

Research Article

Modeling and Analysis of the Fuzzy-Fractional Chaotic Financial System Using the Extended He–Mohand Algorithm in a Fuzzy-Caputo Sense

Mubashir Qayyum ¹, Efaza Ahmad,¹ Aneeza Tahir,¹ and Saraswati Acharya ²

¹Department of Sciences and Humanities, National University of Computer and Emerging Sciences, Lahore, Pakistan

²Kathmandu University School of Science, Dhulikhel, Nepal

Correspondence should be addressed to Saraswati Acharya; saraswati.acharya@ku.edu.np

Received 22 May 2023; Revised 20 October 2023; Accepted 24 October 2023; Published 4 November 2023

Academic Editor: Vasudevan Rajamohan

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This paper contains modeling of a fuzzy-fractional financial chaotic model based on triangular fuzzy numbers (TFNs) to predict the idea that long-term dependency and uncertainty both have an impact on the financial market. For solution purposes, the He–Mohand algorithm is proposed where homotopy perturbation is hybrid with Mohand transform in a fuzzy-Caputo sense. In analysis, solutions and corresponding errors at upper and lower bounds are estimated. The obtained numerical results are displayed in tables to show the reliability and efficiency of the proposed methodology. Upper bound errors range from 10^{-6} to 10^{-12} and lower bound errors from 10^{-6} to 10^{-11} . For graphical analysis, system profiles are illustrated as two-dimensional and three-dimensional plots at diverse values of fractional parameters and time to comprehend the physical behavior of the proposed fuzzy-fractional model. These plots demonstrate that the interest rate, price index, and investment demand decrease with the increase of the value of the r -cut at the lower bound. At the upper bound, this behavior is totally opposite. The chaotic behavior of the system at smaller values of saving rate, elasticity of demands, and per-investment cost is greater in contrast to their larger value. Analysis reveals that the proposed methodology (He–Mohand algorithm) provides a new way of understanding the complicated structure of financial systems and provides new insights into the dynamics of financial markets. This algorithm has potential applications in risk management, portfolio optimization, and trading strategies.

1. Introduction

Financial modeling [1] is the process of building mathematical models to represent the performance of portfolios, financial assets, or businesses. Its aim is to provide a quantitative representation of the underlying financial conditions and to make predictions about future performance. Numerous uses for these models are possible, such as scenario analysis, valuation, planning, risk assessment, and forecasting. Several quantitative techniques are used in financial modeling. Monte Carlo simulation [2], statistical analysis [3], sensitivity analysis [4], discounted cash flow [5], time-series analysis [6], and regression analysis [7] are some of them. It also requires in-depth knowledge of accounting principles, economic theory, and financial markets. There

are various categories of financial models such as option pricing models [8], discounted cash flow models [9], chaotic financial models [10], portfolio optimization models [11], and credit risk models [12]. Each model has strengths and shortcomings and the best model relies on the application at hand as well as the facts that are accessible.

A fuzzy differential equation (FDE) is a type of differential equation in which some of the initial conditions or parameters are represented by fuzzy sets [13]. In various situations, the values of these initial conditions or parameters are not known precisely, but rather within a particular range. With the use of fuzzy differential equations, we can simulate the uncertainty that arises from such circumstances. These equations have many applications in various fields, such as fluid dynamics [14], kinetics [15], finance [16],

economics [17], and biology [18]. Fuzzy Volterra integrodifferential equations [19], fuzzy Fisher model [20], fuzzy population growth model [21], first-order linear fuzzy differential equations [22], and fuzzy singular integrodifferential models [23] are some examples of fuzzy differential equations.

Fuzzy differential equations are also modeled in fractional derivative form. An extension of a derivative to a noninteger order is known as a fractional derivative. It is described using fractional calculus [24], which is a field of mathematics that manages noninteger order integrals and derivatives. Many physical models including the Lotka–Volterra population equation [25], coupled Schrodinger system [26], Oldroyd 6-constant fluid [27], Grey system models [28], tumor models [29], and Wu–Zhang system [30] exploit a fractional derivative approach. We can model and examine real-world problems that include uncertainty and fractional order derivatives by utilizing fuzzy-fractional differential equations (FFDEs). Oceanography [31], biological population model [32], COVID-19 model [33], heat equation [34], and Fisher’s equation [35] are some areas that employ FFDEs. The Caputo fractional derivative [36] is one of the more popular definitions of the fractional derivative. It is commonly used in physics and engineering. It is described as a modification of the Riemann–Liouville derivative [37] that prevents the singular behavior at the lower limit of the integral. The Caputo fractional derivative is especially beneficial in modeling systems with memory or hereditary properties, such as Brinkman-type fluid [38], Korteweg–de Vries system [39], mosaic disease model [40], relaxation-oscillation equations [41], plant disease model [42], Casson nanofluid model [43], and chaotic system [44].

Due to the nonlocality of fractional derivatives as well as the complexity of fuzzy sets, solving FFDEs can be challenging. In literature, several techniques have been utilized to solve them. Rexma Sherine et al. [45] used the fuzzy-fractional Laplace transform method to estimate the spread of the generalized monkeypox virus model. The fractional differential transform method and Hilbert space method were combined by Najafi and Allahviranloo [46] to solve fuzzy impulsive fractional differential equations. The Legendre spectral method to find the solution of the fuzzy-fractional coronavirus model is adopted by Alderremy et al. [47]. To examine fuzzy-fractional differential equations, Alijani et al. [48] applied Spline collocation methods. Alaroud et al. [49] employed an analytical numerical technique on fuzzy-fractional Volterra integrodifferential equations. The Chebyshev spectral method was utilized by Kumar et al. [50] to analyze the fuzzy-fractional Fredholm–Volterra integrodifferential equation. A powerful tool to solve nonlinear fractional differential equations in fuzzy form is the He–Mohand method [51]. It is an efficient technique that provides a practical method for solving differential models by combining the homotopy perturbation technique (HPM) and the Mohand transform. Thus, in order to solve the fuzzy-fractional chaotic financial system, we have created an extended HPM hybrid using the Mohand transform.

The format of this article is as follows. In Section 2, preliminaries are given in which Mohand transform, Caputo fractional derivative, and its Mohand transform, fuzzy sets, and triangular fuzzy sets are defined. Section 3 is focused on the modeling of the fuzzy-fractional financial chaotic model. The solution framework based on the He–Mohand algorithm is presented in Section 4, whereas, the theoretical analysis of the proposed scheme is presented in Section 5. The focus of Section 6 is on the application and solution of the given system. Results of the study are discussed in Section 7 and, Section 8 provides some important conclusions of the study.

2. Preliminaries

Definition 1. [52]

The Mohand transform \mathcal{M} of the function $\tilde{\mathcal{G}}(\tau)$ for $\tau \geq 0$ is given by the following equation:

$$\mathcal{M}\{\tilde{\mathcal{G}}(\tau)\} = \mathcal{K}(\tau) = p^2 \int_0^{\infty} \tilde{\mathcal{G}}(\tau) e^{-p\tau} d\tau, \quad p \in [k_1, k_2], \quad (1)$$

where $k_1, k_2 > 0$ can be finite or infinite. The parameter p factors the variable τ in the argument of function $\tilde{\mathcal{G}}$.

Definition 2. The inverse Mohand transform \mathcal{M}^{-1} of the function $\mathcal{K}(\tau)$ is as follows:

$$\mathcal{M}^{-1}\{\mathcal{K}(\tau)\} = \frac{1}{2\pi i} \int_{\gamma-i\infty}^{\gamma+i\infty} \frac{1}{p^2} \mathcal{K}(\tau) e^{p\tau} dp, \quad p \in [k_1, k_2]. \quad (2)$$

Definition 3. [53]

For a function $\tilde{\mathcal{G}}(\tau)$, the Caputo fractional derivative ${}^C\mathbb{D}_\tau^\gamma$ is defined as follows:

$${}^C\mathbb{D}_\tau^\gamma\{\tilde{\mathcal{G}}(\tau)\} = \frac{1}{\Gamma(\varepsilon-\gamma)} \int_0^\tau (\tau-p)^{\varepsilon-\gamma-1} \tilde{\mathcal{G}}^{(\varepsilon)}(p) dp, \quad \varepsilon-1 < \gamma \leq \varepsilon. \quad (3)$$

Definition 4. [54]

The Mohand transform \mathcal{M} in the presence of Caputo fractional derivative (3) can be written as follows:

$$\mathcal{M}\{{}^C\mathbb{D}_\tau^\gamma\tilde{\mathcal{G}}(\tau)\} = p^\gamma \mathcal{M}\{\tilde{\mathcal{G}}(\tau)\} - \sum_{a=0}^{\varepsilon-1} p^{\gamma-a+1} \tilde{\mathcal{G}}^{(a)}(0), \quad \varepsilon-1 < \gamma \leq \varepsilon. \quad (4)$$

Definition 5. [55]

Let \mathbb{R} be a real set. Then, a fuzzy set \tilde{w} in \mathbb{R} can be characterized by a membership function $\mu_{\tilde{w}}$, where, $\mu_{\tilde{w}}: \mathbb{R} \rightarrow [0, 1]$. An r -level set of \tilde{w} is $[\tilde{w}]^r = \{w \in \mathbb{R}: \mu_{\tilde{w}}(w) \geq r\}$ for $r \in [0, 1]$.

