two Weibull distributions under type-II censoring. Al-Hussaini et al. [6] applied both maximum likelihood and Bayes estimation methods on a two-component mixture of the Gompertz distribution based on type-I and type-II censoring. Jaheen [7] adopted both maximum
likelihood and Bayesian approach to discuss the problem of estimating the parameters using the finite mixture of two exponential distributions based on record statistics. Shawky and Bakoban [8] adopted both maximum likelihood and Bayesian approach to estimate the parameters of the model, reliability, and failure rate functions of two-component finite mixtures of exponentiated gamma distribution. AbuZinadah [9] used maximum likelihood and Bayes estimation methods to estimate the parameters, reliability, and hazard functions of a mixture of exponentiated Pareto and exponential distribution under complete and type-II censoring schemes. Prakash [10] adopted the Bayes method to estimate the parameters from the mixture of two Weibull distributions based on informative and noninformative priors. Zhang et al. [11] proposed a mixture Weibull proportional hazard model to predict the failure of a mechanical system with multiple failure modes. They estimated the mixed model parameters by combining historical lifetime and monitoring data of all failure modes. ALgfary [12] introduced the finite mixture of two exponentiated Kumaraswamy (MEKum) distributions and obtained the maximum likelihood estimates for the vector of the parameters of the MEKum distribution. Adham and ALgfary [13] adopted the Bayesian approach to estimate the vector of parameters of the finite mixture of two-component exponentiated Kumaraswamy distribution. They also obtained Bayesian predictive density functions of future observations from the MEKum distribution. Ateya and Al Khald [14] studied the finite mixture of truncated generalized Cauchy distribution based on type-I, type-II, and progressively typeII censored samples. Aslam et al. [15] studied the threecomponent mixture of exponential, Rayleigh, Pareto, and Burr type-XII distributions in relation to reliability analysis. Tahir et al. [16] studied the properties of the three-component mixture of Rayleigh distributions based on doubly censored lifetime data. Kalantan and Alrewely [17] studied the two-component Laplace mixture model and obtained the estimates of the parameters using maximum likelihood and method of moments. Recently, Tahir et al. [18] also studied the three-component mixture of exponential distributions from the Bayesian perspective based on type-II doubly censoring sampling scheme. Kharazmi et al. [19] studied the 2-component mixture of Topp-Leone distribution and obtained classical and Bayes estimators based on the complete sample and the references cited therein.

Wide applicability of mixture modeling motivates us to develop a two-component mixture of Marshall-Olkin extended Weibull distribution for efficient modeling of breaking stress of carbon fibers, survival times in days of 72 guinea pigs infected datasets, survival times in weeks of 33 patients suffering from acute myelogenous leukemia, and COVID-19 data belonging to Canada of 36 days. In this article, the primary objective of the paper is twofold. First, we obtain maximum likelihood estimators and corresponding approximate confidence intervals (CIs) of the unknown parameters of the 2-component mixture of the Marshall-Olkin extended Weibull (MOEW) distribution for type-II censored data. Next, we consider the Bayes estimation method. The Bayes estimators have been derived and
evaluated under the assumption of two loss functions using independent gamma priors. Symmetric $100(1-\tau) \%$ twosided Bayes credible intervals are also obtained, and they are compared with classical CIs. To the best of our knowledge, the 2-component mixture of the MOEW distribution is not discussed before using the aforementioned methods of estimation. Through this paper, we purport to provide some guidelines on selecting the best estimator that may be of significant interest to applied statisticians/practitioners/ engineers.

The organization of this paper is as follows. The description of the model along with basic properties is reported in Section 2. We use the maximum likelihood estimation method based on type-II censoring as a part of frequentist methodology for parameter estimation in Section 3. We have also taken into account approximate CIs in the same section. In Section 4, we have derived the Bayes estimators of the unknown parameters of the model under squared error loss (SEL) and linear exponential (LINEX) loss functions using gamma priors for both scale and shape parameters. We have also obtained two-sided Bayes probability intervals in the same section. The simulation study is carried out in Section 5. For illustrative purposes, four real datasets are analyzed in Section 6. Finally, concluding remarks are presented in Section 7.

## 2. Model Description

The probability density function (pdf) of the Marshal-l-Olkin extended Weibull distribution (MOEW) for a random variable $X$ is defined by (see Ghitany et al. [20] and Zhang and Xie [21])

$$
\begin{equation*}
f(x)=\frac{\alpha \lambda \theta x^{\theta-1} e^{-\lambda x^{\theta}}}{\left(1-\bar{a} e^{-\lambda x^{\theta}}\right)^{2}}, \quad x \geq 0, \alpha, \lambda, \theta>0 \tag{1}
\end{equation*}
$$

where $\bar{a}=1-a$, and the cumulative distribution function (cdf) of the distribution is

$$
\begin{equation*}
F(x)=\frac{1-e^{-\lambda x^{\theta}}}{1-\bar{\alpha} e^{-\lambda x^{\theta}}}, \quad x \geq 0 \tag{2}
\end{equation*}
$$

The hazard rate function of MOEW takes the form

$$
\begin{equation*}
h(x)=\frac{\lambda \theta x^{\theta-1}}{1-\bar{\alpha} e^{-\lambda x^{\theta}}}, \quad x \geq 0 \tag{3}
\end{equation*}
$$

A density function for the mixture of two components' densities with mixing proportions $\left(p_{j}, j=1,2\right)$ is

$$
\begin{align*}
g(x) & =\sum_{j=1}^{2} p_{j} f_{j}(x), \quad \text { where } f_{j}(x)  \tag{4}\\
& =\alpha_{j} \theta_{j} \lambda_{j} x^{\theta_{j}-1} e^{-\lambda_{j} x^{\theta_{j}}}\left(1-\bar{\alpha}_{j} e^{\lambda_{j} x^{\theta_{j}}}\right)^{-2}
\end{align*}
$$

where $p_{j}$ are the mixing proportions, satisfying the conditions $\sum_{j=1}^{2} p_{j}=1$ and $p_{j} \geq 0$; all of them are unknowns, and the pdf is plotted in Figure 1. The cdf for the mixture model is

$$
\begin{equation*}
G(x)=\sum_{j=1}^{2} p_{j} F_{j}(x) \tag{5}
\end{equation*}
$$

where

$$
\begin{equation*}
F_{j}(x)=\frac{\left(1-e^{-\lambda_{j} x^{\theta_{j}}}\right)}{\left(1-\bar{\alpha} e^{\lambda_{j} x^{\theta_{j}}}\right)} \tag{6}
\end{equation*}
$$

The reliability function for the mixture model is

$$
\begin{equation*}
S(x)=\sum_{j=1}^{2} p_{j} R_{j}(x) \tag{7}
\end{equation*}
$$

where

$$
\begin{equation*}
R_{j}(x)=1-\frac{\left(1-e^{-\lambda_{j} x^{\theta_{j}}}\right)}{\left(1-\bar{\alpha} e^{\lambda_{j} x^{\theta_{j}}}\right)} \tag{8}
\end{equation*}
$$

Due to exponentiation of each component by a positive integer, the model is so flexible that shows different shapes of hazard rate function (hrf) of the mixture which is given by

$$
\begin{align*}
& H(x)=\sum_{j=1}^{2} p_{j} h_{j}(x) \\
& h_{j}(x)=\frac{\alpha_{j} \theta_{j} \lambda_{j} x^{\theta_{j}-1} e^{-\lambda_{j} x^{\theta_{j}}}\left(1-\bar{\alpha}_{j} e^{\lambda_{j} x^{\theta_{j}}}\right)}{\left(1-\bar{\alpha} e^{\lambda_{j} x^{\theta_{j}}}\right)-\left(1-\bar{\alpha} e^{\lambda_{j} x^{\theta_{j}}}\right)\left(1-e^{-\lambda_{j} x^{\theta_{j}}}\right)} \tag{9}
\end{align*}
$$

## 3. General Properties of the MOEW Distribution

3.1. Moments. The $r^{\text {th }}$ moment of a finite mixture of the 2 component MOEW distribution is given by

$$
\begin{align*}
\mu_{r}^{\prime}= & \sum_{j=1}^{2} p_{j} \mu_{r}=\int_{0}^{\infty} p_{1} \alpha_{1} \theta_{1} \lambda_{1} x^{\theta_{1}+r-1} e^{-\lambda_{1}}\left(1-\bar{\alpha}_{1} e^{\lambda_{1} x^{\theta_{1}}}\right)^{-2} \\
& +\int_{0}^{\infty}\left(1-p_{1}\right) \alpha_{2} \theta_{2} \lambda_{2} x^{\theta_{2}+r-1} e^{-\lambda_{2}}\left(1-\bar{\alpha}_{2} e^{\lambda_{2} x^{\theta_{2}}}\right)^{-2} \\
= & \frac{p_{1}}{\alpha_{1}}\left[\lambda_{1}\left(-\lambda_{1}\right)^{\left(-r / \theta_{1}\right)} \Gamma\left(\frac{r+\theta_{1}}{\theta_{1}}\right)\right] \\
& +\frac{\left(1-p_{1}\right)}{\alpha_{2}}\left[\lambda_{2}\left(-\lambda_{2}\right)^{\left(-r / \theta_{2}\right)} \Gamma\left(\frac{r+\theta_{2}}{\theta_{2}}\right)\right] \tag{10}
\end{align*}
$$

where $p_{2}=\left(1-p_{1}\right)$.
When $r=1$, the mean is given by

$$
\begin{align*}
\mu_{1}^{\prime}= & \frac{p_{1}}{\alpha_{1}}\left[\lambda_{1}\left(-\lambda_{1}\right)^{\left(-1 / \theta_{1}\right)} \Gamma\left(\frac{1+\theta_{1}}{\theta_{1}}\right)\right] \\
& +\frac{\left(1-p_{1}\right)}{\alpha_{2}}\left[\lambda_{2}\left(-\lambda_{2}\right)^{\left(-1 / \theta_{2}\right)} \Gamma\left(\frac{1+\theta_{2}}{\theta_{2}}\right)\right] \tag{11}
\end{align*}
$$

The moment-generating function of the mixture MOEW distribution is given by

$$
\begin{align*}
\mu_{x}(t) & =\int_{0}^{\infty} e^{t x} f_{j}(x) \mathrm{d} x \\
& =\frac{\left(p_{1} \theta_{1} \lambda_{1} x^{\theta_{1}} e^{-\lambda_{1} x^{\theta_{1}}} / \alpha_{1}\right)+\left(\left(1-p_{1}\right) \theta_{2} \lambda_{2} x^{\theta_{2}} e^{-\lambda_{1} x^{\theta_{2}}} / \alpha_{2}\right)}{x^{2}} . \tag{12}
\end{align*}
$$

3.2. Incomplete Moments. The $r^{\text {th }}$ incomplete moment of a finite mixture of the 2-component mixture distribution is given by

$$
\begin{align*}
T_{r}(z) & =E\left(x^{r}\right)=\int_{0}^{z} x^{r} f(x) \mathrm{d} x=\sum_{j=1}^{2} p_{j} \int_{0}^{z} x^{r} f_{j}(x) \mathrm{d} x \\
& =\sum_{j=1}^{2} p_{j} \mu_{j r}, \tag{13}
\end{align*}
$$

where $T_{j r}=\int_{0}^{z} x^{r} f_{j}(x) \mathrm{d} x$ is the $r^{\text {th }}$ incomplete moment of the $j^{\text {th }}$ component so that the $r^{\text {th }}$ incomplete moment of a finite mixture of the 2-component MOEW distribution is given by

$$
\begin{align*}
T_{r}(z)= & \sum_{j=1}^{2} p_{j}=\int_{0}^{z} p_{1} \alpha_{1} \theta_{1} \lambda_{1} x^{\theta_{1}+r-1} e^{-\lambda_{1}}\left(1-\bar{\alpha}_{1} e^{\lambda_{1} x^{\theta_{1}}}\right)^{-2} \\
& +\int_{0}^{z}\left(1-p_{1}\right) \alpha_{2} \theta_{2} \lambda_{2} x^{\theta_{2}+r-1} e^{-\lambda_{2}}\left(1-\bar{\alpha}_{2} e^{\lambda_{2} x^{\theta_{2}}}\right)^{-2} \mathrm{~d} x \tag{14}
\end{align*}
$$

The first incomplete moment of a finite mixture of the 2component MOEW distribution is given by

$$
\begin{align*}
T_{1}(z)= & \sum_{j=1}^{2} p_{j} \mu_{j 1}=\int_{0}^{z} p_{1} \alpha_{1} \theta_{1} \lambda_{1} x^{\theta_{1}} e^{-\lambda_{1}}\left(1-\bar{\alpha}_{1} e^{\lambda_{1} x^{\theta_{1}}}\right)^{-2} \\
& \cdot \int_{0}^{z}\left(1-p_{1}\right) \alpha_{2} \theta_{2} \lambda_{2} x^{\theta_{2}} e^{-\lambda_{2}}\left(1-\bar{\alpha}_{2} e^{\lambda_{2} x^{\theta_{2}}}\right)^{-2} \mathrm{~d} x \tag{15}
\end{align*}
$$

3.3. Mean Deviations. The mean deviations of the random variable $X$ about the mean, $\mu$, and the median, $M$, are given, respectively, by


Figure 1: pdf of the two-component mixture of the Marshall-Olkin extended Weibull distribution for different values of the parameters, where $c 1=\left(\alpha 1=2, \theta 1=3, \lambda 1=1, \alpha 2=2, \theta 2=3, \lambda 2=1, p_{1}=0.4\right), c 2=\left(\alpha 1=1, \theta 1=5, \lambda 1=2, \alpha 2=2, \theta 2=3, \lambda 2=5, p_{1}=0.3\right)$, and $c 3=$ $\left(\alpha 1=1, \theta 1=5, \lambda 1=2, \alpha 2=4, \theta 2=3, \lambda 2=5, p_{1}=0.2\right)$.

$$
\begin{aligned}
& \delta 1=\int_{x}|x-\mu| f(x, \alpha, \lambda, \theta) \mathrm{d} x \\
& \delta 2=\int_{x}|x-M| f(x, \alpha, \lambda, \theta) \mathrm{d} x
\end{aligned}
$$

where $\delta 1=p_{1} \delta 1_{j}+\left(1-p_{1}\right) \delta 1_{j}$ and $\delta 2=p_{1} \delta 2_{j}+\left(1-p_{1}\right)$ $\delta 2_{j} \quad$ since $\quad \delta 1_{j}=\int_{0}^{\infty}|x-\mu| f_{j}\left(x, \alpha_{j}, \lambda_{j}, \theta_{j}\right) \mathrm{d} x \quad$ and $\delta 2_{j}=\int_{0}^{\infty}|x-M| f_{j}\left(x, \alpha_{j}, \lambda_{j}, \theta_{j}\right) \mathrm{d} x$.

$$
\begin{align*}
\delta 1 & =\int_{0}^{\infty}|x-\mu| f(x, \alpha, \lambda, \theta) \mathrm{d} x, \\
\delta 1 & =\int_{0}^{\infty}|x-\mu|\left(p_{1}\left\{\frac{\alpha_{1} \lambda_{1} \theta_{1}(x)^{\theta_{1}-1} e^{-\lambda_{1}(x)^{\theta_{1}}}}{\left(1-\overline{\alpha_{1}} e^{-\lambda_{1}(x)^{\theta_{1}}}\right)^{2}}\right\}+\left(1-p_{1}\right)\left\{\frac{\alpha_{2} \lambda_{2} \theta_{2}(x)^{\theta_{2}-1} e^{-\lambda_{2}(x)^{\theta_{2}}}}{\left(1-\overline{\alpha_{2}} e^{-\lambda_{2}(x)^{\theta_{2}}}\right)^{2}}\right\}\right) \mathrm{d} x \\
& =p_{1}\left\{\frac{\alpha_{1} \lambda_{1} \theta_{1}(x)^{\theta_{1}-1} e^{-\lambda_{1}(x)^{\theta_{1}}}}{\left(1-\overline{\alpha_{1}} e^{-\lambda_{1}(x)^{\theta_{1}}}\right)^{2}}\right\}+\left(1-p_{1}\right)\left\{\frac{\alpha_{2} \lambda_{2} \theta_{2}(x)^{\theta_{2}-1} e^{-\lambda_{2}(x)^{\theta_{2}}}}{\left(1-\overline{\alpha_{2}} e^{-\lambda_{2}(x)^{\theta_{2}}}\right)^{2}}\right\},  \tag{17}\\
\delta 2 & =\int_{0}^{\infty}|x-M|\left(p_{1}\left\{\frac{\alpha_{1} \lambda_{1} \theta_{1}(x)^{\theta_{1}-1} e^{-\lambda_{1}(x)^{\theta_{1}}}}{\left(1-\overline{\alpha_{1}} e^{-\lambda_{1}(x)^{\theta_{1}}}\right)^{2}}\right\}+\left(1-p_{1}\right)\left\{\frac{\alpha_{2} \lambda_{2} \theta_{2}(x)^{\theta_{2}-1} e^{-\lambda_{2}(x)^{\theta_{2}}}}{\left(1-\overline{\alpha_{2}} e^{-\lambda_{2}(x)^{\theta_{2}}}\right)^{2}}\right\}\right) \mathrm{d} x .
\end{align*}
$$

The median follows from the nonlinear equation $F(x, \alpha, \lambda, \theta)=1 / 2$. So, these quantities reduce to

$$
\begin{align*}
& \delta 1 F(x, \alpha, \lambda, \theta) 2 \mu F(x, \alpha, \lambda, \theta)-2 T 1(\mu), \\
& \delta 2=\mu-2 T 1(M), \tag{18}
\end{align*}
$$

where $T 1(z)$ is the first incomplete moment of $X$ obtained from (15),

$$
\begin{align*}
& \delta 1_{j}=2 \mu_{j} F_{j}\left(x, \alpha_{j}, \lambda_{j}, \theta_{j}\right)-2 T 1\left(\mu_{j}\right), \\
& \delta 2_{j}=\mu_{j}-2 T 1\left(M_{j}\right), \tag{19}
\end{align*}
$$

where

$$
\begin{align*}
\begin{aligned}
\delta 1= & 2 \mu\left(p_{1}\left[\frac{1-e^{-\lambda_{1} x_{0}^{\theta_{1}}}}{1-\overline{\alpha_{1}} e^{-\lambda_{1} x_{0}^{\theta_{1}}}}\right]\right. \\
& +\left(\left(1-p_{1}\right)\left[\frac{1-e^{-\lambda_{2} x_{0}^{\theta_{2}}}}{1-\overline{\alpha_{2}} e^{-\lambda_{2} x_{0}^{\theta_{2}}}}\right]\right)-2 T 1(\mu), \\
\delta 1= & \mu-2 T 1(M) .
\end{aligned} .
\end{align*}
$$

3.4. Average Lifetime and Mean Residual Lifetime Functions. The average lifetime is given by

$$
\begin{align*}
t_{m}= & \int_{0}^{\infty}[1-F(x, \alpha, \lambda, \theta)] \mathrm{d} x, \\
t_{m}= & \int_{0}^{\infty}\left[1-\left\{p_{1}\left[\frac{1-e^{-\lambda_{1} x^{\theta_{1}}}}{1-\overline{\alpha_{1}} e^{-\lambda_{1} x^{\theta_{1}}}}\right]\right.\right.  \tag{21}\\
& \left.\left.+\left(1-p_{1}\right)\left[\frac{1-e^{-\lambda_{2} x^{\theta_{2}}}}{1-\overline{\alpha_{2}} e^{-\lambda_{2} x^{\theta_{2}}}}\right]\right\}\right] \mathrm{d} x .
\end{align*}
$$

The application of mean residual lifetime can be seen in the paper of Guess and Proschan [22]. The mean residual lifetime is given by

$$
\begin{align*}
m\left(x_{0}, \alpha, \lambda, \theta\right) & =E\left(X-x_{0} \mid X \geq x_{0}, \alpha, \lambda, \theta\right) \\
& =\int_{\left(X: X-x_{0}\right)}^{\infty} \frac{\left(X-x_{0}\right) f(x, \alpha,(k+1) \lambda, \theta)}{\operatorname{Pr}\left[X \geq x_{0}\right]} \mathrm{d} x=\left[\operatorname{Pr}\left[X \geq x_{0}\right]\right]^{-1} \int_{0}^{\infty} y f\left(x_{0}+y, \alpha, \lambda, \theta\right) \mathrm{d} y \\
& =\left[1-\left\{p_{1}\left[\frac{1-e^{-\lambda_{1} x_{0}^{\theta_{1}}}}{1-\overline{\alpha_{1}} e^{-\lambda_{1} x_{0}^{\theta_{1}}}}\right]+\left(1-p_{1}\right)\left[\frac{1-e^{-\lambda_{2} x_{0}^{\theta_{2}}}}{1-\overline{\alpha_{2}} e^{-\lambda_{2} x_{0}}}\right]\right\}\right]^{-1}  \tag{22}\\
& =\int_{0}^{\infty} y\left(p_{1}\left\{\frac{\alpha_{1} \lambda_{1} \theta_{1}\left(x_{0}+y\right)^{\theta_{1}-1} e^{-\lambda_{1}\left(x_{0}+y\right)^{\theta_{1}}}}{\left(1-\overline{\alpha_{1}} e^{-\lambda_{1}\left(x_{0}+y\right)^{\theta_{1}}}\right)^{2}}\right\}+\left(1-p_{1}\right)\left\{\frac{\alpha_{2} \lambda_{2} \theta_{2}\left(x_{0}+y\right)^{\theta_{2}-1} e^{-\lambda_{2}\left(x_{0}+y\right)^{\theta_{2}}}}{\left(1-\overline{\alpha_{2}} e^{-\lambda_{2}\left(x_{0}+y\right)^{\theta_{2}}}\right)^{2}}\right\}\right) \mathrm{d} y
\end{align*}
$$

Then, $m\left(x_{0}, \alpha, \lambda, \theta\right) \longrightarrow E(X)$ as $x_{0} \longrightarrow 0$.

$$
\begin{equation*}
H_{R}^{S}(x)=\frac{1}{1-S} \log \left(\int_{x} f(x, \alpha, \lambda, \theta)^{S} \mathrm{~d} x\right) \tag{23}
\end{equation*}
$$

3.5. Rényi Entropy. The Rényi entropy of $X$ with pdf (4) is given by

$$
\begin{equation*}
H_{R}^{S}(x)=\frac{1}{1-S} \log \left(\int_{0}^{\infty}\left(p_{1}\left\{\frac{\alpha_{1} \lambda_{1} \theta_{1}(x)^{\theta_{1}-1} e^{-\lambda_{1}(x)^{\theta_{1}}}}{\left(1-\overline{\alpha_{1}} e^{-\lambda_{1}(x)^{\theta_{1}}}\right)^{2}}\right\}+\left(1-p_{1}\right)\left\{\frac{\alpha_{2} \lambda_{2} \theta_{2}(x)^{\theta_{2}-1} e^{-\lambda_{2}(x)^{\theta_{2}}}}{\left(1-\overline{\alpha_{2}} e^{-\lambda_{2}(x)^{\theta_{2}}}\right)^{2}}\right\}\right)^{S} \mathrm{~d} x\right) \tag{24}
\end{equation*}
$$

It is difficult to obtain $H_{R}^{S}(x)$ in the closed form for a finite mixture of the 2 -component MOEW distribution.
3.6. Shannon Entropy. The Shannon entropy of $X$ is given by

$$
\begin{equation*}
H_{s}(x)=E_{x}\{-\log (f(x, \alpha, \lambda, \theta))\} . \tag{25}
\end{equation*}
$$

Thus, from (4), we can get the log-likelihood function as

$$
\begin{align*}
\log \left[f_{j}\left(x, \alpha_{j}, \lambda_{j}, \theta_{j}\right)\right]= & \log \left(\alpha_{j} \lambda_{j}\right)+\log \left[\theta_{j} x^{\theta_{j}-1}\right]-\lambda_{j} x^{\theta_{j}} \\
& -2 \log \left\{1-\bar{\alpha}_{j} \cdot \exp \left[-\lambda_{j} x^{\theta_{j}}\right]\right\} . \tag{26}
\end{align*}
$$

Thus, the above equation can be rewritten as

$$
\begin{align*}
H_{s}(x)= & -\log \left\{P \left[\left(\alpha_{1} \lambda_{1}\right)+2 E\left[\log \left\{1-\bar{\alpha}_{1} \exp \left[-\lambda_{1} x^{\theta_{1}}\right]\right\}\right]\right.\right. \\
& \left.-E_{x}\left\{\log \left(\theta_{1} x^{\theta_{1}-1}\right)\right\}+\lambda_{1} E\left(x^{\theta_{1}}\right)\right]+q\left[\left(\alpha_{2} \lambda_{2}\right)\right. \\
& +2 E\left[\log \left\{1-\bar{\alpha}_{2} \cdot \exp \left[-\lambda_{2} x^{\theta_{2}}\right]\right\}\right] \\
& \left.\left.-E_{x}\left\{\log \left(\theta_{2} x^{\theta_{2}-1}\right)\right\}+\lambda_{2} E\left(x^{\theta_{2}}\right)\right]\right\} . \tag{27}
\end{align*}
$$

3.7. Distribution of Order Statistics. The pdf of the $r^{\text {th }}$ order statistic for a random sample $x_{1}, x_{2}, x_{3}, \ldots, x_{n}$ from the MOEW distribution with pdf and cdf given in equations (1) and (2), respectively, is given by

$$
\begin{equation*}
f_{r, n}(x)=\frac{n!}{(n-r)!(r-1)!} f(X)[F(X)]^{r-1}[1-F(X)]^{n-r}, \tag{28}
\end{equation*}
$$

where

$$
\begin{equation*}
[1-F(X)]^{n-r}=\sum_{i=0}^{n-r}(-1)^{i}\binom{n-r}{i}[F(X)]^{i} \tag{29}
\end{equation*}
$$

$$
\begin{align*}
f_{r, n}(x)= & \frac{n!}{(n-r)!(r-1)!} \\
& \sum_{i=0}^{n-r}(-1)^{i}\binom{n-r}{i}\left\{\left[p_{1}\left\{\frac{\alpha_{1} \lambda_{1} \theta_{1} x^{\theta_{1}-1} e^{-\lambda_{1} x^{\theta_{1}}}\left(1-e^{-\lambda_{1} x^{\theta_{1}}}\right)^{r+i-1}}{\left(1-\overline{\alpha_{1}} e^{-\lambda_{1} x^{\theta_{1}}}\right)^{r+i+1}}\right\}\right]\right.  \tag{30}\\
& \left.+\left[\left(1-p_{1}\right)\left\{\frac{\alpha_{2} \lambda_{2} \theta_{2} x^{\theta_{2}-1} e^{-\lambda_{2} x^{\theta_{2}}}\left(1-e^{-\lambda_{2} x^{\theta_{2}}}\right)^{r+i-1}}{\left(1-\overline{\alpha_{2}} e^{-\lambda_{2} x^{\theta_{2}}}\right)^{r+i+1}}\right\}\right]\right\}
\end{align*}
$$

3.8. Maximum Likelihood Estimation Based on Type-II Censoring. Here, we discuss the maximum likelihood estimates of the unknown parameters of the 2-component mixture of the MOEW distributions. In a life testing experiment, $n$ items from the above mixture model are employed to test and get terminated when a preassigned number of items, say $r(<n)$, have failed. The samples obtained from such an experiment are called failure-censored

The $r^{\text {th }}$ order statistics for a finite mixture of the 2component MOEW distribution can be obtained by substituting equations (4) and (6) into (28); thus, we have

$$
\begin{align*}
& L\left(\theta_{1}, \theta_{2}, \lambda_{1}, \lambda_{2}, \alpha_{1}, \alpha_{2}, p_{1} \mid \underline{x}\right) \propto \prod_{i=1}^{r} f_{j}\left(x_{i}\right) R_{j}\left(x_{r}\right)^{(n-r)} \\
& \quad=p_{1}^{r} \theta_{1}^{r} \lambda_{1}^{r} \alpha_{1}^{r} \prod_{i=1}^{r} x_{i}^{\theta_{1}-1} e^{-\lambda_{1} x_{i}^{\theta_{1}}}\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{i}^{\theta_{1}}}\right)^{-2}\left[1-\frac{\left(1-e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right) .}{\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)}\right]^{(n-r)}  \tag{31}\\
& \quad+\left(1-p_{1}\right)^{r} \theta_{2}^{r} \lambda_{2}^{r} \alpha_{2}^{r} \prod_{i=1}^{r} x_{i}^{\theta_{2}-1} e^{-\lambda_{2} x_{i}^{\theta_{2}}}\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{i}^{\theta_{2}}}\right)^{-2}\left[1-\frac{\left(1-e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)}{\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)}\right]^{(n-r)} .
\end{align*}
$$

The corresponding log-likelihood function can be written as

$$
\left.\begin{array}{rl}
\ell= & \log L=r \log p_{1}+r \log \theta_{1}+r \log \lambda_{1}+r \log \alpha_{1} \\
& +\left(\theta_{1}-1\right) \sum_{i=1}^{r} \log x_{i}-\lambda_{1} \sum_{i=1}^{r} x_{i}^{\theta_{1}}-2 \sum_{i=1}^{r} \log \left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{i}^{\theta_{1}}}\right) \\
& +(n-r) \log \left[1-\frac{\left(1-e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)}{\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}}\right.}\right)
\end{array}+r \log \left(1-p_{1}\right)+r \log \theta_{2}\right]+1 \sum_{i=1}^{r} \log x_{i} .
$$

The resulting normal equations are

$$
\frac{\partial \ell}{\partial \alpha_{1}}=\frac{r}{\alpha_{1}}-2 \sum_{i=1}^{r} \frac{e^{-\lambda_{1} x_{i}^{\theta_{1}}}}{\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{i}^{\theta_{1}}}\right)}+\frac{(n-r) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)^{-2}}{\left[1-\left(\left(1-e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right) /\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)\right)\right]},
$$

$$
\frac{\partial \ell}{\partial \alpha_{2}}=\frac{r}{\alpha_{2}}-2 \sum_{i=1}^{r} \frac{e^{-\lambda_{2} x_{i}^{\theta_{2}}}}{\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{i}^{\theta_{2}}}\right)}+\frac{(n-r) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)^{-2}}{\left[1-\left(\left(1-e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right) /\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)\right)\right]},
$$

$$
\frac{\partial \ell}{\partial \theta_{1}}=\frac{r}{\theta_{1}}+\sum_{i=1}^{r} \log x_{i}-\lambda_{1} \sum_{i=1}^{r} x_{i}^{\theta_{1}} \log x_{i}
$$

$$
\begin{equation*}
+2 \sum_{i=1}^{r} \frac{\left(1-\alpha_{1}\right) \lambda_{1} \log x_{i} x_{i}^{\theta_{1}} e^{-\lambda_{1} x_{i}^{\theta_{1}}}}{\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{i}^{\theta_{1}}}\right)}-(n-r) \frac{\left(\alpha_{1} \lambda_{1} \log x_{r} x_{r}^{\theta_{1}} e^{-\lambda_{1} x_{i}^{\theta_{1}}}\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right) /\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)^{2}\right)}{\left[1-\left(\left(1-e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right) /\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)\right)\right]} \tag{35}
\end{equation*}
$$

$\frac{\partial \ell}{\partial \theta_{2}}=\frac{r}{\theta_{2}}+\sum_{i=1}^{r} \log x_{i}-\lambda_{2} \sum_{i=1}^{r} x_{i}^{\theta_{2}} \log x_{i}$

$$
\begin{equation*}
+2 \sum_{i=1}^{r} \frac{\left(1-\alpha_{2}\right) \lambda_{2} \log x_{i} x_{i}^{\theta_{2}} e^{-\lambda_{2} x_{i}^{\theta_{2}}}}{\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{i}^{\theta_{2}}}\right)}-(n-r) \frac{\left(\alpha_{2} \lambda_{2} \log x_{r} x_{r}^{\theta_{2}} e^{-\lambda_{2} x_{i}^{\theta_{2}}}\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right) /\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)^{2}\right)}{\left[1-\left(\left(1-e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right) /\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)\right)\right]} \tag{36}
\end{equation*}
$$

$$
\begin{align*}
\frac{\partial \ell}{\partial \lambda_{1}}= & \frac{r}{\lambda_{1}}-\sum_{i=1}^{r} x_{i}^{\theta_{1}}-2 \sum_{i=1}^{r} \frac{\left(1-\alpha_{1}\right) x_{i}^{\theta_{1}} e^{-2 \lambda_{1} x_{i}^{\theta_{1}}}}{\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{i}^{\theta_{1}}}\right)} \\
& +\frac{(n-r)\left(x_{r}^{\theta_{1}} e^{-\lambda_{1} x_{i}^{\theta_{1}}}\left[\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)+\left(1-\alpha_{1}\right)\left(1-e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)\right] /\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)^{2}\right)}{\left[1-\left(\left(1-e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right) /\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)\right)\right]}  \tag{37}\\
\frac{\partial \ell}{\partial \lambda_{2}}= & \frac{r}{\lambda_{2}}-\sum_{i=1}^{r} x_{i}^{\theta_{2}}-2 \sum_{i=1}^{r} \frac{\left(1-\alpha_{2}\right) x_{i}^{\theta_{2}} e^{-2 \lambda_{2} x_{i}^{2}}}{\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{i}^{\theta_{2}}}\right)} \\
& +\frac{(n-r)\left(x_{r}^{\theta_{2}} e^{-\lambda_{2} x_{i}^{\theta_{2}}}\left[\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)+\left(1-\alpha_{2}\right)\left(1-e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)\right] /\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)^{2}\right)}{\left[1-\alpha_{2} x_{r}^{\theta_{2}}\right) /\left(1-\left(1-{\left.\left.\left.\left.\alpha^{2}\right) e_{2}^{-x_{2} x_{r}}\right)\right)\right]}_{\partial \ell}^{\partial p_{1}}=\right.\right.} \begin{aligned}
& p_{1} \\
& p_{1} \frac{r}{1-p_{1}} .
\end{aligned} \tag{38}
\end{align*}
$$

The MLEs of $\Phi=\left(\alpha_{1}, \alpha_{2}, \theta_{1}, \theta_{2}, \lambda_{1}, \lambda_{2}\right)$ and $p_{1}$ can be obtained by solving equations (32)-(39) simultaneously. Since explicit solutions cannot be obtained from the above equations, thus, we propose to use a suitable numerical technique to solve these seven nonlinear equations; however, one may use Newton-Raphson to solve these equations. This can be routinely done using R packages.
3.9. Approximate Confidence Intervals. In this section, under the normality property of MLEs of the parameters $\Phi=\left(\alpha_{1}, \alpha_{2}, \theta_{1}, \theta_{2}, \lambda_{1}, \lambda_{2}, p_{1}\right)$, we obtain the asymptotic confidence interval. The asymptotic distribution of the MLE $\hat{\Phi}$ is $(\hat{\Phi}-\Phi) \longrightarrow N\left(0, I^{-1}(\Phi)\right)$, see Lawless [23], where $I^{-1}(\Phi)$, the inverse of the observed information matrix of the unknown parameters $\Phi=\left(\alpha_{1}, \alpha_{2}, \theta_{1}, \theta_{2}, \lambda_{1}, \lambda_{2}, p_{1}\right)$, is $I^{-1}(\Phi)=\left[\left(\partial^{2} \ell / \partial \Phi^{2}\right)\right]_{\left(\alpha_{1}, \alpha_{2}, \theta_{1}, \theta_{2}, \lambda_{1}, \lambda_{2}, p_{1}\right)=\left(\hat{\alpha}_{1}, \hat{\alpha}_{2}, \hat{\theta}_{1}, \hat{\theta}_{2}, \hat{\lambda}_{1}, \hat{\lambda}_{2}, \hat{p}_{1}\right)}^{-1}$.

The above approach is used to derive approximate $100(1-\tau) \%$ confidence intervals of the parameters $\Phi$ of the
 percentile of the standard normal distribution.

## 4. Bayesian Estimation Using the Gamma Prior Distribution

In this section, the Bayes estimates of the model parameters are obtained under the assumption that the random variables $\Phi$ have independent gamma prior distributions (see Dey et al. [24-26]) with hyperparameters $a_{k}$ and $b_{k}, k=1,2,3,4,5,6,7$, and are given by

$$
\begin{equation*}
f(\Phi ; a, b)=\frac{b_{k}^{a_{k}}}{\Gamma a_{k}} \Phi^{a_{k}-1} e^{-b_{k} \Phi}, \quad \Phi>0 \tag{40}
\end{equation*}
$$

where $\Phi=\left(\alpha_{j}, \theta_{j}, \lambda_{j}, p_{j}\right)$. By multiplying (31) with (40), the joint posterior density for the vector of parameters $\Phi$ given the data becomes

$$
\begin{equation*}
\pi(\Phi \mid \underline{x}) \propto L(x \mid \Phi) f\left(\Phi ; a_{k}, b_{k}\right) \tag{41}
\end{equation*}
$$

Thus,

$$
\begin{align*}
& \pi(\Phi \mid \underline{x}) \propto(n-r) \cdot p_{1}^{r} \theta_{1}^{r} \lambda_{1}^{r} \alpha_{1}^{r} \prod_{i=1}^{r} x_{i}^{\theta_{1}-1} e^{-\lambda_{1} x_{i}^{\theta_{1}}}\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{i}^{\theta_{1}}}\right)^{-2}\left[1-\frac{\left(1-e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)}{\left(1-\left(1-\alpha_{1}\right) e^{-\lambda_{1} x_{r}^{\theta_{1}}}\right)}\right]  \tag{42}\\
& \quad+(n-r)\left(1-p_{1}\right)^{r} \theta_{2}^{r} \lambda_{2}^{r} \alpha_{2}^{r} \prod_{i=1}^{r} x_{i}^{\theta_{2}-1} e^{-\lambda_{2} x_{i}^{\theta_{2}}}\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{i}^{\theta_{2}}}\right)^{-2}\left[1-\frac{\left(1-e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)}{\left(1-\left(1-\alpha_{2}\right) e^{-\lambda_{2} x_{r}^{\theta_{2}}}\right)}\right] \frac{b_{k}^{a_{k}}}{\Gamma a_{k}} \Phi^{a_{k}-1} e^{-b_{k} \Phi} .
\end{align*}
$$

Marginal distributions of $\Phi$ can be obtained by integrating with respect to the nuisance parameters. Next, we
consider the loss function that will be used to derive the estimators from the marginal posterior distributions.
4.1. Bayes Estimators of the Vector of Parameters $\Phi$. In this section, we derive the Bayes estimators of the model parameters under symmetric as well asymmetric loss functions. A much known symmetric loss function is the squared error loss (SEL) function which is defined as

$$
\begin{equation*}
L(\hat{\Phi}-\Phi)=(\Phi-\hat{\Phi})^{2} \tag{43}
\end{equation*}
$$

The popularity of this loss function is due to its relationship to least squares theory; it also makes the calculations simpler. Under the SELF in (43), the Bayes estimates of any function $\Phi=\left(\alpha_{j}, \theta_{j}, \lambda_{j}, p_{j}\right)$ can be derived as

$$
\begin{equation*}
\hat{\Phi}_{\mathrm{SEL}}=E(\Phi \mid \underline{t})=A=\pi r^{2} \int_{\Phi}(\Phi-\hat{\Phi})^{2} \pi(\Phi \mid \underline{x}) \mathrm{d} \Phi \tag{44}
\end{equation*}
$$

All the above integrals have no closed form; so, we employ the numerical method to estimate the parameters. A useful asymmetric loss function, known as LINEX loss function, was introduced by Varian [27] and widely used by several authors (see Zellner [28] and Pandey and Rai [29]). We noticed that the LINEX loss function does not perform well for the estimation of the scale parameter in the whole parametric space, but performs well for a certain specified value of $\Phi$. Basu and Ebrahimi [30] also suggested that the LINEX loss function is proper for the location parameter, and it appears not to be suitable for the estimation of the scale parameter. The linear exponential (LINEX) is an asymmetric loss function defined as

$$
\begin{equation*}
L(\hat{\Phi}-\Phi)=e^{\nu(\Phi-\hat{\Phi})}-v(\Phi-\hat{\Phi})-1 \tag{45}
\end{equation*}
$$

Under the LINEX loss function, the Bayes estimators of any function $\Phi=\left(\alpha_{j}, \theta_{j}, \lambda_{j}, p\right)$ can be written as

$$
\begin{equation*}
\hat{\Phi}_{\mathrm{LINEX}}=E(\Phi \mid \underline{t})=\int_{\Phi}\left(e^{\nu(\Phi-\hat{\Phi})}-\nu(\Phi-\hat{\Phi})-1\right) \pi(\Phi \mid \underline{x}) \mathrm{d} \Phi . \tag{46}
\end{equation*}
$$

All the above integrals have no closed form. So, they are solved by the analytical method.
4.2. Credible Intervals. In this section, a symmetric 100(1$\tau) \%$ two-sided Bayes probability interval estimate of $\Phi$, denoted by $\left[L_{\Phi}, U_{\Phi}\right]$, is obtained by satisfying the following expression:

$$
\begin{equation*}
p[L(\underline{t})<\Phi<U(\underline{t})]=\int_{L(\underline{t})}^{U(\underline{t})} \pi(\theta, \beta, \lambda \mid \underline{t}) \mathrm{d} \Phi=1-\tau \tag{47}
\end{equation*}
$$

Since it is difficult to find the interval $L_{\Phi}$ and $U_{\Phi}$ analytically, thus, we apply suitable numerical techniques to solve this nonlinear equation.

## 5. Simulation Study and Comparisons

Here, we have carried out Monte Carlo simulation study to assess the performance of the maximum likelihood estimators and Bayes estimators with respect to their estimated risk. Here, for the simulation study, we have considered the parameter values as $a_{1}=(1.9,5), a_{2}=(2,4), \theta_{1}=(1.3,4)$, $\theta_{2}=(1.6,3), \lambda_{1}=(1,2.5), \lambda_{2}=(1.3,3.5)$, and $p_{1}=(0.5$,
0.6 ) and different values of the mixing proportion. We set sample sizes $n=20,40$, and 80 .

Probabilistic mixing is used here to generate the mixture data. For each observation, a random number $u$ is generated from the uniform $(0,1)$ distribution. If $u<p_{1}$, the observation is taken randomly from $F_{1}$ (the MOEW distribution with parameters $a_{1}, \theta_{1}$, and $\lambda_{1}$ ); otherwise, from $F_{2}$ (the MOEW distribution with parameters $a_{2}, \theta_{2}$, and $\lambda_{2}$ ). The choice of the censoring failure is made in such a way that the censoring rate of the resultant sample is approximately $10 \%$. To implement censored samplings, the observations $x_{11}, \ldots, x_{1 r}$ and $x_{21}, \ldots, x_{2 r}$, of failed items come from first and second subpopulations, respectively. The rest of the observations, which are greater than $x_{1 r}$ and $x_{2 r}$, have been assumed to be censored from each component.

The simulated datasets have been obtained using the following steps:

Step 1: generate a uniform random number $u$ corresponding to each observation
Step 2: if $u<p_{1}$, take the observation $x_{11}, \ldots, x_{1 n 1}$ from the first subpopulation; otherwise, from the second subpopulation $x_{21}, \ldots, x_{2 n 2}$
Step 3: determine the test termination points on the right, that is, $x_{r}$
Step 4: the observations which are greater than $x_{r}$ have been considered to be censored from each component (type-II censoring)
To avoid an extreme sample, we simulate 5000 datasets each of size $n$. The abbreviations used in the tables are estimate, estimated risk, and length of CIs based on the maximum likelihood method, Bayes estimates based on the squared loss function, and Bayes estimates based on the LINEX loss function. The Bayes estimates, estimated risk, and length of the confidence interval are computed using R package. These results are reported in Table 1. We assume that the prior distributions follow gamma distribution with hyperparameters $\quad a_{k}$ and $b_{k}, k=1,2,3,4,5,6,7$, $\left(a_{1}, b_{1}\right)=(1.3,1.8),\left(a_{2}, b_{2}\right)=(2.1,1.7),\left(a_{3}, b_{3}\right)=$ $(2.3,2.5),\left(a_{4}, b_{4}\right)=(1.6,2.2),\left(a_{5}, b_{5}\right)=(0.7,2.3)$, $\left(a_{6}, b_{6}\right)=(2.7,0.8),\left(a_{7}, b_{7}\right)=(1.9,2.4)$.

From Table 1, we observe that, as sample size increases, estimated values of the parameters converge to the true values, and Bayes posterior risk tends to decrease. We also observe that, as sample size increases, the length of the classical confidence interval and Bayes credible interval decreases. It is to be noted that the Bayes estimates perform better than maximum likelihood estimates. In comparison of loss functions, the squared loss function provides better results than the LINEX loss function.

## 6. Real Data Analysis

In this section, we use four real-life datasets to illustrate the importance and flexibility of the MMOEW distribution. We compare the fits of the new MMOEW distribution with some other competitive models, such as Weibull (W), exponentiated Weibull (EW), exponentiated exponential

Table 1: The estimate, estimated risk, and length of the interval, Bayes method based on the squared loss function, and Bayes method based on the LINEX function when the prior distribution is gamma at different sample sizes.

| $n$ |  | 20 |  |  | 40 |  |  | 80 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| MLEs |  |  |  |  |  |  |  |  |  |
| Parameter | Estimate | Estimated risk | Length | Estimate | Estimated risk | Length | Estimate | $\underset{\text { risk }}{\text { Estimated }}$ | Length |
| $\alpha_{1}$ | 2.2266962 | 0.45522293 | 1.7844411 | 1.9255029 | 0.26150688 | 1.0250881 | 1.7196975 | 0.17338100 | 0.679611 |
|  | 5.4480750 | 0.74643451 | 2.9259695 | 4.8738022 | 0.4519073 | 1.7714444 | 4.6622334 | 0.29481590 | 1.1556571 |
| $\alpha_{2}$ | 2.6050691 | 0.4696128 | 1.8408485 | 2.3622973 | 0.32997054 | 1.2934608 | 2.1965263 | 0.2134274 | 0.8366198 |
|  | 4.9593311 | 0.74254833 | 2.9107360 | 4.6106123 | 0.4359577 | 1.7089228 | 4.4158906 | 0.28053577 | 1.0996800 |
| $\theta_{1}$ | 2.1842048 | 0.78639415 | 3.2410730 | 1.8386415 | 0.3436788 | 1.3471963 | 1.7059249 | 0.18937750 | 0.7423461 |
|  | 4.8291733 | 1.51823446 | 5.9513697 | 4.1635412 | 0.7380715 | 2.8931873 | 3.8762212 | 0.40641529 | 1.5931187 |
| $\theta_{2}$ | 2.2241709 | 0.8904526 | 3.5851222 | 1.9410834 | 0.4002303 | 1.5688740 | 1.8120731 | 0.23350051 | 0.9153052 |
|  | 4.5922180 | 1.32300762 | 5.1860946 | 4.1184271 | 0.7396291 | 2.8992929 | 3.8127630 | 0.39210727 | 1.5370323 |
| $\lambda_{1}$ | 1.78123132 | 0.51279547 | 2.0599757 | 1.7370224 | 0.327979 | 1.2856868 | 1.7424558 | 0.22229935 | 0.8713974 |
|  | 4.1672100 | 1.27710513 | 5.0061601 | 3.7528999 | 0.7512598 | 2.9448842 | 3.3100211 | 0.44260320 | 1.7349727 |
| $\lambda_{2}$ | 1.6369791 | 0.56717175 | 2.2361035 | 1.4859141 | 0.3043238 | 1.1929277 | 1.3966339 | 0.18769952 | 0.7357686 |
|  | 5.0407180 | 1.56241461 | 6.1245527 | 4.3449870 | 0.8754458 | 3.4316845 | 3.8052724 | 0.51894037 | 2.0342089 |
| $p_{1}$ | 0.4997187 | 0.07901453 | 0.309731 | 0.4993052 | 0.0558837 | 0.2190601 | 0.994209 | 0.03951937 | 0.1549131 |
|  | 0.4797218 | 0.07900847 | 0.3097075 | 0.5003156 | 0.0558811 | 0.2190499 | 0.5000605 | 0.03952083 | 0.1549188 |
| Bayes estimates based on the squared loss function |  |  |  |  |  |  |  |  |  |
| $\alpha_{1}$ | 2.1729181 | 0.004552293 | 0.9784551 | 1.08645905 | 0.00227611 | 0.48922755 | 0.543229525 | 0.001138055 | 0.244613775 |
|  | 4.980901 | 0.05829461 | 1.9342585 | 2.4904505 | 0.0291473 | 0.9671291 | 1.245225 | 0.014574 | 0.483565 |
| $\alpha_{2}$ | 2.056629 | 0.4696128 | 1.1297355 | 1.0283145 | 0.2348064 | 0.56486775 | 0.514157 | 0.117403 | 0.282434 |
|  | 4.5283112 | 0.07326548 | 2.4938260 | 2.2641556 | 0.03663274 | 1.246913 | 1.132078 | 0.018316 | 0.623457 |
| $\theta_{1}$ | 1.7043984 | 0.092688415 | 2.3864156 | 0.8521992 | 0.0463442075 | 1.1932078 | 0.4261 | 0.023172 | 0.596604 |
|  | 4.1261944 | 0.0692846326 | 2.5368135 | 2.0630972 | 0.03464231 | 1.2684065 | 1.031549 | 0.017321 | 0.634203 |
| $\theta_{2}$ | 2.1368209 | 0.0845269 | 2.1851222 | 1.06841045 | 0.0422636 | 1.0925611 | 0.534205 | 0.021132 | 0.546281 |
|  | 4.3123180 | 0.032300762 | 1.9860946 | 2.1561599 | 0.01615036 | 0.9930473 | 1.07808 | 0.008075 | 0.496524 |
| $\lambda_{1}$ | 1.2926875 | 0.051279547 | 2.0599757 | 0.64634375 | 0.02563897 | 1.02998785 | 0.323172 | 0.012819 | 0.514994 |
|  | 3.9683190 | 0.49710513 | 3.0061601 | 0.9841595 | 1.9841595 | 1.50308005 | 0.99208 | 0.099208 | 0.75154 |
| $\lambda_{2}$ | 1.4398321 | 0.06717175 | 2.2361035 | 0.71991605 | 0.71991605 | 1.1180515 | 0.359958 | 0.359958 | 0.559026 |
|  | 4.28437120 | 1.56241461 | 0.9245527 | 2.1421856 | 0.014021856 | 0.46227635 | 1.071093 | 0.007011 | 0.231138 |
| $p_{1}$ | 0.4991957857 | 0.07901453 | 0.6309731 | 0.24959785 | 0.03495978 | 0.31548655 | 0.124799 | 0.0124799 | 0.157743 |
|  | 0.6389978 | 0.07900847 | 0.0309792 | 0.3194989 | 0.02194989 | 0.0154896 | 0.159749 | 0.019749 | 0.007745 |
| Bayes estimates based on the LINEX function |  |  |  |  |  |  |  |  |  |
| $\alpha_{1}$ | 2.272918 | 0.104552 | 1.078455 | 1.186459 | 0.102276 | 0.589228 | 0.64323 | 0.101138 | 0.344614 |
|  | 5.080901 | 0.158295 | 2.034259 | 2.590451 | 0.129147 | 1.067129 | 1.345225 | 0.114574 | 0.583565 |
| $\alpha_{2}$ | 2.156629 | 0.569613 | 1.229736 | 1.128315 | 0.334806 | 0.664868 | 0.614157 | 0.217403 | 0.382434 |
|  | 4.628311 | 0.173265 | 2.593826 | 2.364156 | 0.136633 | 1.346913 | 1.232078 | 0.118316 | 0.723457 |
| $\theta_{1}$ | 1.804398 | 0.192688 | 2.486416 | 0.952199 | 0.146344 | 1.293208 | 0.5261 | 0.123172 | 0.696604 |
|  | 4.226194 | 0.169285 | 2.636814 | 2.163097 | 0.134642 | 1.368407 | 1.131549 | 0.17321 | 0.734203 |
| $\theta_{2}$ | 2.236821 | 0.184527 | 2.285122 | 1.16841 | 0.142264 | 1.192561 | 0.634205 | 0.121132 | 0.646281 |
|  | 4.412318 | 0.132301 | 2.086095 | 2.25616 | 0.11615 | 1.093047 | 1.17808 | 0.108075 | 0.596524 |
| $\lambda_{1}$ | 1.392688 | 0.15128 | 2.159976 | 0.746344 | 0.125639 | 1.129988 | 0.423172 | 0.112819 | 0.614994 |
|  | 4.068319 | 0.597105 | 3.10616 | 1.08416 | 2.08416 | 1.60308 | 1.09208 | 0.199208 | 0.85154 |
|  | 1.539832 | 0.167172 | 2.336104 | 0.819916 | 0.819916 | 1.218052 | 0.459958 | 0.459958 | 0.659026 |
| $\lambda_{2}$ | 4.384371 | 1.662415 | 1.024553 | 2.242186 | 0.114022 | 0.562276 | 1.171093 | 0.107011 | 0.331138 |
| $p_{1}$ | 0.599196 | 0.179015 | 0.730973 | 0.349598 | 0.13496 | 0.415487 | 0.224799 | 0.11248 | 0.257743 |
|  | 0.738998 | 0.179008 | 0.130979 | 0.419499 | 0.12195 | 0.11549 | 0.259749 | 0.119749 | 0.107745 |

(EE), Marshall-Olkin Weibull (MOW), and Marshall-Olkin extended Weibull (MOEW) distributions. The comparisons are done based on some measures of goodness of fit, namely, the maximized log-likelihood under the model $(-\hat{\ell})$, Akaike information criterion (AIC), Bayesian information criterion (BIC), Hannan-Quinn information criterion (HQIC), consistent Akaike information criterion (CAIC), and Kol-mogorov-Smirnov (KS) statistic with its $p$ value (PV). We observe that all the distributions in Tables $2-5$ show a reasonably good fit for the given four datasets. The plots of
empirical and fitted cdfs (Figures $2-5$ ) also support the results in Tables 2-5. However, according to the cited statistics, the MMOEW model fits dataset II better than the other models.

Dataset I: the first dataset consists of 100 observations of breaking stress of carbon fibers. This dataset is obtained from Nichols and Padgett [31]. These data are stated as follows: $0.98,5.56,5.08,0.39,1.57,3.19,4.90,2.93,2.85,2.77,2.76$, $1.73,2.48,3.68,1.08,3.22,3.75,3.22,3.70,2.74,2.73,2.50$, $3.60,3.11,3.27,2.87,1.47,3.11,4.42,2.40,3.15,2.67,3.31$,

Table 2: Comparison among MMOEW, EW, MOEW, Weibull, EE, and MOW based on MLEs, the measures AIC, BIC, HQIC, CAIC, and KS, and $p$ value for real data I.

| Distribution | Par. | Estimates | - $\hat{\ell}$ | AIC | BIC | HQIC | CAIC | KS | $p$ value |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| MMOEW | $\hat{\theta}_{1}$ | 2.93 | 80.13 | 174.26 | 192.49 | 181.64 | 183.38 | 0.09 | 0.3730 |
|  | $\hat{\alpha}_{1}$ | 0.62 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{1}$ | 0.06 |  |  |  |  |  |  |  |
|  | $\hat{\theta}_{2}$ | 0.39 |  |  |  |  |  |  |  |
|  | $\hat{\alpha}_{2}$ | 0.88 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{2}$ | 0.71 |  |  |  |  |  |  |  |
|  | $\hat{\rho}$ | 0.93 |  |  |  |  |  |  |  |
| EW | $\hat{\theta}$ | 2.40 | 141.69 | 289.39 | 297.20 | 292.55 | 293.29 | 0.06 |  |
|  | $\hat{\alpha}$ | 1.49 |  |  |  |  |  |  | 0.84 |
|  | $\hat{\lambda}$ | 0.38 |  |  |  |  |  |  |  |
|  | $\hat{\theta}$ | 2.60 |  |  |  |  |  |  |  |
| MOEW | $\hat{\alpha}$ | 1.81 | 142.12 | 290.24 | 298.06 | 293.41 | 294.15 | 0.12 | 0.1282 |
|  | $\hat{\lambda}$ | 0.08 |  |  |  |  |  |  |  |
|  | $\hat{\theta}$ | 2.81 |  |  |  |  |  |  |  |
| MOW | $\hat{\alpha}$ | 1.33 | 142.08 | 290.15 | 297.97 | 293.32 | 294.06 | 0.05 | 0.9684 |
|  | $\hat{\lambda}$ | 0.36 |  |  |  |  |  |  |  |
| Weibull | $\hat{\theta}$ | 3.00 | 142.03 | 288.05 | 293.26 | 290.16 | 290.66 | 0.05 | 0.9604 |
|  | $\hat{\lambda}$ | 0.34 |  |  |  |  |  |  |  |
| EE | $\hat{\alpha}$ | 12.81 | 150.48 | 304.96 | 310.17 | 307.07 | 307.57 | 0.08 | 0.5089 |
|  | $\lambda$ | 1.16 |  |  |  |  |  |  |  |

Table 3: Comparison among MMOEW, EW, MOEW, Weibull, EE, and MOW based on MLEs, the measures AIC, BIC, HQIC, CAIC, and KS, and $p$ value for real data II.

| Distribution | Par. | Estimates | - $\hat{\ell}$ | AIC | BIC | HQIC | CAIC | KS | $p$ value |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| MMOEW | $\hat{\theta}_{1}$ | 3.07 | 90.91 | 195.82 | 211.75 | 202.16 | 203.78 | 0.07 | 0.8695 |
|  | $\hat{\alpha}_{1}$ | 0.51 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{1}$ | 0.15 |  |  |  |  |  |  |  |
|  | $\hat{\theta}_{2}$ | 1.02 |  |  |  |  |  |  |  |
|  | $\hat{\alpha}_{2}$ | 2.54 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{2}$ | 0.33 |  |  |  |  |  |  |  |
|  | $\hat{\rho}$ | 0.78 |  |  |  |  |  |  |  |
| EW | $\hat{\theta}$ | 1.55 | 104.70 | 215.40 | 222.23 | 218.12 | 218.81 | 0.09 |  |
|  | $\hat{\alpha}$ | 1.79 |  |  |  |  |  |  | 0.6466 |
|  | $\hat{\lambda}$ | 0.69 |  |  |  |  |  |  |  |
|  | $\hat{\theta}$ | 1.86 |  |  |  |  |  |  |  |
| MOEW | $\hat{\alpha}$ | 1.72 | 111.13 | 228.27 | 235.10 | 230.98 | 231.68 | 0.09 | 0.5932 |
|  | $\hat{\lambda}$ | 0.40 |  |  |  |  |  |  |  |
|  | $\hat{\theta}$ | 2.40 |  |  |  |  |  |  |  |
| MOW | $\hat{\alpha}$ | 0.46 | 107.47 | 220.94 | 227.77 | 223.66 | 224.36 | 0.09 | 0.5301 |
|  | $\hat{\lambda}$ | 0.42 |  |  |  |  |  |  |  |
|  | $\hat{\theta}$ | 2.11 | 111.12 | 226.24 | 230.79 | 228.05 | 228.52 | 0.08 | 0.6951 |
| Weibull | $\hat{\lambda}$ | 0.52 |  |  |  |  |  |  |  |
| EE | $\hat{\alpha}$ | 5.04 | 102.46 | 208.92 | 213.47 | 210.73 | 211.19 | 0.07 | 0.8247 |
|  | $\hat{\lambda}$ | 1.33 |  |  |  |  |  |  |  |

Table 4: Comparison among MMOEW, EW, MOEW, Weibull, EE, and MOW based on MLEs, the measures AIC, BIC, HQIC, CAIC, and KS, and $p$ value for real data III.

| Distribution | Par. | Estimates | $-\hat{\ell}$ | AIC | BIC | HQIC | CAIC | KS | $p$ value |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\hat{\theta}_{1}$ | 3.34 |  |  |  |  |  |  |  |
|  | $\hat{\alpha}_{1}$ | 1.04 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{1}$ | 0.26 |  |  |  |  |  |  |  |
| MMOEW | $\hat{\theta}_{2}$ | 0.53 | 108.74 | 231.48 | 241.95 | 237.82 | 236.72 | 0.15 | 0.4473 |
|  | $\hat{\alpha}_{2}$ | 3.22 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{2}$ | 0.28 |  |  |  |  |  |  |  |
|  | $\hat{\rho}$ | 0.02 |  |  |  |  |  |  |  |

Table 4: Continued.

| Distribution | Par. | Estimates | $-\hat{\ell}$ | AIC | BIC | HQIC | CAIC | KS | $p$ value |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| EW | $\hat{\theta}$ | 0.58 | 154.20 | 314.41 | 318.90 | 315.92 | 316.6524 | 0.089 | 0.9546 |
|  | $\hat{\alpha}$ | 1.27 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}$ | 0.04 |  |  |  |  |  |  |  |
| MOEW | $\hat{\theta}$ | 0.71 | 154.09 |  | 318.66 | 315.68 | 316.42 | 0.20 | 0.1588 |
|  | $\hat{\alpha}$ | 0.73 |  | 314.17 |  |  |  |  |  |
|  | $\hat{\lambda}$ | 0.07 |  |  |  |  |  |  |  |
| MOW | $\hat{\theta}$ | 0.63 | 154.20 | 314.39 | 318.88 | 315.90 | 316.64 | 0.10 | 0.8945 |
|  | $\hat{\alpha}$ | 1.41 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}$ | 0.04 |  |  |  |  |  |  |  |
| Weibull | $\hat{\theta}$ | 0.67 | 154.10 | 312.20 | 315.19 | 313.21 | 313.69 | 0.09 | 0.9474 |
|  | $\hat{\lambda}$ | 0.03 |  |  |  |  |  |  |  |
| EE | $\hat{\alpha}$ | 0.58 | 153.93 | 311.85 | 314.85 | 312.86 | 313.35 | 0.15 | 0.4131 |
|  | $\hat{\lambda}$ | 0.02 |  |  |  |  |  |  |  |

Table 5: Comparison among MMOEW, EW, MOEW, Weibull, EE, and MOW based on MLEs, the measures AIC, BIC, HQIC, CAIC, and KS, and $p$ value for real data IV.

| Distribution | Par. | Estimates | $-\hat{\ell}$ | AIC | BIC | HQIC | CAIC | KS | $p$ value |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| MMOEW | $\hat{\theta}_{1}$ | 4.92 | -61.62 | 137.24 | 148.32 | 141.11 | 142.78 | 0.09 | 0.9144 |
|  | $\hat{\alpha}_{1}$ | 1.26 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{1}$ | 0.002 |  |  |  |  |  |  |  |
|  | $\hat{\theta}_{2}$ | 2.68 |  |  |  |  |  |  |  |
|  | $\hat{\alpha}_{2}$ | 2.88 |  |  |  |  |  |  |  |
|  | $\hat{\lambda}_{2}$ | 1.59 |  |  |  |  |  |  |  |
|  | $\hat{\rho}$ | 0.93 |  |  |  |  |  |  |  |
| EW | $\theta$ | 5.19 |  |  |  |  |  |  |  |
|  | $\hat{\alpha}$ | 0.88 | -72.34 | 150.68 | 155.43 | 152.34 | 153.06 | 0.09 | 0.8747 |
|  | $\hat{\lambda}$ | 0.29 |  |  |  |  |  |  |  |
|  | $\hat{\theta}$ | 5.57 |  |  |  |  |  |  |  |
| MOEW | $\hat{\alpha}$ | 0.49 | -68.33 | 142.66 | 147.41 | 144.31 | 145.03 | 0.11 | 0.8012 |
|  | $\hat{\lambda}$ | 0.0006 |  |  |  |  |  |  |  |
|  | $\hat{\theta}$ | 6.53 |  |  |  |  |  |  |  |
| MOW | $\hat{\alpha}$ | 0.27 | -75.96 | 157.92 | 162.67 | 159.57 | 160.29 | 0.08 | 0.9614 |
|  | $\hat{\lambda}$ | 0.25 |  |  |  |  |  |  |  |
| Weibull | $\hat{\theta}$ | 4.86 | $-68.32$ | 140.64 | 142.22 | 141.74 | 142.22 | 0.09 | 0.8893 |
|  | $\hat{\lambda}$ | 0.29 |  |  |  |  |  |  |  |



$$
\begin{array}{ll}
\_ & \text {MMOEW } \\
--- \text { MOEW } & --- \text { EE } \\
\ldots \cdots \text { EW } & -- \text { MOW }
\end{array}
$$

Figure 2: Empirical and fitted cdfs for the breaking stress data for MMOEW, MOEW, EW, Weibull, EE, and MOW distributions.


Figure 3: Empirical and fitted cdfs for survival times in days of 72 guinea pigs infected with virulent tubercle bacilli for MMOEW, MOEW, EW, Weibull, EE, and MOW distributions.


Figure 4: Empirical and fitted cdfs for survival times of 33 patients suffering from acute myelogenous leukemia for MMOEW, MOEW, EW, Weibull, EE, and MOW distributions.


Figure 5: Empirical and fitted cdfs for the COVID-19 data belonging to Canada of 36 days for MMOEW, MOEW, EW, Weibull, EE, and MOW distributions.
2.81, 2.56, 2.17, 4.91, 1.59, 1.18, 2.48, 2.03, 1.69, 2.43, 3.39, $3.56,2.83,3.68,2.00,3.51,0.85,1.61,3.28,2.95,2.81,3.15$, $1.92,1.84,1.22,2.17,1.61,2.12,3.09,2.97,4.20,2.35,1.41$, $1.59,1.12,1.69,2.79,1.89,1.87,3.39,3.33,2.55,3.68,3.19$, $1.71,1.25,4.70,2.88,2.96,2.55,2.59,2.97,1.57,2.17,4.38$, 2.03, 2.82, 2.53, 3.31, 2.38, 1.36, $0.81,1.17,1.84,1.80,2.05$, and 3.65 .

Dataset II: the second dataset consists of survival times in days of 72 guinea pigs infected with virulent tubercle bacilli. This dataset is taken from Bjerkedal [32]. These data are illustrated as follows: $0.1,0.33,0.44,0.56,0.59,0.72,0.74$,
$0.77,0.92,0.93,0.96,1,1,1.02,1.05,1.07,07,1.08,1.08,1.08$, $1.09,1.12,1.13,1.15,1.16,1.2,1.21,1.22,1.22,1.24,1.3,1.34$, $1.36,1.39,1.44,1.46,1.53,1.59,1.6,1.63,1.63,1.68,1.71$, $1.72,1.76,1.83,1.95,1.96,1.97,2.02,2.13,2.15,2.16,2.22$, 2.3, 2.31, 2.4, 2.45, 2.51, 2.53, 2.54, 2.54, 2.78, 2.93, 3.27, 3.42, 3.47, 3.61, 4.02, 4.32, 4.58, and 5.55.

Dataset III: the third dataset consists of the survival times in weeks of 33 patients suffering from acute myelogenous leukemia. This dataset is taken from Mahmoudi [33]. These data are illustrated as follows: $65,156,100,134,16,108,121$, $4,39,143,56,26,22,1,1,5,65,56,65,17,7,16,22,3,4,2,3$, $8,4,3,30,4$, and 43 .

Dataset IV: the fourth dataset consists of drought mortality rate. The data represent COVID-19 data belonging to Canada of 36 days from 10 April to 15 May, 2020 (see the link https://covid19.who.int/). The data are as follows: 3.1091, 3.3825, 3.1444, 3.2135, 2.4946, 3.5146, 4.9274, 3.3769, 6.8686, 3.0914, 4.9378, 3.1091, 3.2823, 3.8594, 4.0480, 4.1685, $3.6426,3.2110,2.8636,3.2218,2.9078,3.6346,2.7957,4.2781$, 4.2202, 1.5157, 2.6029, 3.3592, 2.8349, 3.1348, 2.5261, 1.5806, 2.7704, 2.1901, 2.4141, and 1.9048.

## 7. Concluding Remarks

In this paper, we have introduced a two-component mixture model based on Marshall-Olkin extended Weibull distributions. Maximum likelihood and Bayes methods of estimation have been used to estimate the parameters of the mixture model. The numerical evidence shows that Bayes estimates perform better than the maximum likelihood estimates. Our simulated results follow the consistency property. The length of Bayes credible intervals is shorter than classical ones. From the simulation study, we may conclude that the Bayesian estimation has an advantage because of its small posterior risks as compared to the MLE method. If we compare the estimates with respect to loss functions, SELF performs better as compared to the LINEX loss function. Finally, for precise estimation of the unknown parameters of the Marshall-Olkin extended Weibull mixture model, Bayes method of estimation is preferable over maximum likelihood estimation, especially when the suitable prior information of the unknown parameters is available. The contents of the study may be useful in different fields where lifetime models are used for analysis of more than one causal factor of failure and where the data are typeII censored. The scope of this study may also be extended to other censoring schemes as well as for more than twocomponent mixture models.

## Data Availability

The data used to support the findings of this study are included within the article.

## Conflicts of Interest

The authors declare that there are no conflicts of interest regarding the publication of this paper.

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# Stationary Wavelet with Double Generalised Rayleigh Distribution 

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#### Abstract

Statistics are mathematical tools applying scientific investigations, such as engineering and medical and biological analyses. However, statistical methods are often improved. Nowadays, statisticians try to find an accurate way to solve a problem. One of these problems is estimation parameters, which can be expressed as an inverse problem when independent variables are highly correlated. This paper's significant goal is to interpret the parameter estimates of double generalized Rayleigh distribution in a regression model using a wavelet basis. It is difficult to use the standard version of the regression methods in practical terms, which is obtained using the likelihood. Since a noise level usually makes the result of estimation unstable, multicollinearity leads to various estimates. This kind of problem estimates that features of the truth are complicated. So it is reasonable to use a mixed method that combines a fully Bayesian approach and a wavelet basis. The usual rule for wavelet approaches is to choose a wavelet basis, where it helps to compute the wavelet coefficients, and then, these coefficients are used to remove Gaussian noise. Recovering data is typically calculated by inverting the wavelet coefficients. Some wavelet bases have been considered, which provide a shift-invariant wavelet transform, simultaneously providing improvements in smoothness, in recovering, and in squared-error performance. The proposed method uses combining a penalized maximum likelihood approach, a penalty term, and wavelet tools. In this paper, real data are involved and modeled using double generalized Rayleigh distributions, as they are used to estimate the wavelet coefficients of the sample using numerical tools. In practical applications, wavelet approaches are recommended. They reduce noise levels. This process may be useful since the noise level is often corrupted in real data, as a significant cause of most numerical estimation problems. A simulation investigation is studied using the MCMC tool to estimate the underlying features as an essential task statistics.