The following are some conditions for a fuzzy set \tilde{w} to be a fuzzy number:

- (i) \tilde{w} is normal, that is, for $w_0 \in \mathbb{R}$, we have $\mu_{\tilde{w}}(w_0) = 1$

- (ii) \tilde{w} is convex, that is, $\mu_{\tilde{w}}(\nu w_1 + (1 - \nu) w_2) \geq \min\{\mu_{\tilde{w}}(w_1), \mu_{\tilde{w}}(w_2)\}$ for all $w_1, w_2 \in \mathbb{R}$ and $\nu \in [0, 1]$
- (iii) \tilde{w} is semicontinuous
- (iv) The set $\overline{\{w \in \mathbb{R}: \mu_{\tilde{w}}(w) > 0\}}$ is compact

Definition 6. [56]

A fuzzy number \tilde{w} is classified as a triangular fuzzy number (TFN) if it is defined by three numbers (k_1, k_2, k_3) with $k_1 < k_2 < k_3$ such that it forms a triangle. Its membership function is as follows:

$$\mu(w, k_1, k_2, k_3) = \begin{cases} 0, & w \leq k_1, \\ \frac{w - k_1}{k_2 - k_1}, & k_1 \leq w \leq k_2, \\ \frac{k_3 - w}{k_3 - k_2}, & k_2 \leq w \leq k_3, \\ 0, & w \geq k_3. \end{cases} \quad (5)$$

By utilizing the r -cut notion, the interval form of TFN can be expressed as follows:

$$\tilde{w} = [\underline{w}, \overline{w}] = [k_1 + (k_2 - k_1)r, k_3 - (k_3 - k_2)r], \quad (6)$$

where \underline{w} and \overline{w} represent the lower and upper bounds, respectively, for $r \in [0, 1]$.

Definition 7. [56]

A fuzzy number $\tilde{w}(r)$ can also be represented as $\tilde{w} = [\underline{w}, \overline{w}]$ which satisfies the following conditions:

- (i) $\underline{w}(r)$ is a bounded monotonic increasing left continuous function
- (ii) $\overline{w}(r)$ is a bounded monotonic decreasing left continuous function
- (iii) $\underline{w}(r) \leq \overline{w}(r)$ for $r \in [0, 1]$

3. Fuzzy-Fractional Modeling of the Financial Chaotic System

This section is focused on the modeling of the fuzzy-fractional chaotic financial system that is mostly used in the financial and economic sectors. We consider the chaotic financial system given as follows:

$$\begin{aligned} \frac{\partial \mathcal{E}1}{\partial \tau} - \mathcal{E}3(\tau) - \mathcal{E}2(\tau)\mathcal{E}1(\tau) + \mathbb{A}\mathcal{E}1(\tau) &= 0, \\ \frac{\partial \mathcal{E}2}{\partial \tau} - 1 + \mathbb{B}\mathcal{E}2(\tau) + \mathcal{E}1^2(\tau) &= 0, \\ \frac{\partial \mathcal{E}3}{\partial \tau} + \mathcal{E}1(\tau) + \mathbb{C}\mathcal{E}3(\tau) &= 0, \quad \tau > 0, \end{aligned} \quad (7)$$

with conditions

$$\begin{aligned} \mathcal{E}1(0) &= \mathbb{Y}1, \\ \mathcal{E}2(0) &= \mathbb{Y}2, \\ \mathcal{E}3(0) &= \mathbb{Y}3, \end{aligned} \quad (8)$$

where $\mathcal{E}1$, $\mathcal{E}2$, and $\mathcal{E}3$ represent the interest rate, investment demand, and price index, respectively. Moreover, \mathbb{A} denotes the saving amount, \mathbb{B} is the per-investment cost, and the parameter \mathbb{C} presents the elasticity of demands. For a precise understanding of the interest rate, investment demand, and price index, the given system is modeled in fractional form by utilizing Definition 3, which is presented in the following equation:

$$\begin{aligned} \frac{\partial^\gamma \mathcal{E}1}{\partial \tau^\gamma} - \mathcal{E}3(\tau) - \mathcal{E}2(\tau)\mathcal{E}1(\tau) + \mathbb{A}\mathcal{E}1(\tau) &= 0, \\ \frac{\partial^\gamma \mathcal{E}2}{\partial \tau^\gamma} - 1 + \mathbb{B}\mathcal{E}2(\tau) + \mathcal{E}1^2(\tau) &= 0, \\ \frac{\partial^\gamma \mathcal{E}3}{\partial \tau^\gamma} + \mathcal{E}1(\tau) + \mathbb{C}\mathcal{E}3(\tau) &= 0, \quad 0 < \gamma \leq 1, \tau > 0, \end{aligned} \quad (9)$$

where γ represents the fractional parameter in a Caputo sense. To introduce uncertainty in the system, we have incorporated triangular fuzzy numbers in given initial conditions $\mathbb{Y}i$, $i = 1, 2, 3$ by utilizing Definition 5–7. In parametric form, they can be written as $\tilde{\mathbb{Y}}1 = [0 + (0.3 - 0)r, 1 - (1 - 0.3)r]$, $\tilde{\mathbb{Y}}2 = [-1 + (-0.3 + 1)r, 1 - (1 + 0.3)r]$, and $\tilde{\mathbb{Y}}3 = [0 + (0.2 - 0)r, 1 - (1 - 0.2)r]$. Thus, the fuzzy-fractional chaotic financial system is as follows:

$$\begin{aligned} \frac{\partial^\gamma \tilde{\mathcal{E}}1}{\partial \tau^\gamma} - \tilde{\mathcal{E}}3(\tau) - \tilde{\mathcal{E}}2(\tau)\tilde{\mathcal{E}}1(\tau) + \mathbb{A}\tilde{\mathcal{E}}1(\tau) &= 0, \\ \frac{\partial^\gamma \tilde{\mathcal{E}}2}{\partial \tau^\gamma} - 1 + \mathbb{B}\tilde{\mathcal{E}}2(\tau) + \tilde{\mathcal{E}}1^2(\tau) &= 0, \\ \frac{\partial^\gamma \tilde{\mathcal{E}}3}{\partial \tau^\gamma} + \tilde{\mathcal{E}}1(\tau) + \mathbb{C}\tilde{\mathcal{E}}3(\tau) &= 0, \quad 0 < \gamma \leq 1, \tau > 0, \end{aligned} \quad (10)$$

with fuzzy initial conditions

$$\begin{aligned} \tilde{\mathcal{E}}1(0) &= \tilde{\mathbb{Y}}1, \\ \tilde{\mathcal{E}}2(0) &= \tilde{\mathbb{Y}}2, \\ \tilde{\mathcal{E}}3(0) &= \tilde{\mathbb{Y}}3. \end{aligned} \quad (11)$$

The system in (10) along with its fuzzy conditions (11) will be utilized by financial analysts and investors to predict the characteristics of the financial market. It can provide an excellent tool to make informed investment decisions in the areas of financial risk management.

4. Solution Framework Based on the Extended He–Mohand Algorithm for Fuzzy-Fractional Systems

Let us consider a general nonlinear fuzzy-fractional system,

$$\mathbb{D}_\tau^\gamma \tilde{\mathcal{G}}_j(\tau) + \mathcal{L}[\tilde{\mathcal{G}}_j(\tau)] + \mathcal{N}[\tilde{\mathcal{G}}_j(\tau)] = 0, \quad j = 1, \dots, m, \tau > 0, \quad \varepsilon - 1 < \gamma \leq \varepsilon, \quad (12)$$

with fuzzified initial conditions

$$\tilde{\mathcal{G}}_j(0) = \mathcal{A}j, \quad j = 1, \dots, m, \quad (13)$$

where γ is the Caputo fractional parameter, \mathbb{D}_τ^γ is the fractional derivative of $\tilde{\mathcal{G}}_j$, and j represents the total equations of the system. The parameters \mathcal{L} and \mathcal{N} are linear and nonlinear operators, respectively.

Equation (12) can be expressed by using r -cut as follows:

$$[\mathbb{D}_\tau^\gamma \underline{\mathcal{L}}_j(\tau; r), \mathbb{D}_\tau^\gamma \overline{\mathcal{L}}_j(\tau; r)] + [\mathcal{L}[\underline{\mathcal{L}}_j(\tau; r)], \mathcal{L}[\overline{\mathcal{L}}_j(\tau; r)]] + [\mathcal{N}[\underline{\mathcal{L}}_j(\tau; r)], \mathcal{N}[\overline{\mathcal{L}}_j(\tau; r)]] = 0, \quad (14)$$

where $\tilde{\mathcal{G}}_j(\tau; r) = [\underline{\mathcal{L}}_j(\tau; r), \overline{\mathcal{L}}_j(\tau; r)]$. $\underline{\mathcal{L}}_j(\tau; r)$ represents the lower bound solution and $\overline{\mathcal{L}}_j(\tau; r)$ represents the upper bound solution.

First, we take Mohand transform on both sides of (12) as follows:

$$\mathcal{M}_\tau\{\mathbb{D}_\tau^\gamma \tilde{\mathcal{G}}_j(\tau; r)\} + \mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_j(\tau; r)] + \mathcal{N}[\tilde{\mathcal{G}}_j(\tau; r)]\} = 0. \quad (15)$$

Application of Definition 4 gives

$$\mathcal{M}_\tau\{\tilde{\mathcal{G}}_j(\tau; r)\} - \left(\frac{1}{p^\gamma}\right) \sum_{a=0}^{\varepsilon-1} p^{\gamma-a+1} \tilde{\mathcal{G}}_j^{(a)}(0) + \left(\frac{1}{p^\gamma}\right) \mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_j(\tau; r)] + \mathcal{N}[\tilde{\mathcal{G}}_j(\tau; r)]\} = 0. \quad (16)$$

The general homotopy of the system is as follows:

$$\text{Hom: } (1 - q)\left(\mathcal{M}_\tau\{\tilde{\mathcal{G}}_j(\tau; r)\} - \tilde{\mathcal{G}}_{j_0}\right) + q\left(\mathcal{M}_\tau\{\tilde{\mathcal{G}}_j(\tau; r)\} - \left(\frac{1}{p^\gamma}\right) \sum_{a=0}^{\varepsilon-1} p^{\gamma-a+1} \tilde{\mathcal{G}}_j^{(a)}(0) + \left(\frac{1}{p^\gamma}\right) \mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_j(\tau; r)] + \mathcal{N}[\tilde{\mathcal{G}}_j(\tau; r)]\}\right) = 0, \quad (17)$$

with $\tilde{\mathcal{G}}_{j_0}$ as an initial guess and $q \in [0, 1]$. Expansion of $\tilde{\mathcal{G}}_j(\tau; r)$ in power series form w.r.t. q gives

Substituting (18) in (17) and comparing similar coefficients of power of q leads to q^1 where

$$\tilde{\mathcal{G}}_j(\tau; r) = \sum_{n=0}^{\infty} q^n \tilde{\mathcal{G}}_{j_n}(\tau; r). \quad (18)$$

$$\mathcal{M}_\tau\{\tilde{\mathcal{G}}_{j_1}(\tau; r)\} + \tilde{\mathcal{G}}_{j_0} - \left(\frac{1}{p^\gamma}\right) \sum_{a=0}^{\varepsilon-1} p^{\gamma-a+1} \tilde{\mathcal{G}}_{j_0}^{(a)}(0) + \left(\frac{1}{p^\gamma}\right) \mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_{j_0}(\tau; r)] + \mathcal{N}[\tilde{\mathcal{G}}_{j_0}(\tau; r)]\} = 0. \quad (19)$$

At q^2 , we obtain

$$\mathcal{M}_\tau\{\tilde{\mathcal{G}}_{j_2}(\tau; r)\} + \left(\frac{1}{p^\gamma}\right) \mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_{j_1}(\tau; r)] + \mathcal{N}[\tilde{\mathcal{G}}_{j_1}(\tau; r)]\} = 0. \quad (20)$$

In general at q^n , we have

$$\mathcal{M}_\tau\{\tilde{\mathcal{G}}_{j_n}(\tau; \mathfrak{r})\} + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_{j_{n-1}}(\tau; \mathfrak{r})] + \mathcal{N}[\tilde{\mathcal{G}}_{j_{n-1}}(\tau; \mathfrak{r})]\} = 0. \tag{21}$$

Applying the inverse Mohand transform gives the following at q^1 :

$$\tilde{\mathcal{G}}_{j_1}(\tau; \mathfrak{r}) + \mathcal{M}_\tau^{-1}\left\{\tilde{\mathcal{G}}_{j_0} - \left(\frac{1}{p^\gamma}\right)\sum_{a=0}^{\xi-1} p^{\gamma-a+1}\tilde{\mathcal{G}}_j^{(a)}(0) + \left(\frac{1}{s^\gamma}\right)\mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_{j_0}(\tau; \mathfrak{r})] + \mathcal{N}[\tilde{\mathcal{G}}_{j_0}(\tau; \mathfrak{r})]\}\right\} = 0. \tag{22}$$

At q^2 , we have

$$\tilde{\mathcal{G}}_{j_2}(\tau; \mathfrak{r}) + \mathcal{M}_\tau^{-1}\left\{\left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_{j_1}(\tau; \mathfrak{r})] + \mathcal{N}[\tilde{\mathcal{G}}_{j_1}(\tau; \mathfrak{r})]\}\right\} = 0. \tag{23}$$

At q^n , we obtain

$$\tilde{\mathcal{G}}_{j_n}(\tau; \mathfrak{r}) + \mathcal{M}_\tau^{-1}\left\{\left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{\mathcal{L}[\tilde{\mathcal{G}}_{j_{n-1}}(\tau; \mathfrak{r})] + \mathcal{N}[\tilde{\mathcal{G}}_{j_{n-1}}(\tau; \mathfrak{r})]\}\right\} = 0. \tag{24}$$

The approximate solution of (12) is obtained by

$$\tilde{\mathcal{G}}_j = \sum_{n=0}^{\infty} \tilde{\mathcal{G}}_{j_n}(\tau; \mathfrak{r}). \tag{25}$$

The residual function can be calculated by substituting (25) in the given system (12) as

$$\mathcal{R}_{\tilde{\mathcal{G}}_j} = \mathbb{D}_\tau^\gamma \tilde{\mathcal{G}}_j + \mathcal{L}[\tilde{\mathcal{G}}_j] + \mathcal{N}[\tilde{\mathcal{G}}_j]. \tag{26}$$

5. Theoretical Analysis of the Extended He–Mohand Algorithm for Fuzzy-Fractional Systems

Theorem 8. Convergence

Given that a Banach space has $\tilde{\mathcal{G}}_{j_n}(\tau)$ and $\tilde{\mathcal{G}}_j(\tau)$ defined in it for $j = 2, \dots, m$. Then, the obtained approximate solution (25) of a fuzzy-fractional differential system for $\mathbb{S} \in (0,1)$ converges to its exact solution (12).

Proof. Let $\{C_{j_n}\}$ be the sequence of partial sums of (25). In order to show that C_{j_n} is a Cauchy sequence in Banach space, let us consider

$$\begin{aligned} \|C_{j_{n+1}} - C_{j_n}\| &= \|\tilde{\mathcal{G}}_{j_{n+1}}\| \\ &\leq \mathbb{S}\|\tilde{\mathcal{G}}_{j_n}\| \\ &\leq \mathbb{S}^2\|\tilde{\mathcal{G}}_{j_{n-1}}\| \\ &\vdots \\ &\leq \mathbb{S}^{n+1}\|\tilde{\mathcal{G}}_{j_0}\|. \end{aligned} \tag{27}$$

By considering C_{j_n} and C_{j_m} as partial sums, for $n \geq m$ and $n, m \in \mathbb{N}$, triangle inequality property provides

$$\begin{aligned} \|C_{j_n} - C_{j_m}\| &= \|(C_{j_n} - C_{j_{n-1}}) + (C_{j_{n-1}} - C_{j_{n-2}}) \\ &\quad + \dots + (C_{j_{m+1}} - C_{j_m})\| \\ &\leq \|C_{j_n} - C_{j_{n-1}}\| + \|C_{j_{n-1}} - C_{j_{n-2}}\| \\ &\quad + \dots + \|C_{j_{m+1}} - C_{j_m}\|. \end{aligned} \tag{28}$$

Utilizing (27) gives

$$\begin{aligned} \|Cj_n - Cj_m\| &\leq \mathbb{S}^n \|\tilde{\mathcal{G}}j_0\| + \mathbb{S}^{n-1} \|\tilde{\mathcal{G}}j_0\| + \dots + \mathbb{S}^{m+1} \|\tilde{\mathcal{G}}j_0\| \\ &\leq (\mathbb{S}^n + \mathbb{S}^{n-1} + \dots + \mathbb{S}^{m+1}) \|\tilde{\mathcal{G}}j_0\| \\ &\leq \mathbb{S}^{m+1} (\mathbb{S}^{n-m-1} + \mathbb{S}^{n-m-2} + \dots + \mathbb{S} + 1) \|\tilde{\mathcal{G}}j_0\| \\ &\leq \mathbb{S}^{m+1} \left(\frac{1 - \mathbb{S}^{n-m}}{1 - \mathbb{S}} \right) \|\tilde{\mathcal{G}}j_0\|. \end{aligned} \tag{29}$$

Since $0 < \mathbb{S} < 1$, therefore, $1 - \mathbb{S}^{n-m} < 1$. Thus, we have

$$\|Cj_n - Cj_m\| \leq \frac{\mathbb{S}^{m+1}}{1 - \mathbb{S}} \max |\tilde{\mathcal{G}}j_0|. \tag{30}$$

The boundedness of $\tilde{\mathcal{G}}j_0$ implies that

$$\lim_{n,m \rightarrow \infty} \|Cj_n - Cj_m\| = 0. \tag{31}$$

Hence, we proved that Cj_n is a Cauchy sequence in a Banach space. This leads to the convergence of the given scheme. \square

Theorem 9. Error Estimation

The solution of the fuzzy-fractional system (12) has maximum absolute truncation error given as follows:

$$\left| \tilde{\mathcal{G}}j - \sum_{h=0}^m \tilde{\mathcal{G}}j_h \right| \leq \frac{\mathbb{S}^{m+1}}{1 - \mathbb{S}} \|\tilde{\mathcal{G}}j_0\|. \tag{32}$$