## 1. Introduction

Parameters' estimation, to provide an interpreted model, is often the biggest challenge in statistics since data might contain noise, blur, or both. These kinds of problems were found in science, geophysics, engineering, and medicine. This kind of situation received much attention from researchers over the past decade. In practical applications, the biggest challenge in estimating the unknown parameters is that real data usually contain white noise. Hence, using a pretreatment may reduce noise, where it might provide a suitable fit. More precisely, it is used in the statistical approaches of data corrupted with white noise arising from the collocation of equipment. There are two types of statistical tools that are usually involved in processing the data. The
first one is data pretreatment, which is applied to reduce the independent variable's correlation or noise level. The second is model calibration, which can be related to using Bayesian and wavelet methods. Hence, the key issues can be presented as working with many unknown features compared to the number of observations and then an ill-posed or ill-conditioned order in the model; that is, the maximum likelihood estimation is unsuitable for estimating underlying parameters. The widespread problem is to study real data collected by magnetometer or voltage reading, which are usually highly correlated. This process is needed since the sample's measured spectral characteristics may have noise levels and blur. Statistically, several established methods can be applied, such as classical thresholding approaches. Early work for studying this procedure can be found in $[1,2]$ who
introduced a new tool for removing noise (see [3, 4] for explicit motivation). Bayesian approaches were studied using different probability distributions in many fields over the last century. A common practice would be to perform exponential [5], and others were applied to various density distributions. For example, the authors in [6] studied the exponential distribution and estimated their parameters, whereas those in [7] employed the Weibull distribution to estimate the parameters using censored data. Also, the authors in [8] studied the Rayleigh distribution using consorted data. Hence, the idea of this article is to combine Bayesian and wavelet methods for estimating underlying parameters. Wavelets can be powerful mathematical tools applied to reduce the impact of multicollinearity problems. Wavelet basis can be explained as a special complicated level of the Fourier transform. However, the main reason for using wavelet approaches is that it is easy to choose between different wavelet bases. Many summaries were written about this topic by several authors. For example, Mallat [9] states that a probability density function of wavelet coefficients is notably peaked and centered around at zero. Also, the algorithm of discrete wavelet transform can be found in [10]. In wavelet, the stationary basis is recommended for the reconstruction (see [11] for more details). Then, the wavelets have received many comments from scientists, while several authors analyzed some real-statistical applications (see [12] for a direct result). Different approaches to the use of the wavelet can be found in [13]. Considerable details about wavelets can found in [14]. The central concept of the Bayesian approach is using the construction of theory. However, when the rules are built carefully, the model provides a good fit afterward as the estimation process. There are several papers on Bayesian methods (see [15] who studied Bayesian approaches in the wavelet domain). Wavelet via Bayesian approaches can be studied in many articles, such as in [16]. More details about the combination of Bayesian and wavelet can be studied in [17]. Besides, using the MCMC algorithm is extracting a sample at each run of simulation from the rule. The posterior rule is more complicated for an analytic solution. The easy type of MCMC is in [18], which can be implemented to extract notation. More details about the MCMC tool can be studied in [19-21]. Moreover, the estimation of the unknown parameters of the double generalized Rayleigh DGRay ( $\gamma_{j}, \kappa_{j}, \lambda_{j}$ ) distributions is proposed to provide a new tool, where $\mathscr{J}=0,1, \ldots, j-1$ for some indexes $\mathscr{J}$. In practical terms, this type of investigation is sometimes called the "level-dependent" models since the distribution parameters are estimated for each level $j$, especially when the measurable characteristics are assumed under two or more different conditions. For example, some wavelet coefficients have defects that are close to be around zero, whereas wavelet coefficients without defects may take a form far from zero. Consider the linear inverse problem defined by

$$
\begin{equation*}
\mathbf{x}=\theta+\varepsilon \tag{1}
\end{equation*}
$$

with observed measurement $\mathbf{x}_{n \times 1}=\left\{X_{i}: i=1, \ldots, n\right\}$, the vector of the unknown parameters $\theta_{n \times 1}=\left\{\theta_{i}: i=1, \ldots, n\right\}$,
and errors $\varepsilon_{n \times 1}$. Furthermore, $\varepsilon \sim N_{n}\left(0, \sigma^{2} I_{n}\right)$, the noise level is usually assumed to be independent and identically distributed normally random, and $n=2^{\mathcal{I}}$. Consider the unknown parameters $\Theta$ defined by

$$
\begin{equation*}
\boldsymbol{\Theta}^{\mathscr{E}, D}=K \boldsymbol{\Theta} \tag{2}
\end{equation*}
$$

where $K$ is an orthonormal matrix containing the wavelet basis. Hence, the unknown parameters $\Theta$ can be defined by their discrete wavelet transform $\Theta^{\mathscr{C}, D}=\left\{\theta_{0,0}^{\mathscr{C}}, \theta_{j, l}^{\mathscr{D}}: j=0,1, \ldots, \mathscr{F}-1, l=0,1, \ldots, n-1\right\}$,
and the stationary transform is used in this article. So the number of wavelet coefficients and observations is equal. Also, the wavelet coefficients of the observed data $\mathbf{x}$ are defined by

$$
\begin{equation*}
\mathbf{x}^{\mathscr{C}, D}=\boldsymbol{\Theta}^{\mathscr{E}, D}+\varrho \tag{3}
\end{equation*}
$$

where $\mathbf{x}^{\mathscr{C}, D}$ is the set of the wavelet coefficients of $\mathbf{x}$ and $\Theta^{\mathscr{C}, D} \subset \mathbf{R}$ is also the set of $\Theta$, where $\varrho \sim N_{n}\left(0, \sigma^{2} I_{n}\right)$. Leveldependent models play a significant role in wavelet appli-cations-this procedure allows us to investigate the value of unknown parameters at each resolution $j$ of wavelet coefficients. There are numerous methods for specifying values of unknown parameters of the double generalized Rayleigh distributions. Moreover, the MCMC algorithms are implemented to investigate the unknown parameters from complicated or nonstandard posterior distributions [22]. In statistics, there are many tools that can be applied to estimate parameters, such as EM and MCMC algorithms. In this article, two types of methods are supposed; the first one is the posterior mean (PM), and the second is maximum a posteriori (MAP).

Figure 1 illustrates the shape of the double Rayleigh distribution for different values of $\gamma$. It can be seen that as $\gamma \longrightarrow 0$, the density double Rayleigh approaches infinity, and this type of distribution can be used to fit the density of the empirical wavelet coefficients. More precisely, the wavelet coefficients are nearby the zero, which is found using the double generalized Rayleigh distribution with $\gamma=0$ and $0<\kappa \leq 0.5$. In the other words, the density double Rayleigh approaches infinity as $x$ approaches zero when $\kappa \in(0,0.5]$ and $\gamma=0$, which is equivalent to the summary of Mallat. This article is structured as follows: introduction to the double generalized Rayleigh distribution is explained in Section 2. All technical arguments are referred to in Sections 3 and 4. Numerical work confirming their features and simulation study to investigate estimation properties is provided in Sections 5 and 6. Section 7 gives the result of the proposed rule to real data. The final summary and conclusions are presented in Section 8.

## 2. Double Generalized Rayleigh Distribution

The generalized Rayleigh DGRay ( $\gamma_{j}, \kappa_{j}, \lambda_{j}$ ) distribution was proposed by Aykroyd et al. as a generalized distribution. They showed the properties of the model, such as cumulative and survivor functions. Also, Aykroyd et al. [23] showed that the generalized Rayleigh distribution works well to fit data. They also used the Bayesian approaches to estimate
unknown parameters of the generalized Rayleigh distribution. In this paper, a double generalized Rayleigh distribution will be used to model the wavelet coefficients,
equivalent to the density of the wavelet coefficients. Let single wavelet coefficient $\theta_{j, l}^{D}$ at the level $j$ be the probability density function (pdf) given by

$$
f\left(\theta_{j, l}^{D} \mid \lambda_{j}, \kappa_{j}, \gamma_{j}\right)=\left\{\begin{array}{lc}
\frac{\lambda_{j} \kappa_{j}}{2}\left|\theta_{j, l}^{D}\right|\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}-1} \exp \left\{-\frac{\lambda_{j}\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}}}{2}\right\}, & -\sqrt{\gamma_{j}}<\theta_{j, l}^{D}<\sqrt{\gamma_{j}}  \tag{4}\\
0, & O . \mathrm{W}
\end{array}\right.
$$

where |.| is the absolute value. The cumulative distribution function (cdf) is defined by

$$
F\left(\theta_{j, l}^{D} \mid \lambda_{j}, \kappa_{j}, \gamma_{j}\right)= \begin{cases}1-\exp \left\{-\frac{\lambda_{j}\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}}}{2}\right\}, & \theta_{j, l}^{D}>\sqrt{\gamma_{j}}  \tag{5}\\ \exp \left\{-\frac{\lambda_{j}\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}}}{2}\right\}, & \theta_{j, l}^{D}<-\sqrt{\gamma_{j}} \\ 0, & \text { O.W. }\end{cases}
$$

The survivor function (sf) is given by

$$
S\left(\theta_{j, l}^{D} \mid \lambda_{j}, \kappa_{j}, \gamma_{j}\right)= \begin{cases}\exp \left\{-\frac{\lambda_{j}\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}}}{2}\right\}, & \theta_{j, l}^{D}>\sqrt{\gamma_{j}},  \tag{6}\\ 1-\exp \left\{-\frac{\lambda_{j}\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}}}{2}\right\}, & \theta_{j, l}^{D}<-\sqrt{\gamma_{j}} \\ 0, & \text { O.W, }\end{cases}
$$

and the failure function (hrf) is given by

$$
h\left(\theta_{j, l}^{D} \lambda_{j}, \kappa_{j}, \gamma_{j}\right)= \begin{cases}\lambda_{j} \kappa_{j} \theta_{j, l}^{D}\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}-1}, & \theta_{j, l}^{D}>\sqrt{\gamma_{j}},  \tag{7}\\ \frac{\left(\lambda_{j} \kappa_{j} / 2\right)\left|\theta_{j, l}^{D}\right|\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}-1} \exp \left\{-\left(\lambda_{j}\left(t^{2}-\gamma_{j}\right)^{\kappa_{j}} / 2\right)\right\}}{1-\exp \left\{-\left(\lambda_{j}\left(\theta_{j, l}^{D 2}-\gamma_{j}\right)^{\kappa_{j}} / 2\right)\right\}}, & \theta_{j, l}^{D}<\sqrt{\gamma_{j}}, \\ 0, & \text { O.W, }\end{cases}
$$

where $\gamma_{j}>0, \kappa_{j}>0$ and $\lambda_{j}>0$. In some indexes, $\mathscr{J}=\log _{2}(n)$. The parameters $\lambda_{j}$ and $\kappa_{j}$ are shape
parameters, and $\gamma_{j}$ is a location parameter. Setting $\gamma_{j}=0$ and $\kappa_{j}=1$ in (4)-(6), the results of the standard of the double Rayleigh distribution with parameter $\lambda_{j}$ are obtained.

## 3. Bayesian Approach

In statistics, Bayesian tools play important roles, where the approach has two keys. The first one is the likelihood, concocted between observation and unknown parameters, say $p(\mathbf{x} \mid \zeta)$, where $\zeta$ and $\mathbf{x}$ are sets of underlying parameters and observations, respectively. The second key is the prior distribution, say $p(\zeta)$, and then the combining posterior distribution. Assuming the link between the model of $\mathbf{x}$ and the unknown of wavelet coefficients $\left(K^{T} \Theta^{D}\right)$,

$$
\begin{gather*}
p\left(\mathbf{x} \mid \Theta^{D}, \sigma^{2}\right)=\prod_{i=1}^{n}\left(\frac{1}{\sqrt{2 \pi \sigma^{2}}}\right)^{n} \\
\exp \left\{-\frac{1}{2 \sigma^{2}} \sum_{i=1}^{n}\left(x_{i}-\left(K^{T} \Theta^{D}\right)_{i}\right)^{2}\right\}  \tag{8}\\
\mathbf{x}, \Theta^{D} \subset \mathbf{R}^{n} ; \sigma>0
\end{gather*}
$$

$$
\begin{align*}
& p\left(\Theta^{D} \mid \mathbf{x}, \lambda_{\mathcal{F}-1}, \ldots, \lambda_{0}, \kappa_{\mathcal{F}-1}, \ldots, \kappa_{0}\right) \\
& =p\left(\mathbf{x} \mid K^{T} \boldsymbol{\Theta}^{D}\right) p\left(\theta_{\mathcal{F}-1}^{D} \mid \lambda_{\mathcal{F}-1}, \kappa_{\mathcal{F}-1}\right), \ldots, p\left(\theta_{0}^{D} \mid \lambda_{0}, \kappa_{0}\right) \\
& \propto \prod_{i=1}^{n} \exp \left\{-\sqrt{2 \tau} \sum_{i=1}^{n}\left|x_{i}-\left(K^{T} \Theta^{D}\right)_{i}\right|\right\} \times \frac{(2 \tau)^{n / 2} \lambda_{\mathcal{F}-1}^{n_{\mathcal{F}}} \kappa_{\mathcal{F}-1}^{n_{\mathcal{F}}}}{2^{n}} \prod_{1=0}^{n_{\mathcal{F}}}\left|\theta_{\mathcal{F}-1,1}^{D}\right|\left(\theta_{\mathcal{F}-1,1}^{D 2}\right)^{\kappa_{\mathcal{F}-1}-1} \\
& \times \exp \left\{-\frac{\lambda_{\mathcal{F}-1}\left(\theta_{\mathcal{F}-1,1}^{D 2}\right)^{\kappa_{\mathcal{F}-1}}}{2}\right\} \cdots \times \lambda_{0}^{n_{0}} \kappa_{0}^{n_{0}}\left|\theta_{0,0}^{D}\right| \times\left(\theta_{0,0}^{D 2}\right)^{\kappa_{0}-1} \exp \left\{-\frac{\lambda_{0}\left(\theta_{0,0}^{D 2}\right)^{\kappa_{0}}}{2}\right\}, \quad x_{i}, \theta_{j, l}^{D} \in \mathbf{R} ;\left(\lambda_{\mathcal{F}-1}, \ldots, \lambda_{0}, \kappa_{\mathcal{F}-1}, \ldots, \kappa_{0}\right)>0, \sigma>0, \tag{12}
\end{align*}
$$

where $n_{\mathcal{F}-1}, \ldots, n_{0}$ are the size of the coefficients at each level $\mathscr{J}-1, \ldots, 0$. Hence, the value of $\kappa_{j}$ is suggested as $0<\kappa_{j} \leq 0.5$. The main reason for choosing the double generalized Rayleigh is that as the value of $\gamma_{j} \longrightarrow 0$ and $\left|\theta_{j, l}^{D}\right| \longrightarrow 0$, the proposed distribution approaches infinity, which is followed by the saying of Mallat about the interpretation of the wavelet coefficients distribution. Clearly, equation (12) can be used to estimate the unknown parameters $\Theta^{D}$ given $\mathbf{x}, \lambda_{\mathcal{F}-1}, \ldots, \lambda_{0}, \kappa_{\mathcal{F}-1}, \ldots, \kappa_{0}$, and then these unknown parameters can be employed to describe the reconstruction. Hence, the unknown parameters are made up of one set, say with $\quad \zeta=\left\{\theta_{0,0}^{\mathscr{G}}, \theta_{1,0}^{\mathscr{E}}, \theta_{1,1}^{\mathscr{E}}, \ldots, \theta_{\mathcal{G}-1, n-1}^{\mathscr{E}}, \lambda_{\mathcal{g}-1}, \ldots, \lambda_{0}\right.$, $\left.\kappa_{\mathcal{F}-1}, \ldots, \kappa_{0}, \tau\right\}$, and then, the previous form (12) becomes
where $\sigma^{2}$ is the variance of data and can be assumed by

$$
\begin{equation*}
p\left(\sigma^{2} \mid \tau\right)=\tau \exp \left\{-\tau \sigma^{2}\right\}, \quad \tau>0 \tag{9}
\end{equation*}
$$

using equation (2) and the marginal likelihood given by

$$
\begin{align*}
p(x \mid \theta, \tau) & =\int_{0}^{\infty} p\left(x \mid \theta, \sigma^{2}\right) p\left(\sigma^{2} \mid \tau\right) \mathrm{d} \sigma^{2} \\
& =\frac{\sqrt{2 \tau}}{2} \exp \{-\sqrt{2 \tau}|x-\theta|\} \tag{10}
\end{align*}
$$

The result of the previous integration can be found in [24]. The equivalent likelihood is defined by

$$
\begin{array}{r}
p\left(\mathbf{x} \mid \Theta^{D}, \tau\right)=\prod_{i=1}^{n}\left(\frac{\sqrt{2 \tau}}{2}\right)^{n} \\
\exp \left\{-\sqrt{2 \tau} \sum_{i=1}^{n}\left|x_{i}-\left(K^{T} \Theta^{D}\right)_{i}\right|\right\}, \quad \mathbf{x}, \Theta^{D} \subset \mathbf{R}^{n} ; \sigma>0 . \tag{11}
\end{array}
$$

In addition, the posterior distribution for $\Theta^{D}$ given $\mathbf{x}$ is

$$
\begin{equation*}
p(\zeta \mid \mathbf{x}) \propto p(\mathbf{x} \mid \zeta) p(\zeta)=p\left(\mathbf{x} \mid K^{T} \Theta^{D}, \omega\right) p\left(\Theta^{D}\right) p(\omega) p(\tau) \tag{13}
\end{equation*}
$$

where $\Theta^{\mathscr{C}, D}=\left\{\theta_{0,0}^{\mathscr{C}}, \theta_{j, l}^{\mathscr{P}}: j=0,1, \ldots, \mathscr{F}-1, l=0,1, \ldots, n-\right.$ $1\}$ and suppose that $\omega=\{\tau, \lambda, \kappa\}$ at the level $j$. Aykroyd et al. considered gamma prior density for $\lambda$ and $\kappa$ with hyperprior parameters $\left(\alpha_{1}, \beta_{1}\right)$ and ( $\alpha_{2}, \beta_{2}$ ). Also, gamma distribution is proposed for $\tau$ with hyperparameters ( $\alpha_{3}, \beta_{3}$ ), with density function

$$
\begin{equation*}
p\left(\omega_{i} \mid \alpha_{i}, \beta_{i}\right)=\frac{1}{\Gamma\left(\alpha_{i}\right)} \beta_{i}^{\alpha} \omega^{\alpha_{i}-1} \exp \left\{-\beta_{i} \omega\right\}, \quad(\alpha, \beta)>0, i=1,2,3 . \tag{14}
\end{equation*}
$$

Then, the posterior density of the single value of $\theta_{j, l}^{D}$ with parameters $\tau, \lambda_{j}$, and $\kappa_{j}$ at the level $j$, given the data $x$, is given by

$$
\begin{array}{r}
p\left(\theta_{j, l}^{\circledast} \tau, \lambda_{j}, \kappa_{j} \mid x_{i}\right)=\frac{p\left(x_{i} \mid\left(K^{T} \Theta^{D}\right)_{i}, \tau\right) p(\tau) p\left(\lambda_{j}\right) p\left(\kappa_{j}\right)}{\int_{\theta_{j, l}^{D}} \int_{\lambda_{j}} \int_{\kappa_{j}} \int_{\tau} p\left(x_{i} \mid\left(K^{T} \Theta^{D}\right)_{i}, \tau\right) p(\tau) p\left(\lambda_{j}\right) p\left(\kappa_{j}\right) \mathrm{d} \kappa_{j} \mathrm{~d} \lambda_{j} \mathrm{~d} \theta_{j, l}^{D},} \\
p\left(\boldsymbol{\Theta}^{D}, \tau, \lambda_{\mathcal{f - 1}}, \ldots, \lambda_{0}, \kappa_{\mathcal{f}-1}, \ldots, \kappa_{0} \mid \mathbf{x}\right) \propto p\left(\mathbf{x} \mid K^{T} \boldsymbol{\Theta}^{D}, \tau\right) p(\tau) p\left(\boldsymbol{\theta}_{\mathcal{f - 1}}^{D}\right) \ldots  \tag{15}\\
\times p\left(\theta_{0}^{D}\right) p\left(\lambda_{\mathcal{f}-1}\right) \ldots p\left(\lambda_{0}\right) p\left(\kappa_{\mathcal{f - 1}}\right) \ldots p\left(\kappa_{0}\right),
\end{array}
$$

and the joint posterior density given data, $x$, can be written as

$$
\begin{align*}
& p\left(\Theta^{D}, \tau, \lambda_{\mathcal{F}-1}, \ldots, \lambda_{0}, \kappa_{\mathcal{F}-1}, \ldots, \kappa_{0} \mid \mathbf{x}\right) \\
& \quad \propto \frac{(2 \tau)^{n / 2} \tau^{n+\alpha_{1}-1} \lambda_{\mathcal{F}-1}^{n_{\mathcal{F}}+\alpha_{2, \mathcal{F}-1}-1} \ldots \lambda_{0}^{n_{0}+\alpha_{2,0}-1} \kappa_{\mathcal{F}-1}^{n_{\mathcal{F}}+\alpha_{3, \mathcal{F}-1}-1} \ldots \kappa_{0}^{n_{0}+\alpha_{3,0}-1}}{2^{n} \Gamma\left(\alpha_{1, n}\right) \beta_{1, n}^{\alpha_{1, n}} \Gamma\left(\alpha_{2, \mathcal{F}-1}\right) \beta_{2, \mathcal{F}-1}^{\alpha_{,-1}} \ldots \Gamma\left(\alpha_{2,0}\right) \beta_{2,0}^{\alpha_{2,0}} \Gamma\left(\alpha_{3, \mathcal{F - 1}}\right) \beta_{3, \mathcal{F}-1}^{\alpha_{3, \mathcal{L}}} \ldots \Gamma\left(\alpha_{3,0}\right) \beta_{3,0}^{\alpha_{3,0}}} \\
& \quad \exp \left\{-\left(\frac{\tau}{\beta_{1, n}}+\frac{\lambda_{\mathcal{F}-1}}{\beta_{2, \mathcal{F}-1}}+\cdots+\frac{\lambda_{0}}{\beta_{2,0}}+\frac{\kappa_{\mathcal{F}-1}}{\beta_{3, \mathcal{F}-1}}+\cdots+\frac{\kappa_{0}}{\beta_{3,0}}\right)\right\} \prod_{1=0}^{n_{\mathcal{F}-1}}\left|\theta_{\mathcal{J}-1,1}^{D}\right|\left(\theta_{\mathcal{J}-1,1}^{D}\right)^{2 \kappa_{\mathcal{F}-1}-1} \exp \left\{-\frac{\lambda_{\mathcal{F}-1}\left(\theta_{\mathcal{F}-1,1}^{D}\right)^{2 \kappa_{\mathcal{F}-1}}}{2}\right\} \\
& \quad \times \cdots\left|\theta_{0}^{D}\right|\left(\theta_{0}^{D}\right)^{2 \kappa_{0}-1} \exp \left\{-\frac{\lambda_{0}\left(\theta_{0}^{D}\right)^{2 \kappa_{0}}}{2}\right\} \prod_{i=1}^{n} \exp \left\{-\sqrt{2 \tau} \sum_{i=1}^{n}\left|x_{i}-\left(K^{T} \Theta^{D}\right)_{i}\right|\right\}, \quad x_{i}, \theta_{j, 1}^{D} \in \mathbf{R} ;\left(\lambda_{\mathcal{F}-1}, \ldots, \lambda_{0}, \kappa_{\mathcal{F}-1}, \ldots, \kappa_{0}\right)>0, \sigma>0 . \tag{16}
\end{align*}
$$

The hyperprior parameters $\tau$ $\kappa=\left\{\kappa_{1, \mathcal{F}-1}, \ldots, \kappa_{1,0}, \kappa_{2, \mathcal{F}-1}, \ldots, \kappa_{2,0}\right\}$, and $\beta=\left\{\beta_{1, \mathcal{F}-1}, \ldots, \beta_{1,0}, \beta_{2, \mathcal{F}-1}, \ldots, \beta_{2,0}\right\}$ can be fixed, as follows: let the expectation and variance of $\omega_{j}$ at resolution $j$, say $t_{i, j}$ and $r_{i, j}$, where $i=1,2,3$. By solving the following equations

$$
\begin{align*}
E\left(\omega_{j}\right) & =\alpha_{i, j} \beta_{i, j}, \\
\operatorname{Var}\left(\omega_{j}\right)=\alpha_{i, j} \beta_{i, j}^{2}, \quad i & =1,2,3, \tag{17}
\end{align*}
$$

the corresponding hyperprior parameters can be defined as $\alpha_{i, j}=t_{i, j}^{2} / r_{i, j}$ and $\beta_{i, j}=r_{i, j} / t_{i, j}$.

## 4. Stationary Approaches

The vital task in the wavelet approaches is to choose a basis. For more details, the interpretation of the wavelet basis is to start with two functions. The first one is scaling or father function $\phi$, where the main task of this function is to compute the scaling coefficients. The other is a wavelet or mother function $\psi$, where it can be used to calculate the wavelet coefficients. Several wavelet bases are now available
with different degrees of smoothness. However, the Haar basis is a simple version of the wavelet transform. Moreover, there are several established wavelet families demonstrated (see [25-29] for details). Stationary wavelet transforms (SWTs) attracted much attention for many applications over the last few years. In particular, the classical stationary wavelet transform was introduced in [30], while the authors in $[31,32]$ applied at that time as the maximal overlap for discrete wavelet.

In 1995, Nason extended the discrete wavelet and recalled it as the "stationary." In the same year, Ronald and David [33] proposed a new tool: stationary wavelet coefficients and is sometimes referred to as "cycle spinning." In general, the SWT can be described as "fills in the gaps" between the decimated wavelet coefficients; that is, there is no missing computation between two different values of wavelet coefficients. Nason stated that this leads to an overdetermined redundancy of the original data (see the below example for more explanation). The producer gives a shiftinvariant removing noisy tool, which simultaneously shows improvements in reconstruction quality (see Ronald and

David). For example, of the SWT, the Haar wavelet is applied to the data $\mathbf{x}=\left\{x_{1}, x_{2}, x_{3}, x_{4}\right\}$. The first and second sets of the scaling and detail coefficients can be computed:

$$
\begin{align*}
& \theta_{1, l}^{C}=\left[\begin{array}{cccc}
\frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 & 0 \\
0 & \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \\
0 & 0 & \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \\
\frac{\sqrt{2}}{2} & 0 & 0 & \frac{\sqrt{2}}{2}
\end{array}\right] \times\left[\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3} \\
x_{4}
\end{array}\right], \\
& \theta_{0, l}^{C}=\left[\begin{array}{cccc}
\frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2} & 0 \\
0 & \frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2} \\
\frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2} & 0 \\
0 & \frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2}
\end{array}\right] \times\left[\begin{array}{c}
\theta_{1,0}^{C} \\
\theta_{1,1}^{C} \\
\theta_{1,2}^{C} \\
\theta_{1,3}^{C}
\end{array}\right], \\
& \theta_{0, l}^{C}=\left[\begin{array}{cccc}
\frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2} & 0 \\
0 & \frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2} \\
\frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2} & 0 \\
0 & \frac{\sqrt{2}}{2} & 0 & \frac{\sqrt{2}}{2}
\end{array}\right] \times\left[\begin{array}{c}
\theta_{1,0}^{C} \\
\theta_{1,1}^{C} \\
\theta_{1,2}^{C} \\
\theta_{1,3}^{C}
\end{array}\right], \\
& \theta_{1, l}^{D}=\left[\begin{array}{cccc}
\frac{\sqrt{2}}{2} & \frac{-\sqrt{2}}{2} & 0 & 0 \\
0 & \frac{\sqrt{2}}{2} & \frac{-\sqrt{2}}{2} & 0 \\
0 & 0 & \frac{\sqrt{2}}{2} & \frac{-\sqrt{2}}{2} \\
\frac{-\sqrt{2}}{2} & 0 & 0 & \frac{\sqrt{2}}{2}
\end{array}\right] \times\left[\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3} \\
x_{4}
\end{array}\right], \tag{18}
\end{align*}
$$

where $\theta_{1,1}^{C}, \theta_{0,1}^{C}, \theta_{1,1}^{D}$, and $\theta_{0,1}^{D}$ are the matrices of transform at level $j=\{0,1\}$, respectively. Hence, the number $N$ of vanishing moments decreases. This implicates that the smoothness of the corresponding shape decreases. In this paper, Daubechies father function $\phi$ and mother $\psi$ with $N=$

8 vanishing moments are used to provide a smooth reconstruction.

The plotting procedure for the stationary wavelet transform is shown in Figure 2. It can be seen that each level $j$ has the same number of wavelet coefficients. Figure 3 shows the scaling and wavelet functions for Daubechies with $N=8$ vanishing moments. Table 1 shows the wavelet coefficients for Daubechies compact, phase $N=8$. Here, we present the idea of Daubechies, omitting some technical details.

## 5. Numerical Methods

The goal of the Bayesian computation is to extract a posterior sample for some unknown parameters $\zeta$. However, computational statistics can be explained as inverse problems. Some tools can be used to make the estimation more efficient. They include the standard version of the MCMC algorithms, Metropolis-Hastings tools, to extract a random sample from the posterior rule $p(\zeta \mid \mathbf{x})$ in (16). The procedure of the technique for parameter estimation, through the MCMC approach, can be found in [34]; for more information, see $[35,36$ ] and more recent works such as [37].

Figure 4 shows the diagram of the procedure of the proposed methods, where the procedure starts with data, which is corrupted by noise. The data are transformed to wavelet coefficients, which are used to estimate the unknown wavelet coefficients using the suggested method, and then, the underlying signal is calculated by inverting the estimation of wavelet coefficients. The main idea of the MCMC algorithms is that the parameter can take at any valued point in the parameter space $\Omega$, say $\zeta_{i}$ is the value point. Then, at each step, MCMC creates values, say $\zeta_{i}^{1}, \zeta_{i}^{2}, \ldots, \zeta_{i}^{r}$. Each single parameter updates separately in the order that the MCMC algorithms depend on a random walk. More precisely, the general framework of the tool is defined as follows:
(i) Starting with an initial value for $\Theta^{C, D}=0$ and for each level $j=0,1, \ldots, 2^{\mathcal{g}-1}$, that is, for parameters, let $\omega^{0}=\left\{\kappa_{0}^{0}, \kappa_{1}^{0}, \ldots \kappa_{2 \mathcal{F}-1}^{0}, \beta_{0}^{0}, \beta_{1}^{0}, \ldots \beta_{2^{\mathcal{g}-1}}^{0}\right\}$.
(ii) For times $k=1, \ldots, K$.
(1) Generate a new value $\omega_{j}^{*}=\omega_{j}^{k-1}+\varepsilon$, where $\varepsilon \sim N\left(0, \varsigma_{\omega, j}^{2 k-1}\right)$. Hence, the current value of the prior parameters is proposed with a variance parameter for each resolution $j$, which is chosen to obtain an acceptable convergence rate.
(2) Compute the posterior distribution in (16).

For $s=1,2, \ldots, n$, that is, for each wavelet coefficient $\theta_{j, l}^{D}$.
(a) Generate a new wavelet coefficient $\theta_{j, l}^{D *}=\theta_{j, l}^{D s-1}+\varepsilon$, where $\varepsilon \sim N\left(0, \varsigma_{\theta, j}^{2 s-1}\right)$.
(b) Again, compute the posterior distribution in (16).
(c) Generate $u \sim U(0,1)$.
(d) When $\alpha\left(\zeta^{*} \mid \zeta^{\zeta k}\right)=\min \left\{1,\left(p\left(\theta_{0,0}^{\mathscr{D}}, \theta_{1,0}^{\mathscr{D}}, \ldots, \theta_{j, l}^{\mathscr{D} *}, \ldots\right.\right.\right.$, $\theta_{\mathcal{G}-1, n-1}^{\mathscr{D}}, \lambda_{\mathcal{G}-1}, \ldots, \lambda_{j}^{*}, \ldots, \lambda_{0}, \kappa$



Figure 1: Typical data (points) derived from the generalized Rayleigh distribution (dashed line) along with different values of $\gamma$, while $\kappa=0.5$ and $\lambda=10$.


Figure 2: Graphical depiction of the stationary wavelet transform. The first row depicts the data, the second row indicates the wavelet coefficients, and the third shows the correspondence to the detailed wavelets. $h_{l}$ and $g_{l}$ are high- and low-pass quadrature mirror filters.


Figure 3: Plots of the father $\phi$ and mother $\psi$ wavelets with $N=8$ vanishing moments.
$\left.\left.\left.1, \ldots, \lambda_{0}, \kappa_{\mathcal{F}-1}, \ldots, \kappa_{j}^{k-1}, \ldots, \kappa_{0}\right)\right)\right\}>u, \quad$ accept the proposal and set $\theta_{j, l}^{D,}=\theta_{j, l}^{D *}$ and $\omega_{j}^{k}=\omega_{j}^{*}$; else, $\theta_{j, l}^{D s}=$ $\theta_{j, l}^{D s-1}$ and $\omega_{j}^{k}=\omega_{j}^{k-1}$.

Hence, all parameters are generated from the Gaussian distribution, while the current amount of the parameter is the expectation of the normal distribution with updating

Table 1: Orthogonal Daubechies coefficients for filter number 8.

| $l$ | 0 | 1 | 2 | 3 | 4 |
| :--- | :---: | :---: | :---: | :---: | :---: |
| $h_{l}$ | 0.0544158422 | 0.3128715909 | 0.6756307363 | 0.5853546837 | -0.0158291053 |
| $l$ | 5 | 6 | 7 | 8 | 9 |
| $h_{l}$ | -0.2840155430 | 0.0004724846 | 0.1287474266 | -0.0173693010 | -0.0440882539 |
| 1 | 10 | 11 | 12 | 13 | 14 |
| $h_{1}$ | 0.0139810279 | 15 | 0.0087460940 | -0.0048703530 | -0.0003917404 |
| $l$ | 16 | 17 | 18 | 0.0006754494 |  |
| $h_{1}$ | -0.0001174768 | 0 | 0 | 0 | 19 |



Figure 4: Diagram showing the structure of the suggested methods.
variance, which is based on the acceptance rate. It is essential to realize to pick up a random value around the current value, that is, both low and high, with variances $\varsigma_{\zeta, \mathrm{j}}^{2}$ chosen to depend on an acceptance rate. More precisely, choosing any valued point $\omega$ in the parameter space is accepted. The authors in [38] stated determining value is between $20 \%$ and $30 \%$ for an acceptance rate. Hence, we considered the following gamma prior density for the variance of noise $\sigma^{2}$, where the starting point is computed from the finest level of the wavelet coefficients, $\Theta_{\mathcal{f}-1}^{D}$, (see Nason).

Once the sample is collected from the posterior rule, the posterior mean for $\zeta$ can be calculated by

$$
\begin{equation*}
\bar{\zeta}=\bar{\zeta}=\frac{1}{K-M} \sum_{k=M+1}^{K} \zeta^{k}, \tag{19}
\end{equation*}
$$

and also, the posterior variance can be calculated by

$$
\begin{equation*}
\widehat{\sigma}^{2}=\frac{1}{K-M} \sum_{k=M+1}^{K}\left(\zeta^{k}-\bar{\zeta}\right)^{2} \tag{20}
\end{equation*}
$$

where $K$ and $M$ are the number of run and burn-in, respectively. Hence, there is an enormous method to compute the estimate point and interval. For the MAP rule, the previous procedure is changed into a simulated annealing
process of Geman and Geman; this process can answer more quickly than the posterior mean. More accurately the MAP estimate is chosen as the final iteration $\widehat{\theta}_{\mathrm{MAP}}=\widehat{\theta}^{K}$. In other words, sample mean and variance can not be computed. The maximum a posteriori estimator (MAP) is defined as

$$
\begin{equation*}
\hat{\zeta}_{\mathrm{MAP}}=\underset{\zeta}{\arg \max } p(\zeta \mid \mathbf{x})^{K} \tag{21}
\end{equation*}
$$

where $K$ indicates the final iteration of the run of the MCMC algorithms.

## 6. Simulation

The investigation of the proposed rule is considered. Then, the results are compared to some established wavelet-based methods. The authors in [39] introduced four test signals: bumps, Doppler, heavisine, and blocks. Moreover, these functions were corrupted by the independent Gaussian noise $N_{n}\left(0, I_{n} \varsigma_{\theta}^{2}\right)$. Different sizes are studied to investigate the proposed method's performance, which is $n=64$ and 128, where the four test functions were simulated. Also, various wavelet bases were used: Daubechies with $N=8$ applied for the test functions heavisine, Doppler, and bumps, while Haar basis was used for blocks. The starting level was $j_{0}=3$, as recommended in [40]. The average mean squared-error

Table 2: The results of the simulation based on different methods.

(AMSE) evaluated the results of the estimation, which is defined as

$$
\begin{equation*}
\mathrm{AMSE}=\frac{1}{K N} \sum_{j=1}^{K} \sum_{i=1}^{N}\left(\hat{\theta}_{j, i}-\theta_{i}\right)^{2} \tag{22}
\end{equation*}
$$

where $N$ and $K$ are the numbers of the data and the runs of the MCMC algorithms. Moreover, the results from the proposed method are denoted by $\hat{\theta}_{k, i}, i=1, \ldots, N$ at $k$-th run of MCMC algorithms.

The proposed estimators were compared to various methods, such as the Bayesian wavelet thresholding (BAYES.THR) method of Abramovich and Silverman, the ABWS rule of Chipman, Kolaczyk, and McCulloch, and the BAMS rule of Vidakovic and Ruggeri. Table 2 shows the results of the simulation when decimated and the nonstationary wavelet were used. It shows the result of AMSE; for our simulation, two bases are used. The first one is the basis with zero vanishing moments, and the other is the Daubechies' wavelets with $N=8$ vanishing moments. The proposed technique always gives the best reconstructions. The main interest is to improve the result of the reconstruction. This can be seen when the size of the sample is large because extensive observations contain massive information about the feature of the signal. In general, the MAP method provides a fair resolution in the test functions. However, the worst of the results is better than the other of the competed wavelet rules. The biggest problem in the MAP estimate is that the confidence intervals
can not be computed because the latest sample of posterior is picked.

## 7. Application to Medical Data

The suggested method is studied and investigated to a realworld inductance plethysmography data to evaluate the excellent performance of the proposed rules, compared to the state-of-the-art methods. The Department of Anaesthesia at the Bristol Royal Infirmary collected these observations. The number of observations is 2048, equally spaced points. Readers can obtain these data within WaveThresh using data (BabyECG). Also, the structure of the sleep state can be downloaded using data (BabySS). Figure 5 shows the plots of BabyECG and sleep state. Hence, the aim of the investigation of the BabyECG was to specify the sleep state successfully from the observations. These data were studied and investigated by other authors (for example, [41]). The reconstruction of the unbalanced Haar approach (red line) is illustrated in Figure 6. It is not accessible to describe every moment using the unbalanced Haar method or to talk in general about the sleep state for the babies. Figures 7 and 8 show the reconstructions of the underlying feature with the MAP method using the Haar wavelet basis and Daubechies wavelet with $N=8$ vanishing moments. In our reconstruction, the value of the shape $\kappa_{j}$ is set within the interval $(0,0.4)$. Table 3 shows the results of the simulations using the MAP and PM estimators. As the level $j$ decreases, the value of $\kappa$ increases. In contrast, the value of the parameter $\lambda$ is slightly changed.


Figure 5: Plots of BabyECG data (solid line) and sleep state (dashed line).


Figure 6: Plots of the reconstructions using the unbalanced Haar estimator (red line) and BabyECG data (black line).


Figure 7: Plots of the reconstructions using the MAP estimator with a Haar basis.


Figure 8: Plots of the reconstructions using the MAP estimator with $N=8$ vanishing moments.
Table 3: The results of the simulation based on different methods.

| Number of vanishing moments | $\kappa_{4}$ | $\kappa_{5}$ | $\kappa_{6}$ | $\kappa_{7}$ | $\kappa_{8}$ | $\kappa_{9}$ | $\kappa_{10}$ | $\kappa_{11}$ | $\lambda_{4}$ | $\lambda_{5}$ | $\lambda_{6}$ | $\lambda_{7}$ | $\lambda_{8}$ | $\lambda_{9}$ | $\lambda_{10}$ | $\lambda_{11}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| MAP estimator |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $N=0$ | 0.3811 | 0.2259 | 0.2307 | 0.2249 | 0.2259 | 0.2287 | 0.0103 | 0.0121 | 0.0686 | 0.0411 | 0.1228 | 0.0943 | 0.1230 | 0.1798 | 0.1321 | 0.0614 |
| $N=8$ | 0.3478 | 0.2230 | 0.2188 | 0.2243 | 0.2172 | 0.2101 | 0.0003 | 0.0001 | 0.1053 | 0.0648 | 0.1016 | 0.0519 | 0.1430 | 0.0548 | 0.0229 | 0.0941 |
| PM estimator |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $N=0$ | 0.3992 | 0.2220 | 0.2211 | 0.2243 | 0.2201 | 0.2322 | 0.0002 | 0.0013 | 0.0232 | 0.0966 | 0.0178 | 0.0646 | 0.0214 | 0.1528 | 0.1007 | 0.0232 |
| $N=8$ | 0.2718 | 0.2150 | 0.2347 | 0.2215 | 0.2131 | 0.2236 | 0.0003 | 0.0001 | 0.1600 | 0.1376 | 0.1534 | 0.0647 | 0.0257 | 0.1622 | 0.0351 | 0.0476 |

## 8. Conclusion

In this article, we show various ways in which the Bayesian rules and wavelet methods were used successfully in the practical problem. Also, a procedure for estimating the scale parameters, $k$ and $\lambda$, of double generalized Rayleigh was estimated based on the BabyECG sample. This approach was adopted from the wavelet method for the independent level $j$ and Bayesian approaches. Prior probability distributions for the parameters were assumed to be gamma distribution. Bayesian estimates for the points were proposed in the cases of artificial samples under the squared-error loss. The simulation studies are showing that the proposed rules worked well, and the proposed Bayesian estimate performed better than the existing state-of-the-art methods based on signal functions by reducing the AMSE. We discussed the proposed method estimates to estimate the underlying parameters. Numerical results were obtained to compare the theoretical performance results. Some points are observed from numerical results, which are summarized as follows:
(i )From the results in Table 2, the suggested method process provides better excellent results for artificial data.
(ii) Estimation results under the PM and MAP methods provide better estimation than the other established wavelet denoising methods according to the MSE.
(iii) The use of the suggested method allows to describe the main feature of the real data, especially when observations are large.
This paper has confirmed that the wavelet approach provides attractive alternatives to other established wavelet methods, especially when underlying signals are inhomogeneous.

## Data Availability

The data used to support the findings of this study are included within the article.

## Conflicts of Interest

The author declares that there are no conflicts of interest.

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# Combined Effects of Heat and Mass Transfer on MHD Free Convective Flow of Maxwell Fluid with Variable Temperature and Concentration 

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Heat and mass transfer combined effects on MHD natural convection for a viscoelastic fluid flow are investigated. The dynamics of the fluid are controlled by the motion of the plate with arbitrary velocity along with varying temperature and mass diffusion. The non-dimensional forms of the governing equations of the model are developed along with generalized boundary conditions and the resulting forms are solved by the classical integral (Laplace) transform technique/method and closed-form solutions are developed. Obtained generalized results are very important due to their vast applications in the field of engineering and applied sciences; few of them are highlighted here as limiting cases. Moreover, parametric analysis of system parameters $P_{r}, S, K_{c}, G_{T}, G_{c}, M, S_{c}, \lambda$ is done via graphical simulations.

## 1. Introduction

In science and in many engineering applications such as in condensation, evaporation, and chemical process, many transport processes are influenced by the combined action of the buoyancy forces from both heat and mass diffusion. Heat and mass transfer combined effects are studied extensively due to their significant role in chemical processing equipment, oceanic circulation, emergency cooling system of advanced nuclear reactors, cooling process of plastic sheets, formation and dissipation of the fog, processing and drying the food, temperature distribution and moisture of agriculture fields, and production of polymer. In recent years, a lot of practical applications attracted many scientists and engineers to pay a considerable amount of focus to learn the heat and mass effects either analytically or numerically [1-4].In
industrial and engineering processes, most fluids are nonNewtonian. Since the non-Newtonian fluids deal more complexities due to the rheological behavior than Newtonian fluids, distinct models were proposed. The influence of heat and mass transfer in the non-Newtonian fluid is an important subject from the theoretical as well as practical point of view due to its abundant applications in industry and engineering. Common examples include polymer extrusion, the emergency cooling system of nuclear reactors, food processing, thermal welding, to name a few. Convective flow is a selfsustained flow with the effect of the temperature gradient. In literature, different theories are made to see the occurrence of heat and mass transfer in convective flows of different fluids. Mebarek-Oudina et al. [5] investigated the natural convective heat transfer phenomenon of water-based hybrid nanofluid in a porous medium along with the magnetic field. Das et al. [6]
considered the natural convective flow of the electrically conducting fluid past on vertical plate embedded in a permeable medium and explored the impacts of heat and mass transfer. The heat transfer phenomenon of Casson nanofluid flow is taken into account by Abo-Dahab et al. [7]. They analyzed the problem with the convective boundary conditions and discussed the influence of chemical reaction and heat source. Sajad et al.[8] studied the heat transfer and magnetic effect on hybrid nanofluid. Nazish et al. [9] explored the influence of heat and concentration/mass transform with the existence of fields developed by magnetic in the Maxwell fluid model. Ahmad [10] explored the heat transfer for the Maxwell fluid on the stretching plate with the slip boundary on the velocity. They explored the numerical solutions and showed the heat flux effect using the Nusselt number and the Prandtl number. A computational analysis is performed to study the effects of the transverse magnetic field at the unsteady Poiseuille-Rayleigh-Benard flow by Marzougui et al. [11]. The thermal properties' effects on the soil temperature are modeled and investigated numerically by Belatrache [12]. To have more insight about heat and mass transfer mechanisms in fluid flow and their applications, readers are referred to review references [13-16].On the other hand, many researchers paid a significant amount of attention to the study of MHD free convective flows due to its numerous applications in solar and stellar structure, radio propagation, MHD pumps, MHD bearings, aerodynamics, polymer technology, petroleum industry, crude oil purification, glass fiber drawing, etc. In light of these applications, many researchers such as Rajput [17], Gupta [18], and El Amin [19] studied the MHD flow of different fluids. They found the exact solution for velocity, concentration, and temperature by the Laplace transform method. Heat and mass transfer simultaneous effects on MHD flow of Maxwell fluid have been investigated by Nadeem et al. [20]. Recently, the study of the unsteady boundary layer heat transfer of Maxwell viscoelastic fluid was carried out by Zhao et al. [21]. Ahmad [22] studied MHD viscous, with constant density, electro-conducting fluid in the existence of the radiation, thermal diffusion, free convection, and mass transfer flow. These results motivated Chaudhry et al. [23] and they used classical integral transform to obtain the exact solutions of natural MHD convective flow past on an accelerated surface submerged in a permeable medium. Das [24] developed the closed-form solution of the unsteady

MHD natural convection flow on a moving vertical plate accompanied by mass transfer and thermal radiation. Carrying on, Das et al. [25] investigated the time-dependent MHD natural convection flow past a moving vertical plate dipped in a porous medium and studied the different aspects of heat and mass transfer. They discussed the problem with the uniform, oscillating, and impulsive motions of the plate besides considering the constant heat and mass diffusion and implemented the Laplace integral transform to develop the analytic solutions.Motivated by these investigations, the objective of this manuscript is to study the combined effect of heat and mass transfer on MHD Maxwell fluid. Laplace integral transformation is used to obtain the unique solution of temperature, velocity, and concentration under the impact of generalized boundary conditions on temperature, velocity, and concentration. The importance of the problem is highlighted by showing its impact/applications in the field of engineering and applied sciences. The paper is organized into six sections. After the introductory section in Section 2, the dimensionless governing equations are developed. In Section 3, Laplace integral transform is implemented to find the exact solution of the temperature, velocity, and concentration field. In Section 4, some applications in different fields are discussed as limiting cases to justify our results. In Section 5, the effect of physical parameters is analyzed graphically. The concluding observation is listed at the end.

## 2. Problem Formulation

We studied here the motion of the viscoelastic, incompressible, electronically conducting Maxwell fluid due to plate motion with arbitrary velocity $u_{0} f^{\prime}\left(t^{\prime}\right)$. The plate is along $x$-axis and $y$-axis is considered normal on the plate. In the first instance, at $t=0$ the plate and fluid are at temperature $T_{\infty}{ }^{\prime}$ and concentration $C_{\infty}{ }^{\prime}$. With the time $t=0^{+}$, the plate starts to move in its own axis. Then, the level of temperature and concentration takes up to $T_{\infty}\{\hat{\prime}\}+T_{w}\{\hat{\prime}\} h^{\prime}\left(t^{\prime}\right)$ and $C_{\infty}\{\hat{\prime}\}+C_{w}\left\{\prime^{\prime}\right\} g^{\prime}\left(t^{\prime}\right)$ where $f^{\prime}\left(t^{\prime}\right), h^{\prime}\left(t^{\prime}\right)$ and $g^{\prime}\left(t^{\prime}\right)$ are piecewise continuous functions that vanish at $t=0$. Details of different parameters are given in Table 1. Momentum, energy, and concentration equations are formed as follows:

Table 1: Nomenclature.

| Symbol | Quantity |
| :--- | ---: |
| $u$ | Velocity of fluid |
| $B_{0}$ | Magnetic field parameter |
| $q$ | Laplace transforms parameter |
| $D$ | Mass diffusivity |
| $B_{T}$ | Thermal expansion parameter |
| $B_{C}$ | Concentration expansion coefficient |
| $K$ | Thermal conductivity |
| $\rho$ | Density of fluid |
| $\lambda$ | Relaxation time |
| $\sigma$ | Electric conductivity coefficient |
| $\mu$ | Dynamic viscosity |
| $v$ | Kinematic viscosity |
| $c_{p}$ | Specific heat |
| $S$ | Heat source parameter |
| $K_{c}$ | Chemical reaction coefficient |
| $g$ | Gravitational acceleration |
| $S_{C}$ | Schmidt number |
| $M$ | Parameter due to magnetic field |
| $P_{r}$ | Prandtl number |
| $G_{T}$ | Grashof number due to thermal effect |
| $G_{C}$ | Grashof number due to concentration |

$$
\begin{align*}
\left(1+\lambda^{\prime} \frac{\partial}{\partial t^{\prime}}\right) \frac{\partial u^{\prime}\left(\chi^{\prime}, t^{\prime}\right)}{\partial t^{\prime}}= & \nu \frac{\partial^{2} u^{\prime}\left(\chi^{\prime}, t^{\prime}\right)}{\partial \chi^{\prime 2}} \\
& +g B_{T}\left(1+\lambda^{\prime} \frac{\partial}{\partial t^{\prime}}\right)\left(\varphi^{\prime}-\varphi_{\infty}^{\prime}\right)+g B_{C} \\
\left(1+\lambda^{\prime} \frac{\partial}{\partial t^{\prime}}\right)\left(C^{\prime}-C_{\infty}^{\prime}\right) \quad & -\frac{\sigma B_{0}^{2}}{\rho}\left(1+\lambda^{\prime} \frac{\partial}{\partial t^{\prime}}\right) u^{\prime}\left(\chi^{\prime}, t^{\prime}\right) \\
\rho C_{p} \frac{\partial \varphi^{\prime}\left(\chi^{\prime}, t^{\prime}\right)}{\partial t^{\prime}}= & K \frac{\partial^{2} \varphi^{\prime}\left(\chi^{\prime}, t^{\prime}\right)}{\partial \chi^{\prime 2}}-S^{\prime}\left(\varphi^{\prime}-\varphi_{\infty}^{\prime}\right) \\
\frac{\partial C^{\prime}\left(\chi^{\prime}, t^{\prime}\right)}{\partial t^{\prime}}= & D \frac{\partial^{2} C^{\prime}\left(\chi^{\prime}, t^{\prime}\right)}{\partial \chi^{\prime 2}}-K_{c}^{\prime}\left(C^{\prime}-C_{\infty}^{\prime}\right) \tag{1}
\end{align*}
$$

The imposed initial and boundary conditions are

$$
\begin{aligned}
t^{\prime} \leq 0, \quad u^{\prime}\left(\chi^{\prime}, t^{\prime}\right) & =0, \\
\varphi^{\prime}\left(\chi^{\prime}, t^{\prime}\right) & =\varphi_{\infty}^{\prime}, \\
C^{\prime}\left(\chi^{\prime}, t^{\prime}\right) & =C_{\infty}^{\prime}, \quad \chi^{\prime} \geq 0, \\
t^{\prime} \geq 0, \quad u^{\prime}\left(0, t^{\prime}\right) & =u_{0} f^{\prime}\left(t^{\prime}\right), \\
\varphi^{\prime}\left(0, t^{\prime}\right) & =\varphi_{\infty}^{\prime}+\varphi_{w}^{\prime} h^{\prime}\left(t^{\prime}\right), \\
C^{\prime}\left(0, t^{\prime}\right) & =C_{\infty}^{\prime}+C_{w}^{\prime} g^{\prime}\left(t^{\prime}\right), \\
u^{\prime}\left(\chi^{\prime}, t^{\prime}\right) & \longrightarrow 0, \\
\varphi^{\prime}\left(\chi^{\prime}, t^{\prime}\right) & \longrightarrow \varphi_{\infty}^{\prime}, \\
C^{\prime}\left(\chi^{\prime}, t^{\prime}\right) \longrightarrow C_{\infty}^{\prime}, \quad \chi^{\prime} & \longrightarrow \infty .
\end{aligned}
$$

For dimensionless problem, we use the following relations:

$$
\begin{align*}
& t=\frac{u_{0}^{2}}{v} t^{\prime}, \\
& y=\frac{u_{0}}{v} \chi^{\prime}, \\
& S_{c}=\frac{\nu}{D}, \\
& K_{c}=\frac{\nu}{u_{0}^{2}} K_{c}^{\prime}, \\
& \lambda=\frac{u_{0}^{2}}{v} \lambda^{\prime},  \tag{4}\\
& u^{\prime}=u_{0} u,  \tag{5}\\
& C=\frac{C^{\prime}-C_{\infty}^{\prime}}{C_{w}^{\prime}},  \tag{6}\\
& T=\frac{\varphi^{\prime}-\varphi_{\infty}^{\prime}}{\varphi_{w}^{\prime}},  \tag{7}\\
& S=\frac{v}{\rho C_{p} u_{0}^{2} S^{\prime}},  \tag{3}\\
& P_{r}=\frac{\mu C_{p}}{K},  \tag{9}\\
& K_{p}^{\prime}=\frac{u_{0}^{2}}{v^{2}} K_{p} G_{c}=\frac{g\left(B_{C} / \nu\right)}{u_{0}^{3}} C_{w}^{\prime}, \\
& G_{T}=\frac{g B_{T} v}{u_{0}^{3}} \varphi_{w}^{\prime}, \\
& M=\frac{\sigma B_{0}^{2} v}{\rho u_{0}^{2}},  \tag{10}\\
& f=\frac{v}{u_{o}^{2}} f^{\prime},  \tag{11}\\
& g=\frac{v}{u_{o}^{2}} g^{\prime}, \\
& h=\frac{v}{u_{o}^{2}} h^{\prime} . \tag{12}
\end{align*}
$$

After non-dimensionalizing, the governing equations become

$$
\begin{aligned}
\left(1+\lambda \frac{\partial}{\partial t}\right) \frac{\partial u(y, t)}{\partial t}= & \frac{\partial^{2} u(y, t)}{\partial y^{2}}+G_{T}\left(1+\lambda \frac{\partial}{\partial t}\right) T(y, t) \\
& +G_{C}\left(1+\lambda \frac{\partial}{\partial t}\right) C(y, t) \\
& -M\left(1+\lambda \frac{\partial}{\partial t}\right) u(y, t)
\end{aligned}
$$

$$
\begin{aligned}
& \frac{\partial T(y, t)}{\partial t}=\frac{1}{P_{r}} \frac{\partial^{2} T(y, t)}{\partial y^{2}}-S T(y, t), \\
& \frac{\partial C(y, t)}{\partial t}=\frac{1}{S_{c}} \frac{\partial^{2} C(y, t)}{\partial y^{2}}-K_{c} C(y, t) .
\end{aligned}
$$

along the following initial and boundary conditions:

$$
\begin{gathered}
u(y, 0)=0, T(y, 0)=0, C(y, 0)=0, \\
u(0, t)=f(t), T(0, t)=h(t), C(0, t)=g(t), \\
u(y, t) \longrightarrow 0, T(y, t) \longrightarrow 0, C(y, t) \longrightarrow 0, \quad \text { as } y \longrightarrow \infty
\end{gathered}
$$

## 3. Solution of the Problem

3.1. Concentration. Transforming equation (7) after applying the Laplace integral transform and utilizing the corresponding initial condition, we get

$$
\frac{\partial^{2} \bar{C}(y, q)}{\partial y^{2}}-S_{c}\left(K_{c}+q\right) \bar{C}(y, q)=0
$$

The above differential equation solution is

$$
\bar{C}(y, q)=C_{1} e^{-y \sqrt{s_{c}\left(K_{c}+q\right)}}+C_{2} e^{y \sqrt{s_{c}\left(K_{c}+q\right)}} .
$$

The solution of equation (11) with the transformed form of boundary conditions becomes

$$
\bar{C}(y, q)=G(q) e^{-y \sqrt{S_{c}\left(K_{c}+q\right)}} .
$$

Applying the Laplace inverse on equation (12) and using the $L^{-1}\{G(q)\}=g^{\prime}(t)$ with $g(0)=0$, convolution theorem and equation (A.22), the generalized solution for concentration is

$$
\begin{equation*}
C(y, t)=\int_{0}^{t} g^{\prime}(t-s) \Phi\left(y, \sqrt{S_{c}}, s ; K_{c}\right) \mathrm{d} s \tag{13}
\end{equation*}
$$

and $\Phi$ is specified in equation (A.23).
3.2. Temperature Distribution. Implementing the Laplace transform on equation (5) and using the concerned the initial condition, we get

$$
\begin{equation*}
\frac{\partial^{2} \bar{T}(y, q)}{\partial y^{2}}-P_{r}(S+q) \bar{T}(y, q)=0 \tag{14}
\end{equation*}
$$

The solution is

$$
\begin{equation*}
\bar{T}(y, q)=C_{1} e^{-y \sqrt{P_{r}(S+q)}}+C_{2} e^{y \sqrt{P_{r}(S+q)}} \tag{15}
\end{equation*}
$$

After implementing the boundary conditions, equation (15) becomes

$$
\begin{equation*}
\bar{T}(y, q)=H(q) e^{-y \sqrt{P_{r}(S+q)}} \tag{16}
\end{equation*}
$$

The Laplace inverse on equation (16) and using the $L^{-1}\{H(q)\}=h^{\prime}(t)$ with $h(0)=0$, convolution theorem and equation (A.24), the generalized solution for temperature obtained is

$$
\begin{equation*}
T(y, t)=\int_{0}^{t} h^{\prime}(t-s) \Psi\left(y, \sqrt{P_{r}}, s ; S\right) \mathrm{d} s \tag{17}
\end{equation*}
$$

where $\Psi$ is defined in equation (A.25).
3.3. Velocity. Employing the Laplace transform on equation (4) and using the corresponding initial condition on velocity form the following differential equation:

$$
\begin{align*}
\frac{\partial^{2} \bar{u}(y, q)}{\partial y^{2}}- & ((1+\lambda q)(q+M)) \bar{u}(y, q) \\
& \left.=-G_{T}(1+\lambda q) \bar{T}(y, q)\right)-G_{c}(1+\lambda q) \bar{C}(y, q) \tag{18}
\end{align*}
$$

In order to solve equation (18), we use the value of $\bar{C}(y, q), \bar{T}(y, q)$ from equation (12) and equation (16), respectively. With boundary conditions use on velocity, the following solution is obtained:

$$
\begin{align*}
\bar{u}(y, q)= & F(q) e^{-y \sqrt{(q+M)(1+\lambda q)}}+\frac{G_{T} H(q)(1+\lambda q)\left(e^{-y \sqrt{(q+M)(1+\lambda q)}}-e^{-y \sqrt{P_{r}(S+q)}}\right)}{\left(P_{r}(S+q)-(1+\lambda q)(q+M)\right)} \\
& +\frac{G_{c} G(q)(1+\lambda q)\left(e^{-y \sqrt{(q+M)(1+\lambda q)}}-e^{-y \sqrt{S_{c}\left(K_{c}+q\right)}}\right)}{\left(S_{c}\left(K_{c}+q\right)-(1+\lambda q)(q+M)\right)} \tag{19}
\end{align*}
$$

Further simplification reduces equation (19):

$$
\begin{align*}
\bar{u}(y, q)= & F(q) e^{-y \sqrt{\lambda\left(\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right)}} \\
& -\frac{G_{T} H(q)(1+\lambda q)\left(e^{-y \sqrt{\lambda\left[\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right]}}-e^{-y \sqrt{P_{r}(S+q)}}\right)}{\lambda\left(\left(q+\alpha_{4}\right)^{2}-\alpha_{6}^{2}\right)} \\
& -\frac{G_{c} G(q)(1+\lambda q)\left(e^{-y \sqrt{\lambda\left[\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right]}}-e^{-y \sqrt{S_{c}\left(K_{c}+q\right)}}\right)}{\lambda\left(\left(q+\alpha_{7}\right)^{2}-\alpha_{9}^{2}\right)}, \tag{20}
\end{align*}
$$

where $2 \alpha_{1}=(\lambda M+1 / \lambda), \alpha_{2}=(M / \lambda), \alpha_{3}^{2}=\alpha_{2}-\alpha_{1}^{2}, 2 \alpha_{4}=$ $\left(\lambda M+1-P_{r} / \lambda\right), \alpha_{5}=\left(M-P_{r} S / \lambda\right), \quad \alpha_{6}^{2}=\alpha_{4}^{2}-\alpha_{5}, 2 \alpha_{7}=$ $\left(\lambda M+1-S_{c} / \lambda\right), \alpha_{8}=\left(M-S_{c} K_{c} / \lambda\right), \alpha_{9}^{2}=\alpha_{7}^{2}-\alpha_{8}$.

Generalized expression for velocity field is acquired by employing the inverse Laplace transform on equation (20):

$$
\begin{equation*}
u(y, t)=I_{1}+\frac{G_{T}}{\lambda} I_{2}-\frac{G_{T}}{\lambda} I_{3}+\frac{G_{c}}{\lambda} I_{4}-\frac{G_{c}}{\lambda} I_{5} \tag{21}
\end{equation*}
$$

where

$$
\begin{align*}
I_{1} & =L^{-1}\left(F(q) e^{-y \sqrt{\lambda\left(\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right)}}\right)  \tag{22}\\
& =L^{-1}\left(B_{1}(y, q)\right) * L^{-1}\left(B_{2}(q)\right) \\
& =B_{1}(y, t) * B_{2}(t)
\end{align*}
$$

where

$$
B_{1}(y, q)=\left(e^{-\sqrt{\lambda} y \sqrt{\left(\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right)}} / \sqrt{\left(\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right)}\right)
$$ $B_{2}(q)=F(q) \sqrt{\left(\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right)}$.

By using equation (A.20) and equation (A.21), expressions for the $B_{1}(y, t)$ and $B_{2}(t)$ are evaluated as follows:

$$
\left.\begin{array}{rl}
B_{1}(y, t)= & \begin{cases}0, & 0<t<y \sqrt{\lambda}, \\
e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} \sqrt{t^{2}-\lambda y^{2}}\right), & t>y \sqrt{\lambda},\end{cases} \\
B_{2}(t)=\left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(f^{\prime}(t)+\alpha_{1} f(t)\right) \\
& +\left(\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)\right)^{*} f(t), \tag{24}
\end{array}\right\}
$$

where $B_{1}(y, t)$ is given in equation (23) and

$$
\begin{align*}
B_{3}(t)= & \left(h^{\prime}(t)+\lambda H(t)^{*} h^{\prime}(t)\right) \\
& *\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right)  \tag{27}\\
& { }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),
\end{align*}
$$

where

$$
\begin{aligned}
A^{*}= & \frac{\left(\alpha_{1}^{2}+\alpha_{3}^{2}\right)}{\alpha_{4}^{2}-\alpha_{6}^{2}}, \\
B^{*}= & \frac{2\left(\alpha_{4}-\alpha_{6}\right) A^{*}-\left(2 \alpha_{1}-\alpha_{4}-\alpha_{6}\right)}{2 \alpha_{6}}, \\
C^{*} & =1-A^{*}-B^{*} . \\
I_{4}= & L^{-1}\left(\frac{G(q)(1+\lambda q) e^{-y \sqrt{s_{c}\left(K_{c}+q\right)}}}{\left(\left(q+\alpha_{7}\right)^{2}-\alpha_{9}^{2}\right)}\right), \\
= & \left(g(t)+\lambda g^{\prime}(t)\right)^{*}\left(e^{-\alpha_{\tau} t} \cosh \left(\alpha_{9} t\right)\right. \\
& \left.+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right), \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t .}
\end{aligned}
$$

Similarly,

$$
\begin{align*}
I_{5} & =L^{-1}\left(\frac{G(q)(1+\lambda q) e^{-y} \sqrt{\lambda\left(\left(q+\alpha_{1}\right)^{2}+\alpha_{3}^{2}\right)}}{\left(\left(q+\alpha_{7}\right)^{2}-\alpha_{9}^{2}\right)}\right),  \tag{30}\\
& =B_{1}(y, t)^{*} B_{4}(t),
\end{align*}
$$

where $B_{1}(y, t)$ is given in equation (23) and

$$
\begin{align*}
B_{4}(t)= & \left(g^{\prime}(t)+\lambda H(t)^{*} g^{\prime}(t)\right) \\
& *\left(C^{*}+D^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+E^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right) \\
& * e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right), \\
D^{*}= & \frac{\left(\alpha_{1}^{2}+\alpha_{3}^{2}\right)}{\alpha_{7}^{2}-\alpha_{9}^{2}},  \tag{31}\\
E^{*}= & \frac{2\left(\alpha_{7}-\alpha_{9}\right) A^{*}-\left(2 \alpha_{1}-\alpha_{7}-\alpha_{9}\right)}{2 \alpha_{9}}, \\
F^{*}= & 1-D^{*}-E^{*} .
\end{align*}
$$

The above results are obtained for generalized time-dependent boundary conditions on velocity, concentration, and temperature. These results have many applications in engineering and applied science. Now, we will consider and discuss its few applications.

## 4. Applications

4.1. Application 1: $f(t)=H(t), g(t)=H(t), h(t)=H(t)$. This function value shows the motion of the fluid is because of the motion of an infinite plate in its plane with constant velocity. This function has importance in a lot of engineering problems such as signal waves, driving forces that act for a short time only, and impulsive forces acting for an instance such as a hammer blow.Substituting the value of $G(q)=$ ( $1 / q$ ) into equation (12) and applying the Laplace inverse, the expression for concentration is

$$
\begin{equation*}
C(y, t)=\left(\delta(t)+K_{c} t\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} \tag{32}
\end{equation*}
$$

where $\delta($.$) is known as Dirac delta function.$
Embedding the value of $H(q)=(1 / q)$ into equation (16) and taking Laplace inverse make the expression of temperature

$$
\begin{equation*}
T(y, t)=(\delta(t)+S t)^{*} \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t} \tag{33}
\end{equation*}
$$

The equation of velocity

$$
\begin{equation*}
u(y, t)=I_{1}^{I}+\frac{G_{T}}{\lambda} I_{2}^{I}-\frac{G_{T}}{\lambda} I_{3}^{I}+\frac{G_{c}}{\lambda} I_{4}^{I}-\frac{G_{c}}{\lambda} I_{5}^{I} \tag{34}
\end{equation*}
$$

where

$$
\begin{equation*}
I_{1}^{I}=B_{1}(y, t)^{*} B_{2}^{I}(t), \tag{35}
\end{equation*}
$$

$B_{2}^{I}(t)$ is obtained as

$$
\begin{align*}
B_{2}^{I}(t)= & \left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(\delta(t)+\alpha_{1} H(t)\right) \\
& +\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)^{*} H(t), \tag{36}
\end{align*}
$$

and for $B_{1}(y, t)$ (see equation (A.1)).
After substituting the value of $H(q)=(1 / q)$ into equation (25)

$$
\begin{align*}
I_{2}^{I}= & (H(t)+\lambda \delta(t))\left(e^{-\alpha_{4} t} \cosh \left(\alpha_{6} t\right)\right. \\
& \left.+\frac{\left(S-\alpha_{4}\right)}{\alpha_{6}} e^{-\alpha_{4} t} \sinh \left(\alpha_{6} t\right)\right)  \tag{37}\\
& * \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t}
\end{align*}
$$

Equation (26) takes the form after employing the value of $H(q)=(1 / q)$

$$
\begin{equation*}
I_{3}^{I}=B_{1}(y, t)^{*} B_{3}^{I}(t) \tag{38}
\end{equation*}
$$

where

$$
\begin{align*}
B_{3}^{I}(t)= & \left(\delta(t)+\lambda H(t)^{*} \delta(t)\right) \\
& *\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C^{*} e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right)  \tag{39}\\
& * e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),
\end{align*}
$$

and for $B_{1}(y, t)$ (see equation (A.1)).
After substituting the value of $G(q)=(1 / q)$ into equation (29)

$$
\begin{align*}
I_{4}^{I}= & (H(t)+\lambda \delta(t))\left(e^{-\alpha_{7} t} \cosh \left(\alpha_{9} t\right)+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} . \tag{40}
\end{align*}
$$

Similarly, equation (30) after substituting the value of $G(q)=(1 / q)$

$$
\begin{equation*}
I_{5}^{I}=B_{1}(y, t)^{*} B_{4}^{I}(t) \tag{41}
\end{equation*}
$$

and

$$
\begin{align*}
B_{4}^{I}(t)= & \left(\delta(t)+\lambda H(t)^{*} \delta(t)\right) \\
& *\left(D^{*}+E^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+E^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right)  \tag{42}\\
& * e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),
\end{align*}
$$

and for $B_{1}(y, t)$ (see equation (A.1).