Proof. From equation (29), we obtain

$$\|\tilde{\mathcal{G}}j - Cj_m\| \leq \mathbb{S}^{m+1} \left(\frac{1 - \mathbb{S}^{n-m}}{1 - \mathbb{S}} \right) \|\tilde{\mathcal{G}}j_0\|, \tag{33}$$

where $0 < \mathbb{S} < 1 \Rightarrow 1 - \mathbb{S}^{n-m} < 1$. Thus, we have

$$\left| \tilde{\mathcal{G}}j - \sum_{b=0}^m \tilde{\mathcal{G}}j_b \right| \leq \frac{\mathbb{S}^{m+1}}{1 - \mathbb{S}} \|\tilde{\mathcal{G}}j_0\|. \tag{34}$$

Hence proved. \square

6. Application of the Proposed Methodology to the Fuzzy-Fractional Chaotic Financial System

Let us consider the chaotic financial system modeled in fuzzy-fractional form (see (10)) in Section 3,

$$\begin{aligned} \frac{\partial^\gamma \tilde{\mathcal{G}}1}{\partial \tau^\gamma} - \tilde{\mathcal{G}}3(\tau) - \tilde{\mathcal{G}}2(\tau)\tilde{\mathcal{G}}1(\tau) + \mathbb{A}\tilde{\mathcal{G}}1(\tau) &= 0, \\ \frac{\partial^\gamma \tilde{\mathcal{G}}2}{\partial \tau^\gamma} - 1 + \mathbb{B}\tilde{\mathcal{G}}2(\tau) + \tilde{\mathcal{G}}1^2(\tau) &= 0, \\ \frac{\partial^\gamma \tilde{\mathcal{G}}3}{\partial \tau^\gamma} + \tilde{\mathcal{G}}1(\tau) + \mathbb{C}\tilde{\mathcal{G}}3(\tau) &= 0, \quad 0 < \gamma \leq 1, \tau > 0, \end{aligned} \tag{35}$$

with fuzzified conditions (see (11)).

$$\begin{aligned} \tilde{\mathcal{G}}1(0) &= \tilde{\mathbb{Y}}1, \\ \tilde{\mathcal{G}}2(0) &= \tilde{\mathbb{Y}}2, \\ \tilde{\mathcal{G}}3(0) &= \tilde{\mathbb{Y}}3, \end{aligned} \tag{36}$$

where $\tilde{\mathbb{Y}}1 = [0, 0.3, 1]$, $\tilde{\mathbb{Y}}2 = [-1, -0.3, 1]$, and $\tilde{\mathbb{Y}}3 = [0, 0.2, 1]$ are triangular fuzzy numbers.

Solution 10. Initiating Mohand transform and then using the differential property of Mohand transform (4) gives

$$\begin{aligned} p^\gamma \mathcal{M}_\tau \{ \tilde{\mathcal{G}}1(\tau; r) \} - p^{\gamma+1} \tilde{\mathbb{Y}}1 + \mathcal{M}_\tau \{ -\tilde{\mathcal{G}}3(\tau; r) - \tilde{\mathcal{G}}2(\tau; r)\tilde{\mathcal{G}}1(\tau; r) + \mathbb{A}\tilde{\mathcal{G}}1(\tau; r) \} &= 0, \\ p^\gamma \mathcal{M}_\tau \{ \tilde{\mathcal{G}}2(\tau; r) \} - p^{\gamma+1} \tilde{\mathbb{Y}}2 + \mathcal{M}_\tau \{ -1 + \mathbb{B}\tilde{\mathcal{G}}2(\tau; r) + \tilde{\mathcal{G}}1^2(\tau; r) \} &= 0, \\ p^\gamma \mathcal{M}_\tau \{ \tilde{\mathcal{G}}3(\tau; r) \} - p^{\gamma+1} \tilde{\mathbb{Y}}3 + \mathcal{M}_\tau \{ +\tilde{\mathcal{G}}1(\tau; r) + \mathbb{C}\tilde{\mathcal{G}}3(\tau; r) \} &= 0. \end{aligned} \tag{37}$$

Homotopies of abovementioned system for $q \in [0,1]$ are as follows:

$$\begin{aligned} \text{H1: } (1 - q) (\mathcal{M}_\tau \{ \tilde{\mathcal{G}}1(\tau; r) \} - \tilde{\mathcal{G}}1_0) + q \left(\mathcal{M}_\tau \{ \tilde{\mathcal{G}}1(\tau; r) \} - p\tilde{\mathbb{Y}}1 + \left(\frac{1}{p^\gamma} \right) \mathcal{M}_\tau \{ -\tilde{\mathcal{G}}3(\tau; r) - \tilde{\mathcal{G}}2(\tau; r)\tilde{\mathcal{G}}1(\tau; r) + \mathbb{A}\tilde{\mathcal{G}}1(\tau; r) \} \right) &= 0, \\ \text{H2: } (1 - q) (\mathcal{M}_\tau \{ \tilde{\mathcal{G}}2(\tau; r) \} - \tilde{\mathcal{G}}2_0) + q \left(\mathcal{M}_\tau \{ \tilde{\mathcal{G}}2(\tau; r) \} - p\tilde{\mathbb{Y}}2 + \left(\frac{1}{p^\gamma} \right) \mathcal{M}_\tau \{ -1 + \mathbb{B}\tilde{\mathcal{G}}2(\tau; r) + \tilde{\mathcal{G}}1^2(\tau; r) \} \right) &= 0, \\ \text{H3: } (1 - q) (\mathcal{M}_\tau \{ \tilde{\mathcal{G}}3(\tau; r) \} - \tilde{\mathcal{G}}3_0) + q \left(\mathcal{M}_\tau \{ \tilde{\mathcal{G}}3(\tau; r) \} - p\tilde{\mathbb{Y}}3 + \left(\frac{1}{p^\gamma} \right) \mathcal{M}_\tau \{ \tilde{\mathcal{G}}1(\tau; r) + \mathbb{C}\tilde{\mathcal{G}}3(\tau; r) \} \right) &= 0. \end{aligned} \tag{38}$$

Substitution of (18) in (38) gives the following at q^1 :

$$\begin{aligned} \mathcal{M}_\tau\{\tilde{\mathcal{G}}_{1_1}(\tau; r)\} + \tilde{\mathcal{G}}_{1_0} - p\tilde{\mathcal{Y}}_1 + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{-\tilde{\mathcal{G}}_{3_0}(\tau; r) - \tilde{\mathcal{G}}_{2_0}(\tau; r)\tilde{\mathcal{G}}_{1_0}(\tau; r) + \mathbb{A}\tilde{\mathcal{G}}_{1_0}(\tau; r)\} &= 0, \\ \mathcal{M}_\tau\{\tilde{\mathcal{G}}_{2_1}(\tau; r)\} + \tilde{\mathcal{G}}_{2_0} - p\tilde{\mathcal{Y}}_2 + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{-1 + \mathbb{B}\tilde{\mathcal{G}}_{2_0}(\tau; r) + \tilde{\mathcal{G}}_{1_0}^2(\tau; r)\} &= 0, \\ \mathcal{M}_\tau\{\tilde{\mathcal{G}}_{3_1}(\tau; r)\} + \tilde{\mathcal{G}}_{3_0} - p\tilde{\mathcal{Y}}_3 + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{\tilde{\mathcal{G}}_{1_0}(\tau; r) + \mathbb{C}\tilde{\mathcal{G}}_{3_0}(\tau; r)\} &= 0. \end{aligned} \tag{39}$$

Applying the inverse Mohand transform results in

At q^2 , we obtain

$$\begin{aligned} \tilde{\mathcal{G}}_{1_1}(\tau; r) &= -\frac{(\mathbb{A}\tilde{\mathcal{Y}}_1 - \tilde{\mathcal{Y}}_2\tilde{\mathcal{Y}}_1 - \tilde{\mathcal{Y}}_3)\tau^\gamma}{\Gamma(\gamma + 1)}, \\ \tilde{\mathcal{G}}_{2_1}(\tau; r) &= -\frac{(\mathbb{B}\tilde{\mathcal{Y}}_2 + \tilde{\mathcal{Y}}_1^2 - 1)\tau^\gamma}{\Gamma(\gamma + 1)}, \\ \tilde{\mathcal{G}}_{3_1}(\tau; r) &= -\frac{(\mathbb{C}\tilde{\mathcal{Y}}_3 + \tilde{\mathcal{Y}}_1)\tau^\gamma}{\Gamma(\gamma + 1)}. \end{aligned} \tag{40}$$

$$\begin{aligned} \mathcal{M}_\tau\{\tilde{\mathcal{G}}_{1_2}(\tau; r)\} + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{-\tilde{\mathcal{G}}_{3_1}(\tau; r) - \tilde{\mathcal{G}}_{2_1}(\tau; r)\tilde{\mathcal{G}}_{1_1}(\tau; r) + \mathbb{A}\tilde{\mathcal{G}}_{1_1}(\tau; r)\} &= 0, \\ \mathcal{M}_\tau\{\tilde{\mathcal{G}}_{2_2}(\tau; r)\} + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{-1 + \mathbb{B}\tilde{\mathcal{G}}_{2_1}(\tau; r) + \tilde{\mathcal{G}}_{1_1}^2(\tau; r)\} &= 0, \\ \mathcal{M}_\tau\{\tilde{\mathcal{G}}_{3_2}(\tau; r)\} + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{\tilde{\mathcal{G}}_{1_1}(\tau; r) + \mathbb{C}\tilde{\mathcal{G}}_{3_1}(\tau; r)\} &= 0. \end{aligned} \tag{41}$$

Applying the inverse Mohand transform gives

$$\begin{aligned} \tilde{\mathcal{G}}_{1_2}(\tau; r) &= \frac{-\tau^{2\gamma}(-\tilde{\mathcal{Y}}_1(\mathbb{A}^2 - \tilde{\mathcal{Y}}_2(2\mathbb{A} + \mathbb{B}) + \tilde{\mathcal{Y}}_2^2) + \tilde{\mathcal{Y}}_3(\mathbb{A} + \mathbb{C} - \tilde{\mathcal{Y}}_2) + \tilde{\mathcal{Y}}_1^3)}{\Gamma(2\gamma + 1)}, \\ \tilde{\mathcal{G}}_{2_2}(\tau; r) &= \frac{\tau^{2\gamma}(\tilde{\mathcal{Y}}_1^2(2\mathbb{A} + \mathbb{B} - 2\tilde{\mathcal{Y}}_2) + \mathbb{B}(\mathbb{B}\tilde{\mathcal{Y}}_2 - 1) - 2\tilde{\mathcal{Y}}_3\tilde{\mathcal{Y}}_1)}{\Gamma(2\gamma + 1)}, \\ \tilde{\mathcal{G}}_{3_2}(\tau; r) &= \frac{\tau^{2\gamma}(\tilde{\mathcal{Y}}_1(\mathbb{A} + \mathbb{C} - \tilde{\mathcal{Y}}_2) + (\mathbb{C}^2 - 1)\tilde{\mathcal{Y}}_3)}{\Gamma(2\gamma + 1)}. \end{aligned} \tag{42}$$

At q^3 , we obtain

$$\begin{aligned}
 \mathcal{M}_\tau\{\tilde{\mathcal{G}}1_3(\tau; r)\} + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{-\tilde{\mathcal{G}}3_2(\tau; r) - \tilde{\mathcal{G}}2_2(\tau; r)\tilde{\mathcal{G}}1_2(\tau; r) + \mathbb{A}\tilde{\mathcal{G}}1_2(\tau; r)\} &= 0, \\
 \mathcal{M}_\tau\{\tilde{\mathcal{G}}2_3(\tau; r)\} + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{-1 + \mathbb{B}\tilde{\mathcal{G}}2_2(\tau; r) + \tilde{\mathcal{G}}1_2^2(\tau; r)\} &= 0, \\
 \mathcal{M}_\tau\{\tilde{\mathcal{G}}3_3(\tau; r)\} + \left(\frac{1}{p^\gamma}\right)\mathcal{M}_\tau\{\tilde{\mathcal{G}}1_2(\tau; r) + \mathbb{C}\tilde{\mathcal{G}}3_2(\tau; r)\} &= 0.
 \end{aligned}
 \tag{43}$$

Applying the inverse Mohand transform leads to

$$\begin{aligned}
 \tilde{\mathcal{G}}1_3(\tau; r) &= \frac{1}{\Gamma(\gamma+1)^2\Gamma(3\gamma+1)}\tau^{3\gamma} - \tilde{\mathcal{Y}}3((\mathbb{A}^2 + \mathbb{A}\mathbb{C} + \mathbb{C}^2 - 1)\Gamma(\gamma+1)^2 - \tilde{\mathcal{Y}}2((2\mathbb{A} + \mathbb{C})\Gamma(\gamma+1)^2 + \mathbb{B}\Gamma(2\gamma+1)) \\
 &\quad + \Gamma(2\gamma+1) + \tilde{\mathcal{Y}}2^2\Gamma(\gamma+1)^2) \\
 &\quad + \tilde{\mathcal{Y}}1(\Gamma(\gamma+1)^2(\mathbb{A}^3 - \mathbb{A} + \mathbb{B} - \mathbb{C})) \\
 &\quad - \tilde{\mathcal{Y}}2((3\mathbb{A}^2 + \mathbb{A}\mathbb{B} + \mathbb{B}^2 - 1)\Gamma(\gamma+1)^2 + (\mathbb{A}\mathbb{B} + 1)\Gamma(2\gamma+1)) \\
 &\quad + \mathbb{A}\Gamma(2\gamma+1) + \tilde{\mathcal{Y}}2^2((3\mathbb{A} + \mathbb{B})\Gamma(\gamma+1)^2 + \mathbb{B}\Gamma(2\gamma+1)) + \tilde{\mathcal{Y}}2^3(-\Gamma(\gamma+1)^2) \\
 &\quad - \tilde{\mathcal{Y}}1^3(\mathbb{A}\Gamma(2\gamma+1) + (3\mathbb{A} + \mathbb{B})\Gamma(\gamma+1)^2 - \tilde{\mathcal{Y}}2(3\Gamma(\gamma+1)^2 + \Gamma(2\gamma+1))) \\
 &\quad + \tilde{\mathcal{Y}}3\tilde{\mathcal{Y}}1^2(2\Gamma(\gamma+1)^2 + \Gamma(2\gamma+1)), \\
 \tilde{\mathcal{G}}2_3(\tau; r) &= \frac{-1}{\Gamma(\gamma+1)^2\Gamma(3\gamma+1)}\tau^{3\gamma}\tilde{\mathcal{Y}}1^2(\mathbb{A}^2\Gamma(2\gamma+1) + (2\mathbb{A}^2 + 2\mathbb{A}\mathbb{B} + \mathbb{B}^2)\Gamma(\gamma+1)^2) \\
 &\quad - \tilde{\mathcal{Y}}2(\mathbb{A}\Gamma(2\gamma+1) + 2(\mathbb{A} + \mathbb{B})\Gamma(\gamma+1)^2) + \tilde{\mathcal{Y}}2^2(2\Gamma(\gamma+1)^2 + \Gamma(2\gamma+1)) \\
 &\quad - 2\tilde{\mathcal{Y}}3\tilde{\mathcal{Y}}1(\mathbb{A}\Gamma(2\gamma+1) + \Gamma(\gamma+1)^2(\mathbb{A} + \mathbb{B} + \mathbb{C}) - \tilde{\mathcal{Y}}2(\Gamma(\gamma+1)^2 + \Gamma(2\gamma+1))) \\
 &\quad + \mathbb{B}^3\tilde{\mathcal{Y}}2\Gamma(\gamma+1)^2 - \mathbb{B}^2\Gamma(\gamma+1)^2 - 2\tilde{\mathcal{Y}}1^4\Gamma((\gamma+1)^2 + \tilde{\mathcal{Y}}3^2\Gamma(2\gamma+1)), \\
 \tilde{\mathcal{G}}3_3(\tau; r) &= \frac{-1}{\Gamma(3\gamma+1)}\tau^{3\gamma}(\tilde{\mathcal{Y}}1(\mathbb{A}^2 - \tilde{\mathcal{Y}}2(2\mathbb{A} + \mathbb{B} + \mathbb{C}) + \mathbb{A}\mathbb{C} + \mathbb{C}^2 + \tilde{\mathcal{Y}}2^2) + \tilde{\mathcal{Y}}3(-\mathbb{A} + \mathbb{C}^3 - 2\mathbb{C} + \tilde{\mathcal{Y}}2) - \tilde{\mathcal{Y}}1^3).
 \end{aligned}
 \tag{44}$$

The higher-order problems and solutions can be calculated in a similar way. Thus, by adding the terms we can get the required approximate solution. The residual error of system (35) can be observed through (26).

7. Results and Discussion

The main objective of the current study is the solution and analysis of a fuzzy-fractional chaotic financial model that depends upon interest rate, price index, and investment demand. It is a highly nonlinear differential system with a time-fractional derivative. The fuzziness in initial conditions $\tilde{\mathcal{Y}}1$, $\tilde{\mathcal{Y}}2$, and $\tilde{\mathcal{Y}}3$ are incorporated with the help of triangular fuzzy numbers (TFNs). The approximate series solution is calculated for both the upper bound and lower

bound of TFNs through the He–Mohand technique. In this method, the homotopy perturbation method and Mohand transform are combined to tackle the noninteger order derivative and fuzziness. At different values of time, solution and absolute errors are determined. The accuracy of the proposed methodology can be seen from absolute residual and system errors (Tables 1 and 2) that range from 10^{-6} to 10^{-12} at the upper bound and from 10^{-6} to 10^{-11} at the lower bound for fractional parameter $\gamma = 0.89$ and 1.0 and $r = 0.8$.

To analyze the behavior of interest rate, investment demand, and price index across the fuzzy domain, three-dimensional plots are created. From Figure 1, it can be observed that in the case of lower bound solutions, increasing the value of r -cut decreases the rate of interest, investment demand, and price index. On the other hand, the

TABLE 1: Upper and lower bound solutions and errors at $\mathbb{A} = \mathbb{B} = \mathbb{C} = 0.01$, $r = 0.8$, and $\gamma = 0.89$.