Similar result for concentration was obtained by Nehad Ali Shah [26] (equation (35)). Thus, our result supports the result already present in literature.
4.2. Application 2: $f(t)=t, g(t)=t, h(t)=t$. The important concepts of engineering are based around linear functions. They are often used in engineering to explain data and evaluate the lines that best fit the given data sets. It has a lot of applications in engineering and it can be represented in a variety of ways. One of the particular interests is direct variation, which forms many engineering applications such as Hooke's law and Ohm's law. To learn about slope, engineers use linear functions to interpret and understand graphs that describe displacement, velocity, and acceleration. They use these functions to analyze data to learn how to design their engineering products more efficiently, reliably, and safely.For the choice of $F(q), G(q), H(q)$ equal to $\left(1 / q^{2}\right)$ in the appropriate equations and employing the Laplace inverse, the expression of $C(y, t), T(y, t), u(y, t)$, and then $I_{1}^{I I}, I_{2}^{I I}, I_{3}^{I I}, I_{4}^{I I}$ and $I_{5}^{I I}$ changes into, respectively,

$$
\begin{align*}
& C(y, t)=\left(1+K_{c} t\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t}, \\
& T(y, t)=(1+S t)^{*} \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t},  \tag{43}\\
& u(y, t)=I_{1}^{I I}+\frac{G_{T}}{\lambda} I_{2}^{I I}-\frac{G_{T}}{\lambda} I_{3}^{I I}+\frac{G_{c}}{\lambda} I_{4}^{I I}-\frac{G_{c}}{\lambda} I_{5}^{I I},
\end{align*}
$$

where

$$
\begin{equation*}
I_{1}^{I I}=B_{1}(y, t)^{*} B_{2}^{I I}(t) \tag{44}
\end{equation*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{2}^{I I}(t)$ (see equation (A.2)).

$$
\begin{align*}
I_{2}^{I I}= & (t+\lambda)^{*}\left(e^{-\alpha_{4} t} \cosh \left(\alpha_{6} t\right)+\frac{\left(S-\alpha_{4}\right)}{\alpha_{6}} e^{-\alpha_{4} t} \sinh \left(\alpha_{6} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t} \\
I_{3}^{I I}= & B_{1}(y, t)^{*} B_{3}^{I I}(t) \tag{45}
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{3}^{I I}(t)$ (see equation (A.3)).

$$
\begin{align*}
I_{4}^{I I}= & (t+\lambda)^{*}\left(e^{-\alpha_{7} t} \cosh \left(\alpha_{9} t\right)+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t}, \\
I_{5}^{I I}= & B_{1}(y, t)^{*} B_{4}^{I I}(t), \tag{46}
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{4}^{I I}(t)$ (see equation (A.4)).
4.3. Application 3: $f(t)=\sin t, g(t)=\sin t, h(t)=\sin t$. The choice of this function shows the fluid motion due to the oscillation of the plate. It has a lot of applications in physics such as wave motion, other oscillatory motions, and engineering. It is used to model the behavior that repeats. Trigonometric functions are used to calculate angles in many engineering problems. In civil and mechanical engineering, trigonometry is used to calculate torque and forces on objects, which help build bridges and girders. In the construction of bridges, we need to consider the forces which keep the bridges at their balance and trigonometry helps us to calculate the static force which keeps the bridges static. In engineering, trigonometry is used to decompose the forces into horizontal and vertical components that can be analyzed.The expression for concentration after putting the value of $G(q)=\left(1 / q^{2}+1\right)$ into equation (12) is

$$
\begin{equation*}
C(y, t)=\left(\cos t+K_{c} \sin t\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} \tag{47}
\end{equation*}
$$

the expression for temperature become after putting the value of $H(q)=\left(1 / q^{2}+1\right)$ into equation (16) is

$$
\begin{equation*}
T(y, t)=(\cos t+S \sin t)^{*} \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t} \tag{48}
\end{equation*}
$$

and velocity change after substitute the value of $F(q)=\left(1 / q^{2}+1\right)$ into equation (21) is

$$
\begin{equation*}
u(y, t)=I_{1}^{I I I}+\frac{G_{T}}{\lambda} I_{2}^{I I I}-\frac{G_{T}}{\lambda} I_{3}^{I I I}+\frac{G_{c}}{\lambda} I_{4}^{I I I}-\frac{G_{c}}{\lambda} I_{5}^{I I I} \tag{49}
\end{equation*}
$$

where

$$
\begin{equation*}
I_{1}^{I I I}=B_{1}(y, t)^{*} B_{2}^{I I I}(t) \tag{50}
\end{equation*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{2}^{I I I}(t)$ (see equation (A.5)).

$$
\begin{align*}
I_{2}^{I I I}= & (\sin t+\lambda \cos t)^{*}\left(e^{-\alpha_{4} t} \cosh \left(\alpha_{6} t\right)+\frac{\left(S-\alpha_{4}\right)}{\alpha_{6}} e^{-\alpha_{4} t} \sinh \left(\alpha_{6} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t}, \\
I_{3}^{I I I}= & B_{1}(y, t)^{*} B_{3}^{I I I}(t), \tag{51}
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{3}^{I I I}(t)$ (see equation (A.6)).

$$
\begin{align*}
I_{4}^{I I I}= & (\sin t+\lambda \cos t)^{*}\left(e^{-\alpha_{\tau} t} \cosh \left(\alpha_{9} t\right)+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t}, \\
I_{5}^{I I I}= & B_{1}(y, t)^{*} B_{4}^{I I I}(t), \tag{52}
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{4}^{I I I}(t)$ (see equation (A.7)).
4.4. Application 4: $f(t)=e^{t}, g(t)=e^{t}, h(t)=e^{t}$. The exponent functions are used for real-world application as for calculating area, volume, determining growth or decay, and impacts of force. In engineering, it helps them to design, build, and improve the machinery, structure, and equipment. For example, in sound engineering, it is used to calculate sound waves. In basic engineering, it is used to compute the tensile strength, which finds out the amount of stress that a structure can withstand. In aeronautical engineering, it is used to predict how airplanes, rockets, and jets will perform during flight. To determine the kinetic and potential energy, pressure, heat, and airflow of waves behavior, it is very helpful. Nuclear power sources are one of the important things developed by nuclear engineers. They used the exponents to work with extremely small numbers to make the big things happen. Substituting the value of $G(q)=$ ( $1 / q-1$ ) into equation (12), the concentration equation after implementing the Laplace inverse becomes

$$
\begin{equation*}
C(y, t)=\left(e^{t}+K_{c} e^{t}\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} \tag{53}
\end{equation*}
$$

and equation of temperature distribution after putting the value of $H(q)=(1 / q-1)$ into equation (16) and applying Laplace inverse

$$
\begin{equation*}
T(y, t)=\left(e^{t}+S e^{t}\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t} \tag{54}
\end{equation*}
$$

The expression for velocity is

$$
\begin{equation*}
u(y, t)=I_{1}^{I V}+\frac{G_{T}}{\lambda} I_{2}^{I V}-\frac{G_{T}}{\lambda} I_{3}^{I V}+\frac{G_{c}}{\lambda} I_{4}^{I V}-\frac{G_{c}}{\lambda} I_{5}^{I V}, \tag{55}
\end{equation*}
$$

where

$$
\begin{equation*}
I_{1}^{I V}=B_{1}(y, t)^{*} B_{2}^{I V}(t) \tag{56}
\end{equation*}
$$

and $B_{1}(y, t)$ (see equation (A.1)) and $B_{2}^{I V}(t)$ (see equation (A.8)).

$$
\begin{align*}
I_{2}^{I V}= & \left(e^{t}+\lambda e^{t}\right)^{*}\left(e^{-\alpha_{4} t} \cosh \left(\alpha_{6} t\right)+\frac{\left(S-\alpha_{4}\right)}{\alpha_{6}} e^{-\alpha_{4} t} \sinh \left(\alpha_{6} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t}, \\
I_{3}^{I V}= & B_{1}(y, t)^{*} B_{3}^{I V}(t), \tag{57}
\end{align*}
$$

and $B_{1}(y, t)$ (see equation (A.1)) and $B_{3}^{I V}(t)$ (see equation (A.9)).

$$
\begin{align*}
I_{4}^{I V}= & \left(e^{t}+\lambda e^{t}\right)^{*}\left(e^{-\alpha_{7} t} \cosh \left(\alpha_{9} t\right)+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} . \tag{58}
\end{align*}
$$

Similarly,

$$
\begin{equation*}
I_{5}^{I V}=B_{1}(y, t)^{*} B_{4}^{I V}(t) \tag{59}
\end{equation*}
$$

and $B_{1}(y, t)$ (see equation (A.1)) and $B_{4}^{I V}(t)$ (see equation (A.10)).
4.5. Application 5: $f(t)=t e^{t}, g(t)=t e^{t}, h(t)=t e^{t}$. Inserting the $G(q)=\left(1 /(q-1)^{2}\right)$ into equation (12) and applying the Laplace inverse, we get

$$
\begin{equation*}
C(y, t)=\left(e^{t}+\left(K_{c}+1\right) t e^{t}\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} \tag{60}
\end{equation*}
$$

and insert the $H(q)=\left(1 /(q-1)^{2}\right)$ into equation (16) and take Laplace inverse:

$$
\begin{align*}
& T(y, t)=\left(e^{t}+(S+1) t e^{t}\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t}  \tag{61}\\
& u(y, t)=I_{1}^{V}+\frac{G_{T}}{\lambda} I_{2}^{V}-\frac{G_{T}}{\lambda} I_{3}^{V}+\frac{G_{c}}{\lambda} I_{4}^{V}-\frac{G_{c}}{\lambda} I_{5}^{V}
\end{align*}
$$

The $I_{1}^{V}$ takes the form after embedding the $F(q)=(1 /$ $\left.(q-1)^{2}\right)$

$$
\begin{equation*}
I_{1}^{V}=B_{1}(y, t)^{*} B_{2}^{V}(t) \tag{62}
\end{equation*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{2}^{V}(t)$ (see equation (A.11)).

$$
\begin{align*}
I_{2}^{V}= & \left(t e^{t}+\lambda e^{t}(t+1)\right)^{*}\left(e^{-\alpha_{4} t} \cosh \left(\alpha_{6} t\right)+\frac{\left(S-\alpha_{4}\right)}{\alpha_{6}} e^{-\alpha_{4} t} \sinh \left(\alpha_{6} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t}, \\
I_{3}^{V}= & B_{1}(y, t)^{*} B_{3}^{V}(t), \tag{63}
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{3}^{V}(t)$ (see equation (A.12)).

$$
\begin{align*}
I_{4}^{V}= & \left(t e^{t}+\lambda e^{t}(t+1)\right)^{*}\left(e^{-\alpha_{7} t} \cosh \left(\alpha_{9} t\right)+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t}, \\
I_{5}^{V}= & B_{1}(y, t)^{*} B_{4}^{V}(t), \tag{64}
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{4}^{V}(t)$ (see equation (A.13)).
4.6. Application 6: $f(t)=e^{t} \sin t, g(t)=e^{t} \sin t, h(t)=e^{t}$ $\sin t$. The choice of the value of $G(q)=\left(1 /(q-1)^{2}+1\right), H$ $(q)=\left(1 /(q-1)^{2}+1\right), F(q)=\left(1 /(q-1)^{2}+1\right)$ makes the expression

$$
\begin{align*}
C(y, t) & =\left(e^{t}\left(\cos t+\left(1+K_{c}\right) \sin t\right)\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} \\
T(y, t) & =\left(e^{t}(\cos t+(1+S) \sin t)\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t} \\
u(y, t) & =I_{1}^{V I}+\frac{G_{T}}{\lambda} I_{2}^{V I}-\frac{G_{T}}{\lambda} I_{3}^{V I}+\frac{G_{c}}{\lambda} I_{4}^{V I}-\frac{G_{c}}{\lambda} I_{5}^{V I} \\
I_{1}^{V I} & =B_{1}(y, t)^{*} B_{2}^{V I}(t) \tag{65}
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{2}^{V I}(t)$ (see equation (A.14)).

$$
\begin{align*}
I_{2}^{V I}= & \left(e^{t} \sin t(1+\lambda)+e^{t} \cos t\right) \\
& *\left(e^{-\alpha_{4} t} \cosh \left(\alpha_{6} t\right)+\frac{\left(S-\alpha_{4}\right)}{\alpha_{6}} e^{-\alpha_{4} t} \sinh \left(\alpha_{6} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t},  \tag{66}\\
I_{3}^{V I}= & B_{1}(y, t)^{*} B_{3}^{V I}(t),
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{3}^{V I}(t)$ (see equation (A.15)).

$$
\begin{align*}
I_{4}^{V I}= & \left(e^{t} \sin t(1+\lambda)+e^{t} \cos t\right) \\
& *\left(e^{-\alpha_{7} t} \cosh \left(\alpha_{9} t\right)+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} . \tag{67}
\end{align*}
$$

Similarly,

$$
\begin{equation*}
I_{5}^{V I}=B_{1}(y, t)^{*} B_{4}^{V I}(t), \tag{68}
\end{equation*}
$$

where $B_{1}(y, t)$ (see (A.1)) and $B_{4}^{V I}(t)$ (see equation (A.16)).
4.7. Application 7: $f(t)=t \sin t, g(t)=t \sin t, h(t)=t \sin$ $t$. By putting the value of $G(q)=\left(2 q /\left(q^{2}+1\right)^{2}\right)$ into equation (12) and employing Laplace inverse,

$$
\begin{equation*}
C(y, t)=\left(t \cos t+\left(K_{c} t+1\right) \sin t\right)^{*} \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t} \tag{69}
\end{equation*}
$$

By putting the value of $H(q)=\left(2 q /\left(q^{2}+1\right)^{2}\right)$ into equation (16) and Laplace inverse gives the expression

$$
\begin{align*}
& T(y, t)=(t \cos t+(S t+1) \sin t)^{*} \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t} \\
& u(y, t)=I_{1}^{V I I}+\frac{G_{T}}{\lambda} I_{2}^{V I I}-\frac{G_{T}}{\lambda} I_{3}^{V I I}+\frac{G_{c}}{\lambda} I_{4}^{V I I}-\frac{G_{c}}{\lambda} I_{5}^{V I I} \tag{70}
\end{align*}
$$

where

$$
\begin{equation*}
I_{1}^{V I I}=B_{1}(y, t)^{*} B_{2}^{V I I}(t) \tag{71}
\end{equation*}
$$

and $B_{1}(y, t)$ (see equation (A.1)) and $B_{2}^{V I I}(t)$ (see equation (A.17)).

$$
\begin{align*}
I_{2}^{V I I}= & ((t+\lambda) \sin t+\lambda \cos t) \\
& *\left(e^{-\alpha_{4} t} \cosh \left(\alpha_{6} t\right)+\frac{\left(S-\alpha_{4}\right)}{\alpha_{6}} e^{-\alpha_{4} t} \sinh \left(\alpha_{6} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{P_{r}}}{2 \sqrt{t}}\right) e^{-S t},  \tag{72}\\
I_{3}^{V I I}= & B_{1}(y, t)^{*} B_{3}^{V I I}(t),
\end{align*}
$$

where $B_{1}(y, t)$ (see equation (A.1)) and $B_{3}^{V I I}(t)$ (see equation (A.18)).

$$
\begin{align*}
I_{4}^{V I I}= & ((t+\lambda) \sin t+\lambda \cos t) \\
& *\left(e^{-\alpha_{7} t} \cosh \left(\alpha_{9} t\right)+\frac{\left(K_{c}-\alpha_{7}\right)}{\alpha_{9}} e^{-\alpha_{7} t} \sinh \left(\alpha_{9} t\right)\right) \\
& * \operatorname{erfc}\left(\frac{y \sqrt{S_{c}}}{2 \sqrt{t}}\right) e^{-K_{c} t}, \\
I_{5}^{V I I}= & B_{1}(y, t)^{*} B_{4}^{V I I}(t), \tag{73}
\end{align*}
$$

and for $B_{1}(y, t)$ (see equation (A.1)) and $B_{4}^{V I I}(t)$ (see equation (A.19)).These are solutions for the choice of same function for $f(t), g(t)$, and $h(t)$ from the list of functions $H(t), t, \sin t, e^{t}, t e^{t}, t \sin t, \sin t e^{t}$. We can consider the problem with the different choice of function for $f(t), g(t), h(t)$, e.g., $f(t)=t, g(t)=e^{t}, h(t)=H(t)$ and find its solution. For the validation of results, if we take $\lambda=$ $0, G_{T}=1, S=0, h(t)=1-a e^{b t}, g(t)=1$ with choice of $f(t)=H(t) t^{\alpha},(\alpha>0)$ or $\sin t$, in our system of equations (4)-(8), the results obtained are the same as the result
obtained by Nehad Ali shah [27] (choosing the $\epsilon=0, N=$ $G_{C}$ in equation (9)).

## 5. Results and Discussion

The heat and mass transfer study of Maxwell fluid is discussed here. The solutions for dimensionless velocity, concentration, and temperature are assessed by the Laplace transform method. The application of these solutions in different fields of engineering sciences is also discussed. It brings to attention that these results are helpful to solve the complicated problems of engineering and applied science. The behavior of these solutions for velocity, concentration, and temperature profile is depicted graphically. The impacts of different pertinent parameters $\lambda, M, S_{c}, K_{c}, P_{r}, G_{T}, G_{C}, S$ on fluid flow are also deliberated using plots and their physical aspects described. To avoid repetition, only the most significant graphical representations regarding the effects of the concerned parameter will be included here.

The variation in the behavior of velocity and concentration with varying values of Schmidt number $S_{c}$ is illustrated inFigures $1-3$, respectively. An increase in Schmidt number results in the decline in the thickness of the boundary layer of concentration. Since the Schmidt number is defined as the ratio between kinematic viscosity and mass diffusivity, it reduced the concentration as well as velocity profile. In reality, the increase occurring in momentum diffusivity causes a decline in the fluid velocity.

The deviation in temperature profile for varying values of Prandtl number $\operatorname{Pr}$ is demonstrated in Figures 4 and 5. It is depicted that the thermal boundary layer thickness decreases rapidly with the increase in the values of Pr. For a small value of Pr , heat diffuses very quickly in comparison to the velocity. The reason is, the thermal boundary layer thickness in liquid metals is higher than the momentum boundary layer. Finally, Pr can be practiced to expand the percentage of cooling.

Similar effects can be seen for the heat absorption coefficient $S$ on temperature profile with different values of the function $g(t)$ at different time scales depicted in Figure 6 and 7. The thermal buoyancy forces decrease with the increase of $S$ which decreases the fluid temperature.

Impacts of magnetic parameter $M$ displayed in Figure 8 depict the velocity decline with the increase of magnetic parameter values. Physically, when magnetic force is applied to the velocity field, it generates the drag force known as the Lorentz force which opposes the motion of the fluid. Figure 9 shows the impact of Pr on velocity. The increase of $P_{r}$ results in a decrease in velocity. The velocity boundary layer gets thicker due to the low rate of thermal diffusion. Basically, in heat transfer problems, $P_{r}$ control the relative thickness momentum boundary layer.

In Figure 10, the study of the effects of $G_{T}$ on velocity describes the increase in behavior with increment in the values of $G_{T}$. Physically, the result of more induced fluid flows is due to a rise in buoyancy effects which is the result of the increase in $G_{T}$. The depiction of $G_{C}$ on velocity is portrayed in Figure 11. We can see the rise in velocity with the rise in the value of $G_{C}$. The natural convection and


- $S c=0.22$
…. $S c=0.85$
-     - $S c=1.96$
.- - $S c=2.50$
(a)


$$
\begin{array}{ll}
-S c=0.22 & --S c=1.96 \\
S c=0.85 & --S c=2.50
\end{array}
$$

(c)


$-S c=0.22$
….. $S c=0.85$

-     -         - $S c=1.96$
.-. $S c=2.50$
(b)

- $S c=0.22$
$S c=0.85$
(d)

(e)

$$
\begin{array}{ll}
-S c=0.22 & --S c=1.96 \\
\ldots \cdot S c=0.85 & --S c=2.50
\end{array}
$$

(f)

Figure 1: Continued.


Figure 1: Profile of velocity for different values of $S_{c}$ and $M=2.0, \lambda=0.6, S=1.0 .5, K_{c}=0.5, G_{T}=5.0, G_{C}=2.0, P_{r}=0.71$.

--- $S c=1$.
(a)



Sc $=0.60$


- $S c=0.22$
--- $S c=0.96$
‥-Sc= 1.4
(b)


$$
\begin{array}{ll}
-S c=0.22 & --S c=0.96 \\
\ldots \cdot \cdot S c=0.60 & --S c=1.4
\end{array}
$$

(c)
(d)

Figure 2: Continued.


$$
\begin{array}{ll}
-S c=0.22 & ---S c=0.96 \\
\ldots \ldots S c=0.60 & --S c=1.4
\end{array}
$$

(e)


$$
\begin{aligned}
& -\quad S c=0.22 \\
& \ldots c=0.60
\end{aligned}
$$

--- $S c=0.96$
(f)

Figure 2: Profile of concentration for different values of $S_{c}$ and variation of time.



$$
\begin{aligned}
& \text {-... } S c=0.22 \\
& \ldots c=0.60
\end{aligned}
$$

(a)


- $S c=0.22$
$S c=0.60$
-     - $S c=0.96$
.-. $-S c=1.4$

$$
\begin{array}{ll}
-S c=0.22 & ---S c=0.96 \\
\ldots . \cdot S c=0.60 & --S c=1.4
\end{array}
$$

(b)

(c)

- $S c=0.22$
…. $S c=0.60$
--- $S c=0.96$

$$
\cdots-S c=1.4
$$

(d)

Figure 3: Continued.

(e)

(g)


$$
\begin{array}{ll}
-S c=0.22 & --S C=0.96 \\
\cdots \cdot S c=0.60 & --S c=1.4
\end{array}
$$

(f)

(h)

Figure 3: Profile of concentration for different values of $S_{c}$ and variation of time.


Figure 4: Continued.


$---\operatorname{Pr}=5.0$
(c)


$$
\begin{array}{ll}
\ldots \operatorname{Pr}=0.71 & --\operatorname{Pr}=5.0 \\
\ldots \ldots \operatorname{Pr}=1.5 & --\operatorname{Pr}=7.0
\end{array}
$$

(e)

(g)

(d)


$$
\begin{array}{ll}
-\operatorname{Pr}=0.71 & --\operatorname{Pr}=5.0 \\
\ldots . \cdot \operatorname{Pr}=1.5 & --\operatorname{Pr}=7.0
\end{array}
$$

(f)


$$
\begin{array}{ll}
-\operatorname{Pr}=0.71 & --\operatorname{Pr}=5.0 \\
\ldots \cdot \operatorname{Pr}=1.5 & --\operatorname{Pr}=7.0
\end{array}
$$

(h)

Figure 4: Profile of temperature for different values of $P_{r}$ and variation of time.


$$
\begin{aligned}
& \text { Pr }=0.71 \\
& \ldots \ldots \quad \operatorname{Pr}=1.5
\end{aligned}
$$

(a)

—— $\operatorname{Pr}=0.71$
… $\operatorname{Pr}=1.5$
(c)

(e)


$$
\begin{array}{ll}
— & \operatorname{Pr}=0.71 \\
\operatorname{Pr}=1.5 & --\operatorname{Pr}=5.0 \\
\ldots \cdot \operatorname{Pr}=7.0
\end{array}
$$

(b)


- $\operatorname{Pr}=0.71$
-     - $\operatorname{Pr}=5.0$
.- - $\operatorname{Pr}=7.0$
(d)

(f)

Figure 5: Profile of temperature for different values of $P_{r}$ and variation of time.


- $S=0.5$
$S=1.5$

- $S=0.5$
--- $S=3.0$
-.- $S=4.5$
(b)


$$
-S=0.5
$$

$$
---S=3.0
$$

(c)

(e)

$$
\begin{array}{ll}
-S=0.5 & --S=3.0 \\
\ldots \ldots S=1.5 & --S=4.5
\end{array}
$$

(d)

(f)

Figure 6: Continued.


$$
\begin{aligned}
& \_S=0.5 \\
& \cdots \cdots \cdot S=1.5
\end{aligned}
$$

(g)


> _. $S=0.5$
> $\ldots . S=1.5$
(h)

Figure 6: Profile of temperature for different values of $S$ and variation of time.


$$
-S=0.5
$$

$$
\ldots . S=1.5
$$

(a)


- $S=0.5$
$S=1.5$
(c)


$$
\begin{aligned}
& -S=0.5 \\
& \ldots S=1.5
\end{aligned}
$$

(b)

$-S=0.5$
$S=1.5$
(d)

Figure 7: Continued.


Figure 7: Profile of temperature for different values of $S$ and variation of time.


Figure 8: Continued.


Figure 8: Profile of velocity for different values of $M$ and $P_{r}=0.71, \lambda=0.6, S=0.5, S_{c}=0.60, K_{c}=0.5, G_{T}=5.0, G_{C}=2.0$.
acceleration in fluid flow is due to the fact that buoyancy forces bcomes powerful than viscous forces.

Figures 12-14 show the chemical reaction $K_{c}$ parameter effects on velocity and concentration description. The increase in $K_{c}$ decreases the concentration and velocity profiles. Basically, chemical molecular diffusivity and species concentration drop with the higher values of $K_{c}$. The distribution in concentration falls at all the fluid flow field points with the rise in $K_{c}$. The curve for velocity for different values of the Maxwell fluid parameter $\lambda$ is plotted in Figure 15. It is observed that an increase in $\lambda$ produces a significant increase in the momentum boundary layer of the fluid which then increases the velocity. The rise in $\lambda$ will therefore correspond to a fall in fluid viscosity, resulting in it accelerating the flow and hence velocity rising. Further, an increase in $\lambda$ causes a rise in velocity near the plate surface. Although the trend is reversed away from the plate, the Newtonian fluid $(\lambda \longrightarrow 0)$ has a higher velocity.

In Figure 16, velocity profiles are plotted against the heat absorption parameter $S$ values. These curves show that the velocity is a decreasing function of parameter S. Furthermore, due to the absorption of heat, the fluid temperature diminishes and the thermal buoyancy force diminishes. These results have seen a fall in the velocity of a fluid with the increase in the values of S.

Figure 17 shows the behavior of velocity profile for different values of the parameters with a different choice of the function $f(t), g(t), h(t)$. For validation of our results, we consider some special cases of temperature profile already existing in literature and their graphical illustration is depicted in Figures 18-20. Figure 18 shows the temperature decrease with the increase in $P_{r}$ for the variation of time with $g(t)=1-e^{-t}$. The effects of the heat absorption parameter can be observed in Figure (19) which depicts the decline in temperature. The impacts of $g(t)=1-a e^{-b t}$ for different choices of $a$ and $b$ are explained in Figure 20. We see the decline in temperature with the increasing values of $a$ and $b$.

(a)


..... $\operatorname{Pr}=1.5$
-- $\operatorname{Pr}=5.0$
(c)

(e)

- $\operatorname{Pr}=0.71$
$\operatorname{Pr}=1.5$
--- $\operatorname{Pr}=5.0$

(b)

- $\operatorname{Pr}=0.71$
--- $\operatorname{Pr}=5.0$
.- $\operatorname{Pr}=7.0$
(d)


$$
\begin{array}{ll}
\ldots \operatorname{Pr}=0.71 & ---\operatorname{Pr}=5.0 \\
\ldots \ldots \cdot \operatorname{Pr}=1.5 & --\operatorname{Pr}=7.0
\end{array}
$$

(f)

Figure 9: Continued.

(g)

Figure 9: Profile of velocity for different values of $P_{r}$ and $M=0.5, \lambda=0.6, S=0.5, S_{c}=0.60, K_{c}=0.5, G_{T}=5.0, G_{C}=2.0$.


$$
-G_{T}=1.0
$$

$$
---G_{T}=6.0
$$

(a)

$-G_{T}=1.0$
--- $G_{T}=6.0$
.- $-G_{T}=10$
(c)


$$
-G_{T}=1.0
$$

$$
---G_{T}=6.0
$$

$$
\cdots-G_{T}=10
$$

(b)

(d)


Figure 10: Profile of velocity for different values of $G_{T}$ and $M=0.2, \lambda=0.6, S=1.0 .5, K_{c}=0.5, S_{c}=0.60, G_{C}=2.0, P_{r}=0.71$.


$$
\begin{aligned}
& \ldots G_{C}=2.0 \\
& \ldots \ldots
\end{aligned} G_{C}=4.0
$$



$$
\begin{array}{ll}
- & G_{C}=2.0 \\
\cdots \cdots & G_{C}=4.0
\end{array}
$$

(a)
(b)


$$
\begin{array}{ll}
-G_{C}=2.0 & -- \\
\cdots \cdots \cdot G_{C}=6.0 \\
G_{C}=4.0 & --G_{C}=8.0
\end{array}
$$

(c)

--- $G_{C}=6.0$
.-- $G_{C}=8.0$
(e)


$$
\begin{aligned}
& -G_{C}=2.0 \\
& \cdots \cdots \cdot G_{C}=4.0
\end{aligned}
$$

$$
---G_{C}=6.0
$$

$$
\cdots-G_{C}=8.0
$$

(d)

$-G_{C}=2.0$
$G_{C}=4.0$
$G_{C}=6.0$
.-. $G_{C}=8.0$
(f)


$$
\begin{array}{ll}
-G_{C}=2.0 & ---G_{C}=6.0 \\
\cdots \cdots \cdot G_{C}=4.0 & --G_{C}=8.0
\end{array}
$$

(g)

Figure 11: Profile of velocity for different values of $G_{C}$ and $M=0.2, \lambda=0.6, S=0.5, K_{c}=0.5, S_{c}=0.60, G_{T}=5.0, P_{r}=0.71$.


Figure 12: Continued.

(g)

Figure 12: Profile of velocity for different values of $K_{c}$ and $M=0.5, \lambda=0.6, S=0.5, S_{c}=0.60, G_{T}=5.0, G_{C}=2.0, P_{r}=0.71$.


$$
\begin{array}{ll}
-K c=0.5 & --K c=2.0 \\
\ldots \cdot \cdot K c=1.4 & --K c=2.9
\end{array}
$$

(a)

_ - - $K c=2.0$
._.- $K c=2.9$


- $K c=0.5$
$K c=1.4$
--- $K c=2.0$
_.- $K c=2.9$
(b)

(c)
(d)


Figure 13: Profile of concentration for different values of $K_{c}$ and variation of time.


Figure 14: Continued.


Figure 14: Profile of concentration for different values of $K_{c}$ and variation of time.


Figure 15: Continued.


Figure 15: Profile of velocity for different values of $\lambda$ and $M=2.0, G_{C}=2.0, S=1.5, K_{c}=0.5, S_{c}=0.60, G_{T}=10.0, P_{r}=0.5$.


(a)


- $S=1$
…. $S=2$
--- $S=3$
.- $-S=4$
(c)

- $S=1$
--- $S=3$
(e)


$$
\begin{array}{ll}
-S=1 & --S=3 \\
\cdots=2 & --S=4
\end{array}
$$

(b)


- $S=1$
-- $S=3$
-. $-S=4$
(d)


(g)

Figure 16: Profile of velocity for different values of $S$ and $M=2.0, G_{C}=2.0, \lambda=0.6, K_{c}=0.5, S_{c}=0.60, G_{T}=10.0, P_{r}=0.5$.


$$
-M=0.5
$$

$$
M=1.0
$$

$$
\ldots M=1.5
$$

(a)



- $K c=0.5$
$K c=2.5$

$-G_{T}=1.0$
$\cdots \cdots \quad G_{T}=3.5$
--- $K c=4.5$
.-. $K c=6.5$
(b)
--- $G_{T}=6.0$
(d)


Figure 17: Profile of velocity for different values of $M, S, S_{c}, K_{c}, G_{T}, G_{C}$ and different choice of function for $f(t), g(t), h(t)$.


$$
\begin{aligned}
& \text { Pr }=0.71 \\
& \ldots . \operatorname{Pr}=1.5
\end{aligned}
$$



$$
\begin{array}{ll}
— & \operatorname{Pr}=0.71 \\
\ldots . \cdot \operatorname{Pr}=1.5 & --\operatorname{Pr}=5.0 \\
& --\operatorname{Pr}=7.0
\end{array}
$$

(b)

Figure 18: Temperature profile for different values of $P_{r}$ with $S=0.5$ and $g(t)=1-e^{-t}$.


Figure 19: Temperature profile for different values of $S$ with $P_{r}=0.71$ and $g(t)=1-e^{-t}$.


Figure 20: Temperature profile for different values of $a, b$ with $P_{r}=0.71, S=0.5$ and $g(t)=1-a e^{-b t}$.

## 6. Conclusions

A thorough investigation of MHD Maxwell fluid motion has been studied here under the effects of different parameters. The exact solutions are obtained for concentration, temperature, and velocity which satisfied the described initial and boundary conditions. Laplace transform is employed to obtain the exact solution and the behavior of different parameters on the flow of fluid along with different boundary conditions is investigated. Effects of chemical reaction coefficient, Schmidt number, and different boundary conditions on concentration, effects of Prandtl number, heat
source, Newtonian heating, etc. on temperature, Magnetic parameter, Schmidt number, thermal Grashof number, relaxation parameter, mass Grashof number, Prandtl number, heat source, and chemical reaction impacts on the fluid motion are discussed. The results obtained are as follows:
(1) The boundary layer thickness of concentration decreases with the increase in the mass diffusivity $S_{c}$ and chemical reaction parameter $K_{C}$.
(2) The thermal boundary layer decreases with the increase in momentum boundary layer due to $P_{r}$ and heat absorption $S$.
(3) Lorentz force effects due to $M$, momentum boundary layer effects due to $P_{r}$, mass diffusivity effects due to $S_{c}$, chemical reaction $K_{c}$, and heat absorption decrease the velocity with the increase of these parameters.
(4) Increase in buoyancy forces due to $G_{T}$ and $G_{C}$ stimulates the speed of fluid flow.
(5) A rise in relaxation time $\lambda$ reduces the fluid viscosity and results in acceleration of fluid flow.

## Appendix

$$
\begin{align*}
& B_{1}(y, t)= \begin{cases}0, & 0<t<y \lambda, \\
e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} \sqrt{t^{2}-\lambda y^{2}}\right), & t>y \lambda,\end{cases}  \tag{A.1}\\
& B_{2}^{I I}(t)=\left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(1+\alpha_{1} t\right)+\left(\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)\right)^{*} t,  \tag{A.2}\\
& \left.B_{3}^{I I}(t)=(1+\lambda H(t))^{*}\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C^{*} e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right)\right)^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),  \tag{A.3}\\
& B_{4}^{I I}(t)=\left(1+\lambda H(t)^{*}\left(D^{*}+E^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+F^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right)^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),\right.  \tag{A.4}\\
& B_{2}^{I I I}(t)=\left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(\cos t+\alpha_{1} \sin t\right)+\left(\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)\right)^{*} \sin t,  \tag{A.5}\\
& B_{3}^{I I I}(t)=\left(\cos t+\lambda H(t)^{*} \cos t\right)^{*}\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C^{*} e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right)^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),  \tag{A.6}\\
& B_{4}^{I I I}(t)=\left(\cos t+\lambda H(t)^{*} \cos t\right)^{*}\left(D^{*}+E^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+F^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right){ }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),  \tag{A.7}\\
& B_{2}^{I V}(t)=\left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(\left(H(t)+\alpha_{1}\right)^{*} e^{t}\right)+\left(\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)\right)^{*} e^{t},  \tag{A.8}\\
& B_{3}^{I V}(t)=\left(e^{t}+\lambda H(t)^{*} e^{t}\right)^{*}\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C^{*} e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right){ }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),  \tag{A.9}\\
& B_{4}^{I V}(t)=\left(e^{t}+\lambda H(t)^{*} e^{t}\right)^{*}\left(D^{*}+E^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+F^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right)^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right),  \tag{A.10}\\
& B_{2}^{V}(t)=\left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(e^{t}+\left(1+\alpha_{1}\right) t e^{t}\right)+\left(\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)\right)^{*} t e^{t},  \tag{A.11}\\
& B_{3}^{V}(t)=\left(e^{t}(1+t)+\lambda H(t)^{*} e^{t}(1+t)\right)^{*}\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C^{*} e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right)  \tag{A.12}\\
& { }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right), \\
& B_{4}^{V}(t)=\left(e^{t}(1+t)+\lambda H(t)^{*} e^{t}(1+t)\right)^{*}\left(D^{*}+E^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+F^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right)  \tag{A.13}\\
& { }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right) . \\
& B_{2}^{V I}(t)=\left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(e^{t} \cos t+\left(1+\alpha_{1}\right) e^{t} \sin t\right)+\left(\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)\right)  \tag{A.14}\\
& { }^{*} e^{t} \sin t . \\
& B_{3}^{V I}(t)=\left(e^{t} \cos t+e^{t} \sin t+\lambda H(t)^{*}\left(e^{t} \cos t+e^{t} \sin t\right)\right)^{*}\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C^{*} e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right)  \tag{A.15}\\
& { }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right) .
\end{align*}
$$

$$
\begin{align*}
& \begin{aligned}
B_{4}^{V I}(t)= & \left(e^{t} \cos t+e^{t} \sin t+\lambda H(t)^{*}\left(e^{t} \cos t+e^{t} \sin t\right)\right)^{*}\left(D^{*}+E^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+F^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right) \\
& { }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right) .
\end{aligned}  \tag{A.16}\\
& B_{2}^{V I I}(t)=\left(e^{-\alpha_{1} t}\left(\left(i \alpha_{3}\right) I_{1}\left(i \alpha_{3} t\right)+\delta(t)\right)\right)^{*}\left(H(t)+\left(1+\alpha_{1}\right) t \sin t\right)+\left(\alpha_{3}^{2} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right)\right)  \tag{A.17}\\
& { }^{*} t \sin t, \\
& B_{3}^{V I I}(t)=\left((t \sin t+t \cos t)+\lambda H(t)^{*}(\sin t+\lambda \cos t)\right)^{*}\left(A^{*}+B^{*} e^{-\left(\alpha_{4}+\alpha_{6}\right) t}+C^{*} e^{-\left(\alpha_{4}-\alpha_{6}\right) t}\right)  \tag{A.18}\\
& { }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right), \\
& B_{4}^{V I I}(t)=\left((t \sin t+t \cos t)+\lambda H(t)^{*}(\sin t+\lambda \cos t)\right)^{*}\left(D^{*}+E^{*} e^{-\left(\alpha_{7}+\alpha_{9}\right) t}+F^{*} e^{-\left(\alpha_{7}-\alpha_{9}\right) t}\right)  \tag{A.19}\\
& { }^{*} e^{-\alpha_{1} t} I_{0}\left(i \alpha_{3} t\right), \\
& L^{-1}\left[\frac{e^{-b \sqrt{q^{2}-a^{2}}}}{\sqrt{q^{2}-a^{2}}}\right]= \begin{cases}0, & 0<t<b, b>0, \\
I_{0}\left(a \sqrt{t^{2}-b^{2}}\right), & t>b, \operatorname{Re}(q)>\operatorname{Re}(a),\end{cases}  \tag{A.20}\\
& L^{-1}[F(q+m)]=f(t) e^{-m t},  \tag{A.21}\\
& L^{-1}\left[\frac{\exp (-y \sqrt{q+a})}{q-b}\right]=e^{b t} \Phi(y ; t, a+b),  \tag{A.22}\\
& \Phi(y, t ; a)=\frac{1}{2}\left\{e^{y \sqrt{a}} \operatorname{ercf}\left(\frac{y}{2 \sqrt{t}}+\sqrt{a t}\right)+e^{-y \sqrt{a}} \operatorname{ercf}\left(\frac{y}{2 \sqrt{t}}-\sqrt{a t}\right)\right\},  \tag{A.23}\\
& L^{-1}\left[\frac{\exp (-y \sqrt{q+a})}{q-b}\right]=e^{b t} \Psi(y ; t, a+b),  \tag{A.24}\\
& \Psi(y, t ; a)=\frac{1}{2}\left\{e^{y \sqrt{a}} \operatorname{ercf}\left(\frac{y}{2 \sqrt{t}}+\sqrt{a t}\right)+e^{-y \sqrt{a}} \operatorname{ercf}\left(\frac{y}{2 \sqrt{t}}-\sqrt{a t}\right)\right\} . \tag{A.25}
\end{align*}
$$

## Data Availability

The data used to support the findings of this study are included within the article.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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# Generalized Type-I Hybrid Censoring Scheme in Estimation Competing Risks Chen Lifetime Populations 

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#### Abstract

Different types of censoring scheme are investigated; however, statistical inference on censoring scheme which can save the ideal test time and the minimum number of failures is needed. The generalized type-I hybrid censoring scheme (GHCS) solves this problem. Competing the risk models under the GHCS when time to failure has Chen lifetime distribution (CD) is adopted in this research with consideration of only two cases of failure. Partially step-stress accelerated life tests (ALTs) are applied to obtain enough failure times in a small period to achieve a highly reliable product. The problem of parameter estimation under maximum likelihood (ML) and Bayes methods is discussed. The asymptotic confidence interval as well as the Bayes credible interval is constructed. The validity of theoretical results is assessed and compared through simulation study. Finally, brief comments are reported to describe the behaviour of the estimation results.


## 1. Introduction

Information about the lifetime products is presented in complete or censored data with respect to time or cost considerations. The complete failure time data is used when all the units under the test fail through the determined period of time. However, the censoring failure time data is used when some units under the test fail through a determined period of time. Various types of censoring are available and the common types are called type-I and type-II censoring schemes. The first scheme has a prefixed test time and a random number of failures but the second scheme has a prefixed number of failures and a random test time. In serval cases of censoring, the test is required to run joint case of type-I and type-II censoring schemes described by the hybrid censoring scheme (HCS). The HCS can be described statistically as follows: suppose ( $\tau, T_{m}$ ) denote the ideal test time and the time of $m$ failure which is used for statistical inference, respectively, and the test is removed at
the only one time of them. Then, HCS is defined under type-I and type-II censoring schemes and is called type-I HCS and type-II HCS. The test is removed at $\min \left(\tau, T_{m}\right)$ in the type-I HCS, but at $\max \left(\tau, T_{m}\right)$ in the type-II HCS, there is more information about the type of censoring presented by [1-3]. Furthermore, type-I censoring scheme or type-I HCS may satisfy the properties that the test has the smallest number of failures or maybe zero. However, type-II censoring scheme or type-II HCS satisfies the properties that the test has the largest number of failures; see [4]. The problem that appeared in these censoring schemes can be overcome in the generalized form of HSC; see [5] as type-I GHCS and type-II GHCS.
(1) In type-I GHCS, suppose $n$ independent units are put under test and the prior integers $r$ and $m$ satisfy that $1 \leq$ $r<m \leq n$ and prior time $\tau$. If the smallest number $r$ is satisfied before $\tau\left(T_{r}<\tau\right)$, then the test is terminated at $\min \left(T_{m}, \tau\right)$ and the observed test times data are given by

$$
\begin{equation*}
\underline{t}=\left(t_{1 ; n}<t_{1 ; n}<\cdots<t_{r ; n}<\cdots<t_{s ; n}\right), \quad r \leq s \leq m \tag{1}
\end{equation*}
$$

However, if the smallest number $r$ does not satisfy before $\tau\left(\tau<T_{r}\right)$, then the test is terminated at $T_{r}$ and the observed test times data are given by

$$
\begin{equation*}
\underline{t}=\left(t_{1 ; n}<t_{1 ; n}<\cdots<t_{r ; n}\right) \tag{2}
\end{equation*}
$$

Finally, if the largest number $m$ is satisfied before $\tau$ ( $T_{m}<\tau$ ), then the test is terminated at $T_{m}$ and the observed test times are given by

$$
\begin{equation*}
\underline{t}=\left(t_{1 ; n}<t_{1 ; n}<\cdots<t_{m ; n}\right) \tag{3}
\end{equation*}
$$

Therefore, the type-I GHCS saves the minimum number which is necessary for statistical inferences.
(2) In type-II GHCS, let $n$ independent units be put under test and the two prior times $\tau_{1}$ and $\tau_{2}$ such that $\tau_{1}<\tau_{2}$ and integer $m$ satisfies $1 \leq m \leq n$. If the required number of failures is observed before $\tau_{1}$ ( $T_{m}<\tau_{1}$ ), then the test is terminated at $\tau_{1}$ and the observed test times data are given by

$$
\begin{equation*}
\underline{t}=\left(t_{1 ; n}<t_{1 ; n}<\cdots<t_{m ; n}<\cdots<t_{s ; n}\right), \quad m \leq s \leq n . \tag{4}
\end{equation*}
$$

On the other hand, if the required number of failures observed satisfies $\tau_{1}<T_{m}<\tau_{2}$, then the test is terminated at $T_{m}$ and the observed test times data are given by

$$
\begin{equation*}
\underline{t}=\left(t_{1 ; n}<t_{1 ; n}<\cdots<t_{m ; n}\right) \tag{5}
\end{equation*}
$$

Finally, if the required number of failures is observed to satisfy $\tau_{1}<\tau_{2}<T_{m}$, then the test is terminated at $\tau_{2}$ and the observed test times are given by

$$
\begin{equation*}
\underline{t}=\left(t_{1 ; n}<t_{1 ; n}<\cdots<t_{s ; n}\right), \quad 1 \leq s \leq m \tag{6}
\end{equation*}
$$

In life testing experiments, the common problem is that units fail due to several fetal risks which are known as competing risks problem.

The effect of any risk factor in the presence of other risk factors need to be assessed. This problem has been discussed early in [6-10] and recently in [11]. Under the consideration of two causes of failure, the competing risks model in the presence of type-I GHCS is presented as follows.

For a randomly selected $n$ independent unit, a life testing experiment with priors integers $r$ and $m, 1 \leq r<m \leq n$, is considered. At each step of the experiment, time $T_{i ; n}$ and the cause of failure $\rho_{i}$ are recorded for $i=1,2, \ldots, d$, where $d$ satisfies $r<d<m$ and $\rho_{i} \in\{1,2\}$. Then, the joint likelihood function of type-I GHCS where $t=\left\{\left(t_{1 ; n}, \rho_{1}\right),\left(t_{2 ; n}, \rho_{2}\right), \ldots\right.$, $\left.\left(t_{d ; n}, \rho_{d}\right)\right\}$ under the competing risks model is reported as

$$
\begin{equation*}
L(\underline{t} \mid \underline{\Theta})=A\left[S_{1}\left(t_{d}\right) S_{2}\left(t_{d}\right)\right]^{(n-d)} \prod_{i=1}^{d}\left[h_{1}\left(t_{i}\right)\right]^{\delta\left(\rho_{i}=1\right)}\left[h_{2}\left(t_{i}\right)\right]^{\delta\left(\rho_{i}=2\right)} \tag{7}
\end{equation*}
$$

where $A=n!/(n-d)!, \quad S()=.1-F(),. \quad h()=.f(.) / F($.$) ,$ $\delta\left(\rho_{i}=j\right)=\left\{\begin{array}{ll}1, & \rho_{i}=j, \\ 0, & \rho_{i} \neq j,\end{array}\right.$ and for $t_{i}=t_{i ; n}, 0<t_{1}<t_{2}<\cdots<$ $t_{m}<\infty$.

To obtain more information about the lifetime of products industrial process, accelerated life tests (ALTs) present a suitable manner for reducing test time rather than using conditions. As we see in [12], ALTs are presented in different types; one of them is constant-stress ALTs, in which the test is kept with a constant level of stress; see [13-15]. The second type is called progressive-stress ALTs, in which the stress is kept with a continuously increasing level; see [16-18]. The third type is called step-stress ALTs, in which the stress level is changed through a prior time or the number of failures; see [19, 20]. Furthermore, the ALTs can be done under the accelerated condition which is known by partial ALTs; see [21-26].

This paper aims to build and analyze type-I GHC competing risks sample under the model of partially stepstress ALTs from Chen lifetime products. The results of statistical analysis are built under maximum likelihood and Bayes method for point and interval estimation. The performances of the developed results are assessed and compared with mean squared error (MSE), average interval length (AL), and probability coverage (PC) through the Monte Carlo study.

This paper is structured as follows: the model formulation and abbreviation are presented in Section 2. The MLEs of model parameters as well as the asymptotic confidence intervals are investigated in Section 3. Bayes estimation with credible intervals is discussed in Section 4. The quality points and interval estimators are assessed via the Monte Carlo study in Section 5. Finally, the discussion and conclusion are presented in Section 6 (Table 1).

## 2. Abbreviation and Model Formulation

### 2.1. Abbreviations

2.1.1. Model Formulation. Suppose $n$ identical units are under life testing and two prior integer numbers $r$ and $m$ satisfy that $1 \leq r<m \leq n$ and prior fixed time $\tau$. The failure times $T_{i}$ and cause of failure $\rho_{i}$ are recorded through the test steps. If the smallest number of failure units $r$ is satisfied before time $\tau$, then the test is terminated at $\min \left(T_{m}, \tau\right)$. However, if $r$ is satisfied after the time $\tau$, then the test is terminated at $T_{r}$. The test is running under conditions until a fixed time $\eta$; then, the test is ruining under accelerated conditions. Considering that, the failure time has an independent CD and two independent causes of failure to satisfy the following assumptions:
(1) The random variable $T_{j i}$ is distributed with $\mathrm{CD}\left(\alpha, \beta_{j}\right), j=1,2$ with PDF given by

$$
\begin{equation*}
f_{j 1}(t)=\alpha \beta_{j} t^{\alpha-1} e^{t^{\alpha}+\beta_{j}\left[1-e^{t^{\alpha}}\right]}, \quad t>0, \alpha, \beta_{j}>0, \tag{8}
\end{equation*}
$$

and CDF is given by

Table 1: List of abbreviations.

| PDF | Probability density function |
| :--- | :---: |
| CDF | Cumulative probability function |
| $t_{i}$ | The $i$-th unit failure time |
| $t_{j i}$ | The $i$-th unit failure time under cause $j, j=1,2$ |
| $f()$. | The PDF of $t_{i}$ |
| $F()$. | The CDF of $t_{i}$ |
| $S()$. | Survival function of $t_{i}$ |
| $h()$. | Hazard failure rate of $t_{i}$ |
| $f_{j}()$. | The PDF of $t_{j i}$ |
| $F_{j}()$. | The CDF of $t_{j i}$ |
| $S_{j}()$. | Survival function of $t_{j i}$ |
| $h_{j}()$. | Hazard failure rate function of $t_{j i}$ |
| $\rho_{i}$ | The cause of failure form $i$-th unit failure time |
| $\mathrm{CD}(\alpha, \beta)$ | Chen distribution with PDF $\alpha \beta t^{\alpha-1} e^{\left\{t^{\alpha}+\beta\left[1-e^{t^{\alpha}}\right]\right\}}$ |
| MLE | Maximum likelihood estimators |
| AV | Average estimate |
| MSE | Mean squared error |
| AL | Average interval length |
| PC | Probability coverage |

$$
\begin{equation*}
F_{j 1}(t)=1-e^{\beta_{j}\left[1-e^{t^{\alpha}}\right]} \tag{9}
\end{equation*}
$$

Also, $S_{j 1}(t)$ and $h_{j 1}(t)$ of $\mathrm{CD}\left(\alpha, \beta_{j}\right)$ are, respectively, given by

$$
\begin{align*}
& S_{j 1}(t)=e^{\beta_{j}\left[1-e^{t^{\alpha}}\right]}, \\
& h_{j 1}(t)=\alpha \beta_{j} t^{\alpha-1} e^{t^{\alpha}} \tag{10}
\end{align*}
$$

(2) The random failure times $T_{i}, i=1,2, \ldots, d$, with the failure times under cause $j T_{j i}$, satisfy $T_{i}=\min \left\{T_{1 i}\right.$, $\left.T_{2 i}\right\}$.
(3) The total lifetime under use and accelerated condition $W$ under accelerated stress change $\eta$ is defined by

$$
W= \begin{cases}T, & T<\eta  \tag{11}\\ \eta+\frac{T-\eta}{\lambda}, & T>\eta\end{cases}
$$

where $\lambda$ is the accelerated factor. The random variable $W$ is distributed with Chen lifetime distribution with PDF and is given by

$$
f_{l}(w)= \begin{cases}0, & w<0  \tag{12}\\ f_{j 1}(w), & 0<w<\eta \\ f_{j 2}(w), & w \geq \eta\end{cases}
$$

where

$$
\begin{equation*}
f_{j 2}(w)=\alpha \lambda \beta_{j}(\eta+\lambda(w-\eta))^{\alpha-1} e^{(\eta+\lambda(w-\eta))^{\alpha}+\beta_{j}\left[1-e^{\left(\eta+\lambda(w-\eta)^{\alpha}\right.}\right]} \tag{13}
\end{equation*}
$$

and $f_{j 1}(z)$ is given by (8). The CDF, $S_{j 2}(z)$, and hazard rate function $h_{j 2}(z)$ are given by

$$
\begin{align*}
& F_{j 2}(w)=1-e^{\beta_{j}\left[1-e^{(\eta+\lambda(w-\eta))^{\alpha}}\right]},  \tag{14}\\
& S_{j 2}(w)=e^{\beta_{j}\left[1-e^{(\eta+\lambda(w-\eta))^{\alpha}}\right]},  \tag{15}\\
& h_{j 2}(w)=\alpha \lambda \beta_{j}(\eta+\lambda(w-\eta))^{\alpha-1} e^{(\eta+\lambda(w-\eta))^{\alpha}} . \tag{16}
\end{align*}
$$

(4) Under competing risks type-I GHC sample and partially step-stress ALTs model, the test is terminated at $T_{r}$ at $\tau<T_{r}$ and $\min \left(\tau, T_{m}\right)$ at $\tau>T_{r}$. Then, the random sample of the total lifetime $W$ is described by

$$
\begin{align*}
\underline{W} & =\left\{\left(W_{1}, \rho_{1}\right)<\left(W_{2}, \rho_{2}\right)<\cdots<\left(W_{k}, \rho_{k}\right)<\eta\right. \\
& \left.<\left(W_{k+1}, \rho_{k+1}\right)<\cdots<\left(W_{m}, \rho_{d}\right)\right\}, \tag{17}
\end{align*}
$$

where $d$ denotes the number of fail units, where $r<d<m$ and $k$ and $m-k$ are the numbers of fail units under using and accelerated conditions, respectively. For this model, we can consider three different cases, $\tau<\eta, \tau=\eta$, or $\tau<\eta$. Hence, the joint likelihood function of the observed values $\underline{w}=\left\{\left(w_{1}, \rho_{1}\right)<\left(w_{2}, \rho_{2}\right)<\cdots<\left(w_{k}, \rho_{k}\right)<\eta<\left(w_{k+1}, \rho_{k+1}\right)\right.$ $\left.<\cdots<\left(w_{m}, \rho_{d}\right)\right\}$ is obtained as follows:

$$
\begin{align*}
L(\underline{t} \mid \underline{\Theta})= & A\left[S_{12}\left(w_{d}\right) S_{22}\left(w_{d}\right)\right]^{(n-d)} \prod_{i=1}^{k}\left[h_{11}\left(w_{i}\right)\right]^{\delta\left(\rho_{i}=1\right)} \\
& {\left[h_{21}\left(w_{i}\right)\right]^{\delta\left(\rho_{i}=2\right)} S_{11}\left(w_{i}\right) S_{21}\left(w_{i}\right) \times } \\
& \prod_{i=k+1}^{d}\left[h_{12}\left(w_{i}\right)\right]^{\delta\left(\rho_{i}=1\right)}\left[h_{22}\left(w_{i}\right)\right]^{\delta\left(\rho_{i}=2\right)} S_{12}\left(w_{i}\right) S_{22}\left(w_{i}\right) . \tag{18}
\end{align*}
$$

## 3. Maximum Likelihood Estimation

When only two independent causes of failure and the test are running under the model of partially step-stress ALTs with type-I GHCS, the test information sample is used to obtain the point and interval MLEs which is reported in this section as follows.
3.1. MLEs. The joint likelihood function (18) under CDF (9) and (14) for the observed type-I GHC sample $\underline{w}=\left\{\left(w_{1}, \rho_{1}\right)\right.$ $\left.<\left(w_{2}, \rho_{2}\right)<\cdots<\left(w_{k}, \rho_{k}\right)<\eta<\left(w_{k+1}, \rho_{k+1}\right)<\cdots<\left(w_{m}, \rho_{d}\right)\right\}$ is given by

$$
\begin{align*}
& L\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right) \propto \alpha^{d} \beta_{1}^{m_{1}} \beta_{2}^{m_{2}} \lambda^{d-k} \prod_{i=1}^{k} w_{i}^{\alpha-1} \prod_{i=k+1}^{d}(\eta+\lambda(w-\eta))^{\alpha-1} \\
& e^{-\left(\beta_{1}+\beta_{2}\right)\left\{(n-d)\left[e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}\right]^{\alpha} \sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}-n\right\}}  \tag{19}\\
& e^{\sum_{i=1}^{k} w_{i}^{\alpha}+\sum_{i=k+1}^{d}(\eta+\lambda(w-\eta))^{\alpha}},
\end{align*}
$$

where integers $k$ and $(d-k)$ are denoted to failure units under using and stress conditions, respectively, and integers
$m_{1}$ and $m_{2}$ denoted failure units under causes ( $\rho_{1}, \rho_{2}$ ). Then, the log-form from (19) is reduced to

$$
\begin{align*}
\ell\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right)= & d \log \alpha+m_{1} \log \beta_{1}+m_{2} \log \beta_{2}+(d-k) \log \lambda-\left(\beta_{1}+\beta_{2}\right) \\
& \times\left\{(n-d) e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}-n\right\} \\
& +(\alpha-1)\left\{\sum_{i=1}^{k} \log w_{i}+\sum_{i=k+1}^{d} \log [\eta+\lambda(w-\eta)]\right\}+\sum_{i=1}^{k} w_{i}^{\alpha}  \tag{20}\\
& +\sum_{i=k+1}^{m}(\eta+\lambda(w-\eta))^{\alpha} .
\end{align*}
$$

$$
\begin{equation*}
0=\frac{\partial\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right)}{\partial \beta_{l}}, \quad l=1,2 \tag{21}
\end{equation*}
$$

The partial derivatives of log-likelihood function (20) are reduced to the likelihood equations solved with some numerical methods to obtain the estimates as follows:
is reduced to

$$
\begin{align*}
& \beta_{1}=\frac{m_{1}}{(n-d) e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}-n},  \tag{22}\\
& \beta_{2}=\frac{m_{2}}{(n-d) e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}-n} . \tag{23}
\end{align*}
$$

Also, is reduced to

$$
\begin{equation*}
0=\frac{\partial\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right)}{\partial \alpha} \tag{24}
\end{equation*}
$$

$$
\begin{align*}
& \frac{d}{\alpha}-\left(\beta_{1}+\beta_{2}\right)\left\{(n-d)\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha} e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}} \log \left(\eta+\lambda\left(w_{d}-\eta\right)\right)\right. \\
&\left.+\sum_{i=1}^{k} w_{i}^{\alpha} e^{w_{i}^{\alpha}} \log w_{i}+\sum_{i=k+1}^{d}\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}} \log \left(\eta+\lambda\left(w_{i}-\eta\right)\right)\right\} \\
&+\sum_{i=1}^{k} \log w_{i}+\sum_{i=k+1}^{d} \log [\eta+\lambda(w-\eta)]+\sum_{i=1}^{k} w_{i}^{\alpha} \log w_{i}  \tag{25}\\
&+\sum_{i=k+1}^{m}(\eta+\lambda(w-\eta))^{\alpha} \log (\eta+\lambda(w-\eta))=0 \\
& 0=\frac{\partial\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right)}{\partial \lambda}
\end{align*}
$$

which is reduced to

$$
\begin{align*}
& \frac{(d-k)}{\lambda}-\left(\beta_{1}+\beta_{2}\right)\left\{\alpha(n-d)\left(w_{d}-\eta\right)\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha-1} e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}\right. \\
& \left.\quad+\alpha \sum_{i=k+1}^{d}\left(w_{i}-\eta\right)\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha-1} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}\right\}  \tag{26}\\
& \quad+(\alpha-1) \sum_{i=k+1}^{d} \frac{(w-\eta)}{\eta+\lambda(w-\eta)}+\alpha \sum_{i=k+1}^{m}\left(w_{i}-\eta\right)\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha-1}=0
\end{align*}
$$

The likelihood equations are reduced to two nonlinear equations which are solved numerically with any iteration method such as Newton Raphson to obtain $\widehat{\alpha}$ and $\widehat{\lambda}$ which are used in (22) and (23) to present $\widehat{\beta}_{1}$ and $\widehat{\beta}_{2}$.
3.2. Interval Estimation. For the parameters vectors $\Theta=(\alpha$, $\left.\beta_{1}, \beta_{2}, \lambda\right)$, the second partial derivatives of (20) with respect to $\Theta_{i}$, where $\Theta_{1}=\alpha, \Theta_{2}=\beta_{1}, \Theta_{3}=\beta_{2}, \Theta_{4}=\lambda$, are given by

$$
\begin{equation*}
\frac{\partial\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right)}{\partial \Theta_{i} \partial \Theta_{l}}, \quad i, l=1,2,3,4 \tag{27}
\end{equation*}
$$

and the Fisher information matrix $\Sigma$ is given by

$$
\begin{equation*}
\Sigma=-E\left(\frac{\partial^{2}\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right)}{\partial \Theta_{i} \partial \Theta_{l}}\right) \tag{28}
\end{equation*}
$$

which is computed as the negative expectation of second partial derivatives (27). The approximate information matrix is used as the approximate form of the Fisher information matrix $\Sigma$ specially in a large sample. The approximate information matrix $\Sigma_{0}$ at the maximum likelihood estimates $\widehat{\Theta}=\left(\widehat{\alpha}, \widehat{\beta}_{1}, \widehat{\beta}_{2}, \widehat{\lambda}\right)$ is given by

$$
\begin{equation*}
\Sigma_{0}=-\left(\frac{\partial^{2}\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right)}{\partial \Theta_{i} \partial \Theta_{l}}\right)_{\widehat{\Theta}=\left\{\widehat{\alpha}_{,}, \widehat{\beta}_{1}, \widehat{\beta}_{2}, \widehat{\lambda}\right\}} \tag{29}
\end{equation*}
$$

The asymptotic normality distribution of estimating $\widehat{\alpha}$, $\widehat{\beta}_{1}, \widehat{\beta}_{2}$, and $\hat{\lambda}$ with mean $\left(\alpha, \beta_{1}, \beta_{2}, \lambda\right)$ and a variance covariance matrix $\Sigma_{0}^{-1}\left(\widehat{\alpha}, \widehat{\beta}_{1}, \widehat{\beta}_{2}, \widehat{\lambda}\right)$ is as follows:

$$
\begin{equation*}
\left(\widehat{\alpha}, \widehat{\beta}_{1}, \widehat{\beta}_{2}, \widehat{\lambda}\right) \text { distributed as } N\left(\left(\alpha, \beta_{1}, \beta_{2}, \lambda\right), \Sigma_{0}^{-1}\left(\widehat{\alpha}, \widehat{\beta}_{1}, \widehat{\beta}_{2}, \widehat{\lambda}\right)\right) . \tag{30}
\end{equation*}
$$

Therefore, $100(1-\xi)$ intervals estimation of parameters vector $\Theta=\left\{\alpha, \beta_{1}, \beta_{2}, \lambda\right\}$ are computed by

$$
\begin{equation*}
\widehat{\Theta}_{i} \mp z_{\xi / 2} \sqrt{\Sigma_{i i}}, \quad i=1,2,3,4, \tag{31}
\end{equation*}
$$

where $\widehat{\Theta}_{i}$ denotes the parameters estimate and value $\Sigma_{i i}$ denotes the diagonal of variance covariance matrix $\Sigma_{0}^{-1}$ with standard normal probability $\xi / 2$.

## 4. Bayesian Approach with MCMC

Information about the model parameters and the information which is obtained from the life sample is used in this section to build the Bayes approach with the MCMC method. Besides, the estimators of parameters of CD and noninformative about accelerated factor are computed under squared error loss (SEL) function and independent prior distributions. Therefore, independent gamma prior is adapted as follows:

$$
\begin{equation*}
\pi_{l}^{*}\left(\Theta_{l}\right) \propto \Theta_{l}^{a_{l}-1} e^{-b_{l} \Theta_{l}}, \quad l=1,2,3, \text { and } \pi_{4}^{*}\left(\Theta_{4}\right) \propto \Theta_{4}^{-1} \tag{32}
\end{equation*}
$$

$$
\begin{equation*}
\pi^{*}(\underline{\Theta})=\prod_{l=1}^{4} \pi_{l}^{*}\left(\Theta_{l}\right) \tag{33}
\end{equation*}
$$

where $\Theta=\left(\alpha, \beta_{1}, \beta_{2}, \lambda\right)$. Then, the posterior distribution of $\Theta$ is defined by

$$
\begin{equation*}
\pi(\underline{\Theta} \mid \underline{w})=\frac{\pi^{*}(\underline{\Theta}) L(\underline{\Theta} \mid \underline{w})}{\int_{\underline{\Theta}} \pi^{*}(\underline{\Theta}) L(\underline{\Theta} \mid \underline{w}) \mathrm{d} \underline{\Theta}} \tag{34}
\end{equation*}
$$

Then, the Bayes estimate for any function $\phi(\underline{\Theta})$ under SEL function is given by

$$
\begin{equation*}
\widehat{\phi}_{B}(\underline{\Theta})=\frac{\int_{\underline{\Theta}} \phi(\underline{\Theta}) \pi^{*}(\underline{\Theta}) L(\underline{\Theta} \mid \underline{w}) \mathrm{d}(\underline{\Theta})}{\int_{(\underline{\Theta})} \pi^{*}(\underline{\Theta}) L(\underline{\Theta} \mid \underline{w}) \mathrm{d}(\underline{\Theta})} . \tag{35}
\end{equation*}
$$

Generally, the ratio in (35) needs numerical approximation to compute, such as numerical integration and Lindley approximation. However, MCMC methods are the important tools that were applied recently with high accuracy and are obtained as follows.
4.1. Gibbs with MH Method. The posterior distribution in (34) with prior distribution (33) and likelihood function (19) is calculated as;

$$
\begin{align*}
& \pi\left(\alpha, \beta_{1}, \beta_{2}, \lambda \mid \underline{w}\right) \propto \alpha^{a_{1}+d-1} \beta_{1}^{a_{2}+m_{1}-1} \beta_{2}^{a_{3}+m_{2}-1} \lambda^{d-k-1} \prod_{i=1}^{k} w_{i}^{\alpha-1} \prod_{i=k+1}^{d}(\eta+\lambda(w-\eta))^{\alpha-1} \\
& \cdot e^{-\left(\beta_{1}+\beta_{2}\right)\left\{(n-d)\left[e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}\right]+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}-n\right\}}  \tag{36}\\
& \cdot e^{-b_{1} \alpha-b_{2} \beta_{1}-b_{3} \beta_{2}+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{m}(\eta+\lambda(w-\eta))^{\alpha}} .
\end{align*}
$$

Then, the conditional PDFs of the posterior distribution is given by

$$
\begin{align*}
& \pi_{1}\left(\alpha\left|\beta_{1}, \beta_{2}, \lambda\right| \underline{w}\right)=\alpha^{a_{1}+d-1} e^{-\left(\beta_{1}+\beta_{2}\right)\left\{(n-d)\left[e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}\right]+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}\right\}} \begin{array}{l}
e^{-b_{1} \alpha+\sum_{i=1}^{k} w_{i}^{\alpha}+\sum_{i=k+1}^{m}(\eta+\lambda(w-\eta))^{\alpha}} \prod_{i=1}^{k} w_{i}^{\alpha-1} \prod_{i=k+1}^{d}(\eta+\lambda(w-\eta))^{\alpha-1}, \\
\pi_{2}\left(\beta_{1}\left|\alpha, \beta_{2}, \lambda\right| \underline{w}\right) \propto \beta_{1}^{a_{2}+m_{1}-1} e^{-\beta_{1}\left\{b_{2}+(n-d)\left[e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}\right]+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left.\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}-n\right\}},\right.} \\
\pi_{3}\left(\beta_{2}\left|\alpha, \beta_{1}, \lambda\right| \underline{w}\right) \propto \beta_{2}^{a_{3}+m_{2}-1} e^{-\beta_{2}\left\{b_{3}+(n-d)\left[e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}\right]+\sum_{i=1}^{k} e^{w_{i}^{\alpha}}+\sum_{i=k+1}^{d} e^{\left.\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}-n\right\}},\right.} \\
\pi_{4}\left(\lambda \mid \alpha, \beta_{1}, \beta_{2}, \underline{w}\right) \propto \lambda^{d-k-1} e^{-\left(\beta_{1}+\beta_{2}\right)}\left[(n-d)\left[e^{\left(\eta+\lambda\left(w_{d}-\eta\right)\right)^{\alpha}}\right]+\sum_{i=k+1}^{k} e^{\left(\eta+\lambda\left(w_{i}-\eta\right)\right)^{\alpha}}\right] \\
\quad+\sum_{i=k+1}^{m}(\eta+\lambda(w-\eta))^{\alpha} \\
\prod_{i=k+1}^{d}(\eta+\lambda(w-\eta))^{\alpha-1}
\end{array}, l
\end{align*}
$$

From equations (37) and (38), the conditional posterior PDFs are reduced to two conditional gamma density equations (38) and (39). Two functions are plotted similar to the normal distribution in (37) and (40). Then, the process of generation from posterior distribution under the conditional posterior distribution by using Gibbs with the MH algorithms with normal proposal distribution [27] is given as follows:
(1) Begin with initial vectors $\Theta^{(0)}=\left(\alpha^{(0)}, \beta_{1}^{(0)}, \beta_{2}^{(0)}, \lambda^{(0)}\right)$ and indicator $\kappa=1$.
(2) The two values $\beta_{1}^{(\kappa)}$ and $\beta_{2}^{(\kappa)}$ are generated from conditional gamma densities (38) and (39).
(3) The two values $\alpha^{(\kappa)}$ and $\lambda^{(\kappa)}$ are generated from conditional densities (37) and (40) by MH algorithms with normal proposal distributions. The symmetric normal distributions are applied with mean $\alpha^{(\kappa-1)}$ or $\lambda^{(\kappa-)}$ and variance obtained from the diagonal of the approximate information matrix, respectively. Also, the generated values are accepted
with acceptance probability $\min \left[1,\left(\pi_{4}\left(\alpha^{(\kappa)} \mid\right.\right.\right.$ $\left.\left.\left.\lambda^{(\kappa-1)}, \beta_{1}^{(\kappa)}, \beta_{2}^{(\kappa)}, \underline{w}\right) / \pi_{4}\left(\alpha^{(\kappa-1)} \mid \quad \lambda^{(\kappa-1)}, \beta_{1}^{(\kappa)}, \beta_{2}^{(\kappa)}, \underline{w}\right)\right)\right]$ or $\min \left[1,\left(\pi_{4}\left(\lambda^{(\kappa)} \mid \alpha^{(\kappa)}, \beta_{1}^{(\kappa)}, \beta_{2}^{(\kappa)}, \underline{w}\right) / \pi_{4}\left(\lambda^{(\kappa-1)} \mid \alpha^{(\kappa)}\right.\right.\right.$, $\left.\left.\left.\beta_{1}^{(\kappa)}, \beta_{2}^{(\kappa)}, \underline{w}\right)\right)\right]$, respectively, with respect to uniform $(0,1)$.
(4) The vector $\Theta^{(\kappa)}=\left(\alpha^{(\kappa)}, \beta_{1}^{(\kappa)}, \beta_{2}^{(\kappa)}, \lambda^{(\kappa)}\right)$ is a built vector with Gibbs manner.
(5) Put $\kappa=1+1$ and then repeat steps $2-4 N$ times.
(6) The Bayes estimates and the corresponding variance are given by $\widehat{\Theta}_{B}=(1 / N-N) \sum_{i=N+1}^{N} \Theta^{(i)}$ and $\operatorname{Var}\left(\widehat{\Theta}_{B}\right)=(1 / N-N) \quad \sum_{i=N+1}^{N} \quad\left(\widehat{\Theta}_{B}-\Theta{ }^{(i)}\right)^{2}$, where $\mathbf{N}$ is the number of iteration used to get stationary distribution.
(7) The Bayes credible intervals can be obtained after ordered values of $\Theta_{l}^{(\kappa)}$ in acceding order of $\Theta_{l}^{(\mathbf{N}+1)}$, $\Theta_{l}^{(\mathrm{N}+2)}, \ldots, \Theta_{l}^{[N]}$ as $\Theta_{l}^{[1]}, \Theta_{l}^{[2]}, \ldots, \Theta_{l}^{[N-\mathrm{N}]}$. Then, the $100(1-\zeta) \%$ symmetric credible intervals are obtained as $\quad\left(\Theta_{l}^{(N-N)(\zeta / 2)}, \Theta_{l}^{(N-N)(1-(\zeta / 2))}\right), \quad l=$ $1,2,3,4$.

TAble 2: The AV and MSE of parameters estimates for $\Theta(2,0.1,0.2,2.0)$.

| $(\eta, \tau)$ | ( $r, m, n$ ) | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| (0.8, 1.0) | (15, 20, 30) | 2.321 | 0.152 | 0.231 | 2.412 | 2.301 | 0.140 | 0.218 | 2.392 | 2.321 | 0.132 | 0.211 | 2.355 |
|  |  | 0.514 | 0.042 | 0.062 | 0.641 | 0.505 | 0.041 | 0.059 | 0.613 | 0.428 | 0.032 | 0.042 | 0.517 |
|  | $(15,25,30)$ | 2.305 | 0.150 | 0.229 | 2.400 | 2.299 | 0.140 | 0.215 | 2.366 | 2.309 | 0.132 | 0.213 | 2.341 |
|  |  | 0.485 | 0.033 | 0.057 | 0.632 | 0.481 | 0.035 | 0.056 | 0.629 | 0.376 | 0.028 | 0.029 | 0.488 |
|  | (20, 25, 30) | 2.317 | 0.147 | 0.224 | 2.389 | 2.294 | 0.136 | 0.209 | 2.359 | 2.290 | 0.124 | 0.208 | 2.337 |
|  |  | 0.479 | 0.029 | 0.049 | 0.628 | 0.478 | 0.029 | 0.050 | 0.624 | 0.371 | 0.024 | 0.024 | 0.480 |
|  | (20, 30, 50) | 2.301 | 0.132 | 0.201 | 2.340 | 2.290 | 0.129 | 0.208 | 2.338 | 2.287 | 0.113 | 0.207 | 2.324 |
|  |  | 0.401 | 0.025 | 0.022 | 0.589 | 0.422 | 0.019 | 0.019 | 0.582 | 0.342 | 0.014 | 0.017 | 0.402 |
|  | (20, 40, 50) | 2.303 | 0.125 | 0.212 | 2.338 | 2.289 | 0.124 | 0.207 | 2.334 | 2.279 | 0.112 | 0.206 | 2.319 |
|  |  | 0.398 | 0.019 | 0.021 | 0.560 | 0.410 | 0.018 | 0.018 | 0.563 | 0.332 | 0.012 | 0.013 | 0.388 |
|  | $(30,40,50)$ | 2.298 | 0.124 | 0.207 | 2.327 | 2.279 | 0.118 | 0.208 | 2.332 | 2.269 | 0.109 | 0.206 | 2.314 |
|  |  | 0.379 | 0.017 | 0.018 | 0.549 | 0.390 | 0.016 | 0.018 | 0.544 | 0.314 | 0.011 | 0.013 | 0.371 |

Table 3: The AV and MSE of parameters estimates for $\Theta(2,0.1,0.2,2.0)$.

| $(\eta, \tau)$ | ( $r, m, n$ ) | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| (1.0, 1.0) | (15, 20, 30) | 2.311 | 0.140 | 0.222 | 2.401 | 2.289 | 0.129 | 0.209 | 2.380 | 2.313 | 0.121 | 0.201 | 2.344 |
|  |  | 0.490 | 0.032 | 0.051 | 0.620 | 0.490 | 0.031 | 0.048 | 0.594 | 0.412 | 0.022 | 0.031 | 0.500 |
|  | (15, 25, 30) | 2.294 | 0.139 | 0.218 | 2.391 | 2.288 | 0.129 | 0.204 | 2.357 | 2.298 | 0.121 | 0.201 | 2.330 |
|  |  | 0.461 | 0.023 | 0.045 | 0.612 | 0.464 | 0.025 | 0.047 | 0.608 | 0.355 | 0.018 | 0.019 | 0.469 |
|  | (20, 25, 30) | 2.307 | 0.138 | 0.215 | 2.379 | 2.285 | 0.126 | 0.200 | 2.348 | 2.288 | 0.114 | 0.200 | 2.329 |
|  |  | 0.460 | 0.018 | 0.038 | 0.609 | 0.456 | 0.018 | 0.041 | 0.601 | 0.352 | 0.014 | 0.013 | 0.461 |
|  | (20, 30, 50) | 2.291 | 0.122 | 0.195 | 2.330 | 2.280 | 0.119 | 0.190 | 2.328 | 2.275 | 0.102 | 0.198 | 2.313 |
|  |  | 0.385 | 0.014 | 0.011 | 0.574 | 0.403 | 0.008 | 0.009 | 0.564 | 0.325 | 0.005 | 0.008 | 0.385 |
|  | ( $20,40,50)$ | 2.295 | 0.115 | 0.203 | 2.327 | 2.278 | 0.113 | 0.198 | 2.324 | 2.268 | 0.100 | 0.195 | 2.308 |
|  |  | 0.378 | 0.010 | 0.011 | 0.542 | 0.401 | 0.009 | 0.009 | 0.544 | 0.315 | 0.004 | 0.007 | 0.368 |
|  | (30, 40, 50) | 2.288 | 0.112 | 0.195 | 2.318 | 2.268 | 0.109 | 0.200 | 2.322 | 2.249 | 0.100 | 0.197 | 2.301 |
|  |  | 0.358 | 0.007 | 0.008 | 0.530 | 0.371 | 0.009 | 0.007 | 0.521 | 0.290 | 0.004 | 0.004 | 0.356 |

TAble 4: The AV and MSE of parameters estimates for $\Theta(2,0.1,0.2,2.0)$.

| $(\eta, \tau)$ | (r,m,n) | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| (1.2, 1.0) | $(15,20,30)$ | 2.341 | 0.162 | 0.252 | 2.430 | 2.322 | 0.160 | 0.239 | 2.412 | 2.342 | 0.153 | 0.231 | 2.374 |
|  |  | 0.545 | 0.073 | 0.095 | 0.671 | 0.535 | 0.074 | 0.090 | 0.645 | 0.460 | 0.061 | 0.071 | 0.550 |
|  | $(15,25,30)$ | 2.325 | 0.170 | 0.249 | 2.419 | 2.318 | 0.161 | 0.236 | 2.386 | 2.328 | 0.151 | 0.233 | 2.360 |
|  |  | 0.515 | 0.073 | 0.088 | 0.672 | 0.507 | 0.065 | 0.086 | 0.661 | 0.399 | 0.059 | 0.061 | 0.510 |
|  | $(20,25,30)$ | 2.337 | 0.165 | 0.244 | 2.410 | 2.310 | 0.156 | 0.228 | 2.379 | 2.310 | 0.143 | 0.227 | 2.358 |
|  |  | 0.509 | 0.060 | 0.081 | 0.660 | 0.508 | 0.058 | 0.081 | 0.658 | 0.399 | 0.055 | 0.056 | 0.503 |
|  | (20, 30, 50) | $2.319$ | 0.150 | 0.224 | 2.365 | 2.310 | 0.148 | 0.228 | 2.355 | 2.305 | 0.133 | 0.227 | 2.345 |
|  |  | 0.431 | 0.055 | 0.052 | 0.610 | 0.452 | 0.050 | 0.049 | 0.605 | 0.372 | 0.044 | 0.048 | 0.437 |
|  | (20, 40, 50) | $2.323$ | 0.145 | 0.233 | 2.358 | $2.307$ | 0.144 | 0.226 | 2.355 | 2.299 | 0.131 | 0.227 | 2.338 |
|  |  | $0.428$ | 0.050 | 0.052 | 0.594 | 0.440 | 0.038 | 0.039 | 0.594 | 0.365 | 0.041 | 0.042 | 0.411 |
|  | $(30,40,50)$ | $2.308$ | $0.141$ | 0.227 | 2.347 | 2.298 | 0.138 | 0.229 | 2.351 | 2.290 | 0.131 | 0.228 | 2.333 |
|  |  | 0.408 | 0.047 | 0.048 | 0.578 | 0.421 | 0.047 | 0.048 | 0.574 | 0.345 | 0.042 | 0.045 | 0.399 |

## 5. Monte Carlo Simulations

Different theoretical results that were developed in this study are assessed and compared in this section through the Monte Carlo study. The numerical computation is formulated to measure the effect of changing the total size $n$ and the sample
size $m$ as well as stress time $\eta$ and censoring time $\tau$ and the effect of parameters change. Different three cases of two time ( $\tau<\eta, \tau=\eta, \tau>\eta$ ) and two values of parameters vector $\Theta=\left(\alpha, \beta_{1}, \beta_{2}, \lambda\right)=\{(2.0,0.1,0.2,2.0),(0.7,1.5,2.0,2.5)\}$ are assumed. Point and interval estimation results are measured. Average (AV) and mean squared error (MSE) are used to

Table 5: The AL and CP of parameters estimates for $\Theta(2,0.1,0.2,2.0)$.

| $(\eta, \tau)$ | $(r, m, n)$ | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| (0.8, 1.0) | $(15,20,30)$ | 4.321 | 0.321 | 0.421 | 5.212 | 4.314 | 0.318 | 0.419 | 5.209 | 4.045 | 0.298 | 0.318 | 5.009 |
|  |  | 0.89 | 0.90 | 0.90 | 0.88 | 0.90 | 0.91 | 0.91 | 0.90 | 0.90 | 0.92 | 0.90 | 0.91 |
|  | $(15,25,30)$ | 4.300 | 0.299 | 0.401 | 5.180 | 4.280 | 0.295 | 0.400 | 5.189 | 4.045 | 0.266 | 0.300 | 4.975 |
|  |  | 0.90 | 0.91 | 0.91 | 0.90 | 0.91 | 0.91 | 0.92 | 0.91 | 0.93 | 0.94 | 0.91 | 0.92 |
|  | $(20,25,30)$ | 4.271 | 0.268 | 0.375 | 5.154 | 4.255 | 0.265 | 0.381 | 5.174 | 4.018 | 0.239 | 0.271 | 4.944 |
|  |  | 0.91 | 0.91 | 0.92 | 0.91 | 0.91 | 0.93 | 0.92 | 0.93 | 0.93 | 0.94 | 0.96 | 0.93 |
|  | $(20,30,50)$ | 4.248 | 0.241 | 0.348 | 5.125 | 4.230 | 0.235 | 0.351 | 5.144 | 4.001 | 0.211 | 0.242 | 4.914 |
|  |  | 0.92 | 0.92 | 0.92 | 0.90 | 0.93 | 0.93 | 0.93 | 0.93 | 0.93 | 0.95 | 0.95 | 0.92 |
|  | (20, 40, 50) | 4.219 | 0.215 | 0.321 | 5.100 | 4.201 | 0.203 | 0.318 | 5.115 | 3.971 | 0.180 | 0.215 | 4.850 |
|  |  | 0.91 | 0.93 | 0.93 | 0.91 | 0.93 | 0.94 | 0.93 | 0.93 | 0.94 | 0.97 | 0.95 | 0.93 |
|  | (30, 40, 50) | 4.185 | 0.194 | 0.300 | 4.990 | 4.150 | 0.164 | 0.270 | 5.070 | 3.941 | 0.149 | 0.176 | 4.817 |
|  |  | 0.92 | 0.93 | 0.93 | 0.92 | 0.93 | 0.94 | 0.93 | 0.93 | 0.94 | 0.93 | 0.95 | 0.94 |

Table 6: The AL and CP of parameters estimates for $\Theta(2,0.1,0.2,2.0)$.

| $(\eta, \tau)$ | (r,m,n) | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| (1.0, 1.0) | $(15,20,30)$ | 4.285 | 0.291 | 0.387 | 5.088 | 4.300 | 0.290 | 0.400 | 5.187 | 4.019 | 0.270 | 0.294 | 4.980 |
|  |  | 0.90 | 0.91 | 0.90 | 0.89 | 0.91 | 0.91 | 0.91 | 0.92 | 0.90 | 0.92 | 0.91 | 0.91 |
|  | $(15,25,30)$ | 4.272 | 0.271 | 0.375 | 5.152 | 4.253 | 0.269 | 0.374 | 5.158 | 4.014 | 0.249 | 0.274 | 4.948 |
|  |  | 0.91 | 0.91 | 0.92 | 0.90 | 0.91 | 0.91 | 0.92 | 0.93 | 0.93 | 0.94 | 0.91 | 0.94 |
|  | (20, 25, 30) | 4.242 | 0.235 | 0.342 | 5.118 | 4.255 | 0.239 | 0.351 | 5.149 | 3.980 | 0.208 | 0.241 | 4.915 |
|  |  | 0.92 | 0.93 | 0.92 | 0.92 | 0.91 | 0.93 | 0.92 | 0.95 | 0.93 | 0.94 | 0.96 | 0.94 |
|  | (20, 30, 50) | 4.219 | 0.212 | 0.318 | 5.100 | 4.204 | 0.208 | 0.328 | 5.113 | 3.971 | 0.185 | 0.214 | 4.875 |
|  |  | 0.93 | 0.92 | 0.94 | 0.90 | 0.93 | 0.93 | 0.97 | 0.93 | 0.93 | 0.95 | 0.97 | 0.92 |
|  | (20, 40, 50) | $4.181$ | 0.185 | 0.284 | 5.079 | 4.172 | 0.169 | 0.280 | 5.081 | 3.940 | 0.151 | 0.179 | 4.821 |
|  |  | 0.96 | 0.92 | 0.93 | 0.91 | 0.93 | 0.93 | 0.93 | 0.93 | 0.94 | 0.97 | 0.95 | 0.94 |
|  | $(30,40,50)$ | 4.158 | 0.166 | 0.274 | 4.965 | 4.118 | 0.135 | 0.241 | 5.040 | 3.915 | 0.118 | 0.148 | 4.782 |
|  |  | 0.92 | 0.94 | 0.93 | 0.92 | 0.93 | 0.94 | 0.92 | 0.93 | 0.94 | 0.95 | 0.95 | 0.94 |

Table 7: The AL and CP of parameters estimates for $\Theta(2,0.1,0.2,2.0)$.

| $(\eta, \tau)$ | (r,m,n) | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| (1.2, 1.0) | $(15,20,30)$ | 4.371 | 0.369 | 0.470 | 5.269 | 4.370 | 0.371 | 0.466 | 5.255 | 4.098 | 0.345 | 0.361 | 5.057 |
|  |  | 0.88 | 0.89 | 0.90 | 0.88 | 0.91 | 0.90 | 0.91 | 0.91 | 0.91 | 0.92 | 0.90 | 0.91 |
|  | $(15,25,30)$ | 4.347 | 0.245 | 0.453 | 5.230 | 4.330 | 0.345 | 0.450 | 5.238 | 4.091 | 0.299 | 0.348 | 5001 |
|  |  | 0.89 | 0.90 | 0.91 | 0.91 | 0.91 | 0.92 | 0.92 | 0.93 | 0.93 | 0.94 | 0.93 | 0.92 |
|  | $(20,25,30)$ | 4.318 | 0.310 | 0.415 | 5.198 | 4.294 | 0.310 | 0.428 | 5.221 | 4.071 | 0.288 | 0.321 | 4.989 |
|  |  | 0.90 | 0.91 | 0.92 | 0.90 | 0.92 | 0.93 | 0.92 | 0.93 | 0.94 | 0.94 | 0.96 | 0.97 |
|  | $(20,30,50)$ | 4.291 | 0.287 | 0.389 | 5.176 | 4.280 | 0.279 | 0.398 | 5.188 | 4.056 | 0.261 | 0.289 | 4.961 |
|  |  | 0.92 | 0.91 | 0.92 | 0.90 | 0.92 | 0.93 | 0.93 | 0.93 | 0.93 | 0.93 | 0.95 | 0.94 |
|  | (20, 40, 50) | 4.270 | 0.265 | 0.372 | 5.164 | 4.245 | 0.253 | 0.359 | 5.164 | 4.008 | 0.220 | 0.266 | 4.894 |
|  |  | 0.92 | 0.93 | 0.90 | 0.91 | 0.93 | 0.94 | 0.93 | 0.92 | 0.94 | 0.97 | 0.94 | 0.93 |
|  | $(30,40,50)$ | 4.221 | 0.243 | 0.351 | 5.011 | 4.194 | 0.210 | 0.320 | 5.123 | 3.990 | 0.187 | 0.516 | 4.866 |
|  |  | 0.92 | 0.90 | 0.93 | 0.91 | 0.93 | 0.94 | 0.93 | 0.93 | 0.94 | 0.94 | 0.95 | 0.96 |

measure point estimate and average lengths (AL) and the probability coverage ( CP ) are used to measure interval estimate. For Bayes estimation with MCMC methods, prior parameters are selected to satisfy the expectation of gamma prior as $E\left(\Theta_{l}\right)=\left(a_{l} / b_{l}\right) \cong \Theta_{l}, l=1,2,3$. Then, informative
prior information (Prior 1) and noninformative prior are obtained when the posterior distribution is proportional with the likelihood function (Prior 0). Also, Chan is built for 11,000 iterations with the first 1000 as bur-in. Average Bayes estimates, mean squared errors (MSEs), coverage

Table 8: The AV and MSE of parameters estimates for $\Theta$ (1.0, $0.8,1.0,2.5$ ).

|  |  |  |  |  |  |  | MCM | Prior 0 |  |  | MCM | Prior 1 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $(r, m, n)$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| $(0.3,0.5)$ | $(15,20,30)$ | 1.235 | 1.001 | 1.321 | 2.821 | 1.214 | 0.982 | 1.301 | 2.805 | 1.201 | 0.854 | 1.241 | 2.754 |
|  |  | 0.254 | 0.221 | 0.243 | 0.647 | 0.241 | 0.213 | 0.232 | 0.641 | 0.185 | 0.167 | 0.200 | 0.425 |
|  | $(15,25,30)$ | 1.219 | 0.985 | 1.307 | 2.809 | 1.200 | 0.968 | 1.280 | 2.791 | 1.189 | 0.842 | 1.229 | 2.742 |
|  |  | 0.237 | 0.205 | 0.218 | 0.622 | 0.221 | 0.190 | 0.214 | 0.628 | 0.165 | 0.155 | 0.193 | 0.417 |
|  | $(20,25,30)$ | 1.221 | 0.982 | 1.313 | 2.814 | 1.201 | 0.955 | 1.275 | 2.792 | 1.178 | 0.840 | 1.214 | 2.735 |
|  |  | 0.227 | 0.201 | 0.212 | 0.619 | 0.218 | 0.183 | 0.208 | 0.614 | 0.145 | 0.147 | 0.190 | 0.409 |
|  | (20, 30, 50) | 1.214 | 0.977 | 1.307 | 2.817 | 1.194 | 0.948 | 1.269 | 2.790 | 1.160 | 0.825 | 1.204 | 2.727 |
|  |  | 0.209 | 0.192 | 0.200 | 0.604 | 0.201 | 0.164 | 0.201 | 0.601 | 0.139 | 0.132 | 0.179 | 0.401 |
|  | (20, 40, 50) | 1.195 | 0.899 | 1.265 | 2.800 | 1.191 | 0.888 | 1.255 | 2.718 | 1.124 | 0.801 | 1.176 | 2.702 |
|  |  | 0.182 | 0.176 | 0.166 | 0.550 | 0.183 | 0.161 | 0.164 | 0.547 | 0.101 | 0.108 | 0.151 | 0.382 |
|  | $(30,40,50)$ | 1.184 | 0.875 | 1.240 | 2.741 | 1.179 | 0.872 | 1.221 | 2.719 | 1.127 | 0.789 | 1.154 | 2.701 |
|  |  | 0.174 | 0.160 | 0.142 | 0.514 | 0.156 | 0.140 | 0.133 | 0.519 | 0.099 | 0.089 | 0.132 | 0.314 |

Table 9: The AV and MSE of parameters estimates for $\Theta$ (1.0, $0.8,1.0,2.5$ ).

| $(\eta, \tau)$ | (r,m,n) | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| $(0.5,0.5)$ | $(15,20,30)$ | 1.219 | 0945 | 1.300 | 2.800 | 1.184 | 0.973 | 1.290 | 2.780 | 1.212 | 0.866 | 1.254 | 2.767 |
|  |  | 0.233 | 0.207 | 0.219 | 0.618 | 0.215 | 0.200 | 0.211 | 0.619 | 0.155 | 0.149 | 0.179 | 0.403 |
|  | $(15,25,30)$ | 1.201 | 0.971 | 1.219 | 2.800 | 1.203 | 0.954 | 1.275 | 2.787 | 1.180 | 0.845 | 1.217 | 2.732 |
|  |  | 0.219 | 0.187 | 0.204 | 0.604 | 0.200 | 0.172 | 0.189 | 0.611 | 0.152 | 0.141 | 0.173 | 0.401 |
|  | $(20,25,30)$ | 1.211 | 0.975 | 1.300 | 2.801 | 1.192 | 0.947 | 1.269 | 2.790 | 1.179 | 0.842 | 1.207 | 2.725 |
|  |  | 0.209 | 0.175 | 0.200 | 0.601 | 0.197 | 0.164 | 0.187 | 0.600 | 0.132 | 0.125 | 0.174 | 0.390 |
|  | (20, 30, 50) | 1.204 | 0.961 | 1.310 | 2.807 | 1.182 | 0.941 | 1.260 | 2.784 | 1.163 | 0.817 | 1.191 | 2.708 |
|  |  | 0.191 | 0.179 | 0.183 | 0.581 | 0.180 | 0.162 | 0.189 | 0.580 | 0.117 | 0.111 | 0.160 | 0.382 |
|  | (20, 40, 50) | 1.190 | 0.880 | 1.260 | 2.803 | 1.190 | 0.878 | 1.247 | 2.725 | 1.132 | 0.715 | 1.160 | 2.650 |
|  |  | 0.161 | 0.156 | 0.149 | 0.535 | 0.168 | 0.145 | 0.145 | 0.538 | 0.089 | 0.079 | 0.138 | 0.365 |
|  | $(30,40,50)$ | 1.176 | 0.871 | 1.233 | 2.742 | 1.165 | 0.869 | 1.217 | 2.731 | 1.130 | 0.776 | 1.151 | 2.690 |
|  |  | 0.155 | 0.143 | 0.128 | 0.500 | 0.141 | 0.128 | 0.117 | 0.504 | 0.081 | 0.076 | 0.114 | 0.301 |

Table 10: The AV and MSE of parameters estimates for $\Theta$ (1.0, $0.8,1.0,2.5$ ).

| $(\eta, \tau)$ | $(r, m, n)$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |
|  | $(15,20,30)$ | 1.235 | 1.001 | 1.321 | 2.821 | 1.214 | 0.982 | 1.301 | 2.805 | 1.201 | 0.854 | 1.241 | 2.754 |
|  |  | 0.254 | 0.221 | 0.243 | 0.647 | 0.241 | 0.213 | 0.232 | 0.641 | 0.185 | 0.167 | 0.200 | 0.425 |
|  | $(15,25,30)$ | 1.219 | 0.985 | 1.307 | 2.809 | 1.200 | 0.968 | 1.280 | 2.791 | 1.189 | 0.842 | 1.229 | 2.742 |
|  |  | 0.237 | 0.205 | 0.218 | 0.622 | 0.221 | 0.190 | 0.214 | 0.628 | 0.165 | 0.155 | 0.193 | 0.417 |
|  | $(20,25,30)$ | 1.221 | 0.982 | 1.313 | 2.814 | 1.201 | 0.955 | 1.275 | 2.792 | 1.178 | 0.840 | 1.214 | 2.735 |
| $(0.3,0.5)$ |  | 0.227 | 0.201 | 0.212 | 0.619 | 0.218 | 0.183 | 0.208 | 0.614 | 0.145 | 0.147 | 0.190 | 0.409 |
|  | $(20,30,50)$ | 1.214 | 0.977 | 1.307 | 2.817 | 1.194 | 0.948 | 1.269 | 2.790 | 1.160 | 0.825 | 1.204 | 2.727 |
|  |  | 0.209 | 0.192 | 0.200 | 0.604 | 0.201 | 0.164 | 0.201 | 0.601 | 0.139 | 0.132 | 0.179 | 0.401 |
|  | $(20,40,50)$ | 1.195 | 0.899 | 1.265 | 2.800 | 1.191 | 0.888 | 1.255 | 2.718 | 1.124 | 0.801 | 1.176 | 2.702 |
|  |  | 0.182 | 0.176 | 0.166 | 0.550 | 0.183 | 0.161 | 0.164 | 0.547 | 0.101 | 0.108 | 0.151 | 0.382 |
|  | $(30,40,50)$ | 1.184 | 0.875 | 1.240 | 2.741 | 1.179 | 0.872 | 1.221 | 2.719 | 1.127 | 0.789 | 1.154 | 2.701 |
|  | 0.174 | 0.160 | 0.142 | 0.514 | 0.156 | 0.140 | 0.133 | 0.519 | 0.099 | 0.089 | 0.132 | 0.314 |  |

Table 11: The AL and CP of parameters estimates for $\Theta$ (1.0, $0.8,1.0,2.5)$.

|  |  |  |  |  |  |  | MCM | Prior 0 |  |  | MCM | Prior 1 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $(r, m, n)$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| $(0.3,0.5)$ | $(15,20,30)$ | 2.824 | 2.541 | 3.104 | 5.512 | 2.801 | 2.549 | 3.097 | 5.489 | 2.612 | 2.325 | 2.900 | 5.342 |
|  |  | 0.89 | 0.88 | 0.89 | 0.87 | 0.89 | 0.90 | 0.90 | 0.88 | 0.90 | 0.90 | 0.90 | 0.90 |
|  | $(15,25,30)$ | 2.804 | 2.525 | 3.077 | 5.500 | 2.785 | 2.525 | 3.077 | 5.468 | 2.591 | 2.301 | 2.879 | 5.324 |
|  |  | 0.90 | 0.90 | 0.90 | 0.90 | 0.91 | 0.91 | 0.91 | 0.91 | 0.91 | 0.92 | 0.91 | 0.92 |
|  | $(20,25,30)$ | 2.781 | 2.507 | 3.059 | 5.480 | 2.766 | 2.508 | 3.059 | 5.455 | 2.579 | 2.284 | 2.862 | 5.304 |
|  |  | 0.91 | 0.91 | 0.92 | 0.92 | 0.91 | 0.93 | 0.92 | 0.92 | 0.93 | 0.94 | 0.98 | 0.93 |
|  | (20, 30, 50) | 2.759 | 2.487 | 3.042 | 5.461 | 2.745 | 2.592 | 3.040 | 5.439 | 2.565 | 2.271 | 2.842 | 5.281 |
|  |  | 0.91 | 0.92 | 0.91 | 0.90 | 0.93 | 0.93 | 0.91 | 0.93 | 0.93 | 0.92 | 0.92 | 0.92 |
|  | (20, 40, 50) | 2.744 | 2.464 | 3.031 | 5.449 | 2.736 | 2.581 | 3.024 | 5.418 | 2.548 | 2.255 | 2.819 | 5.260 |
|  |  | 0.912 | 0.93 | 0.92 | 0.90 | 0.93 | 0.94 | 0.93 | 0.93 | 0.93 | 0.92 | 0.94 | 0.93 |
|  | (30, 40, 50) | 2.719 | 2.439 | 3.008 | 5.413 | 2.714 | 2.547 | 3.004 | 5.400 | 2.521 | 2.228 | 2.801 | 5.239 |
|  |  | 0.92 | 0.92 | 0.93 | 0.92 | 0.92 | 0.94 | 0.93 | 0.94 | 0.94 | 0.93 | 0.95 | 0.95 |

Table 12: The AL and CP of parameters estimates for $\Theta$ (1.0, $0.8,1.0,2.5)$.

| $(\eta, \tau)$ | $(r, m, n)$ | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| $(0.5,0.5)$ | $(15,20,30)$ | 2.801 | 2.514 | 3.076 | 5.500 | 2.783 | 2.515 | 3.072 | 5.471 | 2.593 | 2.304 | 2.879 | 5.319 |
|  |  | 0.90 | 0.90 | 0.89 | 0.89 | 0.90 | 0.90 | 0.91 | 0.90 | 0.90 | 0.91 | 0.90 | 0.92 |
|  | $(15,25,30)$ | 2.782 | 2.507 | 3.058 | 5.483 | 2.769 | 2.507 | 3.061 | 5.454 | 2.576 | 2.279 | 2.864 | 5.308 |
|  |  | 0.92 | 0.91 | 0.90 | 0.91 | 0.92 | 0.91 | 0.92 | 0.91 | 0.91 | 0.92 | 0.93 | 0.93 |
|  | $(20,25,30)$ | 2.756 | 2.486 | 3.035 | 5.464 | 2.748 | 2.492 | 3.041 | 5.441 | 2.568 | 2.269 | 2.847 | 5.292 |
|  |  | 0.92 | 0.93 | 0.92 | 0.93 | 0.91 | 0.93 | 0.94 | 0.92 | 0.93 | 0.95 | 0.96 | 0.95 |
|  | $(20,30,50)$ | 2.743 | 2.461 | 3.020 | 5.443 | 2.738 | 2.582 | 3.025 | 5.428 | 2.547 | 2.256 | 2.821 | 5.255 |
|  |  | 0.93 | 0.92 | 0.93 | 0.93 | 0.93 | 0.93 | 0.93 | 0.93 | 0.93 | 0.95 | 0.94 | 0.93 |
|  | (20, 40, 50) | 2.725 | 2.451 | 3.014 | 5.432 | 2.718 | 2.563 | 3.011 | 5.408 | 2.531 | 2.241 | 2.808 | 5.245 |
|  |  | 0.92 | 0.93 | 0.92 | 0.90 | 0.94 | 0.94 | 0.93 | 0.93 | 0.94 | 0.92 | 0.94 | 0.96 |
|  | (30, 40, 50) | 2.719 | 2.439 | 3.008 | 5.413 | 2.714 | 2.547 | 3.004 | 5.400 | 2.521 | 2.228 | 2.801 | 5.239 |
|  |  | 0.92 | 0.92 | 0.93 | 0.92 | 0.92 | 0.94 | 0.93 | 0.94 | 0.94 | 0.93 | 0.95 | 0.95 |

Table 13: The AL and CP of parameters estimates for $\Theta$ (1.0, $0.8,1.0,2.5$ ).

| $(\eta, \tau)$ | (r,m,n) | MLE |  |  |  | MCMC/Prior 0 |  |  |  | MCMC/Prior 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ | $\alpha$ | $\beta_{1}$ | $\beta_{1}$ | $\lambda$ |
| (0.7, 0.5) | (15, 20, 30) | 2.861 | 2.579 | 3.142 | 5.544 | 2.832 | 2.580 | 3.180 | 5.511 | 2.645 | 2.349 | 2.932 | 5.361 |
|  |  | 0.89 | 0.88 | 0.89 | 0.88 | 0.91 | 0.90 | 0.91 | 0.89 | 0.90 | 0.90 | 0.90 | 0.91 |
|  | $(15,25,30)$ | 2.835 | 2.559 | 3.103 | 5.528 | 2.810 | 2.561 | 3.098 | 5.488 | 2.670 | 2.342 | 2.915 | 5.357 |
|  |  | 0.90 | 0.91 | 0.90 | 0.91 | 0.91 | 0.91 | 0.91 | 0.92 | 0.91 | 0.92 | 0.91 | 0.92 |
|  | $(20,25,30)$ | 2.799 | 2.538 | 3.071 | 5.499 | 2.784 | 2.541 | 3.075 | 5.478 | 2.611 | 2.315 | 2.881 | 5.338 |
|  |  | 0.90 | 0.91 | 0.93 | 0.92 | 0.91 | 0.93 | 0.92 | 0.92 | 0.94 | 0.94 | 0.98 | 0.92 |
|  | (20, 30, 50) | 2.782 | 2.499 | 3.073 | 5.486 | 2.774 | 2.615 | 3.069 | 5.470 | 2.592 | 2.298 | 2.869 | 5.298 |
|  |  | 0.92 | 0.90 | 0.91 | 0.91 | 0.93 | 0.93 | 0.91 | 0.94 | 0.93 | 0.92 | 0.93 | 0.92 |
|  | $(20,40,50)$ | 2.771 | 2.492 | 3.059 | 5.480 | 2.764 | 2.598 | 3.052 | 5.441 | 2.563 | 2.277 | 2.842 | 5.285 |
|  |  | 0.90 | 0.93 | 0.92 | 0.92 | 0.93 | 0.94 | 0.94 | 0.93 | 0.93 | 0.92 | 0.94 | 0.95 |
|  | $(30,40,50)$ | 2.741 | 2.458 | 3.032 | 5.439 | 2.742 | 2.568 | 3.032 | 5.429 | 2.544 | 2.253 | 2.834 | 5.257 |
|  |  | 0.92 | 0.91 | 0.93 | 0.92 | 0.92 | 0.94 | 0.93 | 0.92 | 0.94 | 0.93 | 0.93 | 0.93 |

percentages, and average confidence interval lengths based on 1000 replications are reported. The results are shown in Tables 2-13.

## 6. Conclusions

A suitable censoring scheme and a suitable accelerated type for competing risk populations are developed. The results
are computed when the products have Chen lifetime distribution. The model parameters are estimated with MLEs and Bayes methods; then, the numerical results are reported in Tables 2-13. The following comments are observed for the numerical results.
(1) From all tables, the proposed methods serve well for all choices.
(2) The numerical results which are obtained from MLEs and Bayes with noninformative prior are approximately similar.
(3) The Bayes method under informative prior serves very well compared to the other cases.
(4) The censoring scheme of type-I GHCS has saved the minimum number of failures which present suitable numerical results of all choices
(5) For the effective ( $r, m$ ) sample increases, the MSEs and the AL of different estimators are reduced.
(6) The best results are obtained for the cases of $\eta \cong \tau$.

## Data Availability

No data were used to support the findings of this study.

## Conflicts of Interest

The authors have no conflicts of interest regarding the publication of the paper.

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# Kinematics Analysis of 6-DoF Articulated Robot with Spherical Wrist 

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#### Abstract

The aim of the paper is to study the kinematics of the manipulator. The articulated robot with a spherical wrist has been used for this purpose. The Comau NM45 Manipulator has been chosen for the kinematic model study. The manipulator contains six revolution joints. Pieper's approach has been employed to study the kinematics (inverse) of the robot manipulator. Using this approach, the inverse kinematic problem is divided into two small less complex problems. This reduces the time of analysing the manipulator kinematically. The forward and inverse kinematics has been performed, and mathematical solutions are detailed based on D-H (Denavit-Hartenberg) parameters. The kinematics solution has been verified by solving the manipulator's motion. It has been observed that the model is accurate as the motion trajectory was smoothly followed by the manipulator.


## 1. Introduction

Locomotion is the process of causing a rigid body to move. The body needs force to move. Dynamics is the study of the motion of the body in which forces are modelled which helps the body to move, whereas kinematics is the geometrical study of the motion of the body without considering the forces that can affect the motion of the body.

Kinematics is the motion description of the rigid body. [1] Links are the connectivity body/member between joints. The kinematic chain is a grouping of links connected by joints, as illustrated in Figure 1. In the kinematic chain, the number of DoF (degree of freedom) is equal to the number of joints.

Maintaining a strong connection between the two joints is called the kinematics function of a link. This connection can be described with the following factors:
(i) $a$ : link length
(ii) $\alpha$ : link twist

Link length is measured along the line which is mutually perpendicular to both joints/axes. The perpendicularity in joints always exists except when both joints are parallel. Link
twist is the angle of projection from the previous joint $(i-1)$ to the next joint (i) onto the axis $i-1$ (previous joint); the projection line is parallel to the next joint (axis i). The relationship between link length and twist is described in Figure 2.
(i) A joint axis is formed at the connection of two links. This joint will have two parameters (one for each link) connected to it. These parameters are as follows:
(ii) $d$ : distance between links
(iii) $\theta$ : angle between links

The relative position or distance between the links is called link offset. Figure 3 describes these parameters, in which the joint angle is the angle between the links.

The four parameters demonstrated above are associated with each link. Axes can be aligned using these parameters. The parameters are also known as Denavit-Hartenberg link parameters. These are illustrated in Table 1 below:

The link numbering convention follows from the base of the arm till the last moving link. As mentioned in Figure 4, the first link is the connection between the base and first joint.


Figure 1: Kinematics of the manipulator: (a) open chain; (b) closed chain [2].


Figure 2: Kinematic function of link.


Figure 3: Link and joint parameters. Joint parameters are highlighted in red.

Table 1: Denavit-Hartenberg link parameters.

| Link parameters | Joint parameters |
| :--- | :---: |
| $a:$ link length | $d:$ distance between links |
| $\alpha:$ link twist | $\Theta:$ angle between links |

The parameters mentioned above in Table 1 are used for kinematic modelling of the robot. In kinematics, modelling the geometry of the robot is represented. Homogenous transformation (of the matrix) is commonly used as the definition of the kinematics model (particularly for chains mechanism). As described below,


Figure 4: A 7-DoF manipulator arm [3].

$$
\begin{equation*}
{ }_{n}^{0} T=T_{1} T_{2} T_{3} \ldots T_{i} \ldots T_{n}, \tag{1}
\end{equation*}
$$

where $n$ is the number of links, $T_{i}$ is the link transformation from the $i^{\text {th }}$ joint, and ${ }_{n}^{0} T$ is the final pose for end-effector relative to the base.

There are two main types of kinematic models: forward kinematics and inverse kinematics. In forward kinematics, the length of each link and angle of each joint is given, and through that, position of any point $(x, y, z)$ can be found. In inverse kinematics, the length of each link and position of some points $(x, y, z)$ is given, and the angle of each joint is needed to find to obtain that position.

Several models are developed for kinematic modelling, but the D-H (Denavit-Hartenberg) model [4] is the most popular model. Limitations of the D-H model are discussed, and CPC (completeness and parametric continuity) model and its mapping with the D-H model were proposed [5]. The parametric continuity of the CPC model was achieved by using singularity free line representation.

## 2. Forward Kinematics of Comau NM45

The Comau NM45 [6] is a medium-scale robot. It has 6 degree of freedom joints. It is an articulated arm with a spherical wrist. The wrist joint intersects at one point. Figure 5 shows the manipulator with its link length and working envelope.

Forward kinematics is the study of the manipulator to find out its tip or end-effector position and orientation by using joint values of the manipulator. The first step of performing the forward kinematics is to label link lengths.


Figure 5: Comau NM45 (inline) operative area and link lengths.

This step has been performed in Figure 5. The second step to find the forward kinematics of the manipulator is to assign the frames. The frame assignment is done in Figure 6.

The D-H parameters can be found based on the frame assignment. The modified DH convention has been used for the frame assignment and DH parameters [7, 8]. These parameters are illustrated in Table 2.

The transformation matrix for a link $\mathbf{i}$ is described as follows:

$$
A_{i}=\left[\begin{array}{cccc}
\cos \theta i & -\sin \theta i \cos \alpha i & \sin \theta i \sin \alpha i & \alpha i \cos \theta i  \tag{2}\\
\sin \theta i & \cos \theta i \cos \alpha i & -\cos \theta i \sin \alpha i & \alpha i \sin \theta i \\
0 & \sin \alpha i & \cos \alpha i & \mathrm{~d} i \\
0 & 0 & 0 & 1
\end{array}\right]
$$

$A_{1}$ is the transformation matrix $T_{1}^{0}$ :

$$
T_{1}^{0}=\left[\begin{array}{cccc}
\cos \theta 1 & -\sin \theta 1 \cos \alpha 1 & \sin \theta 1 \sin \alpha 1 & \alpha 1 \cos \theta 1  \tag{3}\\
\sin \theta 1 & \cos \theta 1 \cos \alpha 1 & -\cos \theta 1 \sin \alpha 1 & \alpha 1 \sin \theta 1 \\
0 & \sin \alpha 1 & \cos \alpha 1 & d 1 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

$$
T_{1}^{0}=\left[\begin{array}{cccc}
\cos \theta 1 & 0 & \sin \theta 1 & \alpha 1 \cos \theta 1  \tag{4}\\
\sin \theta 1 & 0 & -\cos \theta 1 & \alpha 1 \sin \theta 1 \\
0 & 1 & 0 & d 1 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

whereas $a 1=l 1=0.4$ and $d 1=0.75$; by placing these values above, the following equation which is the resultant for the transformation between the base and joint 1 is obtained:

$$
T_{1}^{0}=\left[\begin{array}{cccc}
\cos \theta 1 & 0 & \sin \theta 1 & \alpha 1 \cos \theta 1  \tag{5}\\
\sin \theta 1 & 0 & -\cos \theta 1 & \alpha 1 \sin \theta 1 \\
0 & 1 & 0 & 0.75 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

For A2 $=T_{2}^{1}$,

$$
T_{2}^{1}=\left[\begin{array}{cccc}
\cos \theta 2 & -\sin \theta 2 & 0 & \alpha 2 \cos \theta 2  \tag{6}\\
\sin \theta 2 & \cos \theta 2 & 0 & \alpha 2 \sin \theta 2 \\
0 & 0 & 1 & d 2 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

and $\alpha=90$, so $T_{1}^{0}$ will be as follows:


Figure 6: NM45 model with coordinates mapping.

Table 2: Comau NM45 DH parameters.

| Link | Link length $(\alpha i), \mathrm{m}$ | Twist angle $(\alpha i)$, deg | Joint offset | Joint angle $(\theta i)$, deg |
| :--- | :---: | :---: | :---: | :---: |
| 1 | $L 1=0.4$ | 90 | $D 1=0.75$ | $(\theta 1)$ |
| 2 | $L 2=0.75$ | 0 | 0 | $(\theta 2)$ |
| 3 | $L 3=0.25$ | 90 | 0 | $(\theta 3)$ |
| 4 | 0 | -90 | $D 4=0.8124$ | $(\theta 4)$ |
| 5 | 0 | 90 | 0 | $(\theta 5)$ |
| 6 | 0 | 0 | $D 6=0.175$ | $(\theta 6)$ |

whereas $a 2=l 2=0.75$ and $d 2=0$; by placing these values above, the following equation which is the resultant for the transformation between joint 1 and joint 2 is obtained:

$$
T_{2}^{1}=\left[\begin{array}{cccc}
\cos \theta 2 & -\sin \theta 2 & 0 & 0.75 \cos \theta 2  \tag{7}\\
\sin \theta 2 & \cos \theta 2 & 0 & 0.75 \sin \theta 2 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

For $\mathrm{A} 3=T_{3}^{2}$,

$$
T_{3}^{2}=\left[\begin{array}{cccc}
\cos \theta 3 & 0 & \sin \theta 3 & \alpha 3 \cos \theta 3  \tag{8}\\
\sin \theta 3 & 0 & -\cos \theta 3 & \alpha 3 \sin \theta 3 \\
0 & 1 & 0 & d 3 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

whereas $a 3=l 3=0.25$ and $d 3=0$; by placing these values above, the following equation which is the resultant for the transformation between joint 2 and joint 3 is obtained:

$$
T_{3}^{2}=\left[\begin{array}{cccc}
\cos \theta 3 & 0 & \sin \theta 3 & 0.25 \cos \theta 3  \tag{9}\\
\sin \theta 3 & 0 & -\cos \theta 3 & 0.25 \sin \theta 3 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

For $\mathrm{A} 4=T_{4}^{3}$ and $\alpha=90$,

$$
T_{4}^{3}=\left[\begin{array}{cccc}
\cos \theta 4 & 0 & -\sin \theta 4 & \alpha 4 \cos \theta 4  \tag{10}\\
\sin \theta 4 & 0 & \cos \theta 4 & \alpha 4 \sin \theta 4 \\
0 & -1 & 0 & d 4 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

whereas $a 4=0$; by placing the value of $a 4$ in the above equation, the following equation which is the resultant for the transformation between joint 3 and joint 4 is obtained:

$$
T_{4}^{3}=\left[\begin{array}{cccc}
\cos \theta 4 & 0 & -\sin \theta 4 & 0  \tag{11}\\
\sin \theta 4 & 0 & \cos \theta 4 & 0 \\
0 & -1 & 0 & d 4 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

For $\mathrm{A} 4=T_{5}^{4}$,

$$
T_{5}^{4}=\left[\begin{array}{cccc}
\cos \theta 5 & 0 & \sin \theta 5 & \alpha 5 \cos \theta 5  \tag{12}\\
\sin \theta 5 & 0 & -\cos \theta 5 & \alpha 5 \sin \theta 5 \\
0 & 1 & 0 & d 5 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

whereas $a 5=0$ and $d 5=0$; by placing these values in the above equation, the following equation which is the resultant for the transformation between joint 4 and joint 5 is obtained:

$$
T_{5}^{4}=\left[\begin{array}{cccc}
\cos \theta 5 & 0 & \sin \theta 5 & 0  \tag{13}\\
\sin \theta 5 & 0 & -\cos \theta 5 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

For $\mathrm{A} 5=T_{6}^{5}$ and $\alpha=0$,

$$
T_{6}^{5}=\left[\begin{array}{cccc}
\cos \theta 6 & -\sin \theta 6 & 0 & \alpha 6 \cos \theta 6  \tag{14}\\
\sin \theta 6 & \cos \theta 6 & 0 & \alpha 6 \sin \theta 6 \\
0 & 0 & 1 & d 6 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

whereas $a 6=0$; by placing the value of $a 6$ in the above equation, the following equation which is the resultant for the transformation between joint 5 and joint 6 is obtained:

$$
T_{6}^{5}=\left[\begin{array}{cccc}
\cos \theta 6 & -\sin \theta 6 & 0 & 0  \tag{15}\\
\sin \theta 6 & \cos \theta 6 & 0 & 0 \\
0 & 0 & 1 & d 6 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

2.1. Transformation. For simplification, the following has been substituted:

$$
\begin{align*}
& \theta 1=u, \\
& \theta 2=v, \\
& \theta 3=w, \\
& \theta 4=a,  \tag{16}\\
& \theta 5=b, \\
& \theta 6=m .
\end{align*}
$$

For $A 12=T_{2}^{0}=T_{1}^{0} \times T_{2}^{1}$, substitute values from equations (5) and (7):

$$
T_{2}^{0}=T_{1}^{0} X T_{2}^{1}=\left(\begin{array}{cccc}
\cos (u) & 0 & \sin (u) & 0.4 \cos (u)  \tag{17}\\
\sin (u) & 0 & -\cos (u) & 0.4 \sin (u) \\
0 & 1 & 0 & 0.75 \\
0 & 0 & 0 & 1
\end{array}\right) \cdot\left(\begin{array}{cccc}
\cos (v) & -\sin (v) & 0 & 0.75 \cos (v) \\
\sin (v) & \cos (v) & 0 & 0.75 \sin (v) \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1
\end{array}\right)
$$

The resultant transformation between the base and joint 2 is illustrated as follows:

$$
T_{2}^{0}=\left[\begin{array}{cccc}
\cos (u) \cos (v) & -\cos (u) \sin (v) & \sin (u) & \cos (u)(0.75 \cos (v)+0.4)  \tag{18}\\
\cos (v) \sin (u) & -\sin (u) \sin (v) & -\cos (u) & (0.75 \cos (v)+0.4) \sin (u) \\
\sin (v) & \cos (v) & 0 & 0.75 \sin (v)+0.75 \\
0 & 0 & 0 & 1
\end{array}\right]
$$

For A123 $=T_{3}^{0}=T_{1}^{0} \times T_{2}^{1} \times T_{3}^{2}$, for simplicity,

$$
\begin{align*}
\mathrm{m} & =\cos (u)(0.75 \cos (v)+0.4), \\
n & =(0.75 \cos (v)+0.4) \sin (u), \\
u & =0.75 \sin (v)+0.75 \\
T_{3}^{0} & =T_{2}^{0} X T_{3}^{2}=\left(\begin{array}{cccc}
\cos (u) \cos (v) & -\cos (u) \sin (v) & \sin (u) & m \\
\cos (v) \sin (u) & -\sin (u) \sin (v) & -\cos (u) & n \\
\sin (v) & \cos (v) & 0 & o \\
0 & 0 & 0 & 1
\end{array}\right) \cdot\left(\begin{array}{cccc}
\cos (w) & 0 & \sin (w) & 0.25 \cos (w) \\
\sin (w) & 0 & -\cos (w) & 0.25 \sin (w) \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right) . \tag{19}
\end{align*}
$$

The resultant transformation between the base and joint 3 is illustrated as follows:

$$
T_{3}^{0}=\left(\begin{array}{cccc}
\cos (u) \cos (v+w) & \sin (u) & \cos (u) \sin (v+w) & m+0.25 \cos (u) \cos (v+w)  \tag{20}\\
\cos (v+w) \sin (u) & -\cos (u) & \sin (u) \sin (v+w) & n+0.25 \cos (v+w) \sin (u) \\
\sin (v+w) & 0 & -\cos (v+w) & 0+0.25 \sin (v+w) \\
0 & 0 & 0 & 1
\end{array}\right) .
$$

For A456 $=T_{6}^{3}=T_{4}^{3} \times T_{5}^{4} \times T_{6}^{5}$,

$$
\begin{align*}
& A 45=T_{5}^{3}=T_{4}^{3} \times T_{5}^{4}=\left(\begin{array}{cccc}
\cos (a) & 0 & -\sin (a) & 0 \\
\sin (a) & 0 & \cos (a) & 0 \\
0 & -1 & 0 & 0.8124 \\
0 & 0 & 0 & 1
\end{array}\right) \cdot\left(\begin{array}{cccc}
\cos (b) & 0 & \sin (b) & 0 \\
\sin (b) & 0 & -\cos (b) & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right) \text {, } \\
& A 45=T_{5}^{3}=\left(\begin{array}{cccc}
\cos (a) \cos (b) & -\sin (a) & \cos (a) \sin (b) & 0 \\
\sin (a) \cos (b) & \cos (a) & \sin (a) \sin (b) & 0 \\
-\sin (b) & 0 & \cos (b) & 0.8124 \\
0 & 0 & 0 & 1
\end{array}\right) \text {, }  \tag{21}\\
& T_{6}^{3}=T_{4}^{3} \times T_{5}^{4} \times T_{6}^{5}=T_{5}^{3} \times T_{6}^{5}=\left(\begin{array}{cccc}
\cos (a) \cos (b) & -\sin (a) & \cos (a) \sin (b) & 0 \\
\sin (a) \cos (b) & \cos (a) & \sin (a) \sin (b) & 0 \\
-\sin (b) & 0 & \cos (b) & d^{4} \\
0 & 0 & 0 & 1
\end{array}\right) \cdot\left(\begin{array}{cccc}
\cos (m) & -\sin (m) & 0 & 0 \\
\sin (m) & \cos (m) & 0 & 0 \\
0 & 0 & 1 & d^{6} \\
0 & 0 & 0 & 1
\end{array}\right) \text {. }
\end{align*}
$$

The resultant transformation between joint 3 and joint 6 is illustrated as follows:

$$
T_{6}^{3}=\left(\begin{array}{cccc}
\cos (a) \cos (b) \cos (m)-\sin (a) \sin (m) & -\cos (m) \sin (a)-\cos (a) \cos (b) \sin (m) & \cos (a) \sin (b) & d^{6} \cos (a) \sin (b)  \tag{22}\\
\cos (b) \cos (m) \sin (a)+\cos (a) \sin (m) & \cos (a) \cos (m)-\cos (b) \sin (a) \sin (m) & \sin (a) \sin (b) & d^{6} \sin (a) \sin (b) \\
-\cos (m) \sin (b) & \sin (b) \sin (m) & \cos (b) & \cos (b) d^{6}+d^{4} \\
0 & 0 & 0 & 1
\end{array}\right) .
$$

Spherical wrist position can be extracted using the last column of $T_{6}^{0}$ transformation matrix, where

$$
T_{6}^{0}=\left[\begin{array}{cccc}
r_{11} & r_{12} & r_{13} & P_{x}  \tag{23}\\
r_{21} & r_{22} & r_{23} & P_{y} \\
r_{31} & r_{32} & r_{33} & P_{z} \\
0 & 0 & 0 & 1
\end{array}\right]
$$

Which means the $T_{6}^{0}$ the transformation matrix can be represented in terms of $R_{6}^{0}$ and $P_{6}^{0}$ illustrated as follows:

$$
T_{6}^{0}=\left[\begin{array}{cc}
R_{6}^{0} & P_{6}^{0}  \tag{24}\\
0 & 1
\end{array}\right]
$$

and spherical wrist position is illustrated as

$$
P_{6}^{0}=\left[\begin{array}{c}
P x  \tag{25}\\
P y \\
P z
\end{array}\right]
$$

Transformation matrix from base to end-effector is

$$
\begin{align*}
& T_{6}^{0}=T_{1}^{0} \times T_{2}^{1} \times T_{3}^{2} \times T_{4}^{3} \times T_{5}^{4} \times T_{6}^{5}, \\
& T_{6}^{0}=\left[\begin{array}{cccc}
r 11 & r 12 & r 13 & P x \\
r 21 & r 22 & r 23 & P y \\
r 31 & r 32 & r 33 & P z \\
0 & 0 & 0 & 1
\end{array}\right],  \tag{26}\\
& T_{6}^{0}=\left[\begin{array}{cc}
R_{6}^{0} & P_{6}^{0} \\
0 & 1
\end{array}\right] .
\end{align*}
$$

## 3. Inverse Kinematics of Comau NM45

Inverse kinematics is finding the joint values $(\theta 1, \theta 2, \theta 3, \theta 4, \theta 5, \theta 6)$ of the robot arm for the given position ( $p$ ) and orientation ( $o$ ). For inverse kinematics, the inverse orientation $R$ and inverse position $P$ are needed.

The Comau NM45 is an articulated arm with a spherical wrist. For finding anthropomorphic/articulated arm position and joint values $\theta 1, \theta 2, \theta 3$, the inverse position is needed.

Let $O_{c}$ be intersecting the last 3 joints. The motion of joints 4,5 , and 6 will not change the position of $O_{c}$, as stated in Figure 7 [19] This is according to Pieper's approach [7], in which the manipulator is divided to analyse the inverse kinematics [10].

According to $T_{6}^{3}$ matrix from equation (22), the position $P_{6}^{3}$ is always as mentioned in the following equation:

$$
P_{6}^{3}=\left[\begin{array}{c}
0  \tag{27}\\
0 \\
d_{\partial 6}
\end{array}\right]
$$

So, the position $O$ will be as follows:

$$
O=O_{c}^{0}+d_{6} R\left[\begin{array}{l}
0  \tag{28}\\
0 \\
1
\end{array}\right]
$$

where $O$ is the position $P_{6}^{0}$ and R is the orientation which is $R$. The above equation can be written in terms of $O_{c}^{0}$ as follows:

$$
O_{c}^{0}=O-d_{6} R\left[\begin{array}{l}
0  \tag{29}\\
0 \\
1
\end{array}\right]
$$

The first three joints can be found in the following steps. They will determine the position of the manipulator:

$$
\begin{gathered}
O=\left[\begin{array}{c}
o_{x} \\
o_{y} \\
o_{z}
\end{array}\right], \\
O_{c}^{0}=\left[\begin{array}{l}
x_{c} \\
y_{c} \\
z_{c}
\end{array}\right], \\
{\left[\begin{array}{l}
x_{c} \\
y_{c} \\
z_{c}
\end{array}\right]=\left[\begin{array}{l}
o_{x}-d_{6} r_{13} \\
o_{x}-d_{6} r_{23} \\
o_{x}-d_{6} r_{33}
\end{array}\right] .}
\end{gathered}
$$



Figure 7: Kinematics decoupling (Spong, Hutchinson, and Vidyasagar, 2006).

For the orientation, the last three joints orientation is needed. The following equation shows the overall rotation of the manipulator in terms of $R_{3}^{0}$ and $R_{6}^{3}$ :

$$
\begin{equation*}
R=R_{3}^{0} R_{6}^{3} \tag{31}
\end{equation*}
$$

Rearrangement of equation (31) will yield $R_{6}^{3}$ below, through which the last three joint angles can be found:

$$
\begin{equation*}
R_{6}^{3}=R_{3}^{0-1} R \longrightarrow \theta 4, \theta 5, \theta 6 \tag{32}
\end{equation*}
$$

For articulated manipulator, the first three joints tell the position, as illustrated in Figure 8.

Projection of wrist onto $x_{0}, y_{0}$ plane has been shown in Figure 9.

This projection yields the triangle through which the angle value for $\theta 1$ can be found as follows:

$$
\begin{equation*}
\theta 1=\mathrm{A} \tan 2\left(x_{c}, y_{c}\right) \tag{33}
\end{equation*}
$$

If the wrist is rotated, then it will result in the following equation:

$$
\begin{equation*}
\theta 1=A \tan 2\left(x_{c}, y_{c}\right)+\pi \tag{34}
\end{equation*}
$$

Another projection, as mentioned in Figure 10, on the plane formed with link 2 and link 3 can help to find the value of joint angle $2\left(\theta_{2}, 2\right)$ and joint angle $3\left(\theta_{3}, 3\right)$.

Law of cosines can be applied to obtain the joint angle 3 $\left(\theta_{3}, 3\right)$, as follows:

$$
\begin{equation*}
\cos \theta_{3}=\frac{r^{2}+s^{2}-a_{2}^{2}-a_{3}^{2}}{2 a_{2} a_{3}} \tag{35}
\end{equation*}
$$

By substituting the values of $r$ and $s$, these can be extracted by using Figure 8:

$$
\begin{equation*}
\cos \theta_{3}=\frac{\left(x_{c}^{2}+y_{c}^{2}-d^{2}\right)+\left(z_{c}-d_{1}\right)^{2}-a_{2}^{2}-a_{3}^{2}}{2 a_{2} a_{3}} \tag{36}
\end{equation*}
$$

As NM45 2.0 is inline, no shoulder is offset and, hence, $d=0$.


Figure 8: Comau NM45 (articulated arm) first three joints (Spong, Hutchinson, and Vidyasagar, 2006).


Figure 9: Projection of wrist centre onto x 0 , y 0 plane (Spong, Hutchinson, and Vidyasagar, 2006).


Figure 10: Projection on the plane (Link2 12, Link3 13) (Spong, Hutchinson, and Vidyasagar, 2006).

So,

$$
\begin{equation*}
\cos \theta_{3}=\frac{x_{c}^{2}+y_{c}^{2}+\left(z_{c}-d_{1}\right)^{2}-a_{2}^{2}-a_{3}^{2}}{2 a_{2} a_{3}} \tag{37}
\end{equation*}
$$

where $a_{2}=L_{2}$ according to DH labelled figure, and

$$
\begin{align*}
& \cos \theta_{3}=\frac{x_{c}^{2}+y_{c}^{2}+\left(z_{c}-d_{1}\right)^{2}-L_{2}^{2}-a_{3}^{2}}{2 L_{2} a_{3}},  \tag{38}\\
& \sin \theta_{3}=\sqrt{1-\cos \theta_{3}^{2}} .
\end{align*}
$$

Also,

$$
\begin{equation*}
\tan \theta_{3}=\sqrt{\frac{1-\cos \theta_{3}^{2}}{\cos \theta_{3}}}=\frac{\sqrt{1-\mathrm{D}^{2}}}{\mathrm{D}} \tag{39}
\end{equation*}
$$

The value of $\theta_{3}$ can be written in terms of atan 2 and $D$ as follows:

$$
\begin{equation*}
\theta_{3}=\operatorname{atan} 2\left(D, \pm \sqrt{1-\mathrm{D}^{2}}\right) \tag{40}
\end{equation*}
$$

where + is for elbow up and - is for elbow down.
The projection in Figure 11 has been drawn onto the link 2 and link 3 plane to find $\theta_{2}$ :

$$
\begin{align*}
& \theta_{2}=h 1-h 2 \\
& h 1=\operatorname{atan} 2(r, s)=\operatorname{atan} 2\left(\sqrt{x_{c}^{2}+y_{c}^{2}}, Z_{c}-d_{1}\right)  \tag{41}\\
& h_{2}=\operatorname{atan} 2\left(a_{2}+a_{3} \cos \theta_{3}, a_{3} \sin \theta_{3}\right)
\end{align*}
$$

So, the value of $\theta_{2}$ can be written as follows:

$$
\begin{align*}
\theta_{2}= & \operatorname{atan} 2\left(\sqrt{x_{c}^{2}+y_{c}^{2}, z_{c}-d_{1}}\right)  \tag{42}\\
& -\operatorname{atan} 2\left(a_{2}+a_{3} \cos \theta 3, a 3 \sin \theta_{3}\right)
\end{align*}
$$

3.1. Finding Spherical Wrist Joint Values (Rotation) Using Euler Angles. The last three joint variables, $\theta 4, \theta 5, \theta 6$, are the Euler angles. So,

$$
\begin{align*}
& \theta 4=\varnothing \\
& \theta 5=\theta  \tag{43}\\
& \theta 6=\varphi
\end{align*}
$$

These angles are concerning the coordinate frame $o 3, x 3, y 3, z 3$. Now we need to calculate the transformation from $T_{4}^{3}$ to $T_{6}^{5}$ which is $T_{6}^{3}$ :

$$
\begin{equation*}
T_{6}^{3}=\mathrm{A} 5 \mathrm{XA} 5 \mathrm{XA} 6=T_{4}^{3} \times T_{5}^{4} \mathrm{XT}_{6}^{5} \tag{44}
\end{equation*}
$$

where $T_{6}^{3}$ can be written in terms of $R_{6}^{3}$ and $O_{6}^{3}$ mentioned as follows:

$$
T_{6}^{3}=\left[\begin{array}{cc}
R_{6}^{3} & O_{6}^{3}  \tag{45}\\
0 & 1
\end{array}\right]
$$

where $O_{6}^{3}$ is the position.


Figure 11: Projection onto 12 and 13 plane to find $\theta_{2} . \theta_{2}$ can be written in terms of angles $h 1$ and $h 2$.

For the spherical wrist, the ZYZ Euler transformation is needed. $\mathbf{R}_{6}^{3}$ is captured from equation (22), so $\mathbf{R}_{6}^{3}$ can be written as

$$
\left.\begin{array}{rl}
R_{\mathrm{ZYZ}} & =R_{6}^{3}=\left[\begin{array}{cc}
\cos (\varnothing) \cos (\theta) \cos (\varphi)-\sin (\varnothing) \sin (\varphi) & -\cos (\varphi) \sin (\varnothing)-\cos (\varnothing) \cos (\theta) \sin (\varphi) \\
\cos (\theta) \cos (\varnothing) \sin (\theta) \\
-\cos (\varphi) \sin (\theta) & \cos (\varnothing) \sin (\varphi) \\
\cos (\varnothing) \cos (\varphi)-\cos (\theta) \sin (\varnothing) \sin (\varphi) & \sin (\varnothing) \sin (\theta) \\
R_{6}^{3} & =\left[\begin{array}{lll}
r 11 & r 12 & r 13 \\
r 21 & r 22 & r 23 \\
r 31 & r 32 & r 33
\end{array}\right],
\end{array}\right. \\
\text { cos }(\theta)
\end{array}\right],
$$

Whereas angles $a, b, m$ are $\varnothing, \theta, \varphi$, respectively.
The $\cos \theta$ can be extracted from $r 33$ from the following equation:

$$
\begin{align*}
& \cos (b)=\cos \theta \\
& \sin (b)=\sin \theta \sqrt{1-\cos ^{2} \theta} \tag{47}
\end{align*}
$$

The above $\cos \theta$ and $\sin \theta$ can be written in form of $\tan$ to find the value of $\theta$ :

$$
\begin{align*}
& \theta=\tan ^{-1}\left(\frac{\sqrt{1-\cos ^{2} \theta}}{r 33}\right) \\
& \theta=\tan ^{-1}\left(\frac{\sqrt{1-r 33^{2}}}{r 33}\right) \tag{48}
\end{align*}
$$

For nonsingular case, if both $r 13$ and $r 23$ are not zero, then equation (49) will yield the value of $\theta$ :

$$
\begin{equation*}
\theta=a \tan 2\left(r 33, \pm \sqrt{1-r 33^{2}}\right) \tag{49}
\end{equation*}
$$

or

$$
\begin{align*}
\tan \theta & =\frac{\sqrt{r 13^{2}+r 23^{2}}}{r 33}=\frac{\sqrt{(\cos (\varnothing) \sin (\theta))^{2}+(\sin (\varnothing) \sin (\theta))^{2}}}{\cos (\theta)} \\
\theta & =\tan ^{-1}\left(\frac{\sqrt{r 13^{2}+r 23^{2}}}{r 33}\right) \tag{50}
\end{align*}
$$

Equation (51) shows the value of $\theta$ in terms of $r 13, r 23$ and $r 33$ :

$$
\begin{equation*}
\theta=a \tan 2\left(r 33, \pm \sqrt{r 13^{2}+r 23^{2}}\right) \tag{51}
\end{equation*}
$$

The last column of $R_{6}^{3}$ from equation (46) can help to yield the value of $\varnothing$ :

$$
\begin{equation*}
\tan \varnothing=\frac{r 23}{r 13}=\frac{\sin (\varnothing) \sin (\theta)}{\cos (\varnothing) \sin (\theta)} \tag{52}
\end{equation*}
$$

$\varnothing$ can be represented in terms of $r 13$ and $r 23$, as follows:


Figure 12: The trajectory of Comau NM45.


Figure 13: Joint trajectory of Comau NM45.

$$
\begin{equation*}
\varnothing=a \tan 2(r 13, r 23) \tag{53}
\end{equation*}
$$

For $\varphi$,

$$
\begin{equation*}
\tan \varphi=\frac{r 32}{-r 21}=\frac{\sin (\theta) \sin (\varphi)}{-(-\cos (\varphi) \sin (\theta))} \tag{54}
\end{equation*}
$$

The $\varphi$ can be represented in terms of $r 31$ and $r 32$ mentioned as follows:

$$
\begin{equation*}
\varphi=a \tan 2(-r 31, r 32) \tag{55}
\end{equation*}
$$

For catering singularity, if $\sin (\theta)>0$, whereas $0<\theta<\pi$, then solution will be

$$
\begin{align*}
& \varnothing=a \tan 2(-r 13,-r 23) \\
& \theta=a \tan 2\left(r 33, \pm \sqrt{r 13^{2}+r 23^{2}}\right)  \tag{56}\\
& \varphi=a \tan 2(r 31,-r 32)
\end{align*}
$$

If $\sin (\theta)<0$, whereas $-\pi<\theta<0$, then the solution will be

$$
\begin{align*}
& \varnothing=a \tan 2(-r 13,-r 23), \\
& \theta=a \tan 2\left(r 33, \pm \sqrt{r 13^{2}+r 23^{2}}\right),  \tag{57}\\
& \varphi=a \tan 2(r 31,-r 32) .
\end{align*}
$$

The last three joint values have been resolved by using the above equations of $\varnothing, \theta$, and $\varphi$.