τ	Solution			Absolute errors				
	$\tilde{\mathcal{G}}_1$	$\tilde{\mathcal{G}}_2$	$\tilde{\mathcal{G}}_3$	$ \mathcal{R} _{\tilde{\mathcal{G}}_1}^-$	$ \mathcal{R} _{\tilde{\mathcal{G}}_2}^-$	$ \mathcal{R} _{\tilde{\mathcal{G}}_3}^-$	$ \mathcal{R} _{\tilde{\mathcal{G}}-\text{system}}^-$	
$\overline{\mathcal{G}}(\tau; r)$	0.1	0.246633	-0.313071	0.127044	1.28×10^{-8}	1.74×10^{-10}	6.07×10^{-9}	6.35×10^{-9}
	0.2	0.251791	-0.205296	0.098220	4.90×10^{-7}	3.20×10^{-10}	2.46×10^{-7}	2.45×10^{-7}
	0.3	0.256299	-0.103982	0.070471	4.04×10^{-6}	5.89×10^{-8}	2.14×10^{-6}	2.08×10^{-6}
	0.4	0.260353	-0.006756	0.043282	1.77×10^{-5}	5.08×10^{-7}	9.97×10^{-6}	9.41×10^{-6}
	0.5	0.264054	0.087470	0.016437	5.51×10^{-5}	2.37×10^{-6}	3.28×10^{-5}	3.01×10^{-5}
$\underline{\mathcal{G}}(\tau; r)$	0.1	0.484377	0.065320	0.297071	5.15×10^{-8}	1.83×10^{-8}	1.24×10^{-8}	2.74×10^{-8}
	0.2	0.520534	0.150092	0.238334	2.07×10^{-6}	6.96×10^{-7}	5.03×10^{-7}	1.09×10^{-6}
	0.3	0.553101	0.225697	0.178841	1.79×10^{-5}	5.73×10^{-6}	4.39×10^{-6}	9.35×10^{-6}
	0.4	0.582844	0.294436	0.118016	8.27×10^{-5}	2.54×10^{-5}	2.04×10^{-5}	4.28×10^{-5}
	0.5	0.609988	0.357481	0.055758	2.69×10^{-4}	8.02×10^{-5}	6.71×10^{-5}	1.39×10^{-4}

TABLE 2: Upper and lower bound solutions and errors at $\mathbb{A} = \mathbb{B} = \mathbb{C} = 0.01$, $r = 0.8$, and $\gamma = 1.0$.

τ	Solution			Absolute errors				
	$\tilde{\mathcal{G}}_1$	$\tilde{\mathcal{G}}_2$	$\tilde{\mathcal{G}}_3$	$ \mathcal{R} _{\tilde{\mathcal{G}}_1}^-$	$ \mathcal{R} _{\tilde{\mathcal{G}}_2}^-$	$ \mathcal{R} _{\tilde{\mathcal{G}}_3}^-$	$ \mathcal{R} _{\tilde{\mathcal{G}}-\text{system}}^-$	
$\overline{\mathcal{G}}(\tau; r)$	0.1	0.245020	-0.345490	0.135598	6.78×10^{-11}	8.22×10^{-12}	2.54×10^{-11}	3.38×10^{-11}
	0.2	0.249716	-0.251312	0.110736	8.35×10^{-9}	9.35×10^{-10}	3.26×10^{-9}	4.18×10^{-9}
	0.3	0.254131	-0.157456	0.085443	1.37×10^{-7}	1.41×10^{-8}	5.57×10^{-8}	6.90×10^{-8}
	0.4	0.258305	-0.063910	0.059746	9.86×10^{-7}	9.25×10^{-8}	4.17×10^{-7}	4.98×10^{-7}
	0.5	0.262273	0.029330	0.033669	4.50×10^{-6}	3.82×10^{-7}	1.99×10^{-6}	2.29×10^{-6}
$\underline{\mathcal{G}}(\tau; r)$	0.1	0.473284	0.039131	0.313990	5.06×10^{-11}	1.29×10^{-10}	8.29×10^{-11}	8.77×10^{-11}
	0.2	0.505566	0.115083	0.264750	5.78×10^{-9}	1.50×10^{-8}	1.06×10^{-8}	1.04×10^{-8}
	0.3	0.536854	0.187749	0.212381	8.75×10^{-8}	2.30×10^{-7}	1.81×10^{-7}	1.66×10^{-7}
	0.4	0.567106	0.257040	0.156990	2.74×10^{-7}	1.54×10^{-6}	1.35×10^{-6}	1.15×10^{-6}
	0.5	0.596234	0.322897	0.098685	2.36×10^{-6}	6.46×10^{-6}	6.47×10^{-6}	5.10×10^{-6}

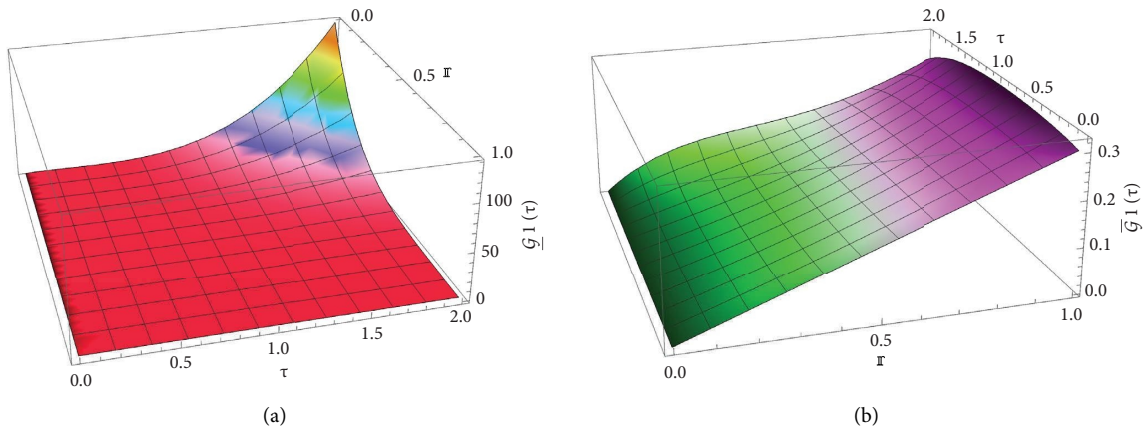


FIGURE 1: Continued.

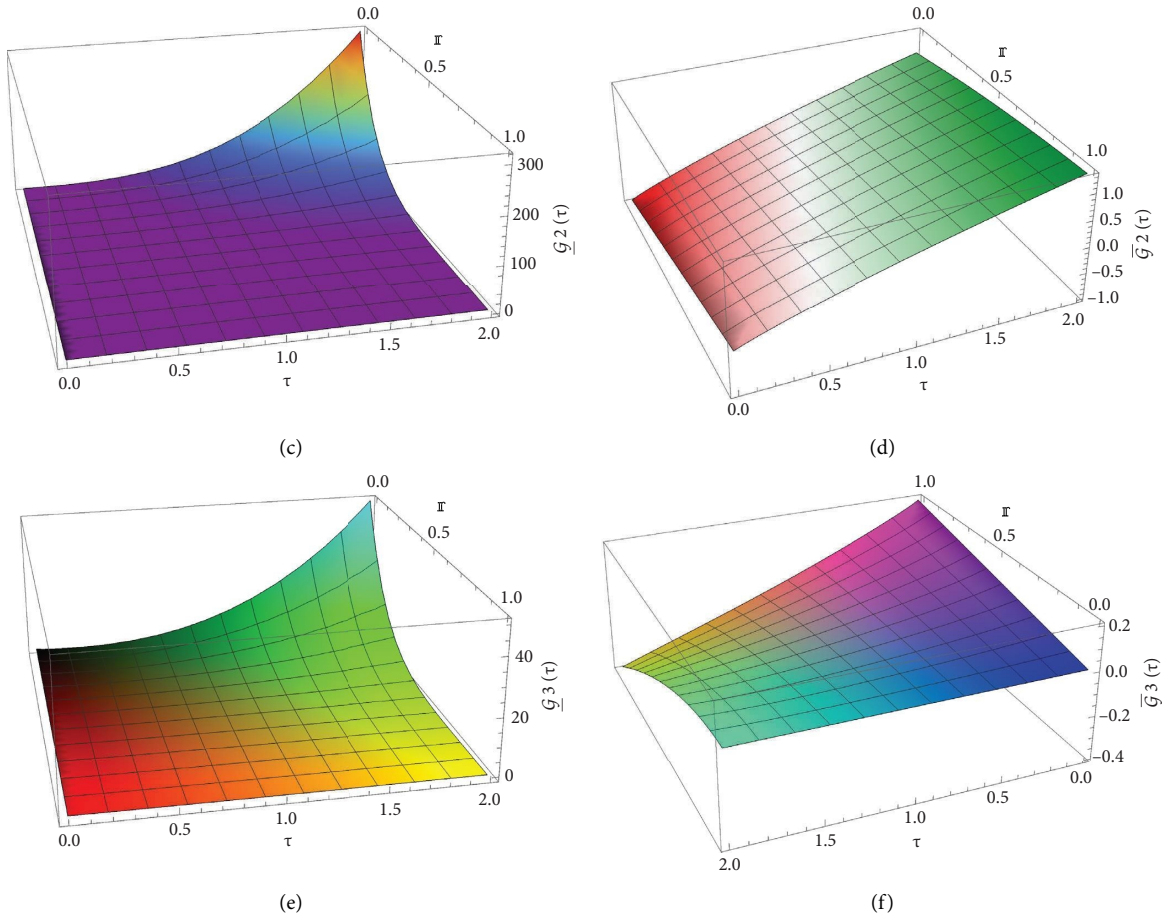


FIGURE 1: 3D fuzzy upper and lower bound solutions at $\mathbb{A} = 0.20$, $\mathbb{B} = 0.10$, $\mathbb{C} = 0.11$, and $\gamma = 0.51$.