For the trajectory planning, the program was written in Matlab by using forward and inverse kinematics equations mentioned in the above sections. The planned trajectory required the smooth motion of the end-effector. The joint angles were calculated using the inverse kinematics equations. Figure 12 shows the trajectory planning and robot motion along with the trajectory for the Comau NM45. The robot was able to follow the trajectory smoothly.

Figure 13 shows the mapping of the values of joints to move on 100 points to follow the trajectory. It shows that the achievement was smooth as there are no sudden spikes in the joint values.

## 4. Conclusion

The modified DH convention has been used to perform the forward kinematics for the manipulator. The kinematics decoupling has been used to perform the inverse kinematics. The manipulator was divided into two parts to make the inverse kinematics problem simpler. The first 3 joints were resolved by using a geometrical approach, whereas the last three joints were resolved using the algebraic approach. The resultant kinematics solution was applied on the manipulator, and it was able to follow a test trajectory successfully. A similar approach can be used while solving the articulated robot with a spherical wrist. The techniques are applied for the 6 -DoF robot.

## Data Availability

All the data used are presented as part of paper.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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# A Dynamic Intersecting Arrangement Model Based on Isolated Draw Zones for Stope Structure Optimization during Sublevel Caving Mining 

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#### Abstract

In this paper, stope structure optimization during sublevel caving mining is considered under the condition that the isolated draw zones (IDZs) are nonstandard ellipsoid, which is realized by dynamically adjusting the arrangement of IDZs and quantifying the degree of intersection of IDZs according to an ore profit and loss calculation model. A dynamic intersecting arrangement model based on IDZs was proposed, which can dynamically adjust the sublevel height and drift spacing according to the ore-rock bulk flow parameters, economic indicators, occurrence condition of the ore body, drilling machine, and so forth. Based on the model, the range of drift spacing, the lower volume of crestal residual ore, and the higher volume of mixing waste rock are calculated. By deducing the function of ore profit and loss, a calculation model for ore profit and loss is established to quantify the degree of intersection of IDZs and determine the best stope structure. Using the constructed dynamic intersecting arrangement model, a stope structure of -213 m to -303 m in the Yanqianshan Iron Mine was designed, with a sublevel height of 22.5 m and a drift spacing of 20.5 m . A physical drawing model was designed, and three physical simulation experiment schemes were conducted to compare and analyse the ore loss and dilution of the intersecting arrangement model and the traditional tangent arrangement model. The results showed that the loss rate decreased by $3.66 \%$ and the dilution rate increased by only $0.22 \%$, thus verifying the effectiveness and applicability of the model to optimize the stope structure.


## 1. Introduction

The sublevel caving method is one of the oldest methods for underground mining. This method has many advantages with regard to safety and mechanization, so it is widely used in various kinds of metallic mines worldwide [1, 2]. With the extension of deep underground mining and the transformation of open-pit mines to underground mines, the application of this mining method is continuously growing, developing towards intelligent mining equipment and stope structure optimization [3]. The stope structure, including sublevel height, drift spacing, and caving spacing, is the main factor that causes ore loss and dilution. An optimal stope structure can adapt to the fluidity of ore-rock, thus reducing ore loss and dilution.

The sublevel caving method is used for mining between sublevels from top to bottom. When ore is extracted, overlying waste rock fills the void created by ore extraction, so the ore is always surrounded by the overlying waste rock. The caved ore is drawn from drifts under loose cover-layerrock, so ore loss and dilution readily occur [4-6]. This is a significant disadvantage of this method. Obviously, the optimal stope structure from a scientific drawing model can accelerate the flow of ore and control the mixing of waste rock, thus mitigating the shortcomings of serious ore loss and dilution [5, 7]. Therefore, relevant studies that can help improve the recovery and utilization of mineral resources and increase the economic benefit of mines are very significant.

To optimize the stope structure to reduce ore loss and dilution in sublevel caving mines, scholars have proposed different basic theories; those theories with practical value include ellipsoid draw theory and stochastic medium draw theory $[8,9]$. These theories study the movement trends of caved ore and overlying waste rock but adopt different research methods [10]. Ellipsoid draw theory experimentally proves that isolated draw zones (IDZs) (the geometrical locus defined by the initial location of the caved ore that is recovered from a drawpoint in underground mining) are approximately ellipsoids [11]. Then, the properties of IDZs are studied and used to explain the flow characteristics of the ore-rock bulk. Stochastic medium draw theory simplifies the ore-rock bulk into a continuous flowing random medium and then studies the movement process of the ore-rock bulk through probability theory [12, 13]. In the final analysis, these theories are ultimately used to optimize caving mining schemes and to determine reasonable stope structures. Based on ellipsoid draw theory, Kvapil [14] expanded from a single IDZ and used IDZ spatial arrangement theory to design stope structures. This design principle later evolved into two stope structures: wide drift spacing structure and high sublevel structure. In wide drift spacing structure, the upper and lower ellipsoids are directly tangent, while the left and right discharge ellipsoids are separated and tangent to the upper and lower ellipsoids (see Figure 1(a)). Similarly, in high sublevel structure, the left and right ellipsoids are directly tangent to each other, while the upper and lower discharge ellipsoids are separated and tangent to both the left and right ellipsoids (see Figure 1(b)). According to the design principle, the sublevel height of caving mines worldwide is 15 m to 30 m , and the drift spacing is generally slightly less than or equal to the sublevel height. However, under these stope structures, a certain proportion of ore is outside the IDZ (see Figure 1); therefore, this cannot be regarded as the optimal stope structure, and it has been proven that these stope structures do not yield a good recovery effect in practical mining applications. Some mines, such as the Xiadian Gold Mine in China and the Kamoto Copper Mine in Zambia, have recoveries of less than $80 \%$ [15]. Yuan et al. [16] analysed the coincident relationship between the caved ore, crestal residual ore, and IDZ and concluded that, within a certain temporal period and spatial area, increasing the distance between the drawpoints reduces the recovery rate of ore. Wang et al. [17] numerically simulated drift spacings of 12 m to 20 m , and the results showed that there were no obvious changes in the waste in-ore rate as the drift spacing increased, but the ore recovery rate decreased from $63 \%$ to $46 \%$ with increasing drift spacing. Tan et al. [18] discussed the wide drift spacing structure based on the ellipsoid drawing theory and pointed out the defects of the design scheme. The above studies are all based on qualitative descriptions of stope structure and analyse the influence of drift spacing on ore recovery. The specific stope structure is still determined according to mining experience and experiments. Moreover, for the whole mine, the stope structure is fixed and cannot be dynamically adjusted.


Figure 1: The conventional tangent arrangement model based on IDZs [14]. (a) Wide drift spacing structure. (b) High sublevel structure.

When mining by the sublevel caving method, the shape of the IDZ is affected by many factors, including not only the physical properties of particles, such as the ore-rock bulk geometry, lumpiness, concrete gradation, humidity, and loose factor, but also the mechanical properties of particles, such as the lateral pressure and friction [19]. Therefore, the shape of the IDZ is not a regular ellipsoid. Marano et al. [ 9,20 ] believe that the IDZ is composed of different shapes, such as an ellipsoid in the upper part and a parabolic rotating body in the lower part. Janelid and Kvapil [3, 21, 22] proved that the ellipsoid is an approximate ellipsoid, but there are cases in which the upper part is thick and the lower half is thin and vice versa. The nonstandard ellipsoid shape of the IDZ means that the tangent arrangement model cannot be completely applied to the design of the stope structure.

In this study, stope structure optimization during sublevel caving mining is considered under the condition that the IDZs are nonstandard ellipsoids, which is realized by dynamically adjusting the arrangement of IDZs and quantifying the degree of intersection of IDZs according to an ore profit and loss calculation model. A dynamic intersecting arrangement model based on IDZs which can dynamically adjust the stope structure according to the ore-rock bulk flow parameters, economic indicators, occurrence conditions of ore bodies, drilling machines, and so forth is proposed. Then, taking the ore profit and loss as the research object, a calculation model is established to quantify the degree of intersection of the IDZs and determine the best sublevel height and drift spacing. Finally, with the dynamic intersecting arrangement model, the optimal stope structure of the Yanqianshan Iron Mine is designed; moreover, three experimental schemes using the Yanqianshan Iron Mine as the background were carried out to compare the tangent arrangement model based on IDZs and the intersecting arrangement model based on IDZs.

## 2. The Intersecting Arrangement Model Based on IDZs

The design and optimization of the stope structure generally follow the principle that the shape of the caved ore heap is
consistent with that of the IDZ [14], which means that the IDZ should contain as much caved ore and residual ore as possible. According to the conclusion of Yuan and Liu [16, 23], if the distance between drawpoints is reduced so that the IDZs intersect, the crestal residual ore will be reduced (see Figure 2), which has a significant impact on the ore recovery rate. Therefore, the intersecting arrangement model based on IDZs can better reflect the principle that the shape of the caved ore heap is consistent with that of the IDZ, especially when the IDZ is a nonstandard ellipsoid.

Stochastic medium draw theory of Ren [12] simplifies the ore-rock bulk into a continuous flowing random medium, which better reflects the shape of the IDZ. Based on stochastic medium draw theory, for the end ore drawing of the sublevel caving method, the shape of IDZ is given by

$$
\begin{equation*}
\frac{\left(x-k z^{(\alpha / 2)}\right)^{2}}{\beta z^{\alpha}}+\frac{y^{2}}{\beta_{1} z^{\alpha_{1}}}=\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \ln \frac{2 H}{z}[12] \tag{1}
\end{equation*}
$$

where $\alpha, \beta, \alpha_{1}$, and $\beta_{1}$ are the ore-rock bulk flow parameters ( $\alpha$ and $\beta$ represent the vertical direction of the drift, and $\alpha_{1}$ and $\beta_{1}$ represent the horizontal direction of the drift); $k$ is the impact coefficient of the ore body and depends on the damping degree of the ore-rock bulk by the ore body (in general, $k=0.1 \sim 0.15$ ); and $H$ is the sublevel height. Equation (1) is mainly used to determine the width of the IDZ in the vertical direction of the drift to design the drift spacing.

When mining by the sublevel caving method, the shape of the IDZ is not a regular ellipsoid, and the tangent arrangement model based on IDZs will reduce the density of the spatial arrangement, causing more ore to be outside the range of the IDZ. Under these conditions, it is more appropriate to adopt the intersecting arrangement model based on IDZs. The drift spacing is shortened so that IDZs $R_{2}$ and $R_{3}$ intersect with the upper and lower IDZs $R_{1}$ and $R_{3}$, respectively, and then the crestal residual ore will be reduced (see Figure 3). Of course, this shortening will promote the mixing of waste rock, but this problem can be addressed by calculating the comprehensive economic benefits of the ore. The best intersection degree of the IDZ corresponds to the maximum economic benefit.

## 3. Optimization of the Stope Structure

The sublevel height is determined by occurrence conditions of the ore body, drilling machine, demolition equipment, and so forth. Once the sublevel height is determined, it is difficult to change. The drift spacing is a quantitative representation of the intersecting arrangement model based on IDZs, and the degree of intersection determines the drift spacing. In this section, on the basis of establishing the intersecting arrangement model based on IDZs, the method for determining the optimal stope structure is given.
3.1. Range of the Drift Spacing. In this section, equations of the four adjacent IDZs are established at the same coordinate according to the shape of the IDZ (equation (1)). On this basis, we derived the calculation method of the width of the IDZ in the vertical direction of the drift and provided the


Figure 2: Stope structure of the intersecting arrangement model based on IDZs.


Figure 3: The intersecting arrangement model based on IDZs. (a) $\alpha$ $<1.44$. (b) $\alpha>1.44$.
determination method for the drift spacing by using the arrangement of the IDZs.

From equation (1) and Figure 3, the equation of the IDZ $R_{1}$ is as follows:

$$
\begin{equation*}
R_{1}: \frac{\left(x-k z^{(\alpha / 2)}\right)^{2}}{\beta z^{\alpha}}+\frac{y^{2}}{\beta_{1} z^{\alpha_{1}}}=\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \ln \frac{2 H}{z}[12] . \tag{2}
\end{equation*}
$$

By transforming the coordinates of the centre point of IDZ $R_{1}$, the equations of $R_{2}, R_{3}$, and $R_{4}$ are as follows:

$$
\begin{equation*}
R_{2}: \frac{\left(x-k(z-H)^{(\alpha / 2)}\right)^{2}}{\beta(z-H)^{\alpha}}+\frac{(y-B)^{2}}{\beta_{1}(z-H)^{\alpha_{1}}}=\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \ln \frac{2 H}{z-H}, \tag{3}
\end{equation*}
$$

$$
\begin{equation*}
R_{3}: \frac{\left(x-k(z-2 H)^{(\alpha / 2)}\right)^{2}}{\beta(z-2 H)^{\alpha}}+\frac{y^{2}}{\beta_{1}(z-2 H)^{\alpha_{1}}}=\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \ln \frac{2 H}{z-2 H}, \tag{4}
\end{equation*}
$$

$$
\begin{equation*}
R_{4}: \frac{\left(x-k(z-H)^{(\alpha / 2)}\right)^{2}}{\beta(z-H)^{\alpha}}+\frac{(y+B)^{2}}{\beta_{1}(z-H)^{\alpha_{1}}}=\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \ln \frac{2 H}{z-H}, \tag{5}
\end{equation*}
$$

where $B$ is the drift spacing, which refers to the distance between two drawpoints; $R_{1}$ is the lower IDZ; $R_{2}$ is the right IDZ; $R_{3}$ is the upper IDZ; and $R_{4}$ is the left IDZ.

For $R_{1}$, along the direction of $\left.x=k z \hat{( } \alpha / 2\right)$, which is the vertical direction of the drift, the profile equation of $R_{1}$ is as follows:

$$
\begin{equation*}
y^{2}=\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1} z^{\alpha_{1}} \ln \frac{2 H}{z} . \tag{6}
\end{equation*}
$$

For $R_{2}$, along the direction of $\left.x=k(z-H) \hat{( } \alpha / 2\right)$, which is the vertical direction of the drift, the profile equation of $R_{2}$ is as follows:

$$
\begin{equation*}
(y-B)^{2}=\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1}(z-H)^{\alpha_{1}} \ln \frac{2 H}{(z-H)} \tag{7}
\end{equation*}
$$

The height of the widest part of $R_{1}$ is denoted by $h$, and the normal slope of $R_{1}$ at $z=h$ is zero; for equation (6), $\mathrm{d} y / \mathrm{d} z=0$. Then,

$$
\begin{equation*}
z=h=2 H e^{-\left(1 / \alpha_{1}\right)} . \tag{8}
\end{equation*}
$$

Next, the maximum and minimum values of the drift spacing are calculated for different IDZ shapes.
(1) $\alpha<1.44$

The widest part of the IDZ is located at the lower part, and the IDZ is similar to an ellipsoid with a thin upper part and thick lower part (see Figure 3(a)).
For $R_{2}$, by equations (7) and (8), the widest part is located at $z$, where

$$
\begin{equation*}
z=2 H e^{-\left(1 / \alpha_{1}\right)}+H . \tag{9}
\end{equation*}
$$

Point $G_{1}$ is defined as follows:

$$
\begin{equation*}
G_{1}\left(2 H e^{-\left(1 / \alpha_{1}\right)}+H,-\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}}+B\right) . \tag{10}
\end{equation*}
$$

For $R_{1}$, by equations (6), (8), and (9), point $F_{1}$ is defined as follows:

$$
\begin{equation*}
F_{1}\left(2 H e^{-\left(1 / \alpha_{1}\right)}+H, \sqrt{\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1} H^{\alpha_{1}}\left(1+2 e^{-\left(1 / \alpha_{1}\right)}\right)^{\alpha_{1}} \ln \frac{2}{1+2 e^{-\left(1 / \alpha_{1}\right)}}}\right) . \tag{11}
\end{equation*}
$$

When $R_{2}$ is tangent to $R_{1}$ and $R_{3}$, the drift spacing is maximized, and points $G_{1}$ and $F_{1}$ coincide; therefore,

$$
\begin{equation*}
B \max =\sqrt{\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1} H^{\alpha_{1}}\left(1+2 e^{-\left(1 / \alpha_{1}\right)}\right)^{\alpha_{1}} \ln \frac{2}{1+2 e^{-\left(1 / \alpha_{1}\right)}}}+\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}} . \tag{12}
\end{equation*}
$$

When $R_{2}$ is tangent to the $z$-axis, the drift spacing is minimized; therefore,

$$
\begin{equation*}
B \min =\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}}, \tag{13}
\end{equation*}
$$

where $B_{\max }$ is the maximum drift spacing and $B_{\text {min }}$ is the minimum drift spacing.
(2) $\alpha>1.44$

The widest part of the IDZ is located in the upper part, and the IDZ is similar to an ellipsoid with a thick upper part and thin lower part (see Figure 3(b)).

For $R_{1}$, according to equations (6) and (8), the widest part is located at $z$, and point $F_{2}$ is defined as follows:

$$
\begin{equation*}
F_{2}\left(2 H e^{-\left(1 / \alpha_{1}\right)}, \sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}}\right) \tag{14}
\end{equation*}
$$

For $R_{2}$, according to equations (7) and (8), point $G_{2}$ is defined as follows:

$$
\begin{equation*}
G_{2}\left(2 H e^{-\left(1 / \alpha_{1}\right)},-\sqrt{\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1} H^{\alpha_{1}}\left(2 e^{-\left(1 / \alpha_{1}\right)}-1\right)^{\alpha_{1}} \ln \frac{2}{2 e^{-\left(1 / \alpha_{1}\right)}-1}}+B\right) \tag{15}
\end{equation*}
$$

When $R_{2}$ is tangent to $R_{1}$ and $R_{3}$, the drift spacing is maximized, and points $G_{2}$ and $F_{2}$ coincide:

$$
\begin{equation*}
B \max =\sqrt{\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1} H^{\alpha_{1}}\left(2 e^{-\left(1 / \alpha_{1}\right)}-1\right)^{\alpha_{1}} \ln \frac{2}{2 e^{-\left(1 / \alpha_{1}\right)}-1}}+\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}} . \tag{16}
\end{equation*}
$$

When $R_{2}$ is tangent to the $z$-axis, the drift spacing is minimized:

$$
\begin{equation*}
B \min =\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}} \tag{17}
\end{equation*}
$$

In summary, when the IDZ is a nonstandard ellipsoid, the range of the drift spacing is as follows:

$$
\begin{align*}
& {\left[\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}}\right.} \\
& \left.\sqrt{\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1} H^{\alpha_{1}}\left(1+2 e^{\left.-\left(1 / \alpha_{1}\right)\right)^{\alpha_{1}} \ln \frac{2}{1+2 e^{-\left(1 / \alpha_{1}\right)}}}\right.}\right]  \tag{18}\\
& \left.+\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}}\right] \alpha<1.44, \\
& {\left[\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{\alpha_{1} e}}\right.} \\
& \sqrt{\left(\frac{\alpha+\alpha_{1}}{2}+1\right) \beta_{1} H^{\alpha_{1}}\left(2 e^{-\left(1 / \alpha_{1}\right)}-1\right)^{\alpha_{1}} \ln \frac{2}{2 e^{-\left(1 / \alpha_{1}\right)}-1}} \\
& \left.+\sqrt{\frac{\left(\left(\alpha+\alpha_{1} / 2\right)+1\right) \beta_{1}(2 H)^{\alpha_{1}}}{}}\right] \quad \alpha>1.44 .
\end{align*}
$$

3.2. Volume Calculation. When the sublevel height is determined, with decreasing drift spacing, the arrangement model based on IDZs changes from tangent to intersecting, the amount of crestal residual ore decreases, and the mixing amount of waste rock increases. $V_{o}$ is the lower crestal residual ore volume, and $V_{r}$ is the higher mixing waste rock volume. When $V_{o}$ and $V_{r}$ are calculated, the functional
relations of $V_{o}$ and $B$ as well as those of $V_{r}$ and $B$ are fitted (see Figure 4).

The solution steps are as follows:
Step 1: determine the sublevel height, parameter value, and range of $B$. The sublevel height is determined by occurrence conditions of the ore body, drilling machine, demolition equipment, and so forth. Once the sublevel height is determined, it is difficult to change. The parameters include the ore-rock bulk flow parameters ( $\alpha, \beta, \alpha_{1}$ and $\beta_{1}$ ), the impact coefficient of the ore body $(k)$, and the sublevel height $(H)$. The ore-rock bulk flow parameters can be obtained by laboratory or mining field drawing experiments, and the impact coefficient of the ore body is generally between 0.1 and 0.15 . The range of $B$ can be determined by equation (18) or (19).
Step 2: divide the areas of $V_{o}$ and $V_{r}$. As shown in Figure 3, the areas of $V_{o}$ and $V_{r}$ can be calculated by $V_{o}=1 / 2\left[R_{2}-\left(R_{2} \cap R_{2}^{\prime}\right)-\left(R_{2} \cap R_{1}\right)-\left(R_{2} \cap R_{4}\right)\right]$ and $V_{r}=R_{1} \cap R_{2}$.
Step 3: find the Monte Carlo solution for $V_{o}$ and $V_{r}$. With equations (2)-(5) and the parameters that are determined in step 1, the IDZ boundary is established. Using the Monte Carlo method with the IDZ boundary, the membership function, which is used to judge whether a point is within the boundary of the IDZ, is written. Statistical points in the $V_{o}$ and $V_{r}$ areas and $V_{o}$ and $V_{r}$ are determined $[24,25]$.
Step 4: determine the functional relationship between $V_{o}$ and $B$ as well as that between $V_{r}$ and $B$. Data analysis is performed on $V_{o}$ and $B$ as well as $V_{r}$ and $B$, and then the functional relationship between $V_{o}$ and $B$ and that between $V_{r}$ and $B$ are fitted.

$$
\begin{align*}
& V o=f_{1}(B), \\
& V r=f_{2}(B) \tag{20}
\end{align*}
$$

3.3. Economic Evaluation. The direct purpose of mining enterprises is to exploit ore, and the final profit product is the ore concentrate. To determine the optimal intersecting arrangement model based on IDZs, the economic benefit of the concentrate is taken as the final evaluation target. First, the calculation formula for ore profit and the calculation formula for ore loss are established. Then, combined with


Figure 4: Technical route of the volume calculation.
equation (20), the ore profit and loss are calculated. Finally, the ore profit and loss are analysed comprehensively, and the optimal solution is determined to obtain the best stope structure.
3.3.1. Ore Profit and Loss. (1) Ore Profit. The crestal residual ore from drawing to dressing eventually becomes concentrate; in this process, the profit mainly considers the concentrate profit and cost loss [26].

The concentrate profit refers to the value of the concentrate produced after the ore is processed by the beneficiation plant [27].

The ore is processed by the beneficiation plant and finally becomes concentrate and tailings. According to the conservation of quality, one ton of ore is processed into concentrate and tailings, so $1=Q_{c}+Q_{t}$. The conservation of the metal content is

$$
\begin{align*}
1 \times c_{o} & =Q_{c} \times c_{c}+Q_{t} \times c_{t}[27]  \tag{21}\\
Q_{c} & =\frac{c_{o}-c_{t}}{c_{c}-c_{t}} \tag{22}
\end{align*}
$$

The profit from converting a ton of ore into concentrate is

$$
\begin{equation*}
p_{1}=Q_{c}=\frac{\left(c_{o}-c_{t}\right)}{\left(c_{c}-c_{t}\right)} \times P \tag{23}
\end{equation*}
$$

where $Q_{c}$ is the quality of the concentrate, $Q_{t}$ is the quality of tailings, $C_{o}$ is the ore grade, $C_{c}$ is the concentrate grade, $C_{t}$ is the tailings grade, $p_{1}$ is the profit of converting a ton of ore into concentrate, and $P$ is the concentrate price.

The cost loss refers to the costs incurred by ore from drawing to dressing due to upgrading, ore blending, transportation, and so forth. Mining enterprises can statistically determine the actual technical and economic indicators of the year to obtain the cost loss per ton of ore, which is represented by $p_{2}$ [28].

Based on the above analysis, the profit generated by recovering one ton of crestal residual ore is $p_{1}-p_{2}$, and, with equation (23),

$$
\begin{equation*}
p=\frac{\left(c_{o}-c_{t}\right)}{\left(c_{c}-c_{t}\right)} \times P-p_{2} \tag{24}
\end{equation*}
$$

where $p$ is the ore profit.
(2) Ore Loss. The mixing of waste rock results in ore dilution, and the loss caused by this dilution mainly considers the loss of metal from tailings and the cost of waste rock mining and processing [26].

The loss of metal from tailings refers to waste rock that is of no grade or is lower than tailings grade; when this rock is treated as tailings, it needs to be matched with some lowgrade ore, causing metal loss [27].

For one ton of waste rock in tailings, the resulting metal loss is

$$
\begin{equation*}
Q_{m}=\frac{1}{1-c_{t}} \times c_{t}[27] \tag{25}
\end{equation*}
$$

where $Q_{m}$ is the metal quality.
The amount of metal lost due to one ton of waste rock in tailings is expressed as the amount of concentrate:

$$
\begin{equation*}
Q_{m}=Q_{c} \times c_{c} . \tag{26}
\end{equation*}
$$

The conservation of the metal content is

$$
\begin{equation*}
\frac{1}{1-c_{t}} \times c_{t}=Q_{c} \times c_{c} . \tag{27}
\end{equation*}
$$

The loss of metal from converting a ton of waste rock into tailings is converted into concentrate:

$$
\begin{equation*}
Q_{c}=\frac{c_{t}}{c_{c} \times\left(1-c_{t}\right)} \tag{28}
\end{equation*}
$$

The loss of metal from tailings is

$$
\begin{equation*}
l_{1}=\frac{c_{w}}{c_{j} \times\left(1-c_{w}\right)} \times P \tag{29}
\end{equation*}
$$

where $l_{1}$ is the loss of metal from tailings.
The cost of waste rock mining and processing refers to the expenses associated with mining, processing, transportation and stacking, and so forth incurred by the mixture of waste rock. Mining enterprises can calculate the actual technical and economic indicators of the year to obtain the cost loss per ton of waste rock, which is represented by $l_{2}$.

Based on the above analysis, the loss generated by mixing one ton of waste rock is $l_{1}+l_{2}$, and, with equation (29),

$$
\begin{equation*}
l=\frac{c_{w}}{c_{j} \times\left(1-c_{w}\right)} \times P+l_{2}, \tag{30}
\end{equation*}
$$

where $l$ is the ore loss.
3.3.2. Optimal Path Spacing. When the spatial arrangement model based on IDZs changes from a tangent arrangement to an intersecting arrangement, as the drift spacing decreases, the amount of crestal residual ore decreases but the amount of mixed waste rock increases. A smaller amount of crestal residual ore increases the profitability of the mine, and a larger amount of mixing waste rock increases the mine loss. With equations (24) and (30), the economy of the intersecting arrangement model based on IDZs can be calculated as

$$
\begin{equation*}
s=V_{O} \times \rho_{O} \times p-V_{r} \times \rho_{r} \times l, \tag{31}
\end{equation*}
$$

where $\rho_{o}$ is the ore density and $\rho_{r}$ is the rock density.
By equations (20), (31) can be rewritten as follows:

$$
\begin{equation*}
s=f_{1}(B) \times \rho_{o} \times p-f_{2}(B) \times \rho_{r} \times l . \tag{32}
\end{equation*}
$$

Differentiating equation (32) with respect to $B$ yields the drift spacing with the highest benefit, that is, the optimal intersection degree of IDZs.

## 4. Application of the Model

4.1. Engineering Background. The Yanqianshan Iron Mine is located in the Anshan mining area in Northeast China, the largest mining area currently in China. After the previous open-pit mining phase, there were approximately 357.27 million tons of geological reserves outside the open-pit boundary. In 2009, the mine officially entered the transition period from open pit to underground mining, which lasted for 5 years. After the transition period, the mine was fully converted to underground mining and the sublevel caving method was adopted (see Figure 5). The stope structure parameters are as follows: the sublevel height is 18 m and the drift spacing is 20 m . Under the conditions of the stope structure, the loss rate and dilution rate are approximately $15 \%$ and $18 \%$, respectively, and there is still room for improvement [29, 30].

In 2019, the design of the mining scheme for elevations of -213 m to -303 m in the Yanqianshan Iron Mine began in preparation for production over the next five years. Under the premise that the mining equipment meets the
production needs, the intent is to divide the ore body from elevations of -213 m to -303 m into four sublevels; then, the sublevel height is 22.5 m . This can reduce the cost of preliminary mining and the cutting of a sublevel (see Figure 6). However, the drift spacing has not been matched with the sublevel height, and the drift spacing has been included in the design and experimentation of the Yanqianshan Iron Mine.

According to the intersecting arrangement model based on IDZs, the optimal stope structure of elevations from -213 m to -303 m in the Yanqianshan Iron Mine is determined. Moreover, three experimental schemes are designed to compare and analyse the differences between the conventional tangent arrangement model based on IDZs and the intersecting arrangement model based on IDZs.
4.2. Stope Structure Determination. According to the intersecting arrangement model based on IDZs and combined with the stope characteristics of the Yanqianshan Iron Mine, the sublevel height is 22.5 m . The next step is to calculate the optimal drift spacing of elevations from - 213 m to -303 m .

The ore-rock bulk flow parameters are determined by ore drawing experiments, with reference to stochastic medium draw theory of Ren. To improve the consistency between the experiment and the mining site, the ore from the stope is used as the experimental material, and the similarity ratio is set to $1: 100$. After three experiments, the experimental results are averaged, and the ore-rock bulk flow parameters of the $-213 m$ to -303 m sublevels of the Yanqianshan Iron Mine are as follows: $\alpha=1.21, \beta=1.63, \alpha_{1}=1.32$, and $\beta_{1}=1.98$. Based on the mine design, the sublevel height $H$ is 22.5 m , and $k=0.1$.

Because $\alpha=1.21$, the IDZ of the Yanqianshan Iron Mine is similar to an ellipsoid with a thin upper part and thick lower part; according to equation (18), the range of the drift spacing is [ $17.9106 \mathrm{~m}, 22.4647 \mathrm{~m}$ ].

Using the calculation formula of $V_{o}$ and $V_{r}$, the values of $\alpha, \beta, \alpha_{1}, \beta_{1}, k$ and H are input, and then $V_{o}$ and $V_{r}$ corresponding to different drift spacings are obtained. Data analysis is performed on $V_{o}$ and $B$ as well as $V_{r}$ and $B$, and then the functional relationship between $V_{o}$ and $B$ as well as that between $V_{r}$ and $B$ are fit (see Figure 7).

$$
\begin{align*}
& V_{o}=-15.08 \times B^{2}+467.37 \times B-1108.36,  \tag{33}\\
& V_{r}=13.90 \times B^{2}-791.98 \times B+9087.28 \tag{34}
\end{align*}
$$

The economic and technical indicators obtained from statistical analysis of the production and economic situation in 2019 by the Technical Department of the Yanqianshan Iron Mine are shown in Table 1.

According to equation (24), the profit generated by recovering one ton of crestal residual ore is 51.09 USD/t; $p=51.09$ USD/t.

According to equation (30), the loss generated by mixing one ton of waste rock is $24.2 \mathrm{USD} / \mathrm{t} ; l=24.2 \mathrm{USD} / \mathrm{t}$.


Figure 5: The mining status model of the Yanqianshan Iron Mine.


Figure 6: Underground mining structure of the Yanqianshan Iron Mine.

The amount of crestal residual ore recovered is $2 V_{o}$, and the amount of waste rock is $4 V_{r}$ (see Figure 3). The ore density is measured as $5.0 \mathrm{t} / \mathrm{m}^{3}$, and the rock density is $3.5 \mathrm{t} /$ $\mathrm{m}^{3}$. According to equations (32)-(34),

$$
\begin{equation*}
s=-12413.47 \times B^{2}+507102.13 \times B-3645031.58 \tag{35}
\end{equation*}
$$

According to equation (35), the derivative with respect to $B$ yields $B=20.43 \mathrm{~m}$. Considering the construction and operation of the mine, the optimal approach spacing is set to 20.5 m .
4.3. Physical Simulation Experiments. Based on the above analysis, the stope structure from -213 m to -303 m in the Yanqianshan Iron Mine is determined, the sublevel height is 22.5 m , and the drift spacing is 20.5 m . To verify the
rationality of the scheme, physical simulation experiments are carried out to compare and analyse the loss and dilution degree of ore under the three conditions of the stope structure: the intersecting model, tangent model, and original stope structure.
4.3.1. Materials and Device. Magnetite and dolomite are used as the simulation materials in the experiments, in which magnetite is regarded as the caved ore and dolomite is regarded as the overlying waste rock. The lumpiness ratio of magnetite to stope caved ore is $1: 100$, which is obtained by screening at the crushing station, and the magnetite size is 1 mm to 3 mm . Dolomite has a slightly larger size than magnetite, ranging from 2 mm to 3 mm . Due to the fluidity of the ore-rock bulk, the size of overlying waste rock is larger


Figure 7: Fitting of the drift spacing and volume. (a) $B$ and $V_{O}$. (b) $B$ and $V_{r}$.

Table 1: Economic and technical indicators of the Yanqianshan Iron Mine in 2019.

| Economic and technical indicators | Symbols \& units | Value |
| :--- | :---: | :---: |
| Ore grade | $C o \%$ | 34 |
| Concentrate grade | $C c \%$ | 65.5 |
| Tailings grade | $C t \%$ | 10 |
| Concentrate price | $P \mathrm{USD} / \mathrm{t}$ | 125.7 |
| Cost loss | $p_{2} \mathrm{USD} / \mathrm{t}$ | 2.96 |
| Cost of waste rock | $l 2 \mathrm{USD} / \mathrm{t}$ | 2.96 |

than that of caved ore, which can reduce the mixing of waste rock and reduce ore dilution.

A self-designed physical stope model (see Figure 8) with a geometric scale of $1: 100$ is used to compare the loss and dilution degree of ore under the three stope structure conditions. A 10 mm thick transparent organic glass plate is used to simulate the stope work-plane. The work-plane has five sublevels, each sublevel has 4 or 5 drawpoints, and the drawpoints are in a staggered arrangement shaped like a rhombus. In this experiment, three organic glass plates are made to simulate three groups of different stope structures. An aluminum frame is used to hold the organic glass plates in place to form the stope model. The left, right, and back sides of the stope model are covered with a 10 mm thick transparent organic glass plate, which is convenient for loading experimental materials and observing the experimental process. There are slots on the left and right sides of the organic glass plate at a distance of 2 cm from the front stope work-plane to simulate caving space.
4.3.2. Schemes. As presented in Table 2, three physical simulation schemes are designed and 9 groups of experiments are conducted to compare and analyse the loss and dilution degree of the ore under the three stope structure conditions. The objective is to verify the rationality and
superiority of the intersecting arrangement model based on IDZs.
4.3.3. Process. The experiment is divided into three steps: filling, drawing, and weighing. During filling, a steel plate should be inserted into the slots of the organic glass plates on the left and right sides to form the caving space, magnetite should be used to fill the caving space, and dolomite should be used to fill the outside of the caving space. Magnetite and dolomite are added simultaneously up to sublevel 5; then, the steel plate is drawn out, and dolomite is continually added to a filling height greater than 2 times the height of a sublevel. The cut-off grade drawing method of the Yanqianshan Iron Mine is simulated for drawing, and the cut-off grade is $20 \%$. To facilitate the experiment operation, the cut-off grade is converted into the waste rock mixing rate, which is calculated to be $26 \%$. For each drawing, the magnetite and dolomite weights are determined, and when the weight of the dolomite divided by the weight of the magnetite and dolomite reaches $26 \%$, drawing at this drawpoint is stopped. Drawing is carried out from top to bottom and from left to right in turn. In the process, the overlying dolomite should be added in a timely manner to maintain a steady drawing pressure. After drawing, the magnetite and dolomite are separated to prepare for the next experiment.

In the drawing process, with the drawing of the ore, the overlying waste rock flows downward, resulting in ore and rock mixtures. When the ore and waste rock are mixed to the cut-off grade, drawing at this drawpoint is stopped. Through the transparent organic glass plate, part of the crestal residual ore can be seen (see Figure 9(a)). When the overlying waste rock is exfoliated layer by layer, residual ore can be seen (see Figure 9(b)). As ore drawing continues, the residual ore moves down with the overlying waste rock and is concentrated in the lowest sublevel.


Figure 8: Physical stope model of ore drawing.

Table 2: Experimental schemes.

| Number | Name | Sublevel height $(\mathrm{m})$ | Drift spacing $(\mathrm{m})$ |
| :--- | :---: | :---: | :---: |
| 1 | Original | 18 | 20 |
| 2 | Intersecting model | 22.5 | 20.5 |
| 3 | Tangent model | 22.5 | 22.5 |



Figure 9: Drawing phenomenon and effect. (a) Front view. (b) Top view.
4.3.4. Experimental Results and Analysis. Through comparative experiments under different IDZ arrangement models, the ore loss rate and the ore dilution rate are obtained. A comparison of the drawing effects under different IDZ arrangement models is presented in Table 3 and Figure 10.

Table 3 shows that, compared to the original stope structure, a sublevel height of 22.5 m can effectively reduce the ore loss and dilution rate, which proves the feasibility of large stope structure. When the sublevel height is 22.5 m , the drift spacing is set as 22.5 m in the tangent model, and the
ore loss rate and ore dilution rate are $14.18 \%$ and $12.08 \%$, respectively. However, when the drift spacing is set as 20.5 m in the intersecting model, the ore loss rate and ore dilution rate are $10.52 \%$ and $12.30 \%$, respectively, and the loss rate is significantly reduced, but the dilution rate changes slightly. Therefore, the intersecting arrangement model based on IDZs can have a beneficial drawing effect. It is further verified that the intersecting arrangement model based on IDZs can reduce the crestal residual ore and improve the recovery of the ore. Although there is also mixing of waste rock, it has little effect on the ore dilution.

Table 3: Comparison of the drawing effects with different IDZ arrangement models.

| Number | Stope height $(\mathrm{m})$ | Drift spacing $(\mathrm{m})$ | Loss rate $(\%)$ | Dilution rate $(\%)$ |
| :--- | :---: | :---: | :---: | :---: |
| 1 | 18 | 20 | 17.96 | 20.45 |
| 2 | 22.5 | 20.5 | 10.52 | 12.30 |
| 3 | 22.5 | 22.5 | 14.18 | 12.08 |



Figure 10: A comparison of the drawing effects with different IDZ arrangement models. (a) Loss rate. (b) Dilution rate.


Figure 11: Schematic of the dynamic intersecting arrangement model based on IDZs for stope structure optimization.

## 5. Discussion and Conclusion

The concept of the intersecting arrangement of IDZs was first proposed by Kvapil [3]; later, Laubscher [31] developed and established interaction theory. However, these studies were based on the ellipsoid draw theory, and no reasonable evidence has been provided to quantify the intersecting arrangement of IDZs [32]. In this paper, based on stochastic medium draw theory, a dynamic intersecting arrangement model based on IDZs was established for the case in which the shape of the IDZ is a nonstandard ellipsoid. The dynamic intersecting arrangement model can dynamically adjust the stope structure according to the ore-rock bulk flow parameters, economic indicators, occurrence conditions of the ore body, drilling machine, and so forth. Then, taking the ore profit and loss as the research object, a calculation model was established to quantify the degree of intersection of the IDZs and to determine the best stope structure. The schematic of the study setup is as follows (see Figure 11).

From the analysis of this model (see Figure 11), the main results are as follows:
(1) A dynamic intersecting arrangement model based on IDZs is established
(2) The range of the drift spacing $\left[B_{\min }, B_{\max }\right]$ is determined
(3) With the Monte Carlo algorithm, the calculation method of the ore recovery volume $\left(V_{o}\right)$ and rock mixed volume $\left(V_{r}\right)$ when the IDZs intersect is provided
(4) A calculation model is established for the ore profit and loss
Ore profit: $p=\left(\left(c_{o}-c_{t}\right) /\left(c_{c}-c_{t}\right)\right) \times P-p_{2}$ Ore loss: $l=\left(c_{w} / c_{j} \times\left(1-c_{w}\right)\right) \times P+l_{2}$
(5) A method for determining the drift spacing: $s=V_{O} \times \rho_{O} \times p-V_{r} \times \rho_{r} \times l=f_{1}(B) \times \rho_{o} \times p-$ $f_{2}(B) \times \rho_{r} \times l$ is established, and the drift spacing is determined by finding the derivative with respect to B

According to the intersecting arrangement model based on IDZs, the stope structure of elevations from -213 m to -303 m in the Yanqianshan Iron Mine in China was determined; the optimal drift spacing is 20.5 m , and the sublevel height is 22.5 m . Nine physical experiments under three stope structure conditions, namely, the intersecting model, tangent model, and original stope structure, were carried out. The experimental results showed that the sublevel height of 22.5 m can effectively reduce the ore loss and dilution rate; at a drift spacing of 20.5 m , the ore loss rate decreased by $3.66 \%$, but the dilution rate changed little, with an increment of $0.22 \%$. This finding verifies the rationality and superiority of the intersecting arrangement model based on IDZs to optimize the stope structure.

An intersecting arrangement model based on Kvapil and Laubscher's interaction theory was proposed in this paper. This model is suitable for the case in which the shape of the IDZ is a nonstandard ellipsoid, and the model is more in line
with actual mining conditions. The accurate shape of the IDZ is the basis for stope structure optimization, and future research should focus on the measurement of the shape of the IDZ. Moreover, the model does consider the mechanical properties of the movement of ore-rock bulk when dynamically adjusting the arrangement of IDZs. In future research, mechanical factors should be added to develop a stope structure optimization software program that dynamically adjusts the intersecting arrangement to better guide mine production.

## Data Availability

The data used to support the findings of this study are included within the article.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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# Managing a Dual-Channel Supply Chain with Fairness and Channel Preference 

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#### Abstract

This paper investigates a dual-channel supply chain in which a manufacturer sells the product via an offline retailer or online store. The manufacturer sets the wholesale and online price, and the retailer decides the retail price with the retailer's fairness preference and consumer's online channel preference. Through investigating the combined impacts of fairness preference and channel preference on the enterprises' operational strategies, this paper obtains some meaningful results. If a manufacturer thinks over the fairness preference, he decreases the wholesale price to mitigate a loss of retailer and benefit the supply chain design. The manufacturer intends to set up the online channel with a lower acceptance as the fairness preference grows. However, the gains from enhanced online channel acceptance cannot compensate for the manufacturer's loss by the fairness effect that benefits the retailer. Moreover, the manufacturer cannot neglect the retailer's fairness preference generating a "lose-lose" case for both members.


## 1. Introduction

With the increasing innovation of information technology and vigorous publicity by commercial organizations, consumers are getting used to online shopping. According to China Statistical Yearbook-2020, online sales in China reached 155.2 billion dollars in 2019, increasing $16.5 \%$ over the previous year (https://www.chinainternetwatch.com/ $30232 /$ retail $\% 202019 /$ ). To seize the Internet opportunity, many enterprises have set up the supply chain composed of online and offline channels such as Apple, Nike, Zara, and Huawei, and many scholars study dual-channel supply chain under different cases. A few papers focus on how the consumer's online channel preference impacts the dualchannel strategy (Chiang et al. [1], Xu et al. [2], and Zhang et al. [3]), while others try to find the operational strategy for the dual-channel supply chain (Guo et al. [4] and Zhu et al. [5]). There is also a lot of research (Chen et al. [6], Cui et al. [7], Guan et al. [8], Niu et al. [9], and Pan et al. [10]) studying
the impact of fairness preference in the supply chain, especially the supply chain design decisions and coordination. However, the combined effect of the retailer's fairness and consumer's online channel preference is ignored in the dualchannel design, which motivates our research. To fill this gap, our paper considers a dual-channel system in which the retailer has fairness preference, and the consumers are differentiated in online channel acceptance. The primary purposes of our research are to address the following issues:
(1) Varying with consumer's online channel preference and fairness concern, how do the manufacturer and retailer decide their operational strategy, including pricing and dual-channel design? Also, how do the related parameters affect the manufacturer's and retailer's strategy?
(2) With the increases in the consumer's online channel preference, can the manufacturer's profit be better when the retailer has a fairness preference?
(3) Can the manufacturer ignore the retailer's fairness preference? What strategy would be adopted by the retailer if the manufacturer ignores it?

To tackle these issues, the paper establishes a dualchannel decision model composed of a manufacturer (he) and a retailer (she). The model has three circumstances: the retailer without fairness preference, the manufacturer considering her fairness preference, and ignoring her fairness preference. The paper studies the synergistic impact of consumer online channel acceptance and retailer's fairness preference on members' pricing decisions and profits. It compares and analyzes optimal decisions under the above three circumstances. Different from other literatures that independently examine the influence of supply chain enterprises' decisions from the perspective of consumer's online channel preference or retailers' fairness, this paper obtains some meaningful conclusions as follows: (i) If a manufacturer thinks over the fairness preference, he decreases his wholesale price to mitigate a loss of retailer. The increase in consumer online channel acceptance would strengthen the trend. However, the manufacturer also intends to build an online channel with a lower consumer online channel acceptance to decrease his dependence on the retailer. (ii) Whether the manufacturer considers or ignores fairness preference, the increase of consumer online channel acceptance cannot compensate for the manufacturer's loss due to the retailer's fairness effect. The manufacturer's interest concessions via wholesale price reduction caused by the consideration of the retailer's fairness preference exceed the benefit from the increase of online shopping. (iii) If the manufacturer overlooks the retailer's fairness, they will fall into a "lose-lose" case. The retailer would price higher to make up for the online market's loss, which also injures her profit. This also brings harm to the sale resulting in the reduction of the manufacturer's profit.

The rest of the paper is arranged as follows. Section 2 is a review of relevant literature with two streams. Section 3 establishes a game model with one leading manufacturer and one following retailer. It examines the members' pricing decisions and profits when the retailer has no fairness preference as a benchmark case. Section 4 consists of two situations as comparative cases, where the manufacturer considers or ignores the retailer's fairness preference. The enterprises' decisions and profits in two different situations are obtained. Section 5 analyzes the manufacturer's decisions on setting up the online channel under different scenarios. Section 6 makes some comparisons between different situations. Section 7 concludes the paper.

## 2. Literature Review

This paper is related to two literature streams: the online channel preference and the retailer's fairness preference in the supply chain.

The first related stream of the literature focuses on the online channel preference that includes Zhang et al. [3], Guo et al. [4], Zhu et al. [5], Cai [11], Cao [12], Dumrongsiri et al. [13], Dan et al. [14], Gao et al. [15], Hua et al. [16], Huang
and Swaminathan [17], Li et al. [18], Lu and Liu [19], Ranjan and Jha [20], and Yoo and Lee [21]. For instance, Zhang et al. [3] find that the channel selections are determined by the consumer's online acceptance. Guo et al. [4] show that the dual channel may make traditional retailers feel threatened. Dumrongsiri et al. [13] show that the uncertain demand obviously affects the optimal pricing and the manufacturer's incentive to seek an online channel. Hua et al. [16] demonstrate that consumer acceptance of online purchases can positively affect the manufacturer's online pricing and lead to longer delivery lead times. Li et al. [18] find that, with the high consumer's online acceptance, the manufacturer provides customized products under the decentralized and centralized supply chain structures. Gao et al. [15] and Lu and Liu [19] investigate how the consumer's online acceptance affects the dual-channel excepted profits and obtain that the consumer's online acceptances positively affect the manufacturer's profit, while the retailer's gain is the opposite. Ranjan and Jha [20] think over the effect of retailers' risk attitude and demand uncertainty on the pricing mechanism among dual-channel members.

The second related stream of the literature concentrates on fairness preference in supply chain management including Chen et al. [6], Cui et al. [7], Guan et al. [8], Niu et al. [9], Caliskan-Demirag et al. [22], Du et al. [23], Ho et al. [24], Katok and Pavlov [25], Q. H. Li and B. Li [26], Nie and Du [27], and Du et al. [28]. Cui et al. [7] introduce members' fairness preference into the dyadic supply chain and conclude that the channel can be coordinated by using a wholesale price contract. Guan et al. [8] extend the analysis of the impact of peer-fairness or Nash bargaining fairness preference on the pricing of supply chain enterprises and the overall efficiency of the supply chain. Niu et al. [9] study the role of channel strength and fairness preference in online channel opening and find that supplier fairness concerns are not conducive to online channel development. Chen et al. [6] and Pan et al. [10] show that the enhancement of the retailer's rights will make her pay attention to the fairness of her own profits, which may improve the supply chain's performance. Caliskan-Demirag et al. [22] research channel coordination issues with different fairness concerns through a simple wholesale price and other nonlinear demands. Du et al. [23] analyze the impact of fairness preference behavior on enterprise decisions and channel efficiency and show that although channel efficiency will be reduced, fairness preference still coordinates the supply chain in specific cases. Q. H. Li and B. Li [26] study the channel issues with retailer's fairness preference and show the fairness concern's negative impacts on channel efficiency.

The above literature studies the influence of supply chain members' decision-making in the sense of consumer's online channel preference or retailers' fairness. Unlike the related research, our paper first studies the combining effect of both consumer's online channel preference differences and the retailer's fairness preference on supply chain decisions. Therefore, our paper theoretically enriches the supply chain design research, which is more practical. It helps decision-makers be aware of the critical role of fairness preference behavior and online consumer heterogeneity in
decision-making. Moreover, our research allows decisionmakers to use these factors to reach supply chain efficiency improvement and provide guidance for the enterprises.

## 3. Benchmark Model

3.1. Model Description. This paper develops a dual-channel supply chain, including one manufacturer (he) and one retailer (she). There is a Stackelberg game between the manufacturer and the retailer, with the manufacturer dominating the supply chain and the retailer being the follower. The manufacturer wholesales the product at the wholesale price $w$ to the retailer, with the cost $c$, while the retailer sells it to consumers at a retail price $p_{r}$. With the popularity of online shopping, the manufacturer opens up an online channel to increase market penetration with online price $p_{e}$ and gain more profits. In turn, the manufacturer has a dual-channel structure, consisting of an offline channel and an online channel as shown in Figure 1.

The demand function refers to the analyses of Chiang et al. [1], Dan et al. [14], and Huang and Swaminathan [17]. The manufacturer's online channel demand function $q_{e}$ and the retailer's offline channel demand function $q_{r}$ are as follows:

$$
\begin{align*}
& q_{e}=\theta a-b p_{e}+\lambda p_{r},  \tag{1}\\
& q_{r}=(1-\theta) a-b p_{r}+\lambda p_{e} \tag{2}
\end{align*}
$$

where $a$ indicates the basic demand of the market. $\theta(0<\theta<1)$ measures the consumer's online channel preference and $1-\theta$ indicates the consumer's offline channel acceptance. $b$ denotes the influence factor of the channel price on demand, while $\lambda$ denotes the influence factor of the cross-price with $b>\lambda$.

The manufacturer's rationality means that the maximization of its profit is the decision-making process's de-cision-making goal. The fairness preference of a retailer indicates that the retailer is concerned with maximizing its utility when making decisions. The retailer's fairness preference model draws on the FS model [29], and its utility function is the following:

$$
\begin{equation*}
U_{r}=\prod_{r}-\alpha \max \left(\prod_{m}-\prod_{r}, 0\right)-\beta \max \left(\prod_{r}-\prod_{m}, 0\right) . \tag{3}
\end{equation*}
$$

The coefficients of $\alpha$ and $\beta$ are the disadvantage and advantage inequity aversion, respectively. Qin and Yang [30] show that, in the Stackelberg game of the two-tier supply chain, the member playing the leadership role can reap more than half of the overall profits. Thus, in this study, only the disadvantage inequity averseness of the retailer is considered. The retailer's utility function is as follows:

$$
\begin{equation*}
U_{r}=\prod_{r}-\alpha \max \left(\prod_{m}-\prod_{r}, 0\right) \tag{4}
\end{equation*}
$$

The relevant symbols are shown in Table 1.


Figure 1: The dual-channel supply chain.

Table 1: Nomenclature.

| Notation | Definition |
| :--- | :---: |
| $\Pi_{m}$ | Manufacturer's profit |
| $\Pi_{r}$ | Retailer's profit |
| $U_{m}$ | Manufacturer's utility |
| $U_{r}$ | Retailer's utility |
| $w$ | Wholesale price |
| $p_{e}$ | Online channel price |
| $p_{r}$ | Offline channel price |
| $c$ | Manufacturer's cost |
| $\theta$ | Consumer's online channel preference |
| $a$ | Total market demand |
| $b$ | Price elasticity |
| $\lambda$ | Cross-price elasticity |
| $\alpha$ | Coefficient of retailer fairness preference |

3.2. The Retailer without Fairness Preference. In this context, the profits functions of the manufacturer and the retailer are as follows:

$$
\begin{align*}
& \prod_{m}=(w-c) q_{r}+\left(p_{e}-c\right) q_{e}  \tag{5}\\
& \prod_{r}=\left(p_{r}-w\right) q_{r} \tag{6}
\end{align*}
$$

and, with backward induction to solve this problem, it can be easily verified that when the retailer has no fairness preference, the equilibrium pricing strategies and profits for the manufacturer and the retailer are

$$
\begin{align*}
& p_{e}=\frac{(\lambda+b \theta-\lambda \theta) a+\left(b^{2}-\lambda^{2}\right) c}{2\left(b^{2}-\lambda^{2}\right)},  \tag{7}\\
& w=\frac{(b+\lambda \theta-b \theta) a+\left(b^{2}-\lambda^{2}\right) c}{2\left(b^{2}-\lambda^{2}\right)},  \tag{8}\\
& p_{r}=\frac{\left[3 b^{2}(1-\theta)+\lambda \theta(2 b+\lambda)-\lambda^{2}\right] a+c(b-\lambda)(b+\lambda)^{2}}{4 b\left(b^{2}-\lambda^{2}\right)} . \tag{9}
\end{align*}
$$

According to the above optimal pricing decisions, the optimal profits for the retailer and the manufacturer without fairness preference are

$$
\begin{align*}
& \prod_{r}=\frac{[a(\theta-1)+c(b-\lambda)]^{2}}{16 b}  \tag{10}\\
& \prod_{m}=\frac{a^{2} A_{1}-2 a c\left(b^{2}-\lambda^{2}\right) A_{2}+c^{2}(b-\lambda)^{2} A_{3}}{8 b\left(b^{2}-\lambda^{2}\right)} \tag{11}
\end{align*}
$$

where $A_{1}=\left(3 b^{2}+\lambda^{2}\right) \theta^{2}-4 b \lambda \theta(\theta-1)+(1-2 \theta)\left(b^{2}+\lambda^{2}\right)$, $A_{2}=\theta(b-\lambda)+(b+\lambda)$, and $A_{3}=3 b^{2}+\lambda^{2}+4 b \lambda$.

## 4. The Model with the Retailer's Fairness Preference

There are two circumstances when the retailer has a fairness preference. One case is when the manufacturer considers the retailer's fairness preference and the other is when the manufacturer ignores it. The paper analyzes the different cases with Sections 4.1 and 4.2. The subscript * represents the manufacturer's optimal decisions considering the retailer's
fairness preference; the subscript ${ }^{* *}$ means the optimal choices with the manufacturer ignoring the retailer's fairness preference.
4.1. The Manufacturer Considers the Retailer's Fairness Preference. In this situation, the utility functions of the manufacturer and the retailer are

$$
\begin{equation*}
U_{m}=\prod_{m} \tag{12}
\end{equation*}
$$

$$
\begin{equation*}
U_{r}=\prod_{r}-\alpha\left(\prod_{m}-\prod_{r}\right)=(1+\alpha) \prod_{r}-\alpha \prod_{m} \tag{13}
\end{equation*}
$$

Lemma 1. When the manufacturer considers the retailer's fairness preference, the equilibrium pricing strategies and profits for the manufacturer and the retailer are

$$
\begin{align*}
& p_{e}^{*}=\frac{(\lambda+b \theta-\lambda \theta) a+\left(b^{2}-\lambda^{2}\right) c}{2\left(b^{2}-\lambda^{2}\right)},  \tag{14}\\
& w^{*}=\frac{(3 \alpha c+c) b^{3}-((c \lambda+(\theta-1) a) \alpha+(\theta-1) a) b^{2}+\left(\left(2 a \lambda \theta-3 c \lambda^{2}\right) \alpha+a \lambda \theta-c \lambda^{2}\right) b-(-\lambda c+(\theta-1) a) \lambda^{2} \alpha}{2 b\left(b^{2}-\lambda^{2}\right)(1+2 \alpha)},  \tag{15}\\
& p_{r}^{*}=\frac{\left[3 b^{2}(1-\theta)+\lambda \theta(2 b+\lambda)-\lambda^{2}\right] a+c(b-\lambda)(b+\lambda)^{2}}{4 b\left(b^{2}-\lambda^{2}\right)} . \tag{16}
\end{align*}
$$

According to the above optimal pricing, it can be obtained that $\prod_{r}$ and $\prod_{m}$ in this case are

$$
\begin{align*}
& \prod_{r}^{*}=\frac{[a(\theta-1)+c(b-\lambda)]^{2}(4 \alpha+1)}{16 b(2 \alpha+1)}  \tag{17}\\
& \prod_{m}^{*}=\frac{a^{2} B_{1}-2 a c\left(b^{2}-\lambda^{2}\right) B_{2}+c^{2}(b-\lambda)^{2} B_{3}}{8 \mathrm{~b}\left(b^{2}-\lambda^{2}\right)(2 \alpha+1)} \tag{18}
\end{align*}
$$

where $\quad B_{1}=\left[(5 \alpha+3) \theta^{2}+(1+\alpha)(1-2 \theta)\right] b^{2}+(3 \alpha+1)$ $(\theta-1)^{2}+4 b \lambda(1-\theta)(1+2 \alpha), \quad B_{2}=3 \alpha \theta(b-\lambda)+b(\alpha+\theta)$ $+\lambda(3 \alpha-\theta)+b+\lambda$, and $B_{3}=(5 \alpha+3) b^{2}+(3 \alpha+1) \lambda^{2}+4 b \lambda$ $(2 \alpha+1)$.
Proof of Lemma 1. Solving the Stackelberg game, the backward induction is used as follows.

Inserting (6) and (12) into (13) produces

$$
\begin{align*}
U_{r}= & (1+\alpha)\left(p_{r}-w\right)\left((1-\theta) a-b p_{r}+\lambda p_{e}\right) \\
& -\alpha(w-c)\left((1-\theta) a-b p_{r}+\lambda p_{e}\right)  \tag{19}\\
& -\alpha\left(p_{e}-c\right)\left(\theta a-b p_{e}+\lambda p_{r}\right) .
\end{align*}
$$

As $\left(\partial U_{r} / \partial p_{r}\right)=0$ and $\left(\partial^{2} U_{r} / \partial p_{r}^{2}\right)=-2 b(1+\alpha)<0$, the retailer has the only optimal retail price. The first-order partial derivative of $U_{r}$ concerning $p_{r}$ is expressed as follows:
$p_{r}\left(w, p_{e}\right)=\frac{[a(1-\theta)-c(b+\lambda)+2 b w] \alpha+a(1-\theta)+b w-\lambda p_{e}}{2 b(1+\alpha)}$.

To verify that $\prod_{m}$ is a joint concave function on wholesale price $w$ and online price $p_{e}$, the second-order conditions can be obtained:

$$
\left\{\begin{array}{l}
\frac{\partial^{2} \prod_{m}}{\partial w^{2}}=\frac{-b(1+2 \alpha)}{1+\alpha}<0,  \tag{21}\\
\frac{\partial^{2} \prod_{m}}{\partial p_{e}^{2}}=\frac{\lambda^{2}}{(1+\alpha) b}-2 b<0, \\
\frac{\partial^{2} \prod_{m}}{\partial w \partial p_{e}}=\frac{(1+2 \alpha) \lambda}{1+\alpha}>0 .
\end{array}\right.
$$

As the Hessian matrix of the $\prod_{m}$ above formulae is negative, it has a unique optimal solution. Then setting $\left(\delta \Pi_{m} / \delta w\right)$ and ( $\left.\delta \Pi_{m} / \delta p_{e}\right)$ to zero yields

$$
\left\{\begin{array}{l}
w=\frac{\left[a(1-\theta)+3 b c-3 c \lambda+4 \lambda p_{e}\right] \alpha+a(1-\theta)+b c-c \lambda+2 \lambda p_{e}}{2 b(1+2 \alpha)},  \tag{22}\\
p_{e}=\frac{\left(2 a b \theta-a \lambda \theta+2 b^{2} c-3 b c \lambda+4 b w \lambda+c \lambda^{2}+a \lambda\right) \alpha+2 a b \theta-a \lambda \theta+2 b^{2} c-b c \lambda+2 b w \lambda-c \lambda^{2}+a \lambda}{2\left(2 b^{2} \alpha+2 b^{2}-\lambda^{2}\right)} .
\end{array}\right.
$$

Solving (20) and (22) can obtain the optimal pricing as (14)-(16). Then inserting (14)-(16) into (5) and (6) produces the equilibrium profits for the retailer and the manufacturer as (17) and (18). Lemma 1 is proved.
4.2. The Manufacturer Ignores the Retailer's Fairness Preference. In this subsection, the paper assumes that the retailer's fairness preference information is not taken into account by the manufacturer. Under this condition, the
manufacturer still prices considering that the retailer has no fairness preference, which is represented by $w^{* *}=w$ and $p_{e}^{* *}=p_{e}$. However, the retailer has a fairness preference and uses the utility function of equation (13) to make optimal decisions.

Lemma 2. When the manufacturer ignores the retailer's fairness preference, the offline price and the manufacturer's and retailer's profits are

$$
\begin{equation*}
p_{r}^{* *}=\frac{\left(2 a b \lambda \theta-4 a b^{2} \theta+2 a \lambda^{2} \theta+2 b^{2} c \lambda-2 c \lambda^{3}+4 a b^{2}-2 a \lambda^{2}\right) \alpha+\left[3 b^{2}(1-\theta)+\lambda \theta(2 b+\lambda)-\lambda^{2}\right] a+c(b-\lambda)(b+\lambda)^{2}}{4 b(1+\alpha)\left(b^{2}-\lambda^{2}\right)} \tag{23}
\end{equation*}
$$

$\prod_{r}$ and $\prod_{m}$ can be calculated as

$$
\begin{align*}
& \prod_{r}^{* *}=\frac{[a(\theta-1)+c(b-\lambda)]^{2}(1+2 \alpha)}{16 b(1+\alpha)^{2}}  \tag{24}\\
& \prod_{m}^{* *}=\frac{a^{2} C_{1}-2 a c\left(b^{2}-\lambda^{2}\right) C_{2}+c^{2}(b-\lambda)^{2} C_{3}}{8 b\left(b^{2}-\lambda^{2}\right)(1+\alpha)} \tag{25}
\end{align*}
$$

where $C_{1}=(2 \alpha+3) b^{2} \theta^{2}-4 b \lambda \theta(\theta-1)(1+\alpha)+\left[2 \alpha(\theta-1)^{2}\right.$ $\left.+\theta^{2}\right] \lambda^{2}+(1-2 \theta)\left(b^{2}+\lambda^{2}\right), \quad C_{2}=\theta(b-\lambda)(2 \alpha+1)+2 \alpha \lambda b$ $+b+\lambda$, and $C_{3}=[(b+\lambda) \alpha+2 b](b+\lambda)$.

Following the calculating process of the proof of Lemma 1 in Section 4.1, Lemma 2 can be easily proved. The detailed formulae are not provided here.

## 5. Supply Chain Design Analysis

This section analyzes the manufacturer's decisions on setting up the online channel under different scenarios. When the manufacturer ignores the fairness concern of the retailer, his strategy is the same as the retailer who has no fairness preference shown in Section 4.2. Therefore, the two possible situations are as follows.
5.1. The Retailer without Fairness Preference. As the manufacturer's online price is greater than the wholesale price
( $p_{e}>w$ ), this results in $\theta>(1 / 2)$. If $0<\theta<(1 / 2)$, then $p_{e}<w$. The retailer can purchase large quantities of goods from the online channel and then sell them offline to make more profits, contrary to the manufacturer's original intention to set up an online channel.

Moreover, the retailer's offline channel's price should be greater than the manufacturer's wholesale price, which means that $p_{r}>w$ and results in $\theta<((a-c(b-\lambda)) / a)$. If $((a-c(b-\lambda)) / a)<\theta<1$, then $p_{r}<w$. It will bring loss to the retailer and cause the retailer to abandon the offline channel. The dual-channel supply chain may need to be reconstructed if $\theta$ is out of $((1 / 2),((a-c(b-\lambda)) / a))$.

Following the same analytical process above, this paper could find that the range is as same as that when the manufacturer ignores the fairness concern of the retailer. The paper concludes the supply chain design strategy in Lemma 3 as follows.

Lemma 3. If the retailer does not have fairness preference or the manufacturer ignores her preference, the manufacturer intends to build an online channel when $\theta>(1 / 2)$. When $\theta$ approaches $((a-c(b-\lambda)) / a)$, the retailer may choose to quit the supply chain.

Lemma 3 indicates that when the retailer has no fairness preference or the manufacturer ignores it, the manufacturer
chooses to develop an online channel when the consumers are more receptive to them. Furthermore, the retailer is unwilling to participate in the dual-channel operation if the online channel acceptance exceeds a threshold. For the manufacturer, if he continues to use the dual channel to protect market penetration, he may need to provide additional benefits for the retailer.
5.2. The Manufacturer Considers the Retailer's Fairness Preference. Now this paper discusses the supply chain design strategy when the retailer possesses the fairness preference. Following the analytic process in Section 5.1, it can get $\theta>\left(\left(\alpha c\left(b^{2}-\lambda^{2}\right)+\alpha a(b-\lambda)+a b\right) /(a(3 b \alpha-\alpha \lambda+2 b))\right)$ as $p_{e}^{*}>w^{*}$. With $p_{r}^{*}>w^{*}$, it gets $\theta<((a-c(b-\lambda)) / a)$. When the lower bound $\quad\left(\left(\alpha c\left(b^{2}-\lambda^{2}\right)+\alpha a(b-\lambda)\right.\right.$ $+a b) /(a(3 b \alpha-\alpha \lambda+2 b)))$ is used to get the derivation of $\alpha$, $\left((b(b+\lambda)[2 c(b-\lambda)-a]) /\left(a(3 b \alpha-\alpha \lambda+2 b)^{2}\right)\right)<0$ can be obtained. This shows that the manufacturer intends to develop a new road to market as he decreases the wholesale price for the retailer's fairness concerns. The upper bound of $\theta$ remains unchanged because the high consumer's online channel preference may lead to the suspension of the retailer's sales. So, the upper bound of $\theta$ is not affected by fairness preference.

Lemma 4 shows the manufacturer's supply chain design strategy when considering the retailer's fairness preference.

Lemma 4. When the manufacturer takes the retailer's fairness into account, the dual-channel supply chain works if and only if $\left(\left(\alpha c\left(b^{2}-\lambda^{2}\right)+\alpha a(b-\lambda)+a b\right) /(a(3 b \alpha-\alpha \lambda+\right.$ $2 b)))<\theta<((a-c(b-\lambda)) / a)$. With the increase of the fairness preference coefficient, the lower bound of $\theta$ will gradually decrease.

Intuitively, if the retailer focuses on fairness, the manufacturer will transfer some profit to the retailer to eliminate its fairness. However, the manufacturer also wants to offset the offline supply chain's loss by adding an online channel. This brings a decrease in the starting point of opening an online channel. As the acceptance is low, the retailer does not think that the online channel is a threat, and the manufacturer can also get a little profit by adding an online channel. Lemma 4 shows that if a retailer pursues the fairness of her profit excessively, the manufacturer becomes urgent for setting up an online channel, even though consumers have a low preference for an online channel. However, this is not beneficial for the retailer, which means that she should maintain an appropriate level of fairness.

## 6. Comparative Analysis

Given the strategies under the three circumstances mentioned in Section 3 and 4, this paper compares operational strategies between the manufacturer and the retailer to analyze the effects of the retailer's fairness preference and consumer's online channel preference.

### 6.1. Pricing Comparison

Lemma 5. For pricing strategies,
(a) $p_{e}=p_{e}^{*}=p_{e}^{* *}, w^{*}<w=w^{* *}$, and $p_{r}=p_{r}^{*}<p_{r}^{* *}$;
(b) $\left(\partial w^{*} / \partial \alpha\right)<0, \quad\left(\partial w^{*} / \partial \theta\right)<0, \quad\left(\partial p_{r}^{* *} / \partial \alpha\right)>0, \quad$ and $\left(\partial p_{r}^{* *} / \partial \theta\right)<0$.

Proof of Lemma 5. When the manufacturer considers the retailers' fairness preference, $w^{*}-w<0 ; \quad\left(\partial w^{*} / \partial \alpha\right)=$ $\left((a(\theta-1)+c(b-\lambda)) /\left(2 b(1+2 \alpha)^{2}\right)\right) \quad$ and $\quad\left(\partial w^{*} / \partial \theta\right)=$ $-((a[\alpha(b-\lambda)+b]) /(2 b(b+\lambda)(1+2 \alpha)))<0$ can be obtained when formula (15) is used to get the derivations of $\alpha$ and $\theta .\left(\partial w^{*} / \partial \alpha\right)=\left((a(\theta-1)+c(b-\lambda)) /\left(2 b(1+2 \alpha)^{2}\right)\right)<0$ because $\theta<((a-c(b-\lambda)) / a)$ is demonstrated in Lemmas 3 and 4. When the manufacturer ignores the retailer's fairness preference, $\quad p_{r}^{* *}-p_{r}>0 . \quad\left(\partial p_{r}^{* *} / \partial \alpha\right)=((a(1-\theta)-$ $\left.c(b-\lambda)) /\left(4 b(1+\alpha)^{2}\right)\right)>0$ and $\left(\partial p^{* *} / \partial \theta\right)=-((a[b(4 \alpha+$ $3)+\lambda(2 \alpha+1)]) /(4 b(b+\lambda)(1+\alpha)))<0$ can be obtained when formula (23) is used to get the derivation of $\alpha$.

It can be seen from Lemma 5 that when the manufacturer thinks over the retailer's fairness preference, the wholesale price reduces as the fairness preference coefficient increases. If the manufacturer ignores it, the retail price improves with the fairness preference coefficient increasing. Moreover, taking the rise in consumers' online channel preferences into account, the wholesale will decline faster, but the retail price growth slows down. To demonstrate the impact of the fairness preference coefficient on supply chain members' pricing decisions and profits in Lemma 5 more clearly, this paper numerically analyzes the above model and discusses the results. The numerical results are shown in Figures 2 and 3 with $a=100, b=1, \lambda=0.4, c=10$. The parameter values are also applicable to subsequent numerical legends. As shown in Figure 2, it can be seen directly that when $\theta$ is constant, with the increase of $\alpha, w^{*}$ continues to fall. In order to make the dual channels work, the value of $\theta$ should be greater than $1 /$ 2. Moreover, when $\alpha$ is constant, the wholesale price still deceases with the increase of $\theta$. According to Figure 3, when $\theta$ is constant, with the increase of $\alpha$, the retail price improves when the manufacturer ignores the retailer's fairness preference. If $\alpha$ is constant, the retail price decreases with the increase of $\theta$.

When the manufacturer takes the retailer's fairness preference into account, he lowers the wholesale price to maintain the balance of channel profit distribution. When the increase in consumer's online channel preference weakens the retailer's market competitiveness, consumers are willing to choose shopping online. To compensate for the loss of the retailer, the manufacturer's wholesale price will fall more. Moreover, because the manufacturer ignores the retailer's fairness preference, with the fairness preference increase, the retailer's focus on fairness will force her to raise pricing for the sake of maintaining profits. However, with the increase of consumer's online preference, the retailer chooses to reduce the retail price attracting consumers to select the offline channel.


Figure 2: The influence of $\alpha$ and $\theta$ on $w^{*}$.


Figure 3: The influence of $\alpha$ and $\theta$ on $p_{r}^{* *}$.

### 6.2. The Retailer's Profits Comparison

Lemma 6. For the retailer's profits under the above three circumstances,
(a) $\prod_{r}^{*}>\prod_{r}>\prod_{r}^{* *}$
(b) $\left(\partial \prod_{r}^{*} / \partial \alpha\right)>0$, $\left(\partial \prod_{r}^{*} / \partial \theta\right)<0,\left(\partial \prod_{r}^{* *} / \partial \alpha\right)<0$, and $\left(\partial \prod_{r}^{* *} / \partial \theta\right)<0$
Proof of Lemma 6. It can be easily calculated from (10), (17), and (24) that $\prod_{r}^{*}-\prod_{r}>0$ and $\prod_{r}-\prod_{r}^{* *}>0$. $\left(\partial \prod_{r}^{*} / \partial \alpha\right)=$ $\left(\left([a(\theta-1)+c(b-\lambda)]^{2}\right) /\left(8 b(2 \alpha+1)^{2}\right)\right)>0 \quad$ and $\left(\partial \prod_{r}^{*} /\right.$ $\partial \theta)=((a(4 \alpha+1)[a(\theta-1)+c(b-\lambda)]) /(8 b \quad(1+2 \alpha)))<0$ can be obtained when (17) is used to get the derivation of $\alpha$ and $\theta$. $\left(\partial \prod_{r}^{* *} / \partial \alpha\right)=\left(\left(-\alpha[a(\theta-1)+c(b-\lambda)]^{2}\right) /(8 b(1+\right.$ $\left.\left.\alpha)^{3}\right)\right)<0$ and $\left(\partial \prod_{r}^{* *} / \partial \theta\right)=((a(2 \alpha+1)[a(\theta-1)+c(b-$ $\left.\lambda)]) /\left(8 b(1+2 \alpha)^{2}\right)\right)<0$ can be obtained when (24) is used to get the derivation of $\alpha$ and $\theta$.

Lemma 6 shows that the retailer can increase profitability when she focuses on fairness. The manufacturer considers the retailer's fairness preference, which enhances the retailer's bargaining power. He transfers profits to the retailer
in various ways, consistent with the recent literature, such as Caliskan-Demirag et al. [22] and Ho et al. [24]. However, it is obvious that the retailer's profit with the manufacturer considering the retailer's fairness preference improves when the consumer's online acceptance increases. If the manufacturer ignores her fairness preference, the retailer intends to increase profits by increasing her pricing demonstrated in Lemma 5. But this reduces the retailer's sales and results in a decrease in the retailer's profit. Moreover, the retailer's profit decreases faster as the consumer's online channel preference increases. It is easy to understand that the number of consumers choosing to shop offline reduces while the online channel acceptance increases. This decrease in retailer sales leads to a decline in profits if the manufacturer ignores the retailer's fairness preference and does not transfer the earning.

The numerical analysis also shows the above results more intuitively. From Figure 4, if $\theta$ is at a certain value, with the increase of $\alpha$, the retailer's profit with the manufacturer considering her fairness preference improves. However, if $\alpha$ is at a certain value, the retailer's profit decreases with the increase of $\theta$. Similarly, it is obvious from Figure 5 that when $\theta$ is at a certain value, with the increase of $\alpha$, the retailer's profit with the manufacturer ignoring the retailer's fairness preference reduces. If $\alpha$ is at a certain value, the retailer's profit also decreases with the increase of $\theta$.

### 6.3. The Manufacturer's Profits Comparison

Lemma 7. For the above three circumstances,
(a) $\prod_{m}>\prod_{m}^{*}>\prod_{m}^{* *}$;
(b) $\left(\partial \prod_{m}^{*} / \partial \alpha\right)<0, \quad\left(\partial \prod_{m}^{*} / \partial \theta\right)>0, \quad\left(\partial \prod_{m}^{* *} / \partial \alpha\right)<0$, and $\left(\partial \prod_{m}^{* *} / \partial \theta\right)>0$.

Proof of Lemma 7. With formulae (11), (18), and (25), $\prod_{m}-$ $\prod_{m}^{*}>0$ and $\prod_{m}^{*}-\prod_{m}^{* *}>0$ are got. $\left(\partial \prod_{m}^{*} / \partial \alpha\right)=$ $-\left(\left([a(\theta-1)+c(b-\lambda)]^{2}\right) /\left(8 b(1+2 \alpha)^{2}\right)\right)<0$ and $\left(\partial \prod_{m}^{*} /\right.$ $\partial \theta)=\left(\left(a^{2}[\alpha \theta(5 b-3 \lambda)-\alpha(b-3 \lambda)+\theta(3 b-\lambda)-(b-\lambda)]-\right.\right.$ $\left.\left.a c(3 \alpha+1)\left(b-\lambda^{2}\right)\right) /(4 b(1+\alpha)(b+\lambda))\right)>0$ can be obtained when (18) is used to get the derivation of $\alpha$ and $\theta$. Calculating with (25), $\quad\left(\partial \prod_{m}^{* *} / \partial \alpha\right)=-\left(\left([a(\theta-1)+c(b-\lambda)]^{2}\right) / \quad(8 b\right.$ $\left.\left.(1+2 \alpha)^{2}\right)\right)<0$ and $\left(\partial \prod_{m}^{* *} / \partial \theta\right)=\left(\left(a^{2}[\theta(3 b-\lambda)+2 \alpha \lambda+\right.\right.$ $\left.(b-\lambda)(2 \alpha \theta-1)]-a c(2 \alpha+1)\left(b^{2}-\lambda^{2}\right)\right) /(4 b(1+\alpha) \quad(b+$ $\lambda)))>0$.

Lemma 7 shows that whether the manufacturer considers the retailer's fairness preference or not, the manufacturer's profit will be lower than the case when the retailer has no fairness preference. When the manufacturer thinks over the retailer's fairness preference, he will reduce his profit to add the retailer's profit, thus ensuring the fairness of channel profit distribution. Moreover, the manufacturer's profit when he considers the retailer's fairness preference is higher than that when he ignores it. When the manufacturer ignores the retailer's fairness preference, this may enlarge the retailer's fairness concerns and lead to broader damage. This means that if the retailer has a fairness preference, the manufacturer should consider this behavior more favorable.


Figure 4: The influence of $\alpha$ and $\theta$ on $\prod_{r}^{*}$.


Figure 5: The influence of $\alpha$ and $\theta$ on $\prod_{r}^{* *}$.


Figure 6: The influence of $\alpha$ and $\theta$ on $\prod_{m}^{*}$.

Taking the increase of consumer's online channel preference into account would benefit the manufacturer's profit for both cases. However, the manufacturer's profit that increased by the online channel is smaller than the profit


Figure 7: The influence of $\alpha$ and $\theta$ on $\prod_{m}^{* *}$.
transfer for the retailer's fairness preference. When more consumers intend to shop online, this injures the retailer's profit and magnifies her fairness requirements. To ensure the channel operation, the manufacturer can only choose to give the retailer more incentives.

To visually show the analysis in Lemma 7, the paper applies a numerical example. As shown in Figures 6 and 7, it can be seen that whether the manufacturer considers the retailer's fairness concern or not, the manufacturer's profit will reduce with the increase of $\alpha$ but improve with the increase of $\theta$. The main reason is that the enhancement of fairness preference makes the manufacturer have to give up profit; otherwise, it is difficult to carry out dual-channel cooperation. On the contrary, enhancing consumers' online acceptance enables the manufacturer to obtain more profit from an online channel, leading to an increased profit in the whole dual channels.

## 7. Conclusions

This paper first attempts to study the compound impact of the retailer's fairness preference and consumer's online channel preference on supply chain enterprises' operational strategy. A single manufacturer sells his products by online and offline channels with a retailer in the dual-channel supply chain. To observe supply chain enterprises' operational decisions, this paper uses three circumstances. The benchmark one is that the retailer has no fairness preference, while the manufacturer considers or ignores the retailer's fairness preference as comparative cases. The study shows that the manufacturer intends to build an online channel with lower consumer acceptance when the manufacturer takes the retailer's fairness preference into account. The wholesale price and manufacturer's profit are lower than those in the benchmark case, while the increase in consumer's online channel preference can narrow the gap. The retailer's profit is higher than that without fairness preference because of the manufacturer's interest concession. When the manufacturer ignores the retailer's fairness preference, the retailer will increase offline channel pricing, which hurts the manufacturer's and the retailer's profits,
although the increase of consumer's online channel acceptance benefits the manufacturer. These results demonstrate that the manufacturer needs to take fairness preference into account to avoid a "lose-lose" situation when the retailer possesses a fairness preference. If the acceptance degree of the online channel is high to a certain extent, the manufacturer needs to provide additional revenue for the retailer to ensure the operation of the dual-channel strategy. For the retailer, her fairness preference needs to be maintained at an appropriate level; otherwise, the manufacturer will start online channel construction earlier, which will affect his profit level.

This paper focuses on the retailer's fairness preference, but, in practical operation, all decision-makers may possess fairness preferences, and the fairness preference could be extended to a more elaborate setting. The market demand is determined to be a linear demand in the study. However, the market demand is random and uncertain or changes to dynamic, which also deserves further study.

## Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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# Thermoelastic Characteristics of Functionally Graded Circular 

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#### Abstract

A rotating functionally graded circular disk undergoing a contact load is taken into account to investigate the thermoelastic characteristics. By considering contact force, a pair of partial differential equations is induced as the governing equations based on Hooke's law. The behavior of circular disk modes is described with the variations of contact force and homogeneous thickness. A finite volume model is introduced to obtain approximate solutions for the governing equations because of the complexity of the equations. Contact force is highly influential in the radial direction compared to the circumferential direction in the displacement distribution, while a large radial stress appears near the area of the contact point. In the strain distribution, the magnitude increases as the angle grows near part of the outer boundary in the circular domain. The radial distribution profiles are susceptible to the growth of contact force in nearby area of the outer boundary, whereas the influences on the circumferential direction profiles are trivial. The increase of homogeneous thickness dwindles the radial magnitude of displacement, stress, and strain distribution profiles over nearby area of the outer boundary of the circular domain. As a result, numerical approach demonstrates that contact force and homogeneous thickness are indispensable parameters and provide deep influence on the thermoelastic movements of a rotating circular disk. Thus, the results obtained may be useful to design an appropriate FGM circular disk model for the industrial area by controlling the above parameters.


## 1. Introduction

The analysis of thermoelastic characteristics of a rotating circular disk has been a challenging work due to complex processes generating during operations. Friction, wear, heat generation, and temperature deformation are representative processes developed during the load of a contact force. By the operation of a rotating circular disk, all parameters are connected mutually and complicated interactions progress. The thermoelastic characteristics of a circular disk are under complicated interaction influences and the variation of each parameter may not be ignorable to determine the thermoelastic movements. Thus, the influences of temperature, contact pressure, and blushed wear on the rotating circular disk are interesting and valuable for research.

Ceramic is a typical material used at one surface to resist severe environmental effects such as high temperature, wear,
and friction. The other surface is composed of a metal to ensure higher toughness and thermal conductivity. The complementary functions of each material invoke the introduction of functionally graded materials (FGMs). FGMs are nonhomogeneous composites composed of two or more different material phases and the volume fractions continuously vary with space variables. Thus, FGMs are structures involving successively varying mechanical and thermal properties from one surface to another surface. FGMs were introduced to prosper thermal barrier coatings for propulsion system of space planes to endure high temperature and ensure high thermal conductivity at the beginning [1]. The graded material property provides improved bonding strength in various applications undergoing a high temperature gradient field.

On the pursuit superior properties such as strength, impact toughness, and fiber resistance, fiber metal laminates (FMLs) have drawn attention as promising hybrid
composite materials in last few decades. FMLs are composed of metallic and composite materials. The combination of fiber-reinforced composite layers and composite materials creates FMLs, which exhibit excellent durability. Thus, FMLs are applicable in various industries and are used in aircraft, wind turbine, and automobile. A laminated composite made with thin sheet/grass was investigated by BnvsGaneshGupta et al. [2] for the flexural properties. They evaluated flexural strength and failure analysis using stereo microscope. Zimmermann and Wang [3] analyzed failure modes and characteristics of adhesively and mechanically fastened joints in aircraft composite materials.

Meanwhile, the shape of circular disk is commonly employed for practical use in various industrial applications, and the physical behavior of circular disk models is a matter of interest to many engineering researchers. Obata and Noda [4] and Liew et al. [5] analyzed the thermal stresses of FGM circular cylinder and hollow sphere. Afsar et al. [6] investigated the brittle fracture characteristics of thick-walled FGM cylinders by considering incompatible eigenstrain, which is a cause of thermal residual stress. The characterization that oscillatory movements reduce cutting forces in operating process of a circular disk cutter was presented by Kovalyshen [7], and Huang and Li [8] studied the radial component effects on the bending of thin circular plates using Kirchhoff plate theory. The angular speed of a rotating circular disk varies during the start and stop process in machine operation. Dai et al. [9] investigated the displacement and stress fields of a FGM hollow circular disk subjecting an angular acceleration due to a changing temperature. Static behavior of functionally graded auxetic-porous structures was presented by Rad [10] considering torsional interaction and horizontal friction force. The static response of nonuniform bidirectional functionally graded auxetic-porous material circular plates is carried out based on poroelasticity theory. Friction induced a vibration in rotary contact system, and the vibration causes a misalignment problem. Tadokoro et al. [11] described the stabilizing effect arising from parallel misalignment in circular sliding contact. Infinitesimal theory of plane elasticity and Complementary Functions Method was employed to investigate hydrogen-induced stresses in functionally graded axisymmetric spheres, cylinders, and disks by Yildirim [12]. Sondhi et al. [13] used element-based gradation of material properties over the discretized domain to study linear elastic stress analysis of functionally graded rotating disks.

However, the comprehensive understanding and quantifying for the thermoelastic field of a rotating FGM circular disk may be useful in design FGM cutter or a grinding disk. In general, FGM cutter or a grinding disk experiences the thermal load and inertia force due to rotation of the disk. The previous study [14] demonstrated that proper control of some parameters such as temperature distribution, angular speed, and radial thickness improves thermoselastic characteristics in the design of a circular or grinding disk and reduces failure mechanism such as crack and brittle, but most of previous works progressed their studies without considering contact load even though contact load is a crucial parameter in determination of thermoelastic characteristics. The present study focuses on the analysis of thermoelastic characteristics of a
rotating $\left(\mathrm{Al}_{2} \mathrm{O}_{3} / \mathrm{Al}\right)$ FGM disk subjected to a contact force. Young's modulus, CTE, and density of the FGM disk are assumed to vary exponentially only in the radial direction due to symmetry with respect to the axis of the disk, whereas Poisson's ratio is assumed to be constant because of insignificant effect on the thermoelastic characteristics of the disk. The research addresses the following: (i) the contact force and homogeneous thickness are crucial factors in the determination of the thermoelastic characteristics of FGM circular models, (ii) the radial direction is more susceptible to the variation of the magnitude of contact force and homogeneous thickness, and (iii) the variation of homogeneous thickness is highly influential to thermoelastic characteristics over the area of $0.2<(r-$ $a) /(b-a)<0.8$ in FGM circular domain.

The process for the mathematical approach is presented in Section 2. A pair of partial differential equations is derived based on Hooke's law in Section 2.1. Due to the complexity of the governing equation, a finite volume method is introduced in Section 2.2 to obtain numerical solutions. In Section 3, numerical solutions are displayed and the elastic characteristics are discussed according to the results obtained. The significances of the results are addressed in the conclusions in Section 4.

## 2. Mathematical Modeling

A rotating FGM circular disk with a concentric circular hole is considered (see Figure 1). The origin of the polar coordinate system $r-\theta$ is assumed to be located at the center of the disk and hole. $A$ and $B$, denoted by the dark and the white colors, represent constituent materials of FGM circular disk, as shown in the figure. The distribution of each material varies continuously along the radial direction only. The radii of the hole and outer surface of the disk are designated by $a$ and $b$. The angular velocity $\omega$ can be determined from the relation $\omega=(2 \pi N / 60), N$ represents the revolutions per minute (rpm). Since the material distributions vary along the radial direction only, the FGM disk can be reduced to an axisymmetric problem. Thus, all properties of the disk can be treated as a function of $r$ only. Young's modulus, coefficient of thermal expansion, and the density of the disk are denoted by $E$, $\alpha$, and $\rho$, respectively, and each property is assumed to vary exponentially as follows (for various and detailed distribution profiles, see [15]):

$$
\begin{gather*}
E=E_{0} e^{\beta r},  \tag{1a}\\
\alpha=\alpha_{0} e^{\xi r},  \tag{1b}\\
\rho=\rho_{0} e^{\mu r} \tag{1c}
\end{gather*}
$$

By the assumption that the disk is composed of $100 \%$ material $A$ at the surface of the hole $(r=a)$ and $100 \%$ material $B$ at the outer surface ( $r=b$ ), the constants in equations (1a)-(1c) can be determined as follows:

$$
\begin{align*}
& E_{0}=E_{A} e^{-\beta a}  \tag{2a}\\
& \alpha_{0}=\alpha_{A} e^{-\xi a} \tag{2b}
\end{align*}
$$



Figure 1: (a) Schematic diagram of functionally graded circular disk models. (b) Notations of finite control volumes.

$$
\begin{align*}
\rho_{0} & =\rho_{A} e^{-\mu a}  \tag{2c}\\
\beta & =\frac{1}{a-b} \ln \left(\frac{E_{A}}{E_{B}}\right),  \tag{3a}\\
\xi & =\frac{1}{a-b} \ln \left(\frac{\alpha_{A}}{\alpha_{B}}\right),  \tag{3b}\\
\mu & =\frac{1}{a-b} \ln \left(\frac{\rho_{A}}{\rho_{B}}\right) \tag{3c}
\end{align*}
$$

The subscripts $A$ and $B$ represent the properties of the constituent materials $A$ and $B$, respectively, but the nonsubscripted variables are used to denote the properties of FGM composed of materials $A$ and $B$.
2.1. Thermoelastic Formulation. In the polar coordinate, the relations between the strain components and deformation components are as follows:

$$
\begin{aligned}
\varepsilon_{r} & =\frac{\partial u}{\partial r} \\
\varepsilon_{\theta} & =\frac{1}{r} \frac{\partial v}{\partial \theta}+\frac{u}{r}, \\
\varepsilon_{z} & =0, \\
\gamma_{r \theta} & =\frac{1}{r} \frac{\partial u}{\partial \theta}+\frac{\partial v}{\partial r}-\frac{v}{r}, \\
\gamma_{z \theta} & =0, \\
\gamma_{r z} & =0 .
\end{aligned}
$$

The plane stress exposed with thermal expansion, by Hooke's law, produces the strain-stress relations undergoing thermal expansion ([16], p8):

$$
\begin{align*}
\varepsilon_{r} & =\frac{1}{E}\left[\sigma_{r}-v \sigma_{\theta}\right]+\alpha T \\
\varepsilon_{\theta} & =\frac{1}{E}\left[\sigma_{\theta}-v \sigma_{r}\right]+\alpha T \\
\tau_{r \theta} & =\frac{E}{2(1+v)} \gamma_{r \theta}  \tag{5}\\
\tau_{\theta z} & =0 \\
\tau_{r z} & =0
\end{align*}
$$

$T(r)$ is the function of temperature variation with radius variable $r$. The temperature function can be expressed as

$$
\begin{equation*}
T(r)=c_{1} \ln r+c_{2} \tag{6}
\end{equation*}
$$

under the assumption that the circular disk is subjected by a loading of symmetric temperature to the radial direction only [17]. $c_{1}$ and $c_{2}$ will be determined based on the boundary conditions. The equilibrium equations in polar coordinates are

$$
\begin{align*}
\frac{\partial \sigma_{r}}{\partial r}+\frac{1}{r} \frac{\partial \tau_{r \theta}}{\partial \theta}+\frac{\sigma_{r}-\sigma_{\theta}}{r}+\rho \omega^{2} r & =0  \tag{7a}\\
\frac{\partial \sigma_{\theta}}{\partial \theta}+r \frac{\partial \tau_{r \theta}}{\partial r}+2 \tau_{r \theta} & =0 \tag{7b}
\end{align*}
$$

and the algebra among equations (4), (5), (7a), and (7b) yields the following governing equations:

Table 1: Mechanical and thermal properties used for analyzing thermoelastic characteristics of rotating CM circular disks.

| Material/property | Elastic modulus $(\mathrm{MPa})$ | Thermal expansion coefficient $\left(10^{-6} /{ }^{\circ} \mathrm{C}\right)$ | Density $\left(\mathrm{g} / \mathrm{cm}^{3}\right)$ |
| :--- | :---: | :---: | :---: |
| Substrate $(\mathrm{Al})$ | 71 | 23.1 | 2.70 |
| Bond coating $\left(\mathrm{Al}^{\circ} / \mathrm{Al}_{2} \mathrm{O}_{3}\right)$ | 164.3 | 13.6 | 1.61 |
| Top $\left(\mathrm{Al}_{2} \mathrm{O}_{3}\right)$ | 380 | 8.0 | 0.96 |



Figure 2: Stress distribution profiles in the (a) radial direction and (b) circumferential direction.

$$
\begin{align*}
& \frac{\partial}{\partial r}\left(r \frac{\partial u}{\partial r}\right)+\frac{1-v}{2 r} \frac{\partial^{2} u}{\partial \theta^{2}}-\frac{3-v}{2 r} \frac{\partial v}{\partial \theta}+\frac{1+v}{2} \frac{\partial^{2} v}{\partial r \partial \theta}-\frac{u}{r}  \tag{8a}\\
& \quad+\frac{1-v^{2}}{E} \rho \omega^{2} r^{2}=(v+1) \alpha r \frac{\mathrm{~d} T(r)}{\mathrm{d} r}, \\
& \frac{1-v}{2} \frac{\partial}{\partial r}\left(r \frac{\partial v}{\partial r}\right)+\frac{1}{r} \frac{\partial^{2} v}{\partial \theta^{2}}+\frac{3-v}{2 r} \frac{\partial u}{\partial \theta}+\frac{1+v}{2} \frac{\partial^{2} u}{\partial r \partial \theta}-\frac{1-v}{2} \frac{v}{r}=0 . \tag{8b}
\end{align*}
$$

The governing equations (8a) and (8b) are a pair of partial differential equations in the polar coordinate system expressed with the radial and circumferential displacement distribution profiles under the loading of temperature. The thermoelastic characteristics of a functionally graded rotating circular disk subjected to a contact force are investigated according to the boundary conditions:

$$
\begin{array}{r}
\sigma_{r}(a, \theta)=0, \\
\sigma_{r}(b, \theta-\{0\})=0,  \tag{9}\\
\sigma_{r}(b, 0)=P,
\end{array}
$$

at a contact point.
The boundary condition implies that the circular model is under the loading of thermal expansion over the entire domain and of radial pressure $P$ at the outer boundary contact point $(b, 0)$.
2.2. Finite Volume Formulation. Due to the complexity of the governing equations, a numerical technique is required, and a finite volume method is applied for approximation. The domain is divided up into control volume and integrates the field equations over each control volume (see Figure 1(b)). Equations (8a) and (8b) thus are discretized as follows [18]:

$$
\begin{align*}
& A_{11} u_{i+1, j}+A_{12} u_{i, j+1}+A_{13} u_{i, j}+A_{14} u_{i, j-1}+A_{15} u_{i-1, j} \\
& \quad+B_{11} v_{i, j+1}+B_{12} v_{i, j}+B_{13} v_{i, j-1}+B_{14} v_{i-1, j+1}+B_{15} v_{i-1, j}+B_{16} v_{i-1, j-1}+B_{17} v_{i-2, j+1}+B_{18} v_{i-2, j}+B_{19} v_{i-2, j-1}=f_{i, j}, \\
& A_{21} u_{i, j+1}+A_{22} u_{i, j}+A_{23} u_{i, j-1}+A_{24} u_{i-1, j+1}+A_{25} u_{i-1, j}+A_{26} u_{i-1, j-1}+A_{27} u_{i-2, j+1}+A_{28} u_{i-2, j}+A_{29} u_{i-2, j-1} \\
& \quad+B_{21} v_{i+1, j}+B_{22} v_{i, j+1}+B_{23} v_{i, j}+B_{24} v_{i, j-1}+B_{25} v_{i-1, j}=0 \tag{10}
\end{align*}
$$

$$
\begin{aligned}
& A_{11}=\frac{1-v}{2} \frac{\Delta r}{\Delta \theta} \frac{1}{r_{i, j}}, \\
& A_{12}=\frac{\Delta \theta}{\Delta r} r_{i, j+(1 / 2)} \text {, } \\
& A_{13}=-\frac{\Delta \theta}{\Delta r}\left(r_{i, j+(1 / 2)}+r_{i, j-(1 / 2)}\right)-(1-v) \frac{\Delta r}{\Delta \theta} \frac{1}{r_{i, j}}-\Delta r \Delta \theta \frac{1}{r_{i, j}}, \\
& A_{14}=\frac{\Delta \theta}{\Delta r} r_{i, j-(1 / 2)}, \\
& A_{15}=\frac{1-v}{2} \frac{\Delta r}{\Delta \theta} \frac{1}{r_{i, j}}, \\
& B_{11}=\frac{3}{8}(1+v) \text {, } \\
& B_{12}=-\frac{3}{4}(3-v) \Delta r \frac{1}{r_{i, j}}, \\
& B_{13}=-\frac{7}{16}(1+v) \text {, } \\
& B_{14}=-\frac{1}{2}(1+v) \text {, } \\
& B_{15}=\frac{5}{2}(3-v) \frac{1}{r_{i, j}}, \\
& B_{16}=\frac{9}{16}(1+v) \text {, } \\
& B_{17}=\frac{1}{8}(1+v) \text {, } \\
& B_{18}=-\frac{3}{4}(3-v) \Delta r \frac{1}{r_{i, j}}, \\
& B_{19}=-\frac{1}{8}(1+v) \text {, } \\
& f_{i, j}=(1+v) \alpha r_{i, j} \Delta \theta\left(T_{i, j+(1 / 2)}-T_{i, j-(1 / 2)}\right)-\Delta r \Delta \theta \rho \omega^{2} \frac{1-v^{2}}{E} r_{i, j}^{2}, \\
& A_{21}=\frac{3}{8}(1+v) \text {, } \\
& A_{22}=\frac{3}{4}(3-v) \Delta r \frac{1}{r_{i, j}}, \\
& A_{23}=-\frac{7}{16}(1+v) \text {, } \\
& A_{24}=-\frac{1}{2}(1+v) \text {, } \\
& A_{25}=-(3-v) \Delta r \frac{1}{r_{i, j}}, \\
& A_{26}=\frac{9}{16}(1+v) \text {, } \\
& A_{27}=\frac{1}{8}(1+v) \text {, } \\
& A_{28}=(3-v) \Delta r \frac{1}{r_{i, j}}, \\
& A_{29}=-\frac{1}{8}(1+v) \text {, } \\
& B_{21}=\frac{\Delta r}{\Delta \theta} \frac{1}{r_{i, j}}, \\
& B_{22}=\frac{\Delta \theta}{\Delta r} r_{i, j+(1 / 2)},
\end{aligned}
$$

$$
\begin{align*}
& B_{23}=-\frac{\Delta \theta}{\Delta r}\left(r_{i, j+(1 / 2)}+r_{i, j-(1 / 2)}\right)-\frac{1-v}{2} \Delta r \Delta \theta \frac{1}{r_{i, j}}-2 \frac{\Delta r}{\Delta \theta} \frac{1}{r_{i, j}} \\
& B_{24}=\frac{\Delta \theta}{\Delta r} r_{i, j-(1 / 2)}  \tag{11}\\
& B_{25}=\frac{\Delta r}{\Delta \theta} \frac{1}{r_{i, j}}
\end{align*}
$$

The validation and convergence of the above finite volume method (FVM) were presented in [17], and Mathematica (version 5.0) is used to obtain numerical solutions.

## 3. Numerical Results and Discussion

In this section, the differential equation induced in Section 2.1 and the finite volume formula developed in Section 2.2 are used to evaluate the approximated results of different components of displacement, stress, and strain for an $\left(\mathrm{Al}_{2} \mathrm{O}_{3} / \mathrm{Al}\right)$ FGM circular disk. The mechanical and thermal properties of these ingredient materials are shown in Table 1.

Stress distribution profiles are presented in Figure 2. As shown in Figure 2(a), the circular disk is undergoing tensile radial stress. All areas of the disk except near outer boundary experience trivial influence in radial stress distribution, whereas the contact point suffers from the largest radial stress. But different phase develops in circumferential stress. At normalized values $(r-a) /(b-a)=0.1$ and 0.5 , the magnitude of circumferential stresses is ignorable in comparison with circumferential stress at $(r-a) /(b-a)=0.9$ (see Figure 2(b)). In addition, the entire area of the disk is under the loading of tensile circumferential stress at $(r-a) /(b-a)=0.9$ and the circumferential stress distribution exhibits exponential decay with the increase of angle. The strain distribution profiles are described in Figure 3. Near area of outer boundary in circular domain presents that the magnitude increases with the increase of angle value in the radial strain distribution (see Figure 3(a)). Positive radial strain distribution appears over the disk, and the largest radial strain value occurring at the contact point generates. However, the circumferential strain distribution shows different pattern (see Figure 3(b)). At $(r-a) /(b-a)=0.9$, the circumferential strain deploys to the positive direction, while negative circumferential strain distributions are obtained at $(r-a) /(b-a)=0.1$ and 0.5 . The changing shape of the circumferential strain is a parabola as the angle increases at $(r-a) /(b-a)=0.1$ and 0.5 , whereas the circumferential strain distribution profile fluctuates at the beginning and logarithmic decay appears after $\theta=2$ at $(r-a) /(b-a)=0.9$. The results inform that (i) contact force is highly influential to near area of contact point in radial direction, (ii) the circumference stress is more sensitive to contact force than the radial stress distribution, and (iii) various phenomena appear in circumferential strain distribution depending on the value of $(r-a) /(b-a)$.

Figure 4 presents the influences of contact force on the stress distribution by choosing the representative angles $\theta=$ 0 and 180 and the normalized radius $(r-a) /(b-a)=0.1$
and 0.9. The radial stress distribution is susceptible to the change of contact force, especially near the area of contact point (see Figures 4(a) and 4(b)). Entire domain of the circular disk suffers from a tensile stress and the magnitude of the radial stress is getting larger as the contact force increases. As shown in Figure 4(c), the area near the inner boundary of the disk sits loose to the variation of contact force, but the area near the outer boundary reacts sensitively to the change of contact force, and noticeable impact appears when $\theta=0$ (see Figure $4(\mathrm{~d})$ ). By the increase of contact force, the radial and circumferential stresses exhibit exponential growth near area of contact point, even though minor effects appear over entire circular domain.

The effects of contact force on the strain distributions are described in Figure 5. The reactions of the radial strain distribution to the change of contact force are similar to those of radial stress distribution (see Figures 5(a) and 5(b)). Nearby area of the outer boundary is particularly susceptible to the increase of contact force and the magnitude of the radial strain grows with the increment of contact force. Trivial reaction appears in the circumferential strain distribution to the variation of contact force for $(r-a) /(b-a)=0.1$ (see Figure 5(c)), whereas nearby part of the outer boundary is sensitive to the change of contact force (see Figure 5(d)). Pronounced alterations in the circumferential strain distribution appear on area around $\theta=0$. The numerical solutions imply that (i) the magnitude of contact force is a critical factor to determine the thermoelastic characteristics of circular FGM models and (ii) the radial direction is more susceptible to the variation of the magnitude of contact force.

The influences of homogeneous thickness on thermoelastic characteristics, denoted by H , are explained through Figures 6-8. $\mathrm{H}=1$ implies that the circular disk is composed of homogeneous material. The representative values $\mathrm{H}=0.25,0.5$, and 0.75 are applied. Figure 6 expresses the effects of H on the displacement distributions.

The area between normalized values $(r-a) /(b-a)=$ 0.2 and $(r-a) /(b-a)=0.85$ reveals noticeable reactions to homogeneous thickness variation. The magnitude of the radial displacement increases with the growth of homogeneous part (see Figures 6(a) and 6(b)), but nearby area of the outer boundary displays that the magnitude to the positive direction decreases as the value of H increases. The results imply that the growth of functionally graded area in the circular disk yields more stability in the radial displacement distribution profiles.

As shown in Figures 6(c) and 6(d)), minor influence appears in the circumferential displacement except nearby area of contact force. The circumferential displacement


Figure 3: Strain distribution profiles in the (a) radial direction and (b) circumferential direction.


Figure 4: Effects of contact force on the stress in the radial direction (a) at $\theta=0$ and (b) at $\theta=180$ and in the circumferential direction (c) at $(r-a) /(b-a)=0.1$ and $(\mathrm{d})$ at $(r-a) /(b-a)=0.9$.


--- $P=500$
$\cdots P=700$
(a)

(c)


$$
\begin{array}{ll}
-P=100 & ---P=500 \\
\cdots P=300 & \cdots-P P=700
\end{array}
$$

(b)

(d)

Figure 5: Effects of contact force on the strain in the radial direction (a) at $\theta=0$ and (b) at $\theta=180$ and in the circumferential direction (c) at $(r-a) /(b-a)=0.1$ and $(\mathrm{d})$ at $(r-a) /(b-a)=0.9$.


Figure 6: Continued.


Figure 6: Effects of homogeneous thickness on the displacement in the radial direction (a) at $\theta=0$ and (b) at $\theta=180$ and in the circumferential direction (c) at $(r-a) /(b-a)=0.1$ and (d) at $(r-a) /(b-a)=0.9$.


Figure 7: Continued.


Figure 7: Effects of homogeneous thickness on the stress in the radial direction (a) at $\theta=0$ and (b) at $\theta=180$ and in the circumferential direction (c) at $(r-a) /(b-a)=0.1$ and (d) at $(r-a) /(b-a)=0.9$.



- $H=0.25$
- $H=0.5$

$$
--H=0.75
$$


(b)

Figure 8: Continued.


Figure 8: Effects of homogeneous thickness on the strain in the radial direction (a) at $\theta=0$ and (b) at $\theta=180$ and in the circumferential direction (c) at $(r-a) /(b-a)=0.1$ and (d) at $(r-a) /(b-a)=0.9$.
increases with growth of the value of H over the area around contact force.

Figure 7 presents the effects of H on the stress distributions. The radial distribution exhibits smaller tensile stress over the nearby area of outer boundary with increase of the value of H when $\theta=0$ (see Figure 7(a)). But the circular disk undergoes larger compressive radial stress over area between normalized values $\quad(r-a) /(b-a)=0.1 \quad$ and $(r-a) /(b-a)=0.8$ when $\theta=180$ (see Figure 7(b)). According to the radial stress distributions, the circular disk with larger functionally graded area may not guarantee higher stability over nearby area of outer boundary. Different phase develops in circumferential stress distributions. Over nearby area of the inner boundary, entire domain of the disk experiences compressive circumferential stress for the values of $\mathrm{H}=0.5$ and 0.75 , whereas some part of the disk with $\mathrm{H}=0.25$ undergoes tensile circumferential stress (see Figure 7(c)). Over nearby area of the outer boundary of the disk, the circumferential stress with $\mathrm{H}=0.75$ is trivial in comparison with the value of $\mathrm{H}=0.25$ and 0.5 (see Figure 7(d)). The magnitude of the circumferential stress is higher near contact point as the homogeneous thickness decreases. Based on the circumferential stress distributions, greater functionally graded area affords higher stability in the circular disk. The influences of homogeneous thickness H on the strain distributions are described in Figure 8. Nearby area of the outer boundary of the disk is sensitive to the change of homogeneous thickness in the radial strain distributions. The magnitude of strain decreases with the growth of the value of H (see Figures 8(a) and 8(b)), which signifies that the circular disk with larger homogeneous area is more stable over nearby area of the outer boundary. Similar movements occur in the circumferential strain. As shown in Figures 8(c) and 8(d), the magnitude of the circumferential strain is getting larger according to the decline
of the value of H . The outcomes describe that the circular disk stability with greater functionally graded area is greater in the circumferential strain distribution. The numerical approaches make point of the following: (i) the variation of homogeneous thickness is highly influential to thermoelastic characteristics over the area of $0.2<(r-a) /(b-a)<0.8$ and (ii) the radial direction is more susceptible to the change of homogeneous thickness.

## 4. Conclusions

A rotating FGM circular disk with a concentric circular hole is taken into account to investigate the thermoelastic characteristics. The circular disk is under the loading of a contact force. The effects of the variation of contact force and homogeneous thickness on the components of displacement, stress, and strain have been described. Over nearby area of the contact point, the thermoselastic behavior reacts sensitively to the change of contact force and exhibits exponential growth in the magnitude, whereas the influence of contact force over the other area is greater in the circumferential stress distributions. The outcomes explain that the outer boundary area is exposed to become brittle with the growth of a contact force. In the radial displacement distribution, the growth of functionally graded area grants more stability in the circular disk, while, according to the radial stress distributions, the circular disk with larger functionally graded area may not guarantee higher stability over nearby area of outer boundary. Based on the circumferential distribution profiles, the circular disk with greater functionally area evinces higher stability. Overall, the results demonstrate that contact force and homogeneous thickness are crucial parameters to determine the thermoelastic characteristics of FGM circular disks, and the movements of an FGM circular disk can be controlled by
controlling of these parameters. Therefore, a FGM circular cutter or grinding disk suffering from the loading of a contact force can be designed applying the finite volume analysis used in this investigation to promote proper and reliable thermoelastic characteristics in service.

## Nomenclature

$u$ : Radial displacement component (mm)
$v$ : Circumferential displacement component (mm)
$\varepsilon_{r}$ : Radial strain
$\varepsilon_{\theta}$ : Circumferential strain
$\gamma$ : Shearing strain
$\sigma_{r}:$ Radial stress (MPa)
$\sigma_{\theta}:$ Circumferential stress (MPa)
$\tau$ : Shearing stress
$v$ : Poisson's ratio
$\omega$ : Angular velocity
$N$ : Revolutions per minute (rpm)
E: Young's modulus (MPa)
$\alpha$ : Coefficient of thermal expansion $\left(10^{-6} /{ }^{\circ} \mathrm{C}\right)$
$\rho$ : Density of the disk $\left(\mathrm{g} / \mathrm{cm}^{3}\right)$.

## Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

## Conflicts of Interest

The author declares that there are no conflicts of interest.

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#### Abstract

This article is concerned about the planed rigid body pendulum motion suspended with a spring which is suspended to move on a vertical plane moving uniformly about a horizontal $X$-axis. This model depends on a system containing three generalized coordinates. The three nonlinear differential equations of motion of the second order are obtained to the elastic string length and the oscillation angles $\varphi_{1}$ and $\varphi_{2}$ which represent the freedom degrees for the pendulum motions. It is assumed that the body moves in a rotating vertical plane uniformly with an arbitrary angular velocity $\omega$. The relative periodic motions of this model are considered. The governing equations of motion are obtained using Lagrange's equations and represent a nonlinear system of second-order differential equations that can be solved in terms of generalized coordinates. The numerical solutions are investigated using the approximated fourth-order Runge-Kutta method through programming packages. These solutions are represented graphically to describe and discuss the behavior of the body at any instant for different values of the different physical parameters of the body. The obtained results have been discussed and compared with some previously published works. Some concluding remarks have been presented at the end of this work. The value of this study comes from its wide applications in both civil and military life. The main findings and objectives of the current study are obtaining periodic solutions for the problem and satisfying their accuracy and stabilities through the numerical procedure.


## 1. Introduction

The pendulum motion is studied by many outstanding scientists in the last century due to the wide application of this problem in applied mathematics, physics, and engineering. In [1], El-Barki and others studied the rotary motion of a pendulum model about an elliptic path. They described the problem dynamically and then deduced the equations of motion for this model using Lagrange's equation. The authors defined a small parameter that depends on the different parameters of the moving model. They solved the problem analytically using the small parameter technique and numerically using the Runge-Kutta method to make a comparison between the two sets of solutions. This comparison proved the validity of both obtained solutions.

Ismail in [2] presented a case of relative periodicity motion of a rigidity pendulum model in presence of
multidegrees of freedom. He described the motion dynamically and used the Lagrangian function to describe the motion. The system of equations of motion is obtained. He defined a small parameter and used the small parameter technique to find the approximated periodic solutions of the obtained motion. He achieved computer programs through numerical consideration for proving the validity of the obtained solutions. In [3], the author studied the periodic solutions of a pendulum in a relative case. This case is considered as an especial one from the problem in [2]. The author used Poincare's method to find the approximated solutions. In [4], the author studied the oscillated motion of a simple pendulum model. He used the Lagrangian function for deriving the equations of motion. The processing method of analysis is used to find approximated solutions of the second order. In [5], the elastic pendulum oscillations are given by Vitt and Gorelik in 1933. They give an example of oscillated linear systems with two parametrical couples.

In [6], Lynch presented the three dimensions of elastic pendulum motions in the resonant case. He used the Lagrangian function for describing the motion. In [7], Holm and others studied a resonant elastic pendulum in the case of stepwise precession. In [8], the authors studied the motion of a harmonically excited elastic pendulum in the chaos response case. They derived the equations of motion of the pendulum model using Lagrange's equations. In [9], Amer described the dynamical oscillations of an elastic rigid pendulum in a plane to the equilibrium position. The author considered the plane rotates about the downward vertical fixed axis with uniform velocity. He used the Lagrangian function to deduce the equations of motion of the model. The numerical considerations [10] are considered using one of the numerical methods for searching the accuracy of the solutions.

The phrase diagram procedure is used for studying the stability of the solutions [11]. In [12], Brearley studied a simple pendulum model when its string length is changing uniformly. In [13], Pinsky and others studied the oscillated pendulum model for swing with a length which varies periodically.

Nayfeh in [14] presented many perturbation techniques for solving a lot of problems in mathematics, physics, and engineering. Such techniques are named the multiple scales, small parameter, KBM, processing analysis, and finite element method which are used in solving most of the previous problems. None of the authors thought about the use of the large parameter technique which gives accuring results for the required solutions. In [15], the two freedom degrees motion of a dynamic nonlinear model for an elastic damped pendulum in the inviscid flow of fluid was considered. The system for equations of motion was considered applying the Lagrangian function. The multiple scales technique is used for solving such equations to obtain the approximated solutions. The cases of resonance and steady state were investigated. The graphical representations of the motion were considered to show the behavior of the motion. The stabilities of the motion were studied. In [16, 17], the restricted motion for the harmonically damped elastic pendulum motion of a rigid body in the elliptic path was investigated when the damped coefficients are linear. In [18], the near resonance pendulum motions in the presence of a tunned absorber dynamical model system were considered. The authors in [19] studied the pendulum motion of a rigid body which moves in a plane with a constant angular velocity $\omega$ attached to a damped spring. The obtained solutions are analyzed numerically through computerized programs for showing motion behavior.

In this paper, a new problem is given for the elastic rigid pendulum motion in a vertical plane which rotates about a horizontal fixed axis in space by a uniform angular velocity $\omega$. The importance of this motion comes from its wide applications in physics, engineering, and other fields.

## 2. Formulation of the Problem

In this section, the motion of an elastic pendulum model is considered which consists of a rigid body suspended with a massless spring at point $O_{2}$ which is suspended from the other hand by point $O_{1}$, see Figure 1. Let the coordinate system $O X Y$ rotate about its horizontal axis $O X$ with a uniform angular velocity $\omega$ relative to the pendulum motion. Consider $O O_{1}=h \cos \omega t$ at any instant of the time $t$ such that at $t=0, O O_{1}=h$. Let the point $C$ represent a mass center of the body, $\varphi_{1}$ represent the angle between $O_{1} Y_{1}$ and $O_{1} C$, and $\varphi_{2}$ denote the angle between $O_{2} C$ and the vertical. Assuming $C \xi, C \eta$, and $C \zeta$ are the principal axes of inertia of the body such that $C \zeta$ is perpendicular to the plane $O X Y$.

Thus, the mass center $\left(x_{C}, y_{C}\right)$ of the body to the system OXY becomes

$$
\begin{align*}
& x_{C}=h \cos \omega t+\rho \sin \varphi_{1}+a \sin \varphi_{2}  \tag{1}\\
& y_{C}=\rho \cos \varphi_{1}+a \cos \varphi_{2}, a=O_{2} C
\end{align*}
$$

where $\rho$ is the elastic string length.
The potential and kinetic energies $V$ and $T$ are given as

$$
\begin{align*}
V= & 0.5 k^{2}(\rho-\ell)^{2}-m g\left(\rho \cos \varphi_{1}+a \cos \varphi_{2}\right), \\
T= & 0.5 m\left\{(-h \omega \sin \omega t)^{2}+\dot{\rho}^{2}+\left(\rho \dot{\varphi}_{1}\right)^{2}+\left(a \dot{\varphi}_{2}\right)^{2}\right. \\
& -2 h \omega\left[\dot{\rho} \sin \varphi_{1} \sin \omega t+\rho \dot{\varphi}_{1} \sin \omega t \cos \varphi_{1}+a \dot{\varphi}_{2} \sin \omega t \cos \varphi_{2}\right] \\
& +2 a\left[\dot{\rho} \dot{\varphi}_{2} \sin \left(\varphi_{1}-\varphi_{2}\right)+\dot{\varphi}_{1} \dot{\varphi}_{2} \cos \left(\varphi_{1}-\varphi_{2}\right)\right] \\
& +\omega^{2}\left(h \cos \omega t+\rho \sin \varphi_{1}+a \sin \varphi_{2}\right)^{2}+m^{-1} I_{3} \dot{\varphi}_{2}^{2} \\
& \left.+m^{-1} \omega^{2}\left(I_{1} \sin ^{2} \varphi_{2}+I_{2} \cos ^{2} \varphi_{2}\right)\right\}, \tag{2}
\end{align*}
$$

where $I_{1}, I_{2}$, and $I_{3}$ are the principal inertia moments to the axes $C \xi \eta \zeta, k^{2}$ is the constant of the spring, $\ell$ denotes the unstretched spring length, and $g$ is the acceleration of gravity.

The Lagrangian function $L$ for this model is of the form [15]

$$
\begin{equation*}
L=T-V, \tag{3}
\end{equation*}
$$

where $L$ is a function of $\rho, \varphi_{1}$, and $\varphi_{2}$ and their derivatives. Applying Lagrange's equations

$$
\begin{equation*}
\frac{d}{d t}\left(\frac{\partial L}{\partial \dot{\rho}}\right)=\frac{\partial L}{\partial \rho}, \rho, \varphi_{1}, \varphi_{2} . \tag{4}
\end{equation*}
$$

Making use of equations (1) to (4), the second-order differential equations of the motion are obtained as follows:


Figure 1: The elastic rigid pendulum model.

$$
\begin{align*}
& \ddot{\rho}+a \ddot{\varphi}_{2} \sin \left(\varphi_{1}-\varphi_{2}\right)-a \dot{\varphi}_{2}^{2} \cos \left(\varphi_{1}-\varphi_{2}\right)-\rho \dot{\varphi}_{1}^{2} \\
&-\omega^{2}\left(2 h \cos \omega t+\rho \sin \varphi_{1}+a \sin \varphi_{2}\right) \sin \varphi_{1}-g \cos \varphi_{1} \\
&+k^{2} m^{-1}(\rho-\ell)=0, \\
& \rho \ddot{\varphi}_{1}+2 \dot{\rho} \dot{\varphi}_{1}+a \ddot{\varphi}_{2} \cos \left(\varphi_{1}-\varphi_{2}\right)+a \dot{\varphi}_{2}^{2} \sin \left(\varphi_{1}-\varphi_{2}\right) \\
&-\omega^{2}\left(2 h \cos \omega t+\rho \sin \varphi_{1}+a \sin \varphi_{2}\right) \cos \varphi_{1}+g \sin \varphi_{1}=0, \\
& \ell_{1} \ddot{\varphi}_{2}+\left(2 \dot{\rho} \dot{\varphi}_{1}+\rho \ddot{\varphi}_{1}\right) \cos \left(\varphi_{1}-\varphi_{2}\right)+\left(\ddot{\rho}-\rho \dot{\varphi}_{1}^{2}\right) \sin \left(\varphi_{1}-\varphi_{2}\right) \\
&-\omega^{2}\left(2 h \cos \omega t+\rho \sin \varphi_{1}+a \sin \varphi_{2}\right) \cos \left(\varphi_{1}-\varphi_{2}\right)+g \sin \varphi_{2} \\
&+2 m^{-1} \omega^{2}\left(I_{2}-I_{1}\right) \sin 2 \varphi_{2}=0, \\
& \ell_{1}=a-I_{3} m^{-1} a^{-1} . \tag{5}
\end{align*}
$$

Equations (5) are the differential equations of motion of second order in the three generalized coordinates.

Let the system oscillate in the closing relative equilibrium position, and the following is obtained:

$$
\begin{equation*}
I_{1}=I_{2} \tag{6}
\end{equation*}
$$

The relative equilibrium admits the equality of the initial values for the angles $\varphi_{1}$ and $\varphi_{2}$, and thus

$$
\begin{align*}
\rho & =b+\xi(t) \\
\varphi_{1} & =\varphi_{0}+\varphi(t)  \tag{7}\\
\varphi_{2} & =\varphi_{0}+\psi(t)
\end{align*}
$$

where $b$ is the relative equilibrium for the length of the pendulum string. The quantities $b$ and $\varphi_{0}$ are determined as follows:

$$
\begin{align*}
m^{-1} k^{2}(b-\ell) & =\omega^{2}(a+b) \sin ^{2} \varphi_{0}+g \cos \varphi_{0} \\
g & =\omega^{2}(a+b) \cos \varphi_{0} \tag{8}
\end{align*}
$$

Making use of (7) into (5) and then (6) and (8), one obtains

$$
\begin{align*}
\ddot{\xi}+a_{11} \xi+a_{12} \varphi+a_{13} \psi & =f_{1}, \\
b \ddot{\varphi}+a \ddot{\psi}+b_{11} \xi+b_{12} \varphi+b_{13} \psi & =f_{2},  \tag{9}\\
b \ddot{\varphi}+\ell_{1} \ddot{\psi}+c_{11} \xi+c_{12} \varphi+c_{13} \psi & =f_{3},
\end{align*}
$$

where

$$
\begin{align*}
& a_{11}=m^{-1} k^{2}-\omega^{2}\left(\sin ^{2} \varphi_{0}+2 h \cos \varphi_{0} \cos \omega t\right) \\
& a_{12}=b c_{11} \\
& a_{13}=a c_{11} \\
& b_{11}=c_{11} \\
& b_{12}=\frac{k^{2}(b-\ell)}{m-\omega^{2} b \cos ^{2} \varphi_{0}},  \tag{10}\\
& b_{13}=-\omega^{2} a \cos ^{2} \varphi_{0} \\
& c_{11}=-\omega^{2} \sin \varphi_{0} \cos \varphi_{0} \\
& c_{12}=-\omega^{2} b \cos ^{2} \varphi_{0} \\
& c_{13}=k^{2} m^{-1}(b-\ell)+\omega^{2}\left(2 h \sin \varphi_{0} \cos \omega t-a \cos ^{2} \varphi_{0}\right)
\end{align*}
$$

$$
\begin{align*}
f_{1}= & (\xi+b) \dot{\varphi}^{2}+A \xi \varphi+a \ddot{\psi}(\psi-\varphi)+a \dot{\psi}^{2}+B \psi \varphi+C_{1}, \\
f_{2}= & -\xi \ddot{\varphi}-2 \dot{\xi} \dot{\varphi}+(\psi-\varphi) a \dot{\psi}+c_{11}(\xi+b) \varphi^{2} \\
& +\left(D \xi+2 a_{13} \psi\right) \varphi+C_{2} \\
f_{3}= & \ddot{\xi}(\psi-\varphi)-\dot{\varphi}^{2}(b+\xi)(\psi-\varphi)-\xi \ddot{\varphi}-2 \dot{\xi} \dot{\varphi} \\
& +2 \omega^{2} h \cos \omega t \cos \varphi_{0} \\
& -\left[b+\left(1-\psi^{2}\right)-(b+\xi) \varphi \psi\right] c_{11}-g \sin \varphi_{0} \\
& +\omega^{2} \xi\left(\varphi \cos ^{2} \varphi_{0}-\psi \sin ^{2} \varphi_{0}\right), \tag{11}
\end{align*}
$$

$$
\begin{align*}
A= & -2 c_{11} \\
B= & a \omega^{2} \cos ^{2} \varphi_{0} \\
D= & \frac{B}{a} \\
C_{1}= & -\left(k^{2}-\omega^{2} \sin ^{2} \varphi_{0}\right)+\omega^{2}\left[2 h \sin \varphi_{0} \cos \omega t\right.  \tag{12}\\
& \left.+a\left(\cos ^{2} \omega t+\sin ^{2} \varphi_{0}\right)\right]+g \cos \varphi_{0}+k^{2} \ell \\
C_{2}= & 2 h \omega^{2} \cos \omega t
\end{align*}
$$

$$
-(b+a) c_{11}-g \sin \varphi_{0}
$$



Figure 2: The solution $\xi$ against the time $t$.


Figure 3: The solution $\varphi$ against the time $t$.


Figure 4: The solution $\psi$ against the time $t$.


Figure 5: The stability $\dot{\xi}$ diagram against $\xi$.


Figure 6: The stability $\dot{\varphi}$ diagram against $\varphi$.


Figure 7: The instability $\dot{\psi}$ diagram against $\psi$.

Next, numerical considerations for solving system (9) in three degrees of freedom $\xi, \varphi$, and $\psi$ are presented. The fourth-order Runge-Kutta method [10] is used for satisfying the numerical solutions for this system.

## 3. Numerical Investigations

In this section, the fourth-order Runge-Kutta method is used for solving the problem in the previous sections through computerized data. These solutions are investigated to illustrate and describe the oscillations of this system at different values of the time.

Making use of (10), (11), and (12), system (9) is reformulated in the form:

$$
\begin{equation*}
\ddot{\xi}=\ddot{\xi}(\xi, \varphi, \psi, \dot{\xi}, \dot{\varphi}, \dot{\psi}),(\xi \varphi \psi) \tag{13}
\end{equation*}
$$

where the symbol $(\xi \varphi \psi)$ refers to the equations which are neglected. These functions are determined accordingly to equations (9)-(12). Introducing the following data:

$$
\begin{align*}
m & =10 \mathrm{~kg} \\
g & =9.8 \mathrm{~m} \cdot \mathrm{~s}^{-2} \\
I_{1} & =3 \mathrm{~kg} \cdot \mathrm{~m}^{2} \\
\ell & =0.9 \mathrm{~m} \\
\omega & =2 \mathrm{rad} \cdot \mathrm{~s}^{-1}  \tag{14}\\
a & =0.7 \mathrm{~m} \\
b & =0.5 \mathrm{~m} \\
\varphi_{0} & =0.2 \mathrm{rad} \\
h & =4 \\
t & =0 \longrightarrow 300 \mathrm{~s} .
\end{align*}
$$

The graphical representations for the solutions $\xi, \varphi, \psi, \dot{\xi}, \dot{\varphi}$, and $\dot{\psi}$ are given in Figures 2-7.

## 4. Conclusion

It is concluded that the model of relativistic elastic rigid pendulum motions is considered a wide application problem in many scientific fields. The Lagrangian function was used, and Lagrange's equations were applied for deriving the
system of equations of motion for this problem. Computer programs were achieved applying the fourth-order Run-ge-Kutta method for obtaining the numerical solutions for the considered system. The obtained solutions are sketched at different values of rigid body parameters. From the figures, it is deduced that the approximated solutions are seemed to be periodic in a big interval of the motion. $\xi$ and $\varphi$ solutions represented in Figures 2, 3, 5, and 6 have uniform motion and stable solutions, but $\psi$ is a chaotic and excited solution, see Figures 4 and 7. In all figures, the positive vibration waves come from the motion of the vertical plane above the horizontal axis and vice versa. The solution $\psi$ is not stable since it moves with fast oscillations with small amplitudes in the beginning time and then goes slowly with larger amplitudes. The changing of the values of $h, b$, and $\varphi_{0}$ affects the behavior of the motion and vice versa.

## Data Availability

Data sharing is not applicable to this article as no datasets were generated or analyzed during the current study.

## Conflicts of Interest

The author declares that there are no conflicts of interest.

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# Nonlinear Dynamics and Suppressing Chaos in Magnetic Bearing System 

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#### Abstract

A nonlinear mathematical model of a magnetic bearing system has been obtained by applying a modified conventional identification technique based on the principle of harmonic balance. In this study, we examined the rich nonlinear dynamics of a magnetic bearing system with closed-loop control using phase portraits, Poincaré maps, and frequency spectra. The resulting bifurcation diagram can be used to evaluate the operational range of systems employing nonlinear actuators. Estimates of the largest Lyapunov exponent based on the properties of synchronization revealed the occurrence of chatter vibration indicative of chaotic motion. Various control methods, such as the state feedback control and the injection of dither signals, were then used to quench the chaotic behavior.


## 1. Introduction and System Description

The rotor in magnetic bearing systems is suspended by magnetic bearings to ensure stable rotation at high speeds; however, a closed-loop control system is required to stabilize the system by eliminating vibrations caused by disturbing forces. The characteristics of magnetic bearings are inherently nonlinear due to nonlinearities in electromagnetic forces. The occurrence of large unbalanced forces in rotor bearings can cause nonlinear motion of high amplitude in magnetic bearing systems. Accurate control of the system is required which designers account for the effects of nonlinearities. In a previous work [1], we conducted experiments on an unloaded symmetric rotor with flexible coupler at one end and a bearing comprising two pairs of electromagnets at the other end (Figure 1). It was carried out by applying a series of nonlinear electromagnetic forces to identify a nonlinear model for this system. These rich nonlinear dynamics must be taken into account in the design of magnetic bearing systems. In this study, we sought to predict these nonlinear dynamics by modifying the
conventional identification technique based on the principle of harmonic balance in order to characterize the system with a higher degree of precision. The resulting nonlinear model [1] is obtained as follows:

$$
\begin{align*}
\dot{x}_{1}= & x_{2}  \tag{1a}\\
\dot{x}_{2}= & -d_{1} x_{2}-d_{2} x_{1}+d_{3} x_{3}+d_{4} x_{4}+d_{5} x_{1}^{2}+d_{6} x_{1} x_{3}  \tag{1b}\\
& +d_{7} x_{1} x_{4}+d_{8} x_{3}^{2}+d_{9} x_{4}^{2}+d_{10} x_{1}^{3}, \\
L \dot{x}_{3} & +R x_{3}=K_{A}\left(e_{c 1}+A_{0} \sin \Omega t\right),  \tag{1c}\\
L \dot{x}_{4} & +R x_{4}=K_{A}\left(e_{c 2}+A_{0} \sin \Omega t\right), \tag{1d}
\end{align*}
$$

where

$$
\begin{align*}
& e_{c 1}=K_{H}\left(K_{P} x_{1}+K_{D} x_{2}+K_{I} \int x_{1} \mathrm{~d} t\right)  \tag{2}\\
& e_{c 2}=K_{H}\left(-K_{P} x_{1}-K_{D} x_{2}-K_{I} \int x_{1} \mathrm{~d} t\right)
\end{align*}
$$



Figure 1: Schematic diagram of a magnetic bearing system.
with the limitations

$$
\begin{array}{r}
-3.5 \text { volts }<e_{c 1}<6.5 \text { volts, } \\
-3.5 \text { volts }<e_{c 2}<6.5 \text { volts, }  \tag{3}\\
0.0 \mathrm{amp}<x_{3} \text { and } x_{4}<2.0 \mathrm{amp},
\end{array}
$$

where $x_{1}$ is the displacement of the rotor around the equilibrium point; $x_{2}$ is the velocity of the rotor; $x_{3}$ and $x_{4}$ are coil currents oscillating around the bias current; $e_{c 1}$ and $e_{c 2}$ are the outputs of the PID controller to the two coils; $K_{P}=-55, K_{D}=-0.3$, and $K_{I}=-50$ are the control gains; $K_{A}$ $(=2.4)$ and $K_{H}(=10000)$ represent the gains of the power amplifier and displacement sensor; and $A_{0} \sin \Omega t$ is the forcing voltage generated by a programmable function generator. The procedures used to derive the other coefficients necessary for (1a)-(1d) are listed in Table 1 [1].

This model captures the primary characteristics of the system by comparing the frequency responses from simulations with those from experiments [1]. However, this model has not been subjected to dynamic analysis to determine whether the nonlinear mathematical model derived from experiments can be used to characterize and/ or predict the dynamics in a physical system. Furthermore, the chaotic motion that occurs at the moment the rotor strikes the electromagnet has not been characterized.

In this study, we employed bifurcation diagrams, phase portraits, Poincaré maps, frequency spectra, and Lyapunov exponents to observe periodic and chaotic motions. Across a broad range of parameters, the Lyapunov exponent provides the most powerful means by which to measure the sensitivity of a dynamic system as it pertains to its initial conditions. This approach can be used to determine whether a system is susceptible to chaotic motion. The algorithms used to

Table 1: Identified results [1].

| System parameter | Identified value |
| :--- | :---: |
| $d_{1}$ | 3.066133 |
| $d_{2}$ | $4.3315 \times 10^{3}$ |
| $d_{3}$ | -5.5786683 |
| $d_{4}$ | 6.5562 |
| $d_{5}$ | $1.43371 \times 10^{7}$ |
| $d_{6}$ | $-2.531 \times 10^{4}$ |
| $d_{7}$ | $-4.53203 \times 10^{4}$ |
| $d_{8}$ | 0.34514 |
| $d_{9}$ | -0.40171 |
| $d_{10}$ | $-2.399816 \times 10^{10}$ |
| $L$ | 0.0161203 |
| $R$ | 14.3128231 |

compute Lyapunov exponents associated with smooth dynamic systems are well-established [2-5]. However, a number of nonsmooth dynamic systems possess discontinuities to which this algorithm cannot be applied directly, such as those associated with dry friction, backlash, and saturation. Many studies have proposed methods for the calculation of Lyapunov exponents associated with nonsmooth dynamic systems [6-8]. In this study, we adopted the method developed by Stefanski [8] for estimating the largest Lyapunov exponent in a magnetic bearing system with closed-loop control.

Many practical engineering problems involving chaos require control techniques to convert chaotic attractors into stable periodic orbits. Since the pioneering work of Ott et al. [9], numerous methods for the control of chaos have been devised [10-19]. Improving the performance of a controlled magnetic bearing system and/or eliminating chatter behavior require the conversion of chaotic behavior into periodic motion. Two control methods have been developed
for chaos suppression: state feedback control [11-15] and dither control [16-19].

## 2. System Characteristics: Simulations and Discussion

We conducted a series of numerical simulations based on (1a)-(1d) to clearly elucidate the dynamic characteristics of the system in this study. In (1a)-(1d), the amplitude of the input excitation, $A_{0}$, equals 3.5 V . We employed FORTRAN subroutines in the commercial software package DIVPRK (IMSL) to solve ordinary differential equations [20]. The resulting bifurcation diagram is presented in Figure 2. This figure clearly shows that the first period-doubling bifurcation occurred at approximately $\Omega=19.5 \mathrm{~Hz}$ and chaotic motion appeared at approximately $\Omega=18.5 \mathrm{~Hz}$. Further details of the responses exhibited by the system are presented in Figures 3-5, in which each type of response is characterized by a phase portrait, Poincaré map, and frequency spectrum. Figures 3(a)-3(c) show that the $T_{f}$-period includes the constant term and fundamental components. From Figure 4, we determined that a cascade of period-doubling bifurcations produced a series of subharmonic components, revealing the bifurcations with new frequency components at $\Omega / 2,3 \Omega / 2$, $5 \Omega / 2, \ldots$. The essence of chaotic behavior can be described using Poincaré maps, which present an infinite set of points referred to as a strange attractor. Chaotic motion also presents a broad continuous frequency spectrum. Thus, strange attractors and continuous-type Fourier spectra are generally regarded as strong indictors of chaos, as illustrated in Figures 5(a)-5(c).

## 3. Synchronization and the Lyapunov Exponent

The largest Lyapunov exponent is a useful diagnostic element for the analysis of chaotic systems. Every dynamic system possesses a spectrum of Lyapunov exponents ( $\lambda$ ), which determine length, area, and volume changes in the phase space. In other words, Lyapunov exponents measure the rate of divergence (or convergence) between two adjacent orbits. Chaos can be identified simply by calculating the largest Lyapunov exponent, thereby determining whether nearby trajectories generally diverge ( $\lambda>0$ ) or converge ( $\lambda<0$ ). Any bounded motion in a system containing at least one positive Lyapunov exponent is defined as chaotic, whereas nonpositive Lyapunov exponents indicate periodic motion. There are a number of well-established algorithms that compute the Lyapunov spectrum of smooth dynamic systems [2-5]. However, nonsmooth dynamic systems with discontinuities, such as dry friction, backlash, and saturation, do not allow for the direct application of such algorithms. In this study, we estimated the largest Lyapunov exponent in order to identify the onset of chaotic motion in a controlled magnetic bearing system. Stefanski [8] proposed a simple method for estimating the largest Lyapunov exponent based on properties associated with synchronization. Synchronization controls the response system by accessing the output of the drive system such that the output of the


Figure 2: Bifurcation diagram of the system for $A_{0}=3.5 \mathrm{~V}$.
response system asymptotically follows the output of the drive system. This method is described briefly as follows.