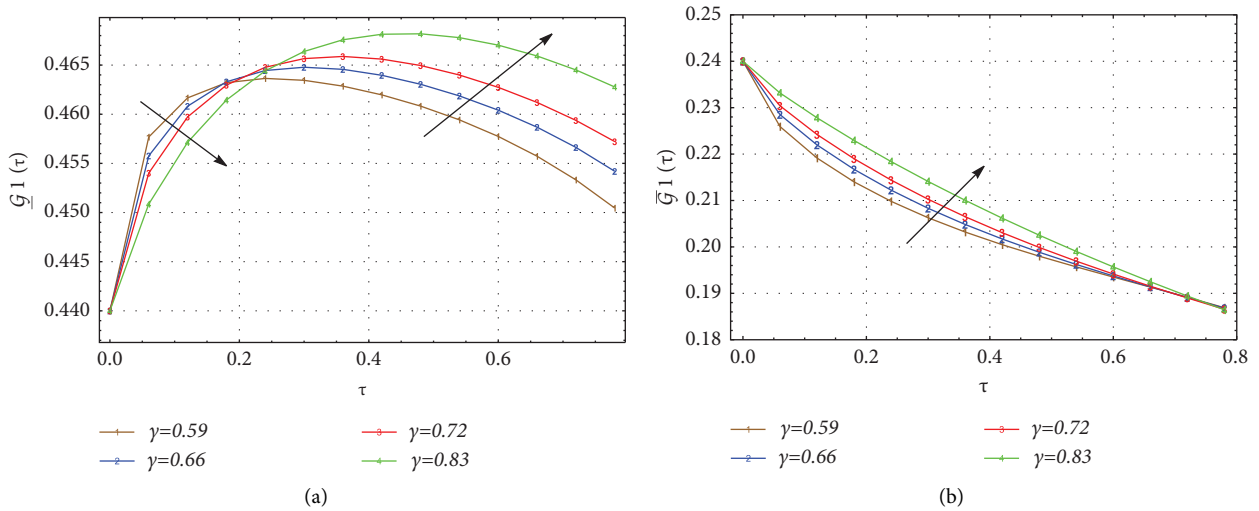


FIGURE 2: Continued.

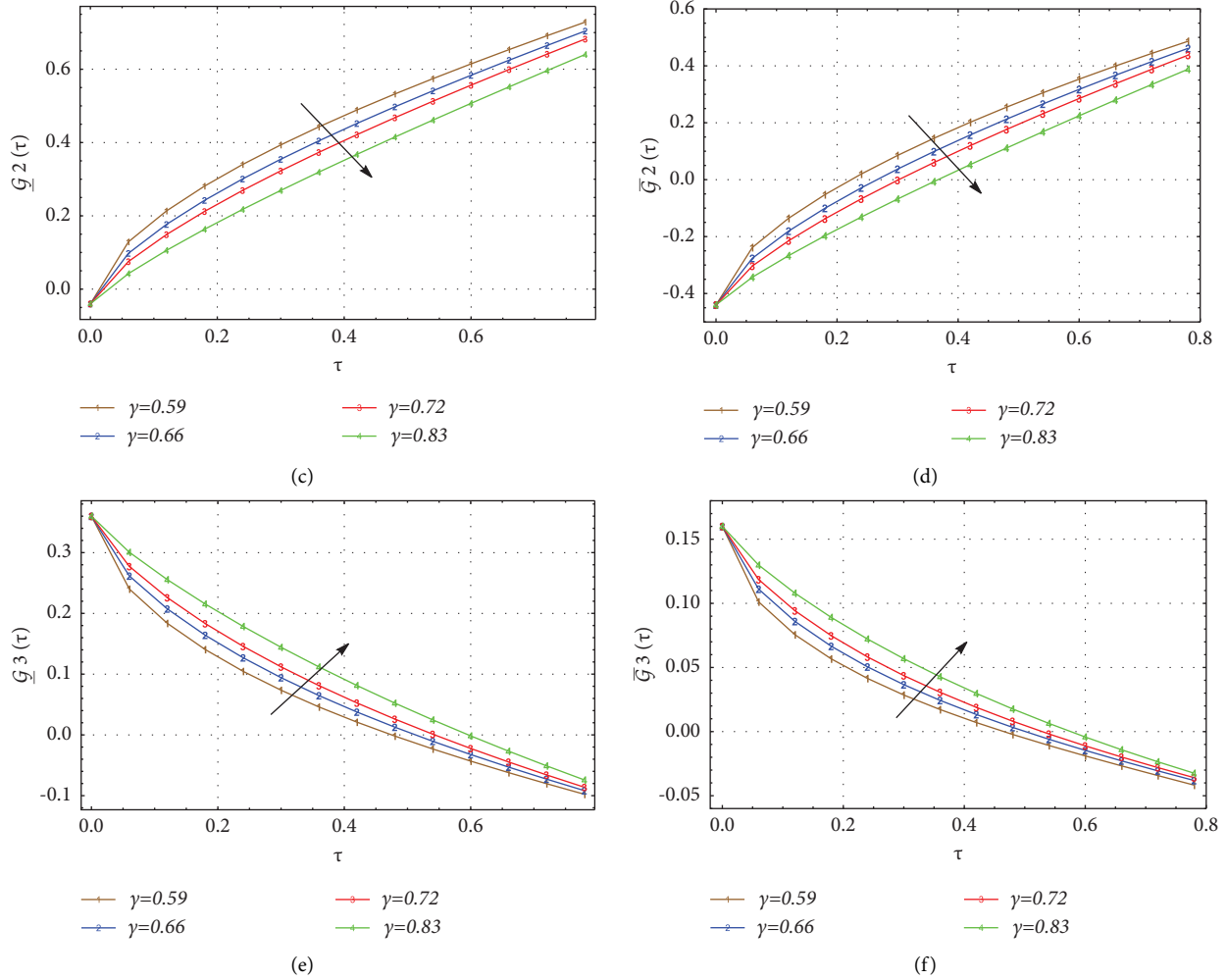


FIGURE 2: 2D fuzzy upper and lower bound solutions at different fractional order γ when $A = 0.5$, $B = 0$, $C = 0.4$, and $r = 0.8$.

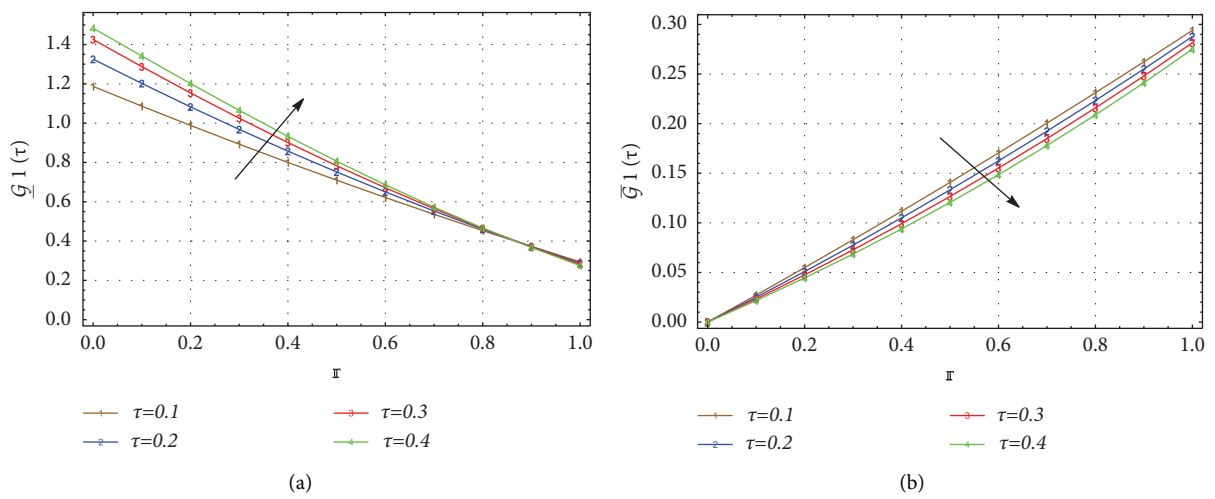


FIGURE 3: Continued.