The dynamic system is decomposed into the following two subsystems:

A drive system

$$
\begin{equation*}
\dot{x}=f(x) . \tag{4}
\end{equation*}
$$

A response system

$$
\begin{equation*}
\dot{y}=f(y) . \tag{5}
\end{equation*}
$$

Consider a dynamic system comprising two identical n-dimensional subsystems, where the response system (5) is combined with coupling coefficient $d$, and the drive system (4) remains the same. The first-order differential equations used to describe such a system are written as follows:

$$
\begin{align*}
\dot{x} & =f(x),  \tag{6}\\
\dot{y} & =f(y)+d(x-y) .
\end{align*}
$$

The condition of synchronization is given by the following inequality:

$$
\begin{equation*}
d>\lambda_{\max } \tag{7}
\end{equation*}
$$

The smallest value of coupling coefficient $d$ in synchronization $d_{s}$ is assumed to be equal to the largest Lyapunov exponent, as follows:

$$
\begin{equation*}
d_{s}=\lambda_{\max } . \tag{8}
\end{equation*}
$$

(6) provides an augmented system based on (1a)-(1d), as follows:

$$
\begin{align*}
\dot{x}_{1}= & x_{2},  \tag{9a}\\
\dot{x}_{2}= & -d_{1} x_{2}-d_{2} x_{1}+d_{3} x_{3}+d_{4} x_{4}+d_{5} x_{1}^{2}+d_{6} x_{1} x_{3} \\
& +d_{7} x_{1} x_{4}+d_{8} x_{3}^{2}+d_{9} x_{4}^{2}+d_{10} x_{1}^{3}, \tag{9b}
\end{align*}
$$



Figure 3: Period-one orbit for $\Omega=20.0 \mathrm{~Hz}$ : (a) phase portrait; (b) Poincaré map; (c) frequency spectrum.

$$
\begin{gather*}
L \dot{x}_{3}+R x_{3}=K_{A}\left(e_{c 1 x}+A_{0} \sin \Omega t\right),  \tag{9c}\\
L \dot{x}_{4}+R x_{4}=K_{A}\left(e_{c 2 x}+A_{0} \sin \Omega t\right),  \tag{9d}\\
\dot{y}_{1}=y_{2}+d\left(x_{1}-y_{1}\right),  \tag{10a}\\
\dot{y}_{2}=-d_{1} y_{2}-d_{2} y_{1}+d_{3} y_{3}+d_{4} y_{4}+d_{5} y_{1}^{2}+d_{6} y_{1} y_{3} \\
+d_{7} y_{1} y_{4}+d_{8} y_{3}^{2}+d_{9} y_{4}^{2}+d_{10} y_{1}^{3}+d\left(x_{2}-y_{2}\right),  \tag{10b}\\
L \dot{y}_{3}+R y_{3}=K_{A}\left(e_{c 1 y}+A_{0} \sin \Omega t\right)+d\left(x_{3}-y_{3}\right),  \tag{10c}\\
L \dot{y}_{4}+R y_{4}=K_{A}\left(e_{c 2 y}+A_{0} \sin \Omega t\right)+d\left(x_{4}-y_{4}\right), \tag{10d}
\end{gather*}
$$

where

$$
\begin{align*}
& e_{c 1 x}=K_{H}\left(K_{P} x_{1}+K_{D} x_{2}+K_{I} \int x_{1} \mathrm{~d} t\right) \\
& e_{c 2 x}=K_{H}\left(-K_{P} x_{1}-K_{D} x_{2}-K_{I} \int x_{1} \mathrm{~d} t\right),  \tag{11}\\
& e_{c 1 y}=K_{H}\left(K_{P} y_{1}+K_{D} y_{2}+K_{I} \int y_{1} \mathrm{~d} t\right), \\
& e_{c 2 y}=K_{H}\left(-K_{P} y_{1}-K_{D} y_{2}-K_{I} \int y_{1} \mathrm{~d} t\right),
\end{align*}
$$

with the following limitations:

$$
\begin{array}{r}
-3.5 \text { volts }<e_{c 1 x}<6.5 \text { volts, } \\
-3.5 \text { volts }<e_{c 2 x}<6.5 \text { volts, } \\
0.0 \mathrm{amp}<x_{3} \text { and } x_{4}<2.0 \mathrm{amp}, \\
-3.5 \text { volts }<e_{c 1 y}<6.5 \text { volts, }  \tag{12}\\
-3.5 \text { volts }<e_{c 2 y}<6.5 \text { volts, } \\
0.0 \mathrm{amp}<y_{3} \text { and } y_{4}<2.0 \mathrm{amp} .
\end{array}
$$

In the next step, we estimate the largest Lyapunov exponent for the selected parametric values in accordance with the method described above. Figure 6 presents the results of numerical calculations showing the estimated largest Lyapunov exponents obtained using the above synchronization method. All of the largest Lyapunov exponents are positive with regard to the forcing frequency ( $\Omega<18.5 \mathrm{~Hz}$ ), indicating that the system exhibits chaotic motion. These results help to enhance our understanding of chatter vibration in a controlled magnetic bearing system.

## 4. Quenching Chatter

Chaotic motion can induce chatter vibrations capable of destroying a magnetic bearing system. Accurate prediction of the behavior of a chaotic system can be beneficial;


Figure 4: Period-two orbit for $\Omega=19.0 \mathrm{~Hz}$ : (a) phase portrait; (b) Poincaré map; (c) frequency spectrum.
however, maximizing the benefits requires the ability to control this behavior. A chaotic system must be transformed into periodic motion in order to work under specific conditions and thereby improve the performance of a dynamic system and/or avoid chatter. In this section, we discuss the means by which chaos can be converted into periodic motion using minimal efforts. We present two methods of controlling chaos: the addition of state feedback control $[12,13]$ and the application of dither signals [16, 17].
4.1. State Feedback Control. Cai et al. [12, 13] proposed a simple, yet effective, state feedback control algorithm. This method for the $n$-dimensional dynamical system is explained briefly as follows:

$$
\begin{equation*}
\dot{\mathbf{x}}=\mathbf{f}(\mathbf{x}, t), \tag{13}
\end{equation*}
$$

where $\mathbf{x}(t) \in R^{n}$ is the state vector and $\mathbf{f}=\left(f_{1}, \ldots, f_{i}, \ldots, f_{n}\right)$, where $f_{i}$ is a linear or a nonlinear function and $\mathbf{f}$ includes at least one nonlinear function. Suppose that $f_{k}(x, t)$ is the key nonlinear function leading to chaotic motion in system (13). We add to the equation only one term of state feedback of an available system variable $x_{m}$ that includes $f_{k}(x, t)$, as follows:

$$
\begin{equation*}
\dot{x}_{k}=f_{k}(\mathbf{x}, t)+G x_{m}, \quad k, m \in\{1,2, \ldots, n\} \tag{14}
\end{equation*}
$$

where $G$ is feedback gain. Other functions retain their original forms. We applied this method to (1a)-(1d) to more clearly elucidate the simple control scheme.

In the absence of state feedback control, (1a)-(1d) exhibits chaotic behavior under the following parameters: $A_{0}=3.5 \mathrm{~V}$ and $\Omega=18.2 \mathrm{~Hz}$. Consider how the addition of state feedback control to the right-hand side of the (1a)-(1d) would affect the results. Figure 7 presents the resulting bifurcation diagram for $\Omega=18.2 \mathrm{~Hz}$ with state feedback control. When $G=0$, (1a)-(1d) displays chaotic motion (see Figure 5). This figure shows that chaotic motion occurs at approximately $G>-7.0$ and disappears at approximately $G \leq-7.0$. When the feedback gain $G$ falls below -7.0 , the system described by (1a)-(1d) exhibits stable periodic motion. After 10 seconds, the control signal is applied to the system to control chaotic oscillation (Figure 8). In so doing, the state feedback control can be used to suppress chaotic motion and improve the performance of a magnetic bearing system with closed-loop control.
4.2. Dither Control. In this section, we describe how chaotic motion can be controlled by injecting an external dither signal to adjust nonlinear terms. A dither is a high-frequency signal introduced to modify the behaviors of nonlinear systems through the elimination of nonlinearities. The ability of a dither signal to average out nonlinearities can be attributed to its high frequency and periodic nature. Dither smoothing techniques were proposed in references [16, 17] to stabilize chaotic systems. Some of the most common dither signals were proposed by Cook [21].
(i) Square-wave dither: the simplest type of dither signal is the square-wave signal as shown in Figure 9, where


Figure 5: Chaotic motion for $\Omega=18.2 \mathrm{~Hz}$ : (a) phase portrait; (b) Poincaré map; (c) frequency spectrum.


Figure 6: Evolutions of the largest Lyapunov exponent.
dither signal takes the constant values $W$ and $-W$ alternately, each holding for a half-period $T / 2$, with $T$ being much smaller than the time constant of the system. The amplitude $W$ is applied in front of the nonlinearity $f(\cdot)$. Thus, the effective value of $\bar{n}$ (the output of the nonlinear element) can be written as [16]

$$
\begin{equation*}
\bar{n}=\frac{1}{2}[f(y+W)+f(y-W)] . \tag{15}
\end{equation*}
$$



Figure 7: Bifurcation diagram of system with state feedback control, where $G$ denotes the feedback gain.

Thus, the system equations can be written as follows:

$$
\begin{equation*}
\dot{y}=\bar{n} . \tag{16}
\end{equation*}
$$

Consider the effect of adding square-wave dither control to the system described in (1a)-(1d), where $\Omega=18.2 \mathrm{~Hz}$. Increasing the amplitude of the squarewave dither signal from $W=0$ to 0.0005 V would


Figure 8: Transforming chaotic motion to a desired period-one orbit at $G=-50$ and $\Omega=18.2 \mathrm{~Hz}$ where the state feedback control signal is introduced after 10 s : (a) time responses; (b) phase portrait of the controlled system.


Figure 9: A square-wave dither signal.
change the chaotic behavior to period-one motion. The resulting bifurcation diagram under $\Omega=18.2 \mathrm{~Hz}$ with a square-wave dither signal is shown in Figure 10. Figure 11 shows the effects 10 seconds after applying a control signal to the system in order to illustrate the effectiveness of this controller in controlling chaotic oscillations.
(ii) Sinusoidal dither: one simple dither signal is a highfrequency sinusoid. Here, the effective value of $n$ is its average over a complete period of the sinusoidal dither signal oscillation:

$$
\begin{equation*}
n=\frac{1}{2 \pi} \int_{0}^{2 \pi} f(x+W \sin \theta) \mathrm{d} \theta \tag{17}
\end{equation*}
$$

Adding a sinusoidal dither signal to (1a)-(1d) yields the following coupled system:

$$
\begin{align*}
& \dot{\tilde{x}}_{1}=\tilde{x}_{2}  \tag{18a}\\
& \dot{\tilde{x}}_{2}=-d_{1} \widetilde{x}_{2}-d_{2} \widetilde{x}_{1}+d_{3} \widetilde{x}_{3}+d_{4} \widetilde{x}_{4}+n,  \tag{18b}\\
& L \dot{\tilde{x}}_{3}+R \widetilde{x}_{3}= K_{A}\left(\widetilde{e}_{c 1}+A_{0} \sin \Omega t\right),  \tag{18c}\\
& L \dot{\tilde{x}}_{4}+R \widetilde{x}_{4}=K_{A}\left(\widetilde{e}_{c 2}+A_{0} \sin \Omega t\right), \tag{18d}
\end{align*}
$$

where


Figure 10: Bifurcation diagram of the system with a square-wave dither, where $W$ denotes the amplitude of the dither signal.

$$
\begin{align*}
\tilde{e}_{c 1}= & K_{H}\left(K_{P} \widetilde{x}_{1}+K_{D} \tilde{x}_{2}+K_{I} \int \widetilde{x}_{1} \mathrm{~d} t\right) \\
\tilde{e}_{c 2}= & K_{H}\left(-K_{P} \widetilde{x}_{1}-K_{D} \widetilde{x}_{2}-K_{I} \int \widetilde{x}_{1} \mathrm{~d} t\right) \\
n= & \frac{1}{2 \pi} \int_{0}^{2 \pi}\left[d_{5}\left(\widetilde{x}_{1}+W \sin \theta\right)^{2}+d_{6}\left(\widetilde{x}_{1}+W \sin \theta\right)\right. \\
& \cdot\left(\widetilde{x}_{3}+W \sin \theta\right)+d_{7}\left(\widetilde{x}_{1}+W \sin \theta\right)\left(\widetilde{x}_{4}+W \sin \theta\right) \\
& \left.+d_{8}\left(\widetilde{x}_{3}+W \sin \theta\right)^{2}+d_{9}\left(\widetilde{x}_{4}+W \sin \theta\right)^{2}+d_{10}\left(\widetilde{x}_{1}+W \sin \theta\right)^{3}\right] \tag{19}
\end{align*}
$$

with the following limitations:

$$
\begin{array}{r}
-3.5 \text { volts }<\tilde{e}_{c 1}<6.5 \text { volts }, \\
-3.5 \text { volts }<\tilde{e}_{c 2}<6.5 \text { volts },  \tag{20}\\
0.0 \mathrm{amp}<\tilde{x}_{3} \text { and } \tilde{x}_{4}<2.0 \mathrm{amp} .
\end{array}
$$

The dither frequency must substantially exceed that of any other frequency used to operate the system in order to ensure that the dither signal does not introduce other undesirable oscillations at the same frequency as the dither signal. For example, Figure 12 presents the resulting


Figure 11: Injecting a square-wave dither signal control which is used to control chaotic motion of the system for $A_{0}=3.5 \mathrm{~V}$ at $\Omega=18.2 \mathrm{~Hz}$. The dither signal ( $W=0.0004$ ) is added after 10 seconds: (a) displacement time series; (b) controlled orbit.


Figure 12: Bifurcation diagram of system with a sinusoidal dither, where $W$ is the amplitude of the dither.


Figure 13: Equivalent nonlinearity $n$ (solid line) described by equation (17). Original nonlinearity $f$ (dashed line) is given in equation (1b).


Figure 14: Controlling chaotic motion to desired period-one orbit for $W=0.003 \mathrm{~V}$ and $\Omega=18.2 \mathrm{~Hz}$ : (a) time responses of displacement. The sinusoidal dither signal is injected after 1 s . (b) phase portrait of controlled system.
bifurcation diagram for a system with a sinusoidal dither in which parameter $\Omega$ is 18.2 Hz and the frequency of the sinusoidal dither is $6000 \mathrm{rad} / \mathrm{s}$. This reveals that a sinusoidal dither with amplitude exceeding 0.0005 V can convert chaotic behavior into stable periodic motion in a magnetic bearing system. The next steps are setting $W=0.008 \mathrm{~V}$ and plotting the effective nonlinearity $n$ and original nonlinearity $f$, as shown in Figure 13. The time response of displacement 10 seconds after the injection of a sinusoidal dither signal is shown in Figure 14(a). The chaotic behavior is converted into period-one motion. Figure $14(\mathrm{~b})$ presents a phase portrait of the controlled system. Note that the behavior of the system is initially chaotic, but gradually becomes periodic following dither injection.

## 5. Conclusions

This work uses the identified nonlinear model examine global bifurcation and chaos control in the magnetic bearing system. Dynamic behavior over the entire range of parameter values is observable in the resulting bifurcation diagram, which reveals that the magnetic bearing system exhibits period-doubling bifurcations and chaotic motions. The largest Lyapunov exponent derived using the properties of synchronization provided the most powerful tool to measure and analyze chaotic motion in such a system. Controlling chaotic motion is an effective way to prevent chatter vibration in magnetic bearing systems. State feedback control is a simple, yet effective, approach to suppressing chaos. It can be implemented by adding feedback associated with a suitable variable into the original system with control gain sufficient to overcome the development of chaos in dynamic systems prone to chaotic behavior. It is also possible to convert chaotic behavior into a periodic orbit through the injection of a dither signal in front of nonlinearities in a chaotic magnetic bearing system.

Our analysis revealed that the proposed nonlinear model is able to predict the occurrence of bifurcation and chaos in an active magnetic bearing system, which means that it is potentially applicable to a wide range of functions in the design of active magnetic bearing systems. Chaotic behavior must be accepted in some situations; however, it is normally deemed undesirable, as it degrades performance and restricts the operating range of electric and mechanical devices. Accordingly, we applied state feedback and dither
control methods to quench chaos, improve the performance of the magnetic bearing system, and prevent the occurrence of chaos behaviors. Furthering the development of magnetically levitated vehicles requires an understanding of their nonlinear dynamic characteristics from the viewpoint of stability, safety, and ride quality at high speeds. Our contribution of this study is that studying nonlinear dynamics and controlling chaotic vibrations in the active magnetic bearing systems will help to advance the development of magnetic transportation systems.

## Data Availability

No data were used to support this study. We only used computer for simulation. Therefore, we can only provide simulation programming, which can be obtained from the corresponding author upon request.

## Conflicts of Interest

The author declares that there are no conflicts of interest regarding the publication of this paper.

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# On a System of $\boldsymbol{k}$-Difference Equations of Order Three 

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In this paper, we deal with the global behavior of the positive solutions of the system of $k$-difference equations $u_{n+1}^{(1)}=$ $\left(\alpha_{1} u_{n-1}^{(1)} / \beta_{1}+\alpha_{1}\left(u_{n-2}^{(2)}\right)^{r_{1}}\right), u_{n+1}^{(2)}=\alpha_{2} u_{n-1}^{(2)} / \beta_{2}+\alpha_{2}\left(u_{n-2}^{(3)}\right)^{r_{2}}, \ldots, u_{n+1}^{(k)}=\alpha_{k} u_{n-1}^{(k)} / \beta_{k}+\alpha_{k}\left(u_{n-2}^{(1)}\right)^{r_{k}}, n \in \mathbb{N}_{0}$, where the initial conditions $u_{-l}^{(i)}(l=0,1,2)$ are nonnegative real numbers and the parameters $\alpha_{i}, \beta_{i}, \gamma_{i}$, and $r_{i}$ are positive real numbers for $i=1,2, \ldots, k$, by extending some results in the literature. By the end of the paper, we give three numerical examples to support our theoretical results related to the system with some restrictions on the parameters.

## 1. Introduction

Recently, many works have been published on rational difference equations, which have an important position in applied sciences. In this process, many rational difference equations have been studied by mathematicians. And so, some equations have frequently been the subject of many articles using generalizations. Many typical examples of these can be found in the literature. For example, in [1], El-Owaidy et al. dealt with global behavior of the difference equation

$$
\begin{equation*}
x_{n+1}=\frac{\alpha x_{n-1}}{\beta+\gamma x_{n-2}^{p}}, \quad n \in \mathbb{N}_{0} \tag{1}
\end{equation*}
$$

with nonnegative parameters and initial conditions. Gumus and Soykan [2] dealt with the dynamical behavior of the positive solutions for a system of rational difference equations of the following form:

$$
\begin{align*}
& u_{n+1}=\frac{\alpha u_{n-1}}{\beta+\gamma v_{n-2}^{p}} \\
&-4 p t  \tag{2}\\
& v_{n+1}=\frac{\alpha_{1} v_{n-1}}{\beta_{1}+\gamma_{1} u_{n-2}^{p}}, \quad n \in \mathbb{N}_{0},
\end{align*}
$$

where the parameters and initial conditions are positive real numbers. Tollu and Yalcinkaya [3] dealt with the dynamical behavior of the positive solutions for the following three-dimensional system of rational difference equations:

$$
\begin{align*}
& u_{n+1}=\frac{\alpha_{1} u_{n-1}}{\beta_{1}+\gamma_{1} v_{n-2}^{p}}, \\
& v_{n+1}=\frac{\alpha_{2} v_{n-1}}{\beta_{2}+\gamma_{2} w_{n-2}^{q}},  \tag{3}\\
& w_{n+1}=\frac{\alpha_{3} w_{n-1}}{\beta_{3}+\gamma_{3} u_{n-2}^{r}}, \quad n \in \mathbb{N}_{0},
\end{align*}
$$

where the parameters and initial conditions are positive real numbers. For more papers on this topic, see, for example, [4-29].

In the present paper, we investigate the global behavior of the positive solutions of the $k$-dimensional system of difference equations:

$$
\begin{align*}
u_{n+1}^{(1)} & =\frac{\alpha_{1} u_{n-1}^{(1)}}{\beta_{1}+\gamma_{1}\left(u_{n-2}^{(2)}\right)^{r_{1}}}, u_{n+1}^{(2)}=\frac{\alpha_{2} u_{n-1}^{(2)}}{\beta_{2}+\gamma_{2}\left(u_{n-2}^{(3)}\right)^{r_{2}}}, \ldots, u_{n+1}^{(k)} \\
& =\frac{\alpha_{k} u_{n-1}^{(k)}}{\beta_{k}+\gamma_{k}\left(u_{n-2}^{(1)}\right)^{r_{k}}}, \quad n \in \mathbb{N}_{0}, \tag{4}
\end{align*}
$$

where the initial conditions $u_{-l}^{(i)}(l=0,1,2)$ are nonnegative real numbers and the parameters $\alpha_{i}, \beta_{i}, \gamma_{i}$, and $r_{i}$ are positive real numbers for $i=1,2, \ldots, k$, by extending some recent results in the literature.

Remark 1. This paper extends the results of studies in the references [1-3]. That is to say, if we take $k=1$, then system (4) reduces equation (1). If we take $k=2$, then system (4) reduces system (2). Finally, if we take $k=3$, then system (4) reduces system (3). So, system (4) is a natural generalization of equation (1), system (2), and system (3).

Note that system (4) can be written as

$$
\begin{align*}
x_{n+1}^{(1)} & =\frac{a_{1} x_{n-1}^{(1)}}{1+\left(x_{n-2}^{(2)}\right)^{r_{1}}}, x_{n+1}^{(2)}=\frac{a_{2} x_{n-1}^{(2)}}{1+\left(x_{n-2}^{(3)}\right)^{r_{2}}}, \ldots, x_{n+1}^{(k)} \\
& =\frac{a_{k} x_{n-1}^{(k)}}{1+\left(x_{n-2}^{(1)}\right)^{r_{k}}}, \quad n \in \mathbb{N}_{0}, \tag{5}
\end{align*}
$$

by the change of variables $u_{n}^{(1)}=\left(\beta_{k} / \gamma_{k}\right)^{1 / r_{k}} x_{n}^{(1)}$, $u_{n}^{(2)}=\left(\beta_{1} / \gamma_{1}\right)^{\left(1 / r_{1}\right)} x_{n}^{(2)}, \ldots, u_{n}^{(k)}=\left(\beta_{k-1} / \gamma_{k-1}\right)^{\left(1 / r_{k-1}\right)} x_{n}^{(k)}$ with $a_{i}=\left(\alpha_{i} / \beta_{i}\right)$ for $i=1,2, \ldots k$. So, we will consider system (5) instead of system (4) from now.

## 2. Preliminaries

Let $I_{1}, I_{2}, \ldots, I_{k}$ be some intervals of real numbers and $f_{1}: I_{1}^{3} \times I_{2}^{3} \times \cdots \times I_{k}^{3} \longrightarrow I_{1}, \quad f_{2}: I_{1}^{3} \times I_{2}^{3} \times \cdots \times I_{k}^{3} \longrightarrow I_{2}$, $\ldots, f_{k}: I_{1}^{3} \times I_{2}^{3} \times \cdots \times I_{k}^{3} \longrightarrow I_{k}$ be continuously differentiable functions. Then, for initial conditions $\left(u_{0}^{(1)}, u_{-1}^{(1)}\right.$, $\left.u_{-2}^{(1)}, u_{0}^{(2)}, u_{-1}^{(2)}, u_{-2}^{(2)}, \ldots, u_{0}^{(k)}, u_{-1}^{(k)}, u_{-2}^{(k)}\right) \in I_{1}^{3} \times I_{2}^{3} \times \cdots \times I_{k}^{3}$, the system of difference equations,

$$
\left\{\begin{array}{l}
u_{n+1}^{(1)}=f_{1}\left(u_{n}^{(1)}, u_{n-1}^{(1)}, u_{n-2}^{(1)}, u_{n}^{(2)}, u_{n-1}^{(2)}, u_{n-2}^{(2)}, \ldots, u_{n}^{(k)}, u_{n-1}^{(k)}, u_{n-2}^{(k)}\right),  \tag{6}\\
u_{n+1}^{(2)}=f_{2}\left(u_{n}^{(1)}, u_{n-1}^{(1)}, u_{n-2}^{(1)}, u_{n}^{(2)}, u_{n-1}^{(2)}, u_{n-2}^{(2)}, \ldots, u_{n}^{(k)}, u_{n-1}^{(k)}, u_{n-2}^{(k)}\right), \quad n \in \mathbb{N}_{0}, \\
\vdots \\
u_{n+1}^{(k)}=f_{k}\left(u_{n}^{(1)}, u_{n-1}^{(1)}, u_{n-2}^{(1)}, u_{n}^{(2)}, u_{n-1}^{(2)}, u_{n-2}^{(2)}, \ldots, u_{n}^{(k)}, u_{n-1}^{(k)}, u_{n-2}^{(k)}\right),
\end{array}\right.
$$

has the unique solution $\left\{\left(u_{n}^{(1)}, u_{n}^{(2)}, \ldots, u_{n}^{(k)}\right)\right\}_{n=-2}^{\infty}$. Also, an equilibrium point of system (6) is a point $\left(\bar{u}^{(1)}, \bar{u}^{(2)}, \ldots, \bar{u}^{(k)}\right)$ that satisfies the following system: $\bar{u}^{(1)}=f_{1}\left(\bar{u}^{(1)}, \bar{u}^{(1)}, \bar{u}^{(1)}, \bar{u}^{(2)}, \bar{u}^{(2)}, \bar{u}^{(2)}, \ldots, \bar{u}^{(k)}, \bar{u}^{(k)}, \bar{u}^{(k)}\right)$, $\bar{u}^{(2)}=f_{2}\left(\bar{u}^{(1)}, \bar{u}^{(1)}, \bar{u}^{(1)}, \bar{u}^{(2)}, \bar{u}^{(2)}, \bar{u}^{(2)}, \ldots, \bar{u}^{(k)}, \bar{u}^{(k)}, \bar{u}^{(k)}\right)$, $\bar{u}^{(k)}=f_{k}\left(\bar{u}^{(1)}, \bar{u}^{(1)}, \bar{u}^{(1)}, \bar{u}^{(2)}, \bar{u}^{(2)}, \bar{u}^{(2)}, \ldots, \bar{u}^{(k)}, \bar{u}^{(k)}, \bar{u}^{(k)}\right)$.

We rewrite system (6) in the vector form

$$
\begin{equation*}
U_{n+1}=F\left(U_{n}\right), \quad n \in \mathbb{N}_{0} \tag{8}
\end{equation*}
$$

where $U_{n}=\left(u_{n}^{(1)}, u_{n-1} \quad{ }^{(1)}, u_{n-2}^{(1)}, u_{n}^{(2)}, u_{n-1}^{(2)}, u_{n-2}^{(2)}, \ldots, u_{n}^{(k)}\right.$, $\left.u_{n-1}^{(k)}, u_{n-2}^{(k)}\right)^{T}, F$ is a vector map such that $F: I_{1}^{3} \times I_{2}^{3} \times \cdots \times$ $I_{k}^{3} \longrightarrow I_{1}^{3} \times I_{2}^{3} \times \cdots \times I_{k}^{3}$, and

$$
F\left(\begin{array}{c}
v_{0}^{(1)}  \tag{9}\\
v_{1}^{(1)} \\
v_{2}^{(1)} \\
v_{0}^{(2)} \\
v_{1}^{(2)} \\
v_{2}^{(2)} \\
\vdots \\
v_{0}^{(k)} \\
v_{1}^{(k)} \\
v_{2}^{(k)}
\end{array}\right)=\left(\begin{array}{c}
f_{1}\left(v_{0}^{(1)}, v_{1}^{(1)}, v_{2}^{(1)}, v_{0}^{(2)}, v_{1}^{(2)}, v_{2}^{(2)}, \ldots, v_{0}^{(k)}, v_{1}^{(k)}, v_{2}^{(k)}\right) \\
v_{0}^{(1)} \\
v_{1}^{(1)} \\
f_{2}\left(v_{0}^{(1)}, v_{1}^{(1)}, v_{2}^{(1)}, v_{0}^{(2)}, v_{1}^{(2)}, v_{2}^{(2)}, \ldots, v_{0}^{(k)}, v_{1}^{(k)}, v_{2}^{(k)}\right) \\
v_{0}^{(2)} \\
v_{1}^{(2)} \\
\vdots \\
f_{k}\left(v_{0}^{(1)}, v_{1}^{(1)}, v_{2}^{(1)}, v_{0}^{(2)}, v_{1}^{(2)}, v_{2}^{(2)}, \ldots, v_{0}^{(k)}, v_{1}^{(k)}, v_{2}^{(k)}\right) \\
v_{0}^{(k)} \\
v_{1}^{(k)}
\end{array}\right) .
$$

It is clear that if an equilibrium point of system (6) is $\left(\bar{u}^{(1)}, \bar{u}^{(2)}, \ldots, \bar{u}^{(k)}\right)$, then the corresponding equilibrium point of system (8) is the point $\bar{U}=\left(\bar{u}^{(1)}, \bar{u}^{(1)}, \bar{u}^{(1)}\right.$, $\left.\bar{u}^{(2)}, \bar{u}^{(2)}, \bar{u}^{(2)}, \ldots, \bar{u}^{(k)}, \bar{u}^{(k)}, \bar{u}^{(k)}\right)^{T}$.

In this study, we denote by $\|\cdot\|$ any convenient vector norm and the corresponding matrix norm. Also, we denote by $U_{0} \in I_{1}^{3} \times I_{2}^{3} \times \cdots \times I_{k}^{3}$ a initial condition of system (8).

Definition 1. Let $\bar{U}$ be an equilibrium point of system (8). Then,
(i) The equilibrium point $\bar{U}$ is called stable if for every $\epsilon>0$ there exists $\delta>0$ such that $\left\|U_{0}-\bar{U}\right\|<\delta$ implies $\left\|U_{n}-\bar{U}\right\|<\varepsilon$, for all $n \geq 0$. Otherwise, the equilibrium point $\bar{U}$ is called unstable.
(ii) The equilibrium point $\bar{U}$ is called locally asymptotically stable if it is stable and there exists $\gamma>0$ such that $\left\|U_{0}-\bar{U}\right\|<\gamma$ and $U_{n} \longrightarrow \bar{U}$ as $n \longrightarrow \infty$.
(iii) The equilibrium point $\bar{U}$ is called a global attractor if $U_{n} \longrightarrow \bar{U}$ as $n \longrightarrow \infty$.
(iv) The equilibrium point $\bar{U}$ is called globally asymptotically stable if it is both locally asymptotically stable and global attractor.
The linearized system of (8) evaluated at the equilibrium

$$
\begin{equation*}
Z_{n+1}=J_{F} Z_{n}, \quad n \in \mathbb{N}_{0}, \tag{10}
\end{equation*}
$$

where $J_{F}$ is the Jacobian matrix of $F$ at the equilibrium $\bar{U}$. The characteristic equation of system (10) about the equilibrium $\bar{U}$ is

$$
\begin{equation*}
P(\lambda)=a_{0} \lambda^{3 k}+a_{1} \lambda^{3 k-2}+\cdots+a_{3 k-1} \lambda+a_{3 k}=0 \tag{11}
\end{equation*}
$$

with real coefficients and $a_{0}>0$.

Theorem 1 (see [30]). Assume that $\bar{U}$ is a equilibrium point of system (8). If all eigenvalues of the Jacobian matrix $J_{F}$ evaluated at $\bar{U}$ lie in the open unit disk $|\lambda|<1$, then $\bar{U}$ is locally asymptotically stable. If one of them has a modulus greater than one, then $\bar{U}$ is unstable.

## 3. Global Stability

In this section, we investigate the stability of the two equilibrium points of system (5). When $a_{i} \in(0,1)$ for $i=1,2, \ldots, k$, the point $\bar{X}_{0}=\left(\bar{x}_{1}^{(1)}, \bar{x}_{1}^{(2)}, \ldots, \bar{x}_{1}^{(k)}\right)=$ $(0,0, \ldots, 0)$ is the unique nonnegative equilibrium point of system (5). When $a_{i} \in(1, \infty)$ for $i=1,2, \ldots, k$, the unique positive equilibrium point of system (5) is

$$
\begin{equation*}
\bar{X}_{a_{i}}=\left(\bar{x}_{2}^{(1)}, \bar{x}_{2}^{(2)}, \ldots, \bar{x}_{2}^{(k)}\right)=\left(\left(a_{k}-1\right)^{\left(1 / r_{k}\right)},\left(a_{1}-1\right)^{\left(1 / r_{1}\right)}, \ldots,\left(a_{k-1}-1\right)^{\left(1 / r_{k-1}\right)}\right) . \tag{12}
\end{equation*}
$$

Theorem 2. The following statements hold:
(i) If $a_{i} \in(0,1)$ for $i=1,2, \ldots, k$, then the equilibrium point $\left(\bar{x}_{1}^{(1)}, \bar{x}_{1}^{(2)}, \ldots, \bar{x}_{1}^{(k)}\right)$ of system (5) is locally asymptotically stable
(ii) If $a_{i} \in(1, \infty)$ for $i=1,2, \ldots, k$, then the equilibrium point $\left(\bar{x}_{1}^{(1)}, \bar{x}_{1}^{(2)}, \ldots, \bar{x}_{1}^{(k)}\right)$ of system (5) is unstable
(iii) If $a_{i} \in(1, \infty)$ for $i=1,2, \ldots, k$, then the positive equilibrium point $\left(\bar{x}_{2}^{(1)}, \bar{x}_{2}^{(2)}, \ldots, \bar{x}_{2}^{(k)}\right)$ of system (5) is unstable

Proof
(i) The characteristic equation of $J_{F}\left(\bar{X}_{0}\right)$ is given by

$$
\begin{equation*}
P(\lambda)=\lambda^{k}\left(\lambda^{2}-a_{1}\right)\left(\lambda^{2}-a_{2}\right), \ldots,\left(\lambda^{2}-a_{k}\right)=0 \tag{13}
\end{equation*}
$$

It is easy to see that if $a_{i} \in(0,1)$ for $i=1,2, \ldots, k$, then all the roots of the characteristic equation (13) lie in the open unit disk $|\lambda|<1$. So, the equilibrium point $\left(\bar{x}_{1}^{(1)}, \bar{x}_{1}^{(2)}, \ldots, \bar{x}_{1}^{(k)}\right)$ of (5) is locally asymptotically stable.
(ii) It is clearly seen that if $a_{i} \in(1, \infty)$ for $i=1,2, \ldots, k$, then some roots of characteristic equation (13) have absolute value greater than one. In this case, the equilibrium point $\left(\bar{x}_{1}^{(1)}, \bar{x}_{1}^{(2)}, \ldots, \bar{x}_{1}^{(k)}\right)$ of (5) is unstable.
(iii) The characteristic polynomial of $J_{F}\left(\bar{X}_{a_{i}}\right)$ is given by

$$
\begin{equation*}
P(\lambda)=\sum_{j=0}^{k}(-1)^{j}\binom{k}{j} \lambda^{3 k-2 j}+(-1)^{k+1} \prod_{i=1}^{k} \frac{r_{i}\left(a_{i}-1\right)}{a_{i}}, \tag{14}
\end{equation*}
$$

where $\binom{k}{j}$ is the binomial coefficient. It is clear that if $k$ is an odd number, then $P(\lambda)$ has a root in interval $(-\infty,-1)$ since

$$
\begin{align*}
P(-1) & =\prod_{i=1}^{k} \frac{r_{i}\left(a_{i}-1\right)}{a_{i}}>0  \tag{15}\\
\lim _{\lambda \longrightarrow-\infty} P(\lambda) & =-\infty
\end{align*}
$$

Also, if $k$ is an even number, then $P(\lambda)$ has a root in interval $(1, \infty)$ since

$$
\begin{align*}
P(1) & =-\prod_{i=1}^{k} \frac{r_{i}\left(a_{i}-1\right)}{a_{i}}<0,  \tag{16}\\
\lim _{\lambda \longrightarrow \infty} P(\lambda) & =\infty
\end{align*}
$$

So, from Theorem 1, we can say that if $a_{i} \in(1, \infty)$ for $i=1,2, \ldots, k$, then the positive equilibrium point $\left(\bar{x}_{2}^{(1)}, \bar{x}_{2}^{(2)}, \ldots, \bar{x}_{2}^{(k)}\right)$ of system (5) is unstable.

Theorem 3. If $a_{i} \in(0,1)$ for $i=1,2, \ldots, k$, then the equilibrium point $\left(\bar{x}_{1}^{(1)}, \bar{x}_{1}^{(2)}, \ldots, \bar{x}_{1}^{(k)}\right)$ of system (5) is globally asymptotically stable.

Proof. From Theorem 2, we know that if $a_{i} \in(0,1)$ for $i=1,2, \ldots, k$, then the equilibrium point $\left(\bar{x}_{1}^{(1)}, \bar{x}_{1}^{(2)}\right.$, $\ldots, \bar{x}_{1}^{(k)}$ ) of system (5) is locally asymptotically stable. Hence, it suffices to show that

$$
\begin{equation*}
\lim _{n \longrightarrow \infty}\left(x_{n}^{(1)}, x_{n}^{(2)}, \ldots, x_{n}^{(k)}\right)=(0,0, \ldots, 0) \tag{17}
\end{equation*}
$$

From system (5), we have that

$$
\begin{gather*}
0 \leq x_{n+1}^{(1)}=\frac{a_{1} x_{n-1}^{(1)}}{1+\left(x_{n-2}^{(2)}\right)^{r_{1}}} \leq a_{1} x_{n-1}^{(1)}, \\
0 \leq x_{n+1}^{(2)}=\frac{a_{2} x_{n-1}^{(2)}}{1+\left(x_{n-2}^{(3)}\right)^{r_{2}}} \leq a_{2} x_{n-1}^{(2)},  \tag{18}\\
\vdots \\
0 \leq x_{n+1}^{(k)}=\frac{a_{k} x_{n-1}^{(k)}}{1+\left(x_{n-2}^{(1)}\right)^{r_{k}}} \leq a_{k} x_{n-1}^{(k)},
\end{gather*}
$$

for $n \in \mathbb{N}_{0}$. From (18), we have by induction

$$
\begin{equation*}
0 \leq x_{2 n-l}^{(i)} \leq a_{i}^{n} x_{-l}^{(i)}, \tag{19}
\end{equation*}
$$

where $x_{-l}^{(i)}(l=0,1)$ for $i=1,2, \ldots, k$ are the initial conditions. Consequently, by taking limits of inequalities in (19) when $a_{i} \in(0,1)$ for $i=1,2, \ldots, k$, we have the limit in (17) which completes the proof.

## 4. Oscillation Behavior and Existence of Unbounded Solutions

In the following result, we are concerned with the oscillation of positive solutions of system (5) about the equilibrium point $\left(\bar{x}_{2}^{(1)}, \bar{x}_{2}^{(2)}, \ldots, \bar{x}_{2}^{(k)}\right)$.

Theorem 4. Assume that $a_{i} \in(1, \infty)$, and let $\left\{\left(x_{n}^{(1)}, x_{n}^{(2)}\right.\right.$, $\left.\left.\ldots, x_{n}^{(k)}\right)\right\}_{n=-2}^{\infty}$ be a positive solution of system (5) such that

$$
\begin{align*}
x_{-2}^{(i)}, x_{0}^{(i)} & \geq \bar{x}_{2}^{(i)},  \tag{20}\\
x_{-1}^{(i)} & <\bar{x}_{2}^{(i)},
\end{align*}
$$

or

$$
\begin{align*}
x_{-2}^{(i)}, x_{0}^{(i)} & <\bar{x}_{2}^{(i)}, \\
x_{-1}^{(i)} & \geq \bar{x}_{2}^{(i)}, \tag{21}
\end{align*}
$$

for $i=1,2, \ldots, k$. Then, $\left\{\left(x_{n}^{(1)}, x_{n}^{(2)}, \ldots, x_{(k)}^{(k)}\right)\right\}_{n=-2}^{\infty}$ oscillates about the equilibrium point $\left(\bar{x}_{2}^{(1)}, \bar{x}_{2}^{(2)}, \ldots, \bar{x}_{2}^{(k)}\right)$ with semicycles of length one.

Proof. Assume that (20) holds. (The case where (21) holds is similar and will be omitted.) From (5), we have

$$
\begin{align*}
x_{1}^{(1)}= & \frac{a_{1} x_{-1}^{(1)}}{1+\left(x_{-2}^{(2)}\right)^{r_{1}}}<\frac{a_{1} \bar{x}_{2}^{(1)}}{1+\left(\bar{x}_{2}^{(2)}\right)^{r_{1}}}=\bar{x}_{2}^{(1)}, x_{1}^{(2)}=\frac{a_{2} x_{-1}^{(2)}}{1+\left(x_{-2}^{(3)}\right)^{r_{2}}}<\frac{a_{2} \bar{x}_{2}^{(2)}}{1+\left(\bar{x}_{2}^{(3)}\right)^{r_{2}}}=\bar{x}_{2}^{(2)}, \vdots x_{1}^{(k)}=\frac{a_{k} x_{-1}^{(k)}}{1+\left(x_{-2}^{(1)}\right)^{r_{k}}}<\frac{a_{k} \bar{x}_{2}^{(k)}}{1+\left(\bar{x}_{2}^{(1)}\right)^{r_{k}}}=\bar{x}_{2}^{(k)}, x_{2}^{(1)} \\
& =\frac{a_{1} x_{0}^{(1)}}{1+\left(x_{-1}^{(2)}\right)^{r_{1}}} \geq \frac{a_{1} \bar{x}_{2}^{(1)}}{1+\left(\bar{x}_{2}^{(2)}\right)^{r_{1}}}=\bar{x}_{2}^{(1)}, x_{2}^{(2)}=\frac{a_{2} x_{0}^{(2)}}{1+\left(x_{-1}^{(3)}\right)^{r_{2}}} \geq \frac{a_{2} \bar{x}_{2}^{(2)}}{1+\left(\bar{x}_{2}^{(3)}\right)^{r_{2}}}=\bar{x}_{2}^{(2)}, \vdots x_{2}^{(k)}=\frac{a_{k} x_{0}^{(k)}}{1+\left(x_{-1}^{(1)}\right)^{r_{k}}} \geq \frac{a_{k} \bar{x}_{2}^{(k)}}{1+\left(\bar{x}_{2}^{(1)}\right)^{r_{k}}}=\bar{x}_{2}^{(k)} . \tag{22}
\end{align*}
$$

Then, the proof follows by induction.
In the following theorem, we show the existence of unbounded solutions for system (5).
Theorem 5. Assume that $a_{i} \in(1, \infty)$ for $i=1,2, \ldots, k$, then system (5) possesses an unbounded solution.

Proof. From Theorem 4, we can assume that, without loss of generality, the solution $\left\{\left(x_{n}^{(1)}, x_{n}^{(2)}, \ldots, x_{n}^{(k)}\right)\right\}_{n=-2}^{\infty}$ of system (5) is such that $x_{2 n-1}^{(i)}<\bar{x}_{2}^{(i)}$ and $x_{2 n}^{(i)}>\bar{x}_{2}^{(i)}$ for $i=1,2, \ldots, k$ and $n \in \mathbb{N}_{0}$. Then, we have

$$
\begin{gather*}
x_{2 n+2}^{(1)}=\frac{a_{1} x_{2 n}^{(1)}}{1+\left(x_{2 n-1}^{(2)}\right)^{r_{1}}}>\frac{a_{1} x_{2 n}^{(1)}}{1+\left(\bar{x}_{2}^{(2)}\right)^{r_{1}}}=\frac{a_{1} x_{2 n}^{(1)}}{1+\left(a_{1}-1\right)}=x_{2 n}^{(1)}, \\
x_{2 n+2}^{(2)}=\frac{a_{2} x_{2 n}^{(2)}}{1+\left(x_{2 n-1}^{(3)}\right)^{r_{2}}}>\frac{a_{2} x_{2 n}^{(2)}}{1+\left(\bar{x}_{2}^{(3)}\right)^{r_{2}}}=\frac{a_{2} x_{2 n}^{(2)}}{1+\left(a_{2}-1\right)}=x_{2 n}^{(2)}, \\
\vdots \\
x_{2 n+2}^{(k)}=\frac{a_{k} x_{2 n}^{(k)}}{1+\left(x_{2 n-1}^{(1)}\right)^{r_{k}}}>\frac{a_{k} x_{2 n}^{(k)}}{1+\left(\bar{x}_{2}^{(1)}\right)^{r_{k}}}=\frac{a_{k} x_{2 n}^{(k)}}{1+\left(a_{k}-1\right)}=x_{2 n}^{(k)}, \\
x_{2 n+3}^{(1)}=\frac{a_{1} x_{2 n+1}^{(1)}}{1+\left(x_{2 n}^{(2)}\right)^{r_{1}}}<\frac{a_{1} x_{2 n+1}^{(1)}}{1+\left(\bar{x}_{2}^{(2)}\right)^{r_{1}}}=\frac{a_{1} x_{2 n+1}^{(1)}}{1+\left(a_{1}-1\right)}=x_{2 n+1}^{(1)}, \\
x_{2 n+3}^{(2)}=\frac{a_{2} x_{2 n+1}^{(2)}}{1+\left(x_{2 n}^{(3)}\right)^{r_{2}}}<\frac{a_{2} x_{2 n+1}^{(2)}}{1+\left(\bar{x}_{2}^{(3)}\right)^{r_{2}}}=\frac{a_{2} x_{2 n+1}^{(2)}}{1+\left(a_{2}-1\right)}=x_{2 n+1}^{(2)}, \\
x_{2 n+3}^{(k)}=\frac{a_{k} x_{2 n+1}^{(k)}}{1+\left(x_{2 n}^{(1)}\right)^{r_{k}}}<\frac{a_{k} x_{2 n+1}^{(k)}}{1+\left(\bar{x}_{2}^{(1)}\right)^{r_{k}}}=\frac{a_{k} x_{2 n+1}^{(k)}}{1+\left(a_{k}-1\right)}=x_{2 n+1}^{(k)}, \tag{23}
\end{gather*}
$$

from which it follows that

$$
\begin{align*}
\lim _{n \longrightarrow \infty}\left(x_{2 n}^{(1)}, x_{2 n}^{(2)}, \ldots, x_{2 n}^{(k)}\right) & =(\infty, \infty, \ldots, \infty) \\
\lim _{n \longrightarrow \infty}\left(x_{2 n+1}^{(1)}, x_{2 n+1}^{(2)}, \ldots, x_{2 n+1}^{(k)}\right) & =(0,0, \ldots, 0), \tag{24}
\end{align*}
$$

which completes the proof.

## 5. Periodicity

In this section, we investigate the existence of period-two solution of system (5).

Theorem 6. If $a_{i}=1$ for $i=1,2, \ldots, k$, then system (5) possesses the prime period-two solution $\ldots,(0,0, \ldots, 0, \varphi),(0,0, \ldots, 0, \psi),(0,0, \ldots, 0, \varphi),(0,0, \ldots, 0, \psi), \ldots$,
with $\varphi, \psi>0$. Furthermore, every solution of system (5) converges to a period-two solution.

Proof. Assume that $a_{i}=1$ for $i=1,2, \ldots, k$, and let $\left\{\left(x_{n}^{(1)}, x_{n}^{(2)}, \ldots, x_{n}^{(k)}\right)\right\}_{n=-2}^{\infty}$ be a solution of system (5). Then, from system (5), we have

$$
\begin{aligned}
x_{2 n+1}^{(1)}= & \frac{x_{2 n-1}^{(1)}}{1+\left(x_{2 n-2}^{(2)}\right)^{r_{1}}}, \\
x_{2 n+2}^{(1)}= & \frac{x_{2 n}^{(1)}}{1+\left(x_{2 n-1}^{(2)}\right)^{r_{1}}}, \\
x_{2 n+1}^{(2)}= & \frac{x_{2 n-1}^{(2)}}{1+\left(x_{2 n-2}^{(3)}\right)^{r_{2}}}, \\
x_{2 n+2}^{(2)}= & \frac{x_{2 n}^{(2)}}{1+\left(x_{2 n-1}^{(3)}\right)^{r_{2}}}, \\
\vdots & \\
x_{2 n+1}^{(k)}= & \frac{x_{2 n-1}^{(k)}}{1+\left(x_{2 n-2}^{(1)}\right)^{r_{k}}}, \\
x_{2 n+2}^{(k)}= & \frac{x_{2 n}^{(k)}}{1+\left(x_{2 n-1}^{(1)}\right)^{r_{k}}},
\end{aligned}
$$

for $n \in \mathbb{N}_{0}$. From (26), we obtain

$$
\begin{aligned}
& x_{2 n-1}^{(1)}=x_{-1}^{(1)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-2}^{(2)}\right)^{r_{1}}}\right) \\
& x_{2 n}^{(1)}=x_{0}^{(1)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-1}^{(2)}\right)^{r_{1}}}\right) \\
& x_{2 n-1}^{(2)}=x_{-1}^{(2)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-2}^{(3)}\right)^{r_{2}}}\right)
\end{aligned}
$$

$$
\begin{equation*}
x_{2 n}^{(2)}=x_{0}^{(2)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-1}^{(3)}\right)^{r_{2}}}\right) \tag{27}
\end{equation*}
$$

$\vdots$

$$
\begin{aligned}
& x_{2 n-1}^{(k)}=x_{-1}^{(k)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-2}^{(1)}\right)^{r_{k}}}\right) \\
& x_{2 n}^{(k)}=x_{0}^{(k)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-1}^{(1)}\right)^{r_{k}}}\right)
\end{aligned}
$$

for $n \in \mathbb{N}_{0}$. If $x_{-l}^{(i)}=0$ for $l=0,1$ and $i=1,2, \ldots, k-1$, then $x_{n}^{(i)}=0$ for $i=1,2, \ldots, k-1$ and $\left(x_{2 n-1}^{(k)}, x_{2 n}^{(k)}\right)=\left(x_{-1}^{(k)}, x_{0}^{(k)}\right)$ for $n \in \mathbb{N}_{0}$. Therefore,
$\ldots,(0,0, \ldots, 0, \varphi),(0,0, \ldots, 0, \psi),(0,0, \ldots, 0, \varphi),(0,0, \ldots, 0, \psi), \ldots$
is a period-two solution of system (5) with $x_{-2}^{(k)}=x_{0}^{(k)}=$ $\varphi>0$ and $x_{-1}^{(k)}=\psi>0$. Furthermore, from (26), we have

$$
\begin{gather*}
x_{2 n+1}^{(1)}-x_{2 n-1}^{(1)}=-\frac{x_{2 n-1}^{(1)}\left(x_{2 n-2}^{(2)}\right)^{r_{1}}}{1+\left(x_{2 n-2}^{(2)}\right)^{r_{1}}} \leq 0, \\
x_{2 n+1}^{(2)}-x_{2 n-1}^{(2)}=-\frac{x_{2 n-1}^{(2)}\left(x_{2 n-2}^{(3)}\right)^{r_{2}}}{1+\left(x_{2 n-2}^{(3)}\right)^{r_{2}}} \leq 0,  \tag{29}\\
\vdots \\
x_{2 n+1}^{(k)}-x_{2 n-1}^{(k)}=-\frac{x_{2 n-1}^{(k)}\left(x_{2 n-2}^{(1)}\right)^{r_{k}}}{1+\left(x_{2 n-2}^{(1)}\right)^{r_{k}}} \leq 0, \\
x_{2 n+2}^{(1)}-x_{2 n}^{(1)}=-\frac{x_{2 n}^{(1)}\left(x_{2 n-1}^{(2)}\right)^{r_{1}}}{1+\left(x_{2 n-1}^{(2)}\right)^{r_{1}}} \leq 0, \\
x_{2 n+2}^{(2)}-x_{2 n}^{(2)}=-\frac{x_{2 n}^{(2)}\left(x_{2 n-1}^{(3)}\right)^{r_{2}}}{1+\left(x_{2 n-1}^{(3)}\right)^{r_{2}}} \leq 0,  \tag{30}\\
\vdots \\
x_{2 n+2}^{(k)}-x_{2 n}^{(k)}=-\frac{x_{2 n}^{(k)}\left(x_{2 n-1}^{(1)}\right)^{r_{k}}}{1+\left(x_{2 n-1}^{(1)}\right)^{r_{k}}} \leq 0 .
\end{gather*}
$$

From (29) and (30), we obtain $x_{2 n+1}^{(i)} \leq x_{2 n-1}^{(i)}$ and $x_{2 n+2}^{(i)} \leq x_{2 n}^{(i)}$ for $i=1,2, \ldots, k$. That is, the sequences $\left(x_{2 n-1}^{(i)}\right)$ and $\left(x_{2 n}^{(i)}\right)$ for $i=1,2, \ldots, k$ are nonincreasing. On the other hand, from (26), we have the inequalities

$$
\begin{aligned}
& x_{2 n-1}^{(1)}= x_{-1}^{(1)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-2}^{(2)}\right)^{r_{1}}}\right) \leq x_{-1}^{(1)}, \\
& x_{2 n}^{(1)}=x_{0}^{(1)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-1}^{(2)}\right)^{r_{1}}}\right) \leq x_{0}^{(1)}, \\
& x_{2 n-1}^{(2)}=x_{-1}^{(2)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-2}^{(3)}\right)^{r_{2}}}\right) \leq x_{-1}^{(2)}, \\
& x_{2 n}^{(2)}=x_{0}^{(2)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-1}^{(3)}\right)^{r_{2}}}\right) \leq x_{0}^{(2)}, \\
& x_{2 n-1}^{(k)}=x_{-1}^{(k)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-2}^{(1)}\right)^{r_{k}}}\right) \leq x_{-1}^{(k)}, \\
& x_{2 n}^{(k)}=x_{0}^{(k)} \prod_{i=0}^{n-1}\left(\frac{1}{1+\left(x_{2 i-1}^{(1)}\right)^{r_{k}}}\right) \leq x_{0}^{(k),}
\end{aligned}
$$

which show the boundedness of the solutions. Hence, the odd-index terms tend to one periodic point and the evenindex terms tend to another periodic point. This completes the proof.

## 6. Numerical Examples

In this section, we give some numerical examples to support our theoretical results related to system (5) with some restrictions on the parameters $a_{i}$ and $r_{i}$ for $i=1,2, \ldots k$.

Example 1. If $k=3, x_{n}^{(1)}=x_{n}, x_{n}^{(2)}=y_{n}, x_{n}^{(3)}=z_{n}, r_{1}=2$, $r_{2}=3$, and $r_{3}=4$ in system (5), we obtain the following system:

$$
\begin{align*}
& x_{n+1}=\frac{a_{1} x_{n-1}}{1+\left(y_{n-2}\right)^{2}}, \\
& y_{n+1}=\frac{a_{2} y_{n-1}}{1+\left(z_{n-2}\right)^{3}},  \tag{32}\\
& z_{n+1}=\frac{a_{3} z_{n-1}}{1+\left(x_{n-2}\right)^{4}} .
\end{align*}
$$

We visualize the solutions of system (32) in Figures 1-3 for the initial conditions $x_{-2}=1.34, x_{-1}=2.13, x_{0}=3.1$, $y_{-2}=0.17, y_{-1}=4.03, y_{0}=2.21, z_{-2}=0.32, z_{-1}=2.76$, and $z_{0}=3.12$.

Example 2. If $k=4, x_{n}^{(1)}=x_{n}, \quad x_{n}^{(2)}=y_{n}, \quad x_{n}^{(3)}=z_{n}$, $x_{n}^{(4)}=p_{n}, r_{1}=2, r_{2}=3, r_{3}=4$, and $r_{4}=5$ in system (5), we obtain the following system:

$$
\begin{align*}
& x_{n+1}=\frac{a_{1} x_{n-1}}{1+\left(y_{n-2}\right)^{2}}, \\
& y_{n+1}=\frac{a_{2} y_{n-1}}{1+\left(z_{n-2}\right)^{3}}, \\
& z_{n+1}=\frac{a_{3} z_{n-1}}{1+\left(p_{n-2}\right)^{4}},  \tag{33}\\
& p_{n+1}=\frac{a_{4} p_{n-1}}{1+\left(x_{n-2}\right)^{5}} .
\end{align*}
$$

We visualize the solutions of system (33) in Figures 4-6 for the initial conditions $x_{-2}=1.34, x_{-1}=2.13, x_{0}=3.1$, $y_{-2}=0.17, \quad y_{-1}=4.03, \quad y_{0}=2.21, \quad z_{-2}=0.32, \quad z_{-1}=2.76$, $z_{0}=3.12, p_{-2}=3.27, p_{-1}=1.33$, and $p_{0}=2.78$.

Example 3. If $k=5, x_{n}^{(1)}=x_{n}, \quad x_{n}^{(2)}=y_{n}, \quad x_{n}^{(3)}=z_{n}$, $x_{n}^{(4)}=p_{n}$, and $x_{n}^{(5)}=q_{n}$ in system (5), we obtain the following system:


Figure 1: The solutions of system (32) when $a_{1}=1.12, a_{2}=1.13$, and $a_{3}=1.14$.


Figure 2: The solutions of system (32) when $a_{1}=1, a_{2}=1$, and $a_{3}=1$.


Figure 3: The solutions of system (32) when $a_{1}=0.91, a_{2}=0.92$, and $a_{3}=0.93$.


Figure 4: The solutions of system (33) when $a_{1}=1.12, a_{2}=1.13, a_{3}=1.14$, and $a_{4}=1.15$.


Figure 5: The solutions of system (33) when $a_{1}=1, a_{2}=1, a_{3}=14$, and $a_{4}=1$.


Figure 6: The solutions of system (33) when $a_{1}=0.91, a_{2}=0.92, a_{3}=0.93$, and $a_{4}=0.94$.


Figure 7: The solutions of system (34) when $a_{1}=1.12, a_{2}=1.13, a_{3}=1.14, a_{4}=1.15$, and $a_{5}=1.16$.


Figure 8: The solutions of system (34) when $a_{1}=1, a_{2}=1, a_{3}=14, a_{4}=1$, and $a_{5}=1$.


Figure 9: The solutions of system (34) when $a_{1}=0.91, a_{2}=0.92, a_{3}=0.93, a_{4}=0.94$, and $a_{1}=0.95$.

$$
\left\{\begin{array}{l}
x_{n+1}=\frac{a_{1} x_{n-1}}{1+\left(y_{n-2}\right)^{2}}  \tag{34}\\
y_{n+1}=\frac{a_{2} y_{n-1}}{1+\left(z_{n-2}\right)^{3}} \\
z_{n+1}=\frac{a_{3} z_{n-1}}{1+\left(p_{n-2}\right)^{4}} \\
p_{n+1}=\frac{a_{4} p_{n-1}}{1+\left(q_{n-2}\right)^{5}} \\
q_{n+1}=\frac{a_{5} q_{n-1}}{1+\left(x_{n-2}\right)^{6}}
\end{array}\right.
$$

with $r_{1}=2, r_{2}=3, r_{3}=4, r_{4}=5$, and $r_{5}=6$. We visualize the solutions of system (34) in Figures 7-9 for the initial conditions $x_{-2}=1.34, x_{-1}=2.13, x_{0}=3.1, y_{-2}=0.17$, $y_{-1}=4.03, \quad y_{0}=2.21, \quad z_{-2}=0.32, \quad z_{-1}=2.76, \quad z_{0}=3.12$, $p_{-2}=3.27, p_{-1}=1.33, p_{0}=2.78, q_{-2}=0.32, q_{-1}=2.16$, and $q_{0}=3.91$.

## 7. Conclusion

In this study, we have generalized some of the results in the literature. As shown in Section 1, equation (1) was developed systematically. By this study, we ended this development. More concretely, we investigated the local asymptotic stability, global asymptotic stability, periodicity, and oscillation behavior of system (5) which is the $k$-dimensional generalization of equation (1). According to our findings, our results are consistent with the results of the paper [1] in the case of $k=1$. Similarly, our results are in line with the results of the papers $[2,3]$ in the case of $k=2$ and $k=3$, respectively.

## Data Availability

The data used to support the findings of this study are available from the first author upon request.

## Conflicts of Interest

The authors declare that there are no conflicts of interest associated with this publication.

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Research Article

# Second-Order Differential Equation: Oscillation Theorems and Applications 

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#### Abstract

Differential equations of second order appear in a wide variety of applications in physics, mathematics, and engineering. In this paper, necessary and sufficient conditions are established for oscillations of solutions to second-order half-linear delay differential equations of the form $\left(\varsigma(y)\left(u^{\prime}(y)\right)^{a}\right)^{\prime}+p(y) u^{c}(\vartheta(y))=0$, for $y \geq y_{0}$, under the assumption $\int^{\infty}(\varsigma(\eta))^{-(1 / a)}=\infty$. Two cases are considered for $a<c$ and $a>c$, where $a$ and $c$ are the quotients of two positive odd integers. Two examples are given to show the effectiveness and applicability of the result.


## 1. Introduction

Differential equations (DEs) have received a lot of attention, and it is an active research area among scientists and engineers $[1-8]$. The DEs have ability to formulate many complex phenomena in various fields such as biology, fluid mechanics, plasma physics, fluid mechanics, and optics; many exact and numerical schemes have been being derived such as [9-15]. Differential equation of second order appears in models as well as in physical applications such as fluid dynamics, electromagnetism, acoustic vibrations and quantum mechanics, biological, physical and chemical phenomena, optimization, mathematics of networks, and dynamical systems (see [16-24]).

In this article, we consider the differential equation

$$
\begin{equation*}
\left(\varsigma(y)\left(u^{\prime}(y)\right)^{a}\right)^{\prime}+p(y) u^{c}(\vartheta(y))=0, \quad \text { for } y \geq y_{0} \tag{1}
\end{equation*}
$$

where $a$ and $c$ are the quotient of two positive odd integers, and the functions $p, \varsigma$, and $\vartheta$ are continuous that satisfy the conditions stated below:
(A1) $\vartheta \in C([0, \infty), \mathbb{R}), \vartheta(y)<y, \lim _{y \rightarrow \infty} \vartheta(y)=\infty$.
(A2) $\varsigma \in C^{1}([0, \infty), \mathbb{R}), \quad p \in C([0, \infty), \mathbb{R}) ; \quad 0<\varsigma(y)$, $0 \leq p(y)$ for all $y \geq 0 ; p(y)$ is not identically zero in any interval $[b, \infty)$.
(A3) $\Upsilon(y)=\int_{y_{1}}^{y} \varsigma^{-1 / a}(\eta) \mathrm{d} \eta$ with $\lim _{y \rightarrow \infty} \Upsilon(y)=\infty$. (A4) the existence of a differentiable function $\vartheta_{0}$ such that $0<\vartheta_{0}(y) \leq \mathcal{\vartheta}(y)$, for $\vartheta_{0}^{\prime}(y) \geq \vartheta_{0}>0$, for $y \geq y_{0}$.
In [25, 26], Baculǐkovǎ and Džurina have considered

$$
\begin{align*}
& \left(\varsigma(y)\left(z^{\prime}(y)\right)^{a}\right)^{\prime}+p(y) u^{c}(\mathcal{V}(y))=0  \tag{2}\\
& z(y)=u(y)+q(y) u(\tau(y)), \quad y \geq y_{0}
\end{align*}
$$

and obtained oscillation criteria for the solutions of (2) using comparison techniques when $a=c=1,0 \leq q(y)<\infty$, and $\lim _{y \rightarrow \infty} \Upsilon(y)=\infty$. In the same technique, Džurina and Dzurina [27] have studied the oscillatory behavior of the solutions of (2) under the assumptions $0 \leq q(y)<\infty$ and $\lim _{y \rightarrow \infty} \Upsilon(y)=\infty$. In [28], Bohner et al. have studied the oscillatory behavior of solutions of (2) under $a=c$, $\lim _{y \rightarrow \infty} \Upsilon(y)<\infty$, and $0 \leq q(y)<1$. Grace et al. [29] have studied the oscillatory behavior of (2) when $a=c$ and $\lim _{y \rightarrow \infty} \Upsilon(y)<\infty \quad$ and $\quad \lim _{y \rightarrow \infty} \Upsilon(y)=\infty \quad$ and $0 \leq q(y)<1$. In [30], Ali has studied the oscillatory behavior of the solutions of (2), under the assumptions $\lim _{y \rightarrow \infty} \Upsilon(y)<\infty$ and $q(y) \geq 0$. Karpuz and Santra [31] have studied the oscillatory behavior of

$$
\begin{equation*}
\left(\varsigma(y)(u(y)+q(y) u(\tau(y)))^{\prime}\right)^{\prime}+p(y) f(u(\vartheta(y)))=0 \tag{3}
\end{equation*}
$$

by considering the assumptions $\lim _{y \rightarrow \infty} \Upsilon(y)<\infty$ and $\lim _{y \rightarrow \infty} \Upsilon(y)=\infty$ for different ranges of the neutral coefficient $q$.

For further work on the oscillation of this type of equations, we refer the readers to the references. Note that the majority of works consider only sufficient conditions, and merely a few consider both necessary and sufficient conditions. Hence, the objective of this work is to establish both necessary and sufficient conditions for the oscillation of solutions of (1) without using the comparison and the Riccati techniques. In this paper, we restrict our attention to the study (1), which includes the class of functional differential equations of neutral type.

Remark 1. When the domain is not specified explicitly, all functional inequalities considered in this paper are assumed to hold eventually, i.e., they are satisfied for all $y$ large enough.

## 2. Necessary and Sufficient Conditions

Lemma 1. Let (A1)-(A3) hold and that $u$ is an eventually positive solution of (1). Then, there exist $y_{1} \geq y_{0}$ and $d>0$ such that

$$
\begin{array}{r}
0<u(y) \leq d \Upsilon(y) \\
\Upsilon(y)\left[\int_{y}^{\infty} p(\zeta) u^{c}(\vartheta(\zeta)) \mathrm{d} \zeta\right]^{1 / a} \leq u(y) \tag{5}
\end{array}
$$

for $y \geq y_{1}$.

Proof. Let $u$ be an eventually positive solution of (1). Then, by (A1), there exists a $y^{*}$ such that $u(y)>0$ and $u(\vartheta(y))>0$ for all $y \geq y^{*}$. From (1) it follows that

$$
\begin{equation*}
\left(\varsigma(y)\left(u^{\prime}(y)\right)^{a}\right)^{\prime}=-p(y) u^{c}(\vartheta(y)) \leq 0 \tag{6}
\end{equation*}
$$

Therefore, $\varsigma(y)\left(u^{\prime}(y)\right)^{a}$ is nonincreasing for $y \geq y^{*}$. Next, we show that $\varsigma(y)\left(u^{\prime}(y)\right)^{a}$ is positive. By contradiction, assume that $\varsigma(y)\left(u^{\prime}(y)\right)^{a} \leq 0$ at a certain time
$y \geq y^{*}$. Using that $p$ is not identically zero on any interval $[b, \infty)$ and by (6), there exists $y_{1} \geq y^{*}$ such that

$$
\begin{equation*}
\varsigma(y)\left(u^{\prime}(y)\right)^{a} \leq \varsigma\left(y_{1}\right)\left(u^{\prime}\left(y_{1}\right)\right)^{a}<0, \quad \text { for all } y \geq y_{1} . \tag{7}
\end{equation*}
$$

Recall that $a$ is the quotient of two positive odd integers. Then,

$$
\begin{equation*}
u^{\prime}(y) \leq\left(\frac{\varsigma\left(y_{1}\right)}{\varsigma(y)}\right)^{1 / a} u^{\prime}\left(y_{1}\right), \quad \text { for } y \geq y_{1} \tag{8}
\end{equation*}
$$

Integrating from $y_{2}$ to $y$, we have

$$
\begin{equation*}
u(y) \leq u\left(y_{1}\right)+\left(\varsigma\left(y_{1}\right)\right)^{1 / a} u^{\prime}\left(y_{1}\right) \Upsilon(y) \tag{9}
\end{equation*}
$$

By (A3), the right-hand side approaches $-\infty$; then, $\lim _{y \rightarrow \infty} u(y)=-\infty$. This is a contradiction to the fact that $u(y)>0$. Therefore, $\varsigma(y)\left(u^{\prime}(y)\right)^{a}>0$ for all $y \geq y^{*}$. From $\varsigma(y)\left(u^{\prime}(y)\right)^{a}$ being nonincreasing, we have

$$
\begin{equation*}
u^{\prime}(y) \leq\left(\frac{\varsigma\left(y_{1}\right)}{\varsigma(y)}\right)^{1 / a} u^{\prime}\left(y_{1}\right), \quad \text { for } y \geq y_{1} \tag{10}
\end{equation*}
$$

Integrating this inequality from $y_{1}$ to $y$ and using that $u$ is continuous,

$$
\begin{equation*}
u(y) \leq u\left(y_{1}\right)+\left(\varsigma\left(y_{1}\right)\right)^{1 / a} u^{\prime}\left(y_{1}\right) \Upsilon(y) \tag{11}
\end{equation*}
$$

Since $\lim _{y \rightarrow \infty} \Upsilon(y)=\infty$, there exists a positive constant $d$ such that (4) holds.

Since $\varsigma(y)\left(u^{\prime}(y)\right)^{a}$ is positive and nonincreasing, $\lim _{y \rightarrow \infty} \varsigma(y)\left(u^{\prime}(y)\right)^{a}$ exists and is nonnegative. Integrating (1) from $y$ to $b$, we have

$$
\begin{equation*}
\varsigma(b)\left(u^{\prime}(b)\right)^{a}-\varsigma(y)\left(u^{\prime}(y)\right)^{a}+\int_{y}^{\infty} p(\eta) u^{c}(\vartheta(\eta)) \mathrm{d} \eta=0 \tag{12}
\end{equation*}
$$

Letting limit as $b \longrightarrow \infty$, we get

$$
\begin{equation*}
\varsigma(y)\left(u^{\prime}(y)\right)^{a} \geq \int_{y}^{\infty} p(\eta) u^{c}(\vartheta(\eta)) \mathrm{d} \eta \tag{13}
\end{equation*}
$$

Then,

$$
\begin{equation*}
u^{\prime}(y) \geq\left[\frac{1}{\varsigma(y)} \int_{y}^{\infty} p(\eta) u^{c}(\vartheta(\eta)) \mathrm{d} \eta\right]^{1 / a} \tag{14}
\end{equation*}
$$

Since $u\left(y_{1}\right)>0$, integrating the above inequality yields

$$
\begin{equation*}
u(y) \geq \int_{y_{1}}^{y}\left[\frac{1}{\varsigma(\eta)} \int_{\eta}^{\infty} p(\zeta) u^{c}(\vartheta(\zeta)) \mathrm{d} \zeta\right]^{1 / a} \mathrm{~d} \eta \tag{15}
\end{equation*}
$$

Since the integrand is positive, we can increase the lower limit of integration from $\eta$ to $y$ and then use the definition of $\Upsilon(y)$ to obtain

$$
\begin{equation*}
u(y) \geq \Upsilon(y)\left[\int_{y}^{\infty} p(\zeta) u^{c}(\vartheta(\zeta)) \mathrm{d} \zeta\right]^{1 / a} \tag{16}
\end{equation*}
$$

which yields (5).

Theorem 1. Assume that there exists a constant $b_{1}$, the quotient of two positive odd integers, such that $0<c<b_{1}<a$.

If (A1)-(A3) hold, then each solution of (1) is oscillatory if and only if

$$
\begin{equation*}
\int_{0}^{\infty} p(\zeta) \Upsilon^{c}(\vartheta(\zeta)) \mathrm{d} \zeta=\infty . \tag{17}
\end{equation*}
$$

Proof. On the contrary, we assume that $u$ is eventually positive solution. So, Lemma 1 holds, and then there exists $y_{1} \geq y_{0}$ such that

$$
\begin{equation*}
u(y) \geq \Upsilon(y) w^{1 / a}(y) \geq 0, \quad \text { for } y \geq y_{1} \tag{18}
\end{equation*}
$$

where

$$
\begin{equation*}
w(y)=\int_{y}^{\infty} p(\zeta) u^{c}(\vartheta(\zeta)) \mathrm{d} \zeta . \tag{19}
\end{equation*}
$$

Computing the derivative of $w$, we have

$$
\begin{equation*}
w^{\prime}(y)=-p(y) u^{c}(\mathcal{Y}(y)) . \tag{20}
\end{equation*}
$$

Thus, $w$ is nonnegative and nonincreasing. Since $u>0$, by (A2), it follows that $p(y) u^{c}(\vartheta(y))$ cannot be identically zero in any interval $[b, \infty)$; thus, $w^{\prime}$ cannot be identically zero, and $w$ cannot be constant on any interval $[b, \infty)$. Therefore, $w(y)>0$ for $y \geq y_{1}$. Computing the derivative, we have

$$
\begin{equation*}
\left(w^{1-b_{1} / a}(y)\right)^{\prime}=\left(1-\frac{b_{1}}{a}\right) w^{-b_{1} / a}(y) w^{\prime}(y) \tag{21}
\end{equation*}
$$

Integrating (21) from $y_{2}$ to $y$ and using that $w>0$, we have

$$
\begin{align*}
w^{1-b_{1} / a}\left(y_{2}\right) & \geq\left(1-\frac{b_{1}}{a}\right)\left[-\int_{y_{2}}^{y} w^{-b_{1} / a}(\zeta) w^{\prime}(\zeta) \mathrm{d} \zeta\right] \\
& =\left(1-\frac{b_{1}}{a}\right)\left[\int_{y_{2}}^{y} w^{-b_{1} / a}(\zeta)\left(p(\zeta) u^{c}(\vartheta(\zeta))\right) \mathrm{d} \zeta\right] . \tag{27}
\end{align*}
$$

Next, we find a lower bound for the right-hand side of (25), independent of the solution $u$. By (4) and (19), we have

$$
\begin{align*}
u^{c}(y) & =u^{c-b_{1}}(y) u^{b_{1}}(y) \geq(d \Upsilon(y))^{c-b_{1}} u^{b_{1}}(y) \\
& \geq(d \Upsilon(y))^{c-b_{1}}\left(\Upsilon(y) w^{1 / a}(y)\right)^{b_{1}}  \tag{23}\\
& =d^{c-b_{1}} \Upsilon^{c}(y) w^{b_{1} / a}(y), \quad \text { for } y \geq y_{2} .
\end{align*}
$$

Since $w$ is nonincreasing, $b_{1} / a>0$, and $\vartheta(\eta)<\eta$, it follows that

$$
\begin{align*}
u^{c}(\vartheta(\eta)) & \geq d^{c-b_{1}} \Upsilon^{c}(\vartheta(\eta)) w^{b_{1} / a}(\vartheta(\eta)) \\
& \geq d^{c-b_{1}} \Upsilon^{c}(\vartheta(\eta)) w^{b_{1} / a}(\eta) \tag{24}
\end{align*}
$$

Going back to (22), we have

$$
\begin{equation*}
w^{1-b_{1} / a}\left(y_{2}\right) \geq\left(1-\frac{b_{1}}{a}\right) d^{c-b_{1}}\left[\int_{y_{2}}^{y} p(\eta) \Upsilon^{c}(\vartheta(\eta)) \mathrm{d} \eta\right] . \tag{25}
\end{equation*}
$$

Since $\left(1-b_{1} / a\right)>0$, by (17) the right-hand side approaches $+\infty$ as $y \longrightarrow \infty$. This contradicts (25) and completes the proof of sufficiency for eventually positive solutions.

The eventually negative solution can be dealt similarly by introducing the variables $v=-u$.

Next, we show the necessity part by a contrapositive argument. If (17) does not hold, then for each $\kappa>0$ there exists $y_{1} \geq y_{0}$ such that

$$
\begin{equation*}
\int_{\eta}^{\infty} p(\zeta) \Upsilon^{c}(\mathcal{\vartheta}(\zeta)) \mathrm{d} \zeta \leq \frac{\kappa^{(1-c / a)}}{2} \tag{26}
\end{equation*}
$$

for all $\eta \geq y_{1}$. We define the set of continuous functions

$$
S=\left\{u \in C([0, \infty)):\left(\frac{\kappa}{2}\right)^{1 / a} \Upsilon(y) \leq u(y) \leq \kappa^{1 / a} \Upsilon(y), \quad y \geq y_{1}\right\} .
$$

We define an operator $\Omega$ on $S$ by

$$
(\Omega u)(y)= \begin{cases}0, & \text { if } y \leq y_{1}  \tag{28}\\ \int_{y_{1}}^{y}\left[\frac{1}{\varsigma(\eta)}\left[\frac{\kappa}{2}+\int_{\eta}^{\infty} p(\zeta) u^{c}(\vartheta(\zeta)) \mathrm{d} \zeta\right]\right]^{1 / a} \mathrm{~d} \eta, & \text { if } y>y_{1}\end{cases}
$$

Note that when $u$ is continuous, $\Omega u$ is also continuous on $[0, \infty)$. If $u$ is a fixed point of $\Omega$, i.e., $\Omega u=u$, then $u$ is a solution of (1).

First, we estimate $(\Omega u)(y)$ from below. By (A3), we have

$$
\begin{equation*}
(\Omega u)(y) \geq \int_{y_{1}}^{y}\left[\frac{1}{\varsigma(\eta)}\left[\frac{\kappa}{2}+0\right]\right]^{1 / a} \mathrm{~d} \eta=\left(\frac{\kappa}{2}\right)^{1 / a} \Upsilon(y) \tag{29}
\end{equation*}
$$

Now, we estimate $(\Omega u)(y)$ from above. For $u$ in $S$, we have $u^{c}(\vartheta(\zeta)) \leq\left(\kappa^{1 / a} \Upsilon(\vartheta(\zeta))\right)^{c}$. Then, by (26),

$$
\begin{align*}
(\Omega u)(y) & \leq \int_{y_{1}}^{y}\left[\frac{1}{\varsigma(\eta)}\left[\frac{\kappa}{2}+\int_{\eta}^{\infty} p(\zeta) u^{c}(\vartheta(\zeta)) \mathrm{d} \zeta\right]\right]^{1 / a} \mathrm{~d} \eta \\
& \leq \kappa^{1 / a} \Upsilon(y) . \tag{30}
\end{align*}
$$

Therefore, $\Omega$ maps $S$ to $S$.
Next, we find a fixed point for $\Omega$ in $S$. Let us define a sequence of functions in $S$ by the recurrence relation

$$
\begin{align*}
& v_{0}(y)=0, \quad \text { for } y \geq y_{0}, \\
& v_{1}(y)=\left(\Omega v_{0}\right)(y)= \begin{cases}0, & \text { if } y<y_{1}, \\
\kappa^{1 / a} \Upsilon(y), & \text { if } y \geq y_{1},\end{cases}  \tag{31}\\
& v_{n+1}(y)=\left(\Omega v_{n}\right)(y), \quad \text { for } n \geq 1, y \geq y_{1} .
\end{align*}
$$

Note that for each fixed $y$, we have $v_{1}(y) \geq v_{0}(y)$. Using mathematical induction, we can show that $v_{n+1}(y) \geq v_{n}(y)$. Therefore, the sequence $\left\{v_{n}\right\}$ converges pointwise to a function $v$. Using the Lebesgue dominated convergence theorem, we can show that $v$ is a fixed point of $\Omega$ in $S$. This shows under assumption (26), there is a nonoscillatory solution that does not converge to zero. This completes the proof.

Theorem 2. Assume that there exists a constant $b_{2}$, the quotient of two positive odd integers such that $0<a<b_{2}<c$. If (A1)-(A4) hold and $\varsigma(y)$ is nondecreasing, then each solution of (1) is oscillatory if and only if

$$
\begin{equation*}
\int_{y_{1}}^{\infty}\left[\frac{1}{\varsigma(s)} \int_{s}^{\infty} p(\zeta) \mathrm{d} \zeta\right]^{1 / a} \mathrm{~d} \eta=\infty \tag{32}
\end{equation*}
$$

Proof. On the contrary, we assume that $u$ is an eventually positive solution that does not converge to zero. Using the same argument as in Lemma 1, there exists $y_{1} \geq y_{0}$ such that $u(\vartheta(y))>0$ and $\varsigma(y)\left(u^{\prime}(y)\right)^{a}$ is positive and nonincreasing. Since $\varsigma(y)>0, u(y)$ is increasing for $y \geq y_{1}$. Using $u(y) \geq u\left(y_{1}\right)$, we have

$$
\begin{equation*}
u^{c}(y) \geq u^{c-b_{2}}(y) u^{b_{2}}(y) \geq u^{c-b_{2}}\left(y_{1}\right) u^{b_{2}}(y) \tag{33}
\end{equation*}
$$

and hence

$$
\begin{equation*}
u^{c}(\vartheta(y)) \geq u^{c-b_{2}}\left(y_{1}\right) u^{b_{2}}(\vartheta(y)), \quad \text { for } y \geq y_{2} \tag{34}
\end{equation*}
$$

Using (34) and $\vartheta(y) \geq \vartheta_{0}(y)$, from (13), we have

$$
\begin{equation*}
\varsigma(y)\left(u^{\prime}(y)\right)^{a} \geq u^{c-b_{2}}\left(y_{1}\right) u^{b_{2}}\left(\vartheta_{0}(y)\right) \int_{y}^{\infty} p(\eta) \mathrm{d} \eta \tag{35}
\end{equation*}
$$

for $y \geq y_{2}$. From $\varsigma(y)\left(u^{\prime}(y)\right)^{a}$ being nonincreasing and $\vartheta_{0}(y) \leq y$, we have

$$
\begin{equation*}
\varsigma\left(\vartheta_{0}(y)\right)\left(u^{\prime}\left(\vartheta_{0}(y)\right)\right)^{a} \geq \varsigma(y)\left(u^{\prime}(y)\right)^{a} . \tag{36}
\end{equation*}
$$

We use this in the left-hand side of (35). Then, dividing by $\varsigma\left(\vartheta_{0}(y)\right) u^{b_{2}}\left(\vartheta_{0}(y)\right)>0$ and raising both sides to the $1 / a$ power, we have

$$
\begin{equation*}
\frac{u^{\prime}\left(\vartheta_{0}((y))\right)}{u^{b_{2} / a}\left(\vartheta_{0}(y)\right)} \geq\left[\frac{u^{c-b_{2}}\left(y_{1}\right)}{\varsigma\left(\vartheta_{0}(y)\right)} \int_{y}^{\infty} p(\eta) \mathrm{d} \eta\right]^{1 / a} \tag{37}
\end{equation*}
$$

for $y \geq y_{2}$. Multiplying the left-hand side by $\vartheta_{0}^{\prime}(y) / \vartheta_{0} \geq 1$ and integrating from $y_{2}$ to $y$, we have

$$
\begin{align*}
& \frac{1}{\vartheta_{0}} \int_{y_{2}}^{y} \frac{u^{\prime}\left(\vartheta_{0}(\eta)\right) \vartheta_{0}^{\prime}(\eta)}{u^{b_{2} / a}\left(\vartheta_{0}(\eta)\right)} \mathrm{d} \eta \\
& \quad \geq u^{c-b_{2}}\left(y_{1}\right) \int_{y_{2}}^{y}\left[\frac{1}{\varsigma\left(\vartheta_{0}(\eta)\right)} \int_{\eta}^{\infty} p(\zeta) \mathrm{d} \zeta\right]^{1 / a} \mathrm{~d} \eta \tag{38}
\end{align*}
$$

On the left-hand side, since $a<b_{2}$, integrating, we have

$$
\begin{align*}
& \frac{1}{\alpha\left(1-b_{2} / a\right)}\left[z^{1-b_{2} / a}\left(\vartheta_{0}(\eta)\right)\right]_{s=y_{2}}^{y}  \tag{39}\\
& \quad \leq \frac{1}{\alpha\left(b_{2} / a-1\right)} z^{1-b_{2} / a}\left(\vartheta_{0}\left(y_{2}\right)\right)<\infty .
\end{align*}
$$

On the right-hand side of (38), we use that $\varsigma\left(\vartheta_{0}(\eta)\right) \leq \varsigma(\eta)$ to conclude that (32) implies the right-hand side approaching $+\infty$, as $y \longrightarrow \infty$, which is a contradiction. Hence, the solution $u$ cannot be eventually positive.

For eventually negative solutions, we use the same change of variables as in Theorem 1 and proceed as above.

To prove the necessity part, we assume that (32) does not hold and obtain an eventually positive solution that does not converge to zero. If (32) does not hold, then for each $\kappa>0$ there exists $y_{1} \geq y_{0}$ such that

$$
\begin{equation*}
\int_{y_{1}}^{\infty}\left[\frac{1}{\varsigma(\eta)} \int_{\eta}^{\infty} p(\zeta) \mathrm{d} \zeta\right]^{1 / a} \mathrm{~d} \eta \frac{\kappa^{(1-c / a)}}{2}, \quad \forall y \geq y_{1} \tag{40}
\end{equation*}
$$

We define the set of continuous function

$$
\begin{equation*}
S=\left\{u \in C([0, \infty)): \frac{\kappa}{2} \leq u(y) \leq \kappa \text { for } y \geq y_{1}\right\} . \tag{41}
\end{equation*}
$$

Then, we define the operator

$$
(\Omega u)(y)= \begin{cases}0, & \text { if } y \leq y_{1}  \tag{42}\\ \frac{\kappa}{2}+\int_{y_{1}}^{y} \frac{1}{\varsigma(\eta)}\left[\int_{\eta}^{\infty} p(\zeta) u^{c}(\vartheta(\zeta)) \mathrm{d} \zeta\right]^{1 / a} \mathrm{~d} \eta, & \text { if } y>y_{1}\end{cases}
$$

Note that if $u$ is continuous, $\Omega u$ is also continuous at $y=y_{1}$. Also, note that if $\Omega u=u$, then $u$ is solution of (1).

First, we estimate $(\Omega u)(y)$ from below. Let $u \in M$, we have $(\Omega u)(y) \geq \kappa / 2+0$, on $\left[y_{1}, \infty\right)$.

Now, we estimate $(\Omega u)(y)$ from above. Let $u \in M$. Then, $u \leq \kappa$ and by (40), we have

$$
\begin{align*}
(\Omega u)(y) & \leq \frac{\kappa}{2}+\kappa^{c / a} \int_{y_{1}}^{y}\left[\frac{1}{\varsigma(\eta)} \int_{\eta}^{\infty} p(\zeta) \mathrm{d} \zeta\right]^{1 / a} \mathrm{~d} \eta  \tag{43}\\
& \leq \frac{\kappa}{2}+\frac{\kappa}{2}=\kappa .
\end{align*}
$$

Therefore, $\Omega$ maps $S$ to $S$. To find a fixed point for $\Omega$ in $S$, we define a sequence of functions by the recurrence relation

$$
\begin{align*}
v_{0}(y) & =0, \quad \text { for } y \geq y_{0}, \\
v_{1}(y) & =\left(\Omega v_{0}\right)(y)=1, \quad \text { for } y \geq y_{1}  \tag{44}\\
v_{n+1}(y) & =\left(\Omega v_{n}\right)(y), \quad \text { for } n \geq 1, y \geq y_{1} .
\end{align*}
$$

Note that for each fixed $y$, we have $v_{1}(y) \geq v_{0}(y)$. Using mathematical induction, we can prove that $v_{n+1}(y) \geq v_{n}(y)$. Therefore, $\left\{v_{n}\right\}$ converges pointwise to a function $v$ in $S$. Then, $v$ is a fixed point of $\Omega$ and a positive solution of (1). The proof is completed.

Example 1. Consider the differential equations

$$
\begin{equation*}
\left(e^{-y}\left(u^{\prime}(y)\right)^{11 / 3}\right)^{\prime}+\frac{1}{y+1}(u(y-2))^{1 / 3}=0 . \tag{45}
\end{equation*}
$$

Here, $\quad a=11 / 3, \quad \varsigma(y)=e^{-y}, \quad \vartheta_{1}(y)=y-2, \quad \Upsilon(y)=$ $\int_{y_{1}}^{y} e^{11 \eta / 3} \mathrm{~d} \eta=(3 / 11)\left(e^{11 y / 3}-e^{11 y_{1} / 3}\right)$, and $c_{1}=1 / 3$. For $b \stackrel{y_{1}}{=} 7 / 3$, we have $0<c<b<a$. To check (17), we have

$$
\begin{align*}
& \int_{0}^{\infty} p(s) \Upsilon^{c}(\mathcal{\vartheta}(s)) \mathrm{d} s \\
& \quad=\int_{0}^{\infty} \frac{1}{\eta+1}\left(\frac{3}{11}\left(e^{11(s-2) / 3}-e^{11 y_{1} / 3}\right)\right)^{1 / 3} \mathrm{~d} s=\infty . \tag{46}
\end{align*}
$$

So, every conditions of Theorem 1 hold true, and therefore, all solutions of (45) are oscillatory or converge to zero.

Example 2. Consider the differential equations

$$
\begin{equation*}
\left(\left(u^{\prime}(y)\right)^{1 / 3}\right)^{\prime}+y(u(y-2))^{7 / 3}=0 \tag{47}
\end{equation*}
$$

Here, $a=1 / 3, \varsigma(y)=1, \vartheta_{1}(y)=y-2$, and $c_{1}=7 / 3$. For $b=5 / 3$, we have $c>b>a$. To check (32), we have

$$
\begin{equation*}
\int_{y_{0}}^{\infty}\left[\frac{1}{\varsigma(s)} \int_{s}^{\infty} p(\zeta) \mathrm{d} \zeta\right]^{1 / a} \mathrm{~d} s=\int_{2}^{\infty}\left[\int_{s}^{\infty} \zeta \mathrm{d} \zeta\right]^{3} \mathrm{~d} s=\infty \tag{48}
\end{equation*}
$$

So, every conditions of Theorem 2 hold true. Thus, all solutions of (47) are oscillatory or converge to zero.

## 3. Conclusion

The aim of this work is to establish necessary and sufficient conditions for the oscillation of solution to second-order half-linear differential equation. The obtained oscillation theorems complement the well-known oscillation results present in the literature. This work, as well as [31-41], leads us to pose an open problem: Can we find necessary and sufficient conditions for the oscillation of solutions to sec-ond-order differential equation

$$
\begin{array}{r}
\left.\left[r(t)\left((y(t)+p(t) y(\tau(t)))^{\prime}\right)^{\gamma}\right]^{\prime}+\sum_{i=1}^{m} q_{i}(t) y^{\alpha_{i}}\left(\tau_{i}(t)\right)\right)=0 \\
\text { for } p \in C\left(\mathbb{R}_{+}, \mathbb{R}\right) ? \tag{49}
\end{array}
$$

## Data Availability

No data were used to support the findings of this study.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

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Research Article

# The Stability Conditions for a Heavy Solid Motion 

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#### Abstract

In this paper, the stability conditions for the rotary motion of a heavy solid about its fixed point are considered. The center of mass of the body is assumed to lie on the moving $z$-axis which is assumed to be the minor axis of the ellipsoid of inertia. The nonlinear equations of motion and their three first integrals are obtained when the principal moments of inertia are distributed as $I_{1}<I_{2}<I_{3}$. We construct a Lyapunov function $L$ to investigate the stability conditions for this motion. We give a numerical example to illustrate the necessary and sufficient conditions for the stability of the body at certain moments of inertia. This problem has many important applications in different sciences.


## 1. Introduction

The problem of stability of a rotary motion of a heavy solid moving in a uniform gravity field or a rigid body moving in a Newtonian one is considered as one of the vital problems in the nonlinear dynamics. These problems were searched in the last half century. These problems are classified according to the bodies containing partially fluid or fully fluid or lie in attracting centers. In [1], the rotational motion about the center of mass for an asymmetric dynamic body with an aspheric cavity filled with liquid is studied. Numerical analysis for changing the kinetic momentum vector of the body was given. The extreme position for the stability of the axis of rotation of a body was obtained. A mass geometry solid of the Earth was attained. In [2], the authors studied the stability conditions of rotary motion for a heavy solid containing cavity filled fully or partially with a viscous fluid. The author in [3] considered the stability problem for a steady rotary motion of a gyro filled with liquid in a cavity. The author achieved the conditions of stability for this case of studied motion. The authors in [4] studied a branching for the stability of permanent rotatory motions of a rigid body filled with a viscous fluid. In [5], the stability problem of a permanent rotary motion of an asymmetric top about a fixed point under the influence of a central Newtonian force field
is considered for different cases. In [6], the evolution of the stability for the rotary motion of a planet containing a liquid core is considered. For some special cases of the orbiting motions for a planet, the stability conditions are obtained in [7]. The work of [8] gave the conditions for stability of a rotating unperturbed rigid body about a fixed point in different problems using the well-known energy first integral.

Moreover, the first integral related to energy was used in [8] for the aim of obtaining sufficient conditions of stability when the governing system of the body has not any perturbation. In [9], the authors evaluated the orbiting for a satellite containing a cavity filled with a viscous fluid. For both the uniform and the Newtonian force fields, the problems of stabilities of the orbiting bodies about fixed points are considered in some steady motion cases in [10]. In [11], the stability spin for a Lagrange top with linear oscillators is considered. In [12], the authors evaluated the plane periodic rotary motion stabilities of a symmetric rigid body moving about a fixed point when the principal moments of inertia $I_{1}=I_{2}=2 I_{3}$. The permanent orbiting stability problem of an asymmetric gyro is considered in [13] when the gyro moved under a Newtonian force field. In [14], the author considered the conditions of stability motion of a
rigid body rotating about the $x$-axis when $I_{1}>I_{2}>I_{3}$. He assumed a moment $\lambda_{1}$ about the $x$-axis and deduced the equations of motion. There are some new applications of nano-/microcomponents related to stability conditions in different application subjects that should be mentioned here. The applications such as the Chebyshev-Ritz technique for static stability and vibration analysis of nonlocal micro-structure-dependent nanostructures are given by Ebrahimi et al. [15].

Our article is searching the conditions of stability of a nonperturbed coupled heavy solid rotating about a fixed point in a uniform field of gravity. The body is assumed to rotate about the minor $z$-axis of the ellipsoid of inertia with couple $\lambda_{3}$. We consider the body rotates with principal moments of inertia to satisfy the condition $I_{1}<I_{2}<I_{3}$. We use the three well-known first integral of the problem for finding the stability conditions using a modified definition of Lyapunov function $L$. A numerical example is considered to show the stability conditions for a certain case.

## 2. Definition of the Problem

In this section, we define the problem of the motion of a heavy solid rotating about the minor $z$-axis of the ellipsoid of inertia. Consider the coordinate frame $O X Y Z$ is the fixed coordinate system in space and $O x y z$ is the fixed system in the body. Let the body rotates with an angular velocity vector $\underline{\omega}=\left(\omega_{1}, \omega_{2}, \omega_{3}\right)$ to the system Oxyz. Assume the couple $\underline{\lambda}=$ $\left(0,0, \lambda_{3}\right)$ to the moving coordinate system Oxyz. Let the mass center of the body lies on the $z$-axis (see Figure 1), and the moments of inertia of the body satisfy the conditions:

$$
\begin{equation*}
I_{1}<I_{2}<I_{3}, \quad z_{0} \neq 0, \tag{1}
\end{equation*}
$$

where $\left(0,0, z_{0}\right)$ is the position of the mass center for the origin.

Thus, the nonlinear system of equations of motion for this case and its first integrals are obtained as follows:

$$
\begin{gather*}
I_{1} \dot{\omega}_{1}+\left(I_{3}-I_{2}\right) \omega_{2} \omega_{3}+\lambda_{3} \omega_{2}=-m g z_{0} \gamma_{2} \\
I_{2} \dot{\omega}_{2}+\left(I_{1}-I_{3}\right) \omega_{3} \omega_{1}-\lambda_{3} \omega_{1}=m g z_{0} \gamma_{1}  \tag{2}\\
I_{3} \dot{\omega}_{3}+\left(I_{2}-I_{1}\right) \omega_{1} \omega_{2}=0 \\
\dot{\gamma}_{1}=\omega_{3} \gamma_{2}-\omega_{2} \gamma_{3} \\
\dot{\gamma}_{2}=\omega_{1} \gamma_{3}-\omega_{3} \gamma_{1}  \tag{3}\\
\dot{\gamma}_{3}=\omega_{2} \gamma_{1}-\omega_{1} \gamma_{2} \\
I_{1} \omega_{1}^{2}+I_{2} \omega_{2}^{2}+I_{3} \omega_{3}^{2}-2 m g z_{o} \gamma_{3}=E_{1} \\
I_{1} \omega_{1} \gamma_{1}+I_{2} \omega_{2} \gamma_{2}+\left(I_{3} \omega_{3}+\lambda_{3}\right) \gamma_{3}=H_{1}  \tag{4}\\
\gamma_{1}^{2}+\gamma_{2}^{2}+\gamma_{3}^{2}=1
\end{gather*}
$$

where the unit vector in the direction of the $Z$-axis is $\underline{\widehat{Z}}=\left(\gamma_{1}, \gamma_{2}, \gamma_{3}\right), m$ is the solid mass, $g$ is the gravity acceleration, $\cdot \equiv \mathrm{d} / \mathrm{d} t$ is the differentiation to time, and $E_{1}$ and $H_{1}$ are the initial energy and momentum constants. Assuming that $m g z_{0}=a$, equation (2) can be rewritten as follows:

$$
\begin{array}{r}
I_{1} \dot{\omega}_{1}=\left(I_{2}-I_{3}\right) \omega_{2} \omega_{3}-\lambda_{3} \omega_{2}-a \gamma_{2} \\
I_{2} \dot{\omega}_{2}=\left(I_{3}-I_{1}\right) \omega_{3} \omega_{1}+\lambda_{3} \omega_{1}+a \gamma_{1}  \tag{5}\\
I_{3} \dot{\omega}_{3}=\left(I_{1}-I_{2}\right) \omega_{1} \omega_{2}
\end{array}
$$

Assuming that the following special solutions satisfy the six nonlinear differential equations of motion (5) and (3),

$$
\begin{array}{r}
\omega_{1}=\omega_{2}=\gamma_{1}=\gamma_{2}=\lambda_{3}=0, \\
\omega_{3}=r=\mathrm{const}  \tag{6}\\
\gamma_{3}=1
\end{array}
$$

This is a case of uniform rotation of the solid about the minor axis of the ellipsoid of inertia ( $z$-axis).

## 3. Stability Investigation

In this section, the stability procedure of the unperturbed solid is investigated. Consider that the variables in (6) change according to

$$
\begin{align*}
\omega_{1} & =\Gamma_{1}, \\
\omega_{2} & =\Gamma_{2}, \\
\omega_{3}=r & +\Gamma_{3},  \tag{7}\\
\gamma_{1} & =\xi_{1}, \\
\gamma_{2} & =\xi_{2}, \\
\gamma_{3}=1 & +\xi_{3} .
\end{align*}
$$

Substituting the variables in (7) into the equations of motion (5) and (3), it gives

$$
\begin{array}{r}
I_{1} \dot{\Gamma}_{1}=\left(I_{2}-I_{3}\right) \Gamma_{2}\left(r+\Gamma_{3}\right)-a \xi_{2}, \\
I_{2} \dot{\Gamma}_{2}=\left(I_{3}-I_{1}\right)\left(r+\Gamma_{3}\right) \Gamma_{1}+a \xi_{1}, \\
I_{3} \dot{\Gamma}_{3}=\left(I_{1}-I_{2}\right) \Gamma_{1} \Gamma_{2}, \\
\dot{\xi}_{1}=\left(r+\Gamma_{3}\right) \xi_{2}-\Gamma_{2}\left(1+\xi_{3}\right),  \tag{8}\\
\dot{\xi}_{2}=\Gamma_{1}\left(1+\xi_{3}\right)-\left(r+\Gamma_{3}\right) \xi_{1}, \\
\dot{\xi}_{3}=\Gamma_{2} \xi_{1}-\Gamma_{1} \xi_{2},
\end{array}
$$

Substituting the variables in (7) into the first integrals (4), it gives

$$
\begin{gather*}
V_{1}=I_{1} \Gamma_{1}^{2}+I_{2} \Gamma_{2}^{2}+I_{3} \Gamma_{3}^{2}+2 I_{3} r \Gamma_{3}-2 a \xi_{3} \\
V_{2}=I_{1} \Gamma_{1} \xi_{1}+I_{2} \Gamma_{2} \xi_{2}+I_{3} \Gamma_{3} \xi_{3}+I_{3} r \xi_{3}  \tag{9}\\
V_{3}=\xi_{1}^{2}+\xi_{2}^{2}+\xi_{3}^{2}+2 \xi_{3}
\end{gather*}
$$

where $V_{1}, V_{2}$, and $V_{3}$ are constants.
We make a linear combination of the relations in (9) to satisfy the Lyapunov and Chetayev stability procedures. the Lyapunov function $L$ is modified to be in the form:

$$
\begin{equation*}
L=V_{1}-2 r V_{2}+\left(I_{3} r^{2}+a\right) V_{3} . \tag{10}
\end{equation*}
$$

Thus, using the relations in (9), we obtain

$$
\begin{equation*}
L=L_{1}+L_{2}+L_{3} \tag{11}
\end{equation*}
$$



Figure 1: Description of forces.
where

$$
\begin{align*}
& L_{1}=I_{1} \Gamma_{1}^{2}-2 I_{1} r \Gamma_{1} \xi_{1}+\left(I_{3} r^{2}+a\right) \xi_{1}^{2} \\
& L_{2}=I_{2} \Gamma_{2}^{2}-2 I_{2} r \Gamma_{2} \xi_{2}+\left(I_{3} r^{2}+a\right) \xi_{2}^{2}  \tag{12}\\
& L_{3}=I_{3} \Gamma_{3}^{2}-2 I_{3} r \Gamma_{3} \xi_{3}+\left(I_{3} r^{2}+a\right) \xi_{3}^{2} .
\end{align*}
$$

We deduce from the relations in (12) that $L_{i}>0, i=1,2,3$, if and only if the following inequalities are satisfied [14], respectively:

$$
\begin{align*}
& \left|\begin{array}{cc}
I_{1} & -I_{1} r \\
-I_{1} r & \left(I_{3} r^{2}+a\right)
\end{array}\right|=I_{1}\left(I_{3} r^{2}+a\right)-I_{1}{ }^{2} r^{2}>0, \\
& \left|\begin{array}{cc}
I_{2} & -I_{2} r \\
-I_{2} r & \left(I_{3} r^{2}+a\right)
\end{array}\right|=I_{2}\left(I_{3} r^{2}+a\right)-I_{2}{ }^{2} r^{2}>0,  \tag{13}\\
& \left|\begin{array}{cc}
I_{3} & -I_{3} r \\
-I_{3} r & \left(I_{3} r^{2}+a\right)
\end{array}\right|=I_{3}\left(I_{3} r^{2}+a\right)-I_{3}{ }^{2} r^{2}=I_{3} a>0 .
\end{align*}
$$

Thus, we obtain from the back substitution

$$
\begin{align*}
& a>0 \\
& I_{3}>I_{2}  \tag{14}\\
& I_{3}>I_{1}
\end{align*}
$$

We note that when condition (1) is satisfied, inequalities (13) and (14) are investigated automatically.

Thus, the conditions in (13) considered the sufficient and necessary conditions for the unperturbed stability motion of the solid concerning the variables $\omega_{1}, \omega_{2}, \omega_{3}, \gamma_{1}, \gamma_{2}$, and $\gamma_{3}$.

## 4. A Simple Application Example

A simple application example can be given in this section. Just formulation the sufficient and necessary conditions for the unperturbed stability motion are studied.

Consider a rigid body of mass $M$ with moments of inertia as follows:

$$
\begin{align*}
I_{1} & =10 \mathrm{~kg} \cdot \mathrm{~m}^{2} \\
I_{2} & =15 \mathrm{~kg} \cdot \mathrm{~m}^{2} \\
I_{3} & =20 \mathrm{~kg} \cdot \mathrm{~m}^{2} \\
M & =100 \mathrm{~kg}  \tag{15}\\
g & =9.81 \mathrm{~m} \cdot \mathrm{~s}^{-2} \\
z_{o} & =50 \mathrm{~m}, \\
r & =10 \mathrm{rad} \cdot \mathrm{~s}^{-1} .
\end{align*}
$$

From (15), we deduce that

$$
\begin{align*}
& a=49050 \mathrm{kgm}^{2} \cdot \mathrm{~s}^{-2}, \\
& \left(I_{3} r^{2}+a\right)=51050 . \tag{16}
\end{align*}
$$

From (13), (14), and (15), we deduce that

$$
\begin{array}{r}
I_{1}\left(I_{3} r^{2}+a\right)-I_{1}{ }^{2} r^{2}=500500>0, \\
I_{2}\left(I_{3} r^{2}+a\right)-I_{2}{ }^{2} r^{2}=743250>0  \tag{17}\\
I_{3} a=981000>0
\end{array}
$$

From this example, we deduce that if condition (1) is satisfied by (15), $L_{i}>0, i=1,2,3$, is satisfied if and only if relation (17) is satisfied.

## 5. Conclusion

The necessary and sufficient conditions for the stability of the rotary motion of a solid about a fixed point are considered when the center of mass of the body lies on the minor $z$-axis of the ellipsoid of inertia. The system of equations of motion and their first integrals are obtained. The modified Lyapunov function $L$ is constructed depending on the three first integrals. The necessary and sufficient conditions are constructed in the case $I_{3}>I_{2}>I_{1}$ that the $x$-axis is the major axis of the ellipsoid of inertia and the $y$-axis is the intermediate axis while the $z$-axis is the minor one. The moments of inertia $I_{\mathrm{i}}$, $i=1,2,3$ are assumed as arbitrary constants that satisfy the above condition, and $\lambda_{3}$ is estimated by substituting (7) into the second equation from (4). This problem has wide applications in satellite motions which assumed moments of inertia according to the above relations [16-18]. A numerical example is given to show the stability conditions for a special case.

## Data Availability

Data sharing not applicable to this article as no datasets were generated or analyzed during the current study.

## Conflicts of Interest

The author declares he has no conflicts of interest.

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# Existence and Uniqueness of Uncertain Fractional Backward Difference Equations of Riemann-Liouville Type 

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#### Abstract

In this article, we consider the analytic solutions of the uncertain fractional backward difference equations in the sense of Riemann-Liouville fractional operators which are solved by using the Picard successive iteration method. Also, we consider the existence and uniqueness theorem of the solution to an uncertain fractional backward difference equation via the Banach contraction fixed-point theorem under the conditions of Lipschitz constant and linear combination growth. Finally, we point out some examples to confirm the validity of the existence and uniqueness of the solution.


## 1. Introduction

Fractional calculus is based on an old idea that has become important and popular in applications only recently. The idea is to generalize integration and differentiation to noninteger orders in order to develop and extend the theory of calculus and to describe a more extensive range of possible doings in reality. During the past decades, fractional differential equations have been widely employed in many fields: mathematical analysis, optics and thermal systems, control engineering, and robotics, see, for example, [1-9].

In recent years, uncertain fractional differential and difference equations and discrete difference equations have become popular in both theory and applications. These represent a new area for researchers which was
developed slowly in their early stages. By using modeling techniques with discrete fractional calculus, some researchers established the existence, uniqueness, monotonicity, multiplicity, and qualitative properties of solutions to uncertain fractional difference equations (UFDEs) in the sense of Riemann-Liouville, Caputo, and $A B$ operators; for further details, see [10-22] and the references cited therein.

The aim of this attempt is to investigate the existence and uniqueness of fractional difference equations in the sense of Riemann-Liouville-like difference operator with assuming Lipschitz condition on its nonlinear term. Our findings are partial continuation of some results obtained in [23-25]. It is worth mentioning that the uncertainty theory of fractional difference equations is used to make the problems have a unique solution almost surely.

## 2. Preliminaries

This section presents some preliminaries, definitions, and facts in the field of uncertainty theory and discrete fractional calculus, see, e.g., [12, 26-28]. Throughout the article, we consider $\mathbb{N}_{a}=\{a, a+1, a+2, \ldots$,$\} for a \in \mathbb{R}$ and the backward jump operator $\rho(r):=r-1$ for $r \in \mathbb{N}_{a}$.

Definition 1 (see [26]). For any function $f: \mathbb{N}_{a} \longrightarrow \mathbb{R}$, the backward difference operator is defined by

$$
\begin{equation*}
\nabla f(t)=f(t)-f(\rho(t)), \quad t \in \mathbb{N}_{a} \tag{1}
\end{equation*}
$$

while the backward sum is given by

$$
\begin{equation*}
\nabla_{a}^{-1} f(t)=\sum_{r=a+1}^{t} f(r), \quad t \in \mathbb{N}_{a+1} \tag{2}
\end{equation*}
$$

Definition 2 (see [26-29]). For any natural number $j$, the $\nabla$-rising factorial function of $t$ is defined by

$$
\begin{equation*}
t^{\bar{j}}=\prod_{\ell=0}^{j-1}(t+\ell), \quad t^{\overline{0}}=1 \tag{3}
\end{equation*}
$$

Moreover, for any $\nu \in \mathbb{R}$, the $\nabla$-rising factorial function is defined by

$$
\begin{equation*}
t^{\bar{\nu}}=\frac{\Gamma(t+\nu)}{\Gamma(t)}, \quad 0^{\bar{v}}=0 \tag{4}
\end{equation*}
$$

for $t \in \mathbb{R} \backslash\{\ldots,-2,-1,0\}$. Also, note that the division by negative integer poles of the gamma function gives zero.

A major property of the rising factorial function is as follows:

$$
\begin{equation*}
\nabla\left(t^{\bar{\nu}}\right)=v t^{\overline{\nu-1}} \tag{5}
\end{equation*}
$$

This implies that $t^{\bar{v}}$ is increasing on $\mathbb{N}_{0}$ such that $v>0$.

Definition 3 (see [13, 14, 26-29]). For any function $f: \mathbb{N}_{a} \longrightarrow \mathbb{R}$, the nabla fractional sum of order $v>0$ in the sense of Riemann-Liouville is defined by

$$
\begin{align*}
\left(\nabla_{a}^{-v} f\right)(t) & =\frac{1}{\Gamma(v)} \sum_{r=a+1}^{t}(t-\rho(r))^{\overline{v-1}} f(r),  \tag{6}\\
\left(\nabla_{a}^{0} f\right)(t) & =f(t) .
\end{align*}
$$

Lemma 1 (see [13, 14, 26-28]). For any function $f$ defined on $\mathbb{N}_{a}$ and any $\nu, \alpha>0$, we have

$$
\begin{equation*}
\left(\nabla_{a}^{-\nu} \nabla_{a}^{-\alpha} f\right)(t)=\left(\nabla_{a}^{-(\alpha+\nu)} f\right)(t)=\left(\nabla_{a}^{-\alpha} \nabla_{a}^{-\gamma} f\right)(t) \tag{7}
\end{equation*}
$$

Lemma 2 (see [13, 14, 26-28]). For any function $f$ defined on $\mathbb{N}_{a}$ and any $\nu>0$, we have

$$
\begin{equation*}
\left(\nabla_{a}^{v} f\right)(t)=\left(\nabla \nabla_{a}^{-(1-v)} f\right)(t) \tag{8}
\end{equation*}
$$

Lemma 3 (see [13, 14, 26-28]). For any function $f$ defined on $\mathbb{N}_{a}$ and any $\nu>0$, we have

$$
\begin{equation*}
\left(\nabla_{a}^{-v} \nabla f\right)(t)=\left(\nabla \nabla_{a}^{-v} f\right)(t)-\frac{(t-a)^{\overline{v-1}}}{\Gamma(\nu)} f(a) \tag{9}
\end{equation*}
$$

Lemma 4 (see [13, 14, 26-28]). For any function $f$ defined on $\mathbb{N}_{a}$ and any $\nu>0$, we have

$$
\begin{align*}
& \left(\nabla_{a}^{-v} \nabla^{v} f\right)(t)=\left(\nabla_{a}^{v} \nabla^{-v} f\right)(t)=f(t), \quad v \notin \mathbb{N}, \\
& \left(\nabla_{a}^{-v} \nabla^{v} f\right)(t)=f(t)-\sum_{k=0}^{v-1} \frac{(t-a)^{\bar{k}}}{k!} \nabla^{k} f(a), \quad v \in \mathbb{N} . \tag{10}
\end{align*}
$$

Lemma 5 (see [13, 14, 26-28]). For any $a \in \mathbb{R}$ and $\nu, \alpha>0$, we have

$$
\begin{equation*}
\nabla_{a}^{-v}(t-a)^{\bar{\alpha}}=\frac{\Gamma(\alpha+1)}{\Gamma(\alpha+v+1)}(t-a)^{\overline{\alpha+v}} \tag{11}
\end{equation*}
$$

Motivated by the definition of the $n$ th-order backward sum for uncertain sequence $\xi_{t}$, we define the $\nu$ th-order backward sum for uncertain sequence $\xi_{t}$ as follows:

Definition 4 (see $[13,14,28]$ ). Let $v>0, a \in \mathbb{R}$, and $\xi_{t}$ be an uncertain sequence indexed by $t \in \mathbb{N}_{a}$. Then, we have

$$
\begin{equation*}
\nabla_{a}^{-v} \xi_{t}=\frac{1}{\Gamma(\nu)} \sum_{r=a+1}^{t}(t-\rho(r))^{\overline{\nu-1}} \xi_{r} \tag{12}
\end{equation*}
$$

which is called the $\nu$ th-order backward fractional sum of uncertain sequence $\xi_{t}$.

Definition 5 (see [13, 14, 28]). For any $\nu>0$, the fractional Riemann-Liouville-like backward difference for uncertain sequence $\xi_{t}$ is defined by

$$
\begin{equation*}
\nabla_{a}^{\nu} \xi_{t}=\nabla_{a}\left(\nabla_{a}^{-(1-\nu)} \xi_{t}\right) \tag{13}
\end{equation*}
$$

Next, we recall the definition of nabla discrete MittagLeffler (ML).

Definition 6 (see [28]). For any $\lambda \in \mathbb{R}$ and $\nu, \mu, \eta \in \mathbb{C}$ with $\operatorname{Re}(\nu)>0$, the two-parameter discrete ML function is defined by

$$
\begin{equation*}
E_{\overline{\nu, \mu}}(\lambda, \eta):=\sum_{\ell=0}^{\infty} \lambda^{\ell} \frac{\eta^{\overline{\ell \nu+\mu-1}}}{\Gamma(\ell \nu+\mu)}, \quad|\lambda|<1 . \tag{14}
\end{equation*}
$$

Particularly, if $\mu=1$, we get the one-parameter discrete ML function:

$$
\begin{equation*}
E_{\bar{\nu}}(\lambda, \eta):=\sum_{\ell=0}^{\infty} \lambda^{\ell} \frac{\eta^{\overline{\ell \nu}}}{\Gamma(\ell v+1)}, \quad|\lambda|<1 . \tag{15}
\end{equation*}
$$

## 3. UFBDE and Existence and Uniqueness Theorem

First, we recall the inverse uncertainty distribution theory.
Definition 7. (see [11]). An uncertainty distribution $\Psi$ is called regular if it is a continuous and strictly increasing function and satisfies

$$
\begin{align*}
& \lim _{x \xrightarrow[\longrightarrow]{ }} \Psi(x)=0 \\
& \lim _{x \longrightarrow+\infty} \Psi(x)=1 . \tag{16}
\end{align*}
$$

Definition 8. (see [11]). Let $\xi$ be an uncertain variable with a regular uncertainty distribution $\Psi$. Then, the inverse function $\Psi^{-1}$ is called the inverse uncertainty distribution of $\xi$.

Example 1. From Definition 8, we deduce that the following:
(i) The inverse uncertainty distribution of a linear uncertain variable $\mathscr{L}(a, b)$ is given by

$$
\begin{equation*}
\Psi^{-1}(\alpha)=(1-\alpha) a+\alpha b \tag{17}
\end{equation*}
$$

(ii) The inverse uncertainty distribution of a normal uncertain variable $\mathscr{L}(e, \sigma)$ is given by

$$
\begin{equation*}
\Psi^{-1}(\alpha)=e+\frac{\sqrt{3} \sigma}{\pi} \ln \left(\frac{\alpha}{1-\alpha}\right) \tag{18}
\end{equation*}
$$

(iii) The inverse uncertainty distribution of a normal uncertain variable $\mathscr{L} \mathcal{O} \mathscr{G} \mathcal{N}(e, \sigma)$ is given by

$$
\begin{equation*}
\Psi^{-1}(\alpha)=\exp (e)+\left(\frac{\alpha}{1-\alpha}\right)^{(\sqrt{3} \sigma / \pi)} \tag{19}
\end{equation*}
$$

Definition 9. (see [11]). We say that an uncertain variable $\xi$ is symmetrical if

$$
\begin{equation*}
\Psi(x)+\Psi(-x)=1 \tag{20}
\end{equation*}
$$

where $\Psi(x)$ is a regular uncertainty distribution of $\xi$.
Remark 1. From Definition 9, we can deduce that the symmetrical uncertain variable has the inverse uncertainty distribution $\Psi^{-1}(\alpha)$ that satiates

$$
\begin{equation*}
\Psi^{-1}(\alpha)+\Psi^{-1}(1-\alpha)=0 \tag{21}
\end{equation*}
$$

Example 2. From Definition 9, we deduce that the following:
(1) The linear uncertain variable $\mathscr{L}(-a, a)$ is symmetrical for any positive real number $a$
(2) The normal uncertain variable $\mathscr{L}(0,1)$ is symmetrical

Definition 10. (see [11]; i.i.d. definition). In statistics and probability theory, a collection of random variables $\xi_{i} s$ is independent and identically distributed (or briefly, i.i.d.) if each random variable $\xi_{i}$ has the same probability distribution as the others and all are mutually independent.

Then, we state the definition of the UFBDE.
Definition 11. An uncertain fractional difference equation is a fractional difference equation which is driven by an uncertain sequence. Moreover, an uncertain fractional backward difference equation for the Riemann-Liouville type is the uncertain fractional difference equation with Rie-mann-Liouville-like backward difference.

Consider the following generalized Riemann-Liouville fractional difference equation:

$$
\begin{equation*}
\left(\nabla_{\alpha-1}^{\alpha} y\right)(t)=G(t, y(t))+H(t, y(t)) \xi_{t}, \tag{22}
\end{equation*}
$$

subject to the initial condition (i.c.)

$$
\begin{equation*}
\left.\left(\nabla_{\alpha-1}^{-(1-\alpha)} y\right)(t)\right|_{t=\alpha}=y(\alpha) \tag{23}
\end{equation*}
$$

where $\nabla_{\alpha-1}^{\alpha}$ denotes fractional Riemann-Liouville-like backward difference with $0<\alpha<1, G, H$ are two real-valued functions defined on $[1, \infty] \times \mathbb{R}, t \in \mathbb{N}_{1} \cap[1, T+1]$, $y(\alpha) \in \mathbb{R}$ is a crisp number, and $\xi_{1}, \xi_{2}, \ldots, \xi_{T+1}$ are ( $T+1$ )-i.i.d. uncertain variables with symmetrical uncertainty distribution $\mathscr{L}(a, b)$.

Remark 2. Observe that the i.i.d. uncertain variables are those uncertain variables that are independent and have the same uncertainty distribution. See [11] for more detail.

Lemma 6. Initial value problem (22) with i.c. (23) is equivalent to the following uncertain fractional sum equation:

$$
\begin{align*}
y(t)= & \frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha) \\
& +\frac{1}{\Gamma(\alpha)} \sum_{r=\alpha+1}^{t}(t-\rho(r))^{\overline{\alpha-1}}\left[G(r, y(r))+H(r, y(r)) \xi_{r}\right] \tag{24}
\end{align*}
$$

for $t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]$.
Proof. The proof is very similar to [29], Lemma 5.1, and [30], Theorem 2, hereby applying the operator $\nabla_{\alpha}^{-\alpha}$ to IVP (22) with Definition 3, Lemma 2, and Lemma 3, so we omit this.

In this paper, the following special linear UFBDE will be considered:

$$
\begin{align*}
\nabla_{\alpha-1}^{\alpha} y(t) & =\lambda y(t)+\lambda \xi_{t}  \tag{25}\\
\left.\nabla_{\alpha-1}^{-(1-\alpha)} y(t)\right|_{t=\alpha} & =y(\alpha) \tag{26}
\end{align*}
$$

for $t \in \mathbb{N}_{1} \cap[1, T+1], n \in \mathbb{N}_{1}$, and $\lambda \in(0,1)$.
Remark 3. The following identity is useful in proving the upcoming theorem. From Definition 3 and Lemma 5, we can deduce for any real number $a$

$$
\begin{align*}
\nabla_{a}^{-\alpha}(t-a+1)^{\bar{\beta}}= & \frac{\Gamma(\beta+1)}{\Gamma(\alpha+\beta+1)}(t-a+1)^{\overline{\alpha+\beta}}  \tag{27}\\
& -\frac{\Gamma(\beta+1)}{\Gamma(\alpha)}(t-a+1)^{\overline{\alpha-1}}
\end{align*}
$$

where $\alpha>0$ and $\beta>-1$.

Theorem 1. For any $t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]$ and $|\lambda|<1$, linear UFBDE (25) with the initial condition (26) has a solution

$$
\begin{equation*}
y(t)=(1-\lambda) E_{\overline{\alpha, \alpha}}(\lambda, t-\alpha+1) y(\alpha)+\xi_{t}, \tag{28}
\end{equation*}
$$

where $\xi_{t}$ is an uncertain sequence with the uncertainty distribution $\mathscr{L}\left(a \cdot \lambda E_{\overline{\alpha, \alpha+1}}(\lambda, t-\alpha), b \cdot \lambda E_{\overline{\alpha, \alpha+1}}(\lambda, t-\alpha)\right)$.

Proof. Applying the operator $\nabla_{\alpha}^{-\alpha}$ to equation (25), we get

$$
\begin{align*}
&\left.\nabla_{\alpha}^{-\alpha}\left(\nabla_{\alpha-1}^{\alpha} y(t)\right)=\lambda \nabla_{\alpha}^{-\alpha} y(t)\right)+\lambda \nabla_{\alpha}^{-\alpha} \xi_{t}  \tag{29}\\
& t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]
\end{align*}
$$

Making use of Lemma 2 and Lemma 3 to the left-hand side of (29), we get

$$
\begin{align*}
\nabla_{\alpha}^{-\alpha}\left(\nabla_{\alpha-1}^{\alpha} y(t)\right)= & \nabla_{\alpha}^{-\alpha}\left(\nabla_{\alpha} \nabla_{\alpha-1}^{-(1-\alpha)} y(t)\right) \\
= & \nabla_{\alpha}\left(\nabla_{\alpha}^{-\alpha} \nabla_{\alpha-1}^{-(1-\alpha)} y(t)\right)-\frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha) \\
= & y(t)-\frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha) \\
& t \in \mathbb{N}_{\alpha+1} \cap[1, T+1] \tag{30}
\end{align*}
$$

It follows from this and equation (29) that

$$
\begin{align*}
& y(t)=\frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha)+\lambda \nabla_{\alpha}^{-\alpha} y(t)+\lambda \nabla_{\alpha}^{-\alpha} \xi_{t}  \tag{31}\\
& t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]
\end{align*}
$$

which is the solution of UFBDE (28).
To derive the solution, we use the Picard approximation recurrence formula with a starting point $y_{0}(t)=$ $\left((t-\alpha+1)^{\frac{\alpha-1}{\alpha-1}} / \Gamma(\alpha)\right) y(\alpha)$ for each $t \in \mathbb{N}_{\alpha-1} \cap[1, T+1]$. The other components can be determined by using the following recurrence relation:

$$
\begin{equation*}
y_{j}(t)=\frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha)+\lambda \nabla_{\alpha}^{-\alpha} y_{j-1}(t)+\lambda \nabla_{\alpha}^{-\alpha} \xi_{t} \tag{32}
\end{equation*}
$$

for $t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]$ and $j \in \mathbb{N}_{1}$. Since $\xi_{1}, \xi_{2}, \ldots, \xi_{T+1}$ are i.i.d. uncertain variables, we write $\xi_{t}=\xi$ in distribution. By using Lemma 5, Remark 3, and the fact that the linear combination of finite independent uncertain variables is an uncertain variable with a positive linear combination coefficient (see Theorems 1.21-1.24 of [11]), we can deduce

$$
\begin{aligned}
y_{1}(t)= & \frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha)+\lambda \nabla_{\alpha}^{-\alpha} y_{0}(t)+\lambda \nabla_{\alpha}^{-\alpha} \xi \\
= & \frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha) \\
& +\lambda\left[\frac{(t-\alpha+1)^{\overline{2 \alpha-1}}}{\Gamma(2 \alpha)}-\frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)}\right] y(\alpha) \\
& +\lambda \frac{(t-\alpha)^{\bar{\alpha}}}{\Gamma(\alpha+1)} \xi, \\
y_{2}(t)= & \frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha)+\lambda \nabla_{\alpha}^{-\alpha} y_{1}(t)+\lambda \nabla_{\alpha}^{-\alpha} \xi \\
= & \frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha) \\
& +\lambda\left[\frac{(t-\alpha+1)^{\overline{2 \alpha-1}}}{\Gamma(2 \alpha)}-\frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)}\right] y(\alpha) \\
& +\lambda^{2}\left[\frac{(t-\alpha+1)^{\overline{3 \alpha-1}}}{\Gamma(3 \alpha)}-\frac{(t-\alpha+1)^{\frac{2 \alpha-1}{2}}}{\Gamma(2 \alpha)}\right] y(\alpha) \\
& +\left[\frac{\lambda(t-\alpha)^{\bar{\alpha}}}{\Gamma(\alpha+1)}+\frac{\lambda^{2}(t-\alpha)^{\overline{2 \alpha}}}{\Gamma(2 \alpha+1)}\right] \xi,
\end{aligned}
$$

$$
\begin{equation*}
\vdots \tag{33}
\end{equation*}
$$

and so on, continuing the process up to the $j$ th term to get

$$
\begin{align*}
y_{j}(t)= & \sum_{k=0}^{j} \lambda^{k}\left[\frac{(t-\alpha+1)^{\overline{k \alpha+\alpha-1}}}{\Gamma((k+1) \alpha)}-\frac{(t-\alpha+1)^{\overline{k \alpha-1}}}{\Gamma(k \alpha)}\right] y(\alpha) \\
& +\sum_{k=1}^{j} \lambda^{k} \frac{(t-\alpha)^{\overline{k \alpha}}}{\Gamma(k \alpha+1)} \xi, \tag{34}
\end{align*}
$$

for each $t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]$. Observe that the two series

$$
\begin{align*}
\sum_{k=0}^{\infty} \lambda^{k}\left[\frac{(t-\alpha+1)^{\overline{k \alpha+\alpha-1}}}{\Gamma((k+1) \alpha)}-\frac{(t-\alpha+1)^{\overline{k \alpha-1}}}{\Gamma(k \alpha)}\right] & =E_{\overline{\alpha, \alpha}}(\lambda, t-\alpha+1)-\lambda E_{\overline{\alpha, \alpha}}(\lambda, t-\alpha+1) \\
& =(1-\lambda) E_{\overline{\alpha, \alpha}}(\lambda, t-\alpha+1),  \tag{35}\\
\sum_{k=1}^{\infty} \lambda^{k} \frac{(t-\alpha)^{\overline{k \alpha}}}{\Gamma(k \alpha+1)} & =\lambda E_{\overline{\alpha, \alpha+1}}(\lambda, t-\alpha),
\end{align*}
$$

are absolutely convergent for $|\lambda|<1$ by the d'Alembert ratio comparison test, and the limitation $\bar{Y}(t):=\lim _{j \rightarrow \infty} y_{j}$ exists. Thus, we have

$$
\begin{array}{r}
\bar{Y}(t)=\lambda E_{\overline{\alpha, \alpha+1}}(\lambda, t-\alpha) \xi+(1-\lambda) E_{\overline{\alpha, \alpha}}(\lambda, t-\alpha+1) y(\alpha), \\
t \in \mathbb{N}_{\alpha+1} \cap[1, T+1] . \tag{36}
\end{array}
$$

On the contrary, taking the limit on both sides of (32) yields

$$
\begin{array}{r}
\bar{Y}(t)=\frac{(t-\alpha+1)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha)+\lambda \nabla_{\alpha}^{-\alpha} \bar{Y}(t)+\lambda \nabla_{\alpha}^{-\alpha} \xi_{t}  \tag{37}\\
t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]
\end{array}
$$

That is, $\bar{Y}(t)$ satisfies equation (31), and hence, $\bar{Y}(t)$ is a solution of equation (25) subject to the initial condition (26). Thus, our proof is completed.

The following theorem provides and confirms the existence and uniqueness of the solution of UFBDEs.

Theorem 2. Assume that $G(t, x)$ and $H(t, x)$ satisfy the Lipschitz condition

$$
\begin{equation*}
|G(t, x)-G(t, y)|+|H(t, x)-H(t, y)| \leq L|x-y| \tag{38}
\end{equation*}
$$

and there is a positive number $L$ that satisfies the following inequality:

$$
\begin{equation*}
L<\frac{\Gamma(\alpha+1) \Gamma(T+1-\alpha)}{\Gamma(T+1)(Q+1)}, \tag{39}
\end{equation*}
$$

where $Q=|a| \vee|b|$. Then, UFBDE (25) subject to the initial condition (26) has a unique solution $y(t)$ for $t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]$ almost surely.

Proof. Define

$$
\begin{align*}
l_{\alpha}^{k} & :=\left\{x ; x=\{x(t)\}_{\alpha+1}^{k}, k \in \mathbb{N}_{1}\right\},\|x\|,  \tag{40}\\
& :=\max _{t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]}|x(t)|,
\end{align*}
$$

where $\{x(t)\}_{\alpha}^{k}$ are finite real sequences which have $k$ terms. It is clear that $\left(l_{\alpha}^{k},\|\cdot\|\right)$ is a Banach space (see [31], Chapter 4). Now, for any $y_{t} \in l_{\alpha}^{k}$, we define the operator $P$ as follows:

$$
\begin{align*}
P y_{t}= & \frac{(t-\alpha)^{\overline{\alpha-1}}}{\Gamma(\alpha)} y(\alpha)+\frac{1}{\Gamma(\alpha)} \sum_{r=\alpha+1}^{t}(t-\rho(r))^{\overline{\alpha-1}}\left[G\left(r, y_{r}\right)\right. \\
& \left.+H\left(r, y_{r}\right) \xi_{r}\right] . \tag{41}
\end{align*}
$$

Since $\xi_{t}\left(t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]\right)$ is an uncertain variable at each time $t$ with the linear uncertainty distribution $\mathscr{L}(a, b)$, we have $\mathscr{M}\left\{\left(\xi_{t}<a\right) \cup\left(\xi_{t}>b\right)\right\}=0$. The inequality $\xi_{t}(\gamma) \leq Q$ (where $Q=|a| \vee|b|$ ) holds almost surely for any $\gamma \in \chi\left\{\left(\xi_{t}<a\right) \cup\left(\xi_{t}>b\right) ; t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]\right\}$, where $\chi$ represents the universal set on the uncertainty space. Then, by making use of the assumptions and Lemma 5, we have, for any $x_{t}, y_{t} \in l_{\alpha}^{k}$,

$$
\begin{align*}
\left\|P x_{t}(\gamma)-P y_{t}(\gamma)\right\|= & \max _{t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]}\left|P x_{t}(\gamma)-P y_{t}(\gamma)\right| \\
\leq & \frac{1}{\Gamma(\alpha)} \max _{t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]} \sum_{r=\alpha+1}^{t}(t-\rho(r))^{\overline{\alpha-1}}, \\
& \times\left(\mid G\left(r, x_{r}(\gamma)-G\left(r, y_{r}(\gamma)|+|\left[H\left(r, x_{r}(\gamma)-H\left(r, y_{r}(\gamma)\right] \xi_{r} \mid\right),\right.\right.\right.\right. \\
\leq & \frac{1}{\Gamma(\alpha)} \max _{t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]} \sum_{r=\alpha+1}^{t}(t-\rho(r))^{\overline{\alpha-1}}, \\
& \times\left(\mid G\left(r, x_{r}(\gamma)-G\left(r, y_{r}(\gamma)|+Q| H\left(r, x_{r}(\gamma)-H\left(r, y_{r}(\gamma) \mid\right),\right.\right.\right.\right. \\
\leq & L(1+Q) \frac{1}{\Gamma(\alpha)} \max _{t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]} \sum_{r=\alpha+1}^{t}(t-\rho(r))^{\frac{\alpha-1}{\alpha-1}}\left|x_{r}(\gamma)-y_{r}(\gamma)\right|,  \tag{42}\\
\leq & L(1+Q)\left\|x_{t}(\gamma)-y_{t}(\gamma)\right\|_{t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]}\left(\nabla_{\alpha}^{-\alpha}(t-\alpha)^{\overline{0}}\right), \\
= & L(1+Q)\left\|x_{t}(\gamma)-y_{t}(\gamma)\right\|_{t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]} \max ^{\max ^{2}}\left(\overline{1} \overline{\Gamma(\alpha+1)}(t-\alpha)^{\bar{\alpha}}\right), \\
\leq & \frac{L(1+Q)(T+1-\alpha)^{\bar{\alpha}}}{\Gamma(\alpha+1)}\left\|x_{t}(\gamma)-y_{t}(\gamma)\right\|, \\
= & \frac{L(1+Q) \Gamma(T+1)}{\Gamma(\alpha+1) \Gamma(T-\alpha+1)}\left\|x_{t}(\gamma)-y_{t}(\gamma)\right\| .
\end{align*}
$$

Now, we can observe that the mapping $P$ is a contraction in $l_{\alpha}^{k}$ almost surely with $0<L<(\Gamma(\alpha+1) \Gamma(T-\alpha+1) /(1+$ Q) $\Gamma(T+1)$ ) (see [31], Chapter 4). Then, by using the Banach contraction mapping theorem (see [31], Chapter 4), we get a unique fixed point $y_{t}(\gamma)$ of $P$ in $l_{\alpha}^{k}$ almost surely. Moreover, $y_{t}(\gamma)=\lim _{j \rightarrow \infty} y_{t}^{j}(\gamma)$, where $y_{t}^{j}(\gamma)=P\left(y_{t}^{j-1}(\gamma)\right)$, with $y_{t}^{0}(\gamma)=\left((t-\alpha)^{\overline{\alpha-1}} / \Gamma(\alpha)\right) y(\alpha)$.

For any given $t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]$, as $G$ and $H$ are Lipschitz continuous functions, the operator $P$ is measurable. Since $y_{t}^{1}(\gamma), y_{t}^{2}(\gamma), \ldots, y_{t}^{j}(\gamma), \ldots$ are uncertain variables and $y_{t}^{0}(\gamma)$ is a real-valued measurable function of uncertain variables, $y_{t}^{0}(\gamma)$ is an uncertain variable by [11], Theorem 1.10. Hence, $y_{t}=\lim _{j \rightarrow \infty} y_{t}^{j}$ is an uncertain variable (see [21], Theorem 3).

Consequently, UFBDE (25) with i.c. (26) has a unique solution $y_{t}$ for $t \in \mathbb{N}_{\alpha+1} \cap[1, T+1]$ almost surely.

## 4. Example Illustrations

This section deals with some examples to confirm the validity of Theorem 2.

Example 3. Consider the following UFBDE:

$$
\nabla_{(-1 / 2)}^{(1 / 2)} y(t)= \begin{cases}\frac{\ln (|y(t)|+1)}{2 t^{3}}+0.5 \xi_{t}, & t \in \mathbb{N}_{1} \cap[1,4]  \tag{43}\\ y\left(\frac{1}{2}\right)=1,\end{cases}
$$

where $\xi_{1}, \ldots, \xi_{4}$ are four i.i.d. uncertain variables with uncertainty distribution $\mathscr{L}(-1,2)$.

According to Lemma 6 with $\alpha=(1 / 2)$, the inverse uncertainty distribution of the solution for UFBDE (43) is the solution of the following sum equation:

$$
y(t)=\frac{(t+(1 / 2))^{-0.5}}{\Gamma(0.5)} y(\alpha)+\frac{1}{\Gamma(0.5)} \sum_{r=(3 / 2)}^{t}(t-\rho(r))^{\overline{-0.5}}
$$

$$
\begin{equation*}
\cdot\left(\frac{\ln (|y(r)|+1)}{4 r^{3}}+0.25 \xi_{r}\right), \quad t \in \mathbb{N}_{0.5} \cap[1,4] \tag{44}
\end{equation*}
$$

Then, for $t \in \mathbb{N}_{0.5} \cap[1,4]$, we have

$$
\begin{align*}
|G(t, x)-G(t, y)|+|H(t, x)-H(t, y)| & =\left|\frac{\ln (|x|+1)}{2 t^{3}}-\frac{\ln (|y|+1)}{2 t^{3}}\right| \\
& =\frac{1}{2 t^{3}}|\ln (|x|+1)-\ln (|y|+1)| \\
& \leq \frac{1}{2(3 / 2)^{3}}| | x|-|y|| \leq \frac{4|x-y|}{27}  \tag{45}\\
\frac{\Gamma(0.5+1) \Gamma(4+1-0.5)}{3 \Gamma(4+1)} & \approx 0.1636>\frac{4}{27}=0.1481
\end{align*}
$$

Thus, UFBDE (43) has a unique solution almost surely by Theorem 2.

Example 4. We consider the following UFBDE:

$$
\begin{equation*}
\nabla_{-0.25}^{0.25} y(t)=\frac{y^{2} t}{40}+\xi_{t}, \quad t \in \mathbb{N}_{1} \cap[1,4] \tag{46}
\end{equation*}
$$

where $\xi_{1}, \xi_{2}, \xi_{3}, \xi_{4}$ are four i.i.d. linear uncertain variables with linear uncertainty distribution $\mathscr{L}(-3,3)$.

According to Lemma 6 with $\alpha=(1 / 4)$, the inverse uncertainty distribution of the solution for UFBDE (46) is the solution of the following sum equation:

$$
\begin{align*}
y(t)= & \frac{(t+(1 / 2))^{\overline{-(3 / 4)}}}{\Gamma(0.25)} y(\alpha) \\
& +\frac{1}{\Gamma(0.25)} \sum_{r=1.25}^{t}(t-\rho(r))^{\overline{-(3 / 4)}}\left(\frac{y^{2} r}{40}+\xi_{r}\right) \tag{47}
\end{align*}
$$

Observe that $|G(t, x)-G(t, y)|+|H(t, x)-H(t, y)|$ is Lipschitz-continuous in $[-20,20]$ with Lipschitz constant 0.1 as follows:

$$
\begin{align*}
& |G(t, x)-G(t, y)|+|H(t, x)-H(t, y)| \leq \frac{1}{40}|x+y||x-y| \\
& \quad=0.1|x-y| \tag{48}
\end{align*}
$$

Also, we have

$$
\begin{equation*}
\frac{\Gamma(0.25+1) \Gamma(3+1-0.25)}{4 \Gamma(3+1)} \approx 0.167>0.1 \tag{49}
\end{equation*}
$$

Consequently, UFBDE (50) has a unique solution almost surely by Theorem 2.

Example 5. Consider the following UFBDE:

$$
\begin{equation*}
\nabla_{-0.5}^{0.5} y(t)=\frac{\sin (t y)}{10+t^{2}}+0.1 \xi_{t}, \quad t \in \mathbb{N}_{1} \cap[1,4] \tag{50}
\end{equation*}
$$

where $\xi_{1}, \xi_{2}, \xi_{3}, \xi_{4}$ are 4 i.i.d. linear uncertain variables with linear uncertainty distribution $\mathscr{L}(-1,1)$.

According to Lemma 6 with $\alpha=(1 / 2)$, the inverse uncertainty distribution of the solution for UFBDE (50) is the solution of the following sum equation:

$$
\begin{align*}
y(t)= & \frac{(t+(1 / 2))^{\overline{-0.5}}}{\sqrt{\pi}} y(\alpha) \\
& +\frac{1}{\sqrt{\pi}} \sum_{r=1.5}^{t}(t-\rho(r))^{-0.5}\left(\frac{\sin (r y)}{10+r^{2}}+0.1 \xi_{r}\right) \tag{51}
\end{align*}
$$

Then, we can directly verify that

$$
\begin{align*}
&|G(t, x)-G(t, y)|+|H(t, x)-H(t, y)| \leq \frac{1}{10}|x-y| \\
& \frac{\Gamma(0.5+1) \Gamma(3+1-0.5)}{2 \Gamma(3+1)} \approx 0.2454>\frac{1}{10}=0.1 \tag{52}
\end{align*}
$$

Consequently, UFBDE (50) has a unique solution almost surely by Theorem 2.

## 5. Conclusion

We have presented analytical solutions to a special type of linear UFBDEs. Moreover, a Lipschitz condition with its constant is given to provide a unique solution almost surely to an UFBDE. It can be seen that our obtained results pave the way for the future works, that is, to investigate the stability analysis and applications of UFBDEs.

## Data Availability

No data were used to support this study.

## Conflicts of Interest

The authors declare that they have no conflicts of interest.

## Authors' Contributions

All authors contributed equally and significantly to writing this article. All authors read and approved the final manuscript.

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[^0]:    Abbreviations
    $\alpha_{1}$ : Reciprocal of the magnetic Prandtl number (-)
    $\mu$ : Dynamic viscosity $\left(\mathrm{Nsm}^{-2}\right)$
    $k^{*}$ : Mean absorption coefficient (-)
    $b: \quad$ Body forces $\left(\mathrm{Nm}^{-3}\right)$
    C: $\quad$ Nanoparticles concentration $\left(\mathrm{kg} \mathrm{m}^{-3}\right)$
    $F$ : $\quad$ Dimensionless velocity function (-)
    T: Temperature (-)
    Nb : Brownian motion parameter (-)
    $N u_{x}$ : Local Nusselt number (-)
    $C_{\infty}$ : Ambient fluid concentration $\left(\mathrm{kg} \mathrm{m}^{-3}\right)$
    $T_{w}$ : Hot fluid temperature (K)