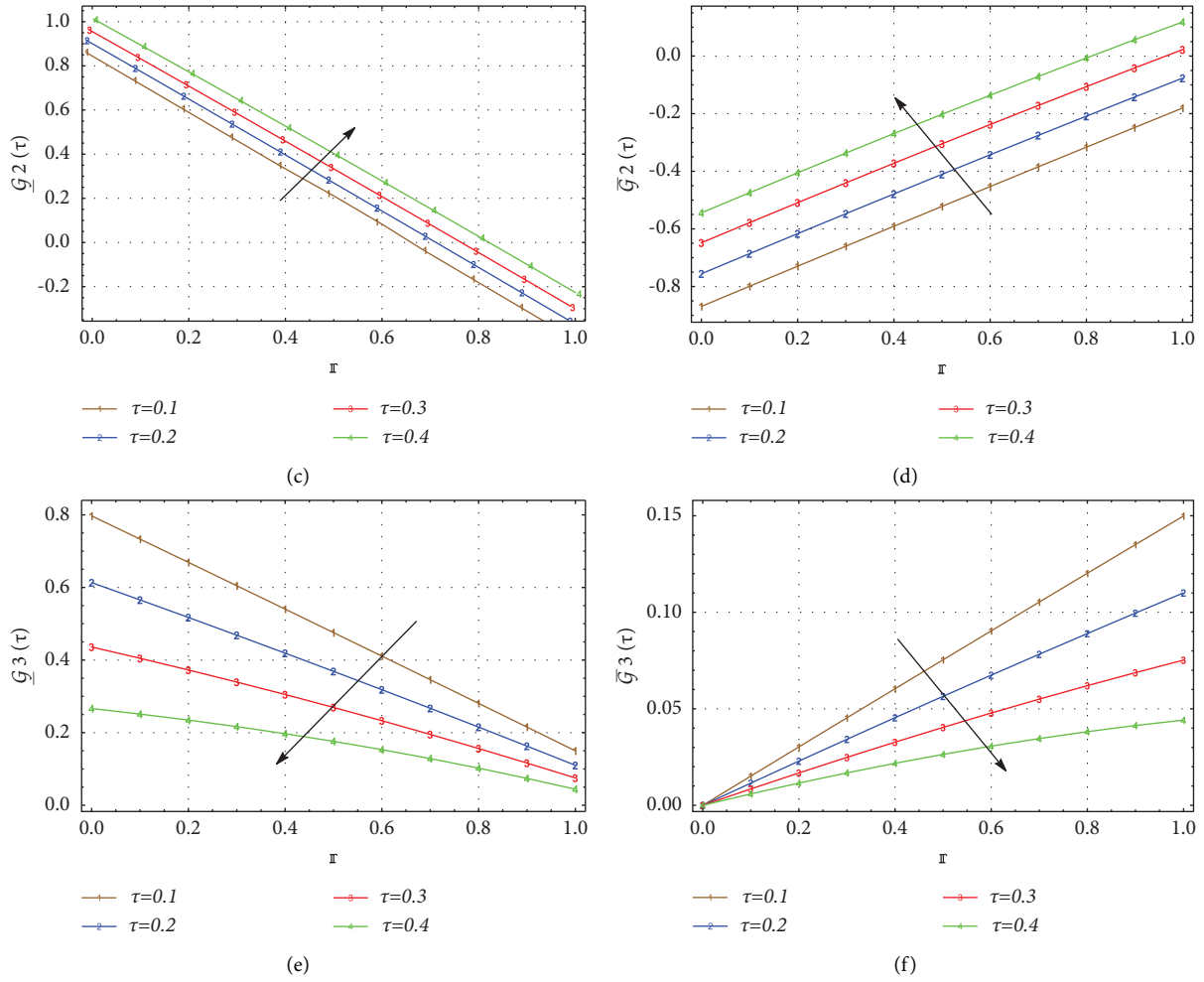


FIGURE 3: 2D fuzzy upper and lower bound solutions at varying values of time τ when $\mathbb{A} = 0.5$, $\mathbb{B} = 0.0$, $\mathbb{C} = 0.5$, and $\gamma = 0.9$.

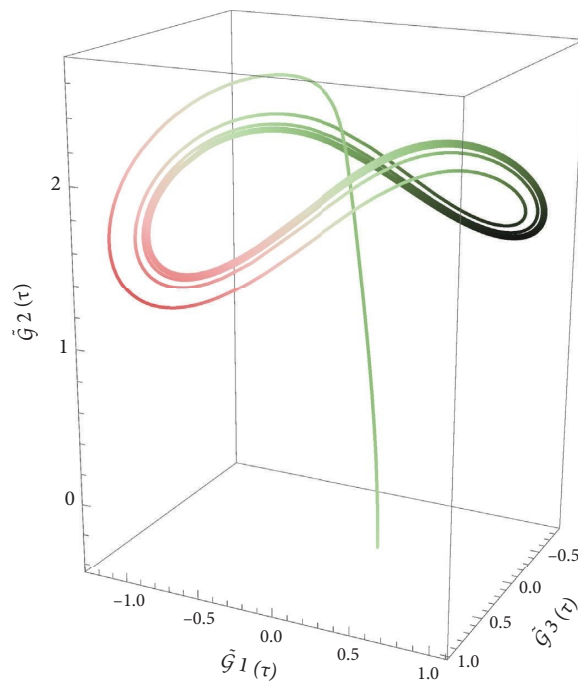


FIGURE 4: Dynamic behavior of interest rate, investment demand, and price index at $\mathbb{A} = 1.1$, $\mathbb{B} = 0.33$, $\mathbb{C} = 0.97$, $\gamma = 0.95$, and $r = 1$.

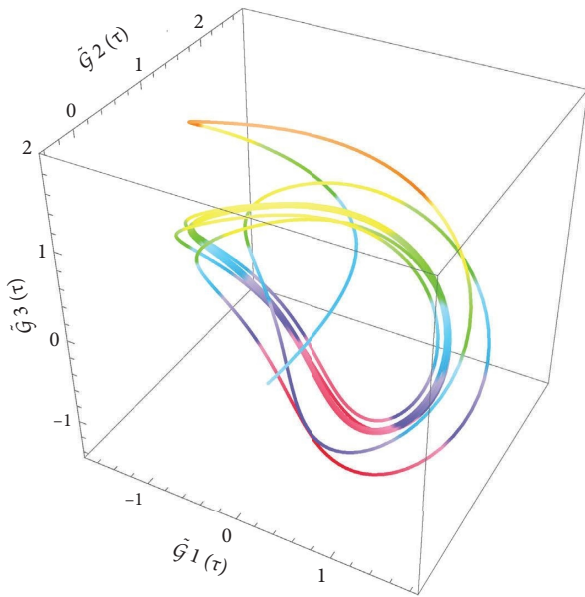


FIGURE 5: Dynamic behavior of interest rate, investment demand, and price index at $\mathbb{A} = 0.9$, $\mathbb{B} = 0.2$, $\mathbb{C} = 0.3$, $\gamma = 0.95$, and $r = 1$.

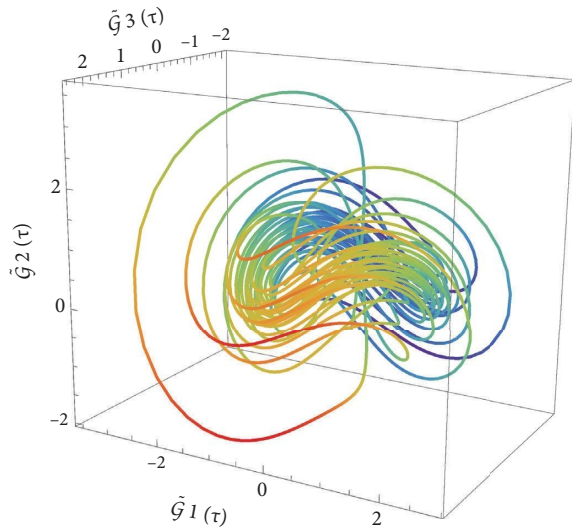


FIGURE 6: Dynamic behavior of interest rate, investment demand, and price index at $\mathbb{A} = 0.43$, $\mathbb{B} = 0.0$, $\mathbb{C} = 0.12$, $\gamma = 0.95$, and $r = 1$.

rate of interest, investment demand, and price index increase with the increase in r -cut value. At $r=0$, the given fuzzy-fractional chaotic system shows maximum uncertainty. The fuzziness in interest rate, investment demand, and price index begins to decline when r expands in its domain as they eventually convert to their crisp form at $r=1$. Figure 2 illustrates the impact of various fractional parameter values with time on the profiles of interest rate, investment demand, and price index in two-dimensional formation. It is seen that at the lower bound solution of interest rate, initially, the solution profile decreases before changing course

after a certain time. In the case of an upper bound, the fractional parameter exhibits a rise in the interest rate. The upper and lower bound profile of the chaotic system increases along with the fractional parameter for the price index. However, the investment demand declines with an increment in the fractional parameter value.

In Figure 3, the behavior of interest rate, investment demand, and price index is demonstrated across the fuzzy domain for different values of the time. It is displayed through arrows that the interest rate at the lower bound rises as time increases. On the other hand, it is declining in the case of the upper bound. At both the upper and lower bounds, overtime the investment demand shows a rise while the price index depreciates. Furthermore, the chaotic patterns of the system at different values of saving amount, per-investment cost, and elasticity in demands with respect to interest rate, investment demand, and price index are illustrated in Figures 4–6 at $\gamma=1$ and fractional parameter $\gamma=0.97$. A significant increase in the chaotic behavior of interest rate, investment demand, and price index is observed as the value of saving rate, elasticity of demands, and per-investment cost decreases.

8. Conclusion

The purpose of this research article is the modeling and analysis of the fuzzy-fractional financial chaotic model. Here, we combine fuzzy logic with fractional calculus through an efficient semianalytical methodology which is known as the He–Mohand algorithm. The time-fractional derivative in the model is considered in the Caputo sense. The triangular fuzzy numbers (TFNs) approach is used to include the uncertainty in the system. Error analysis spanning across the r -cut domain is illustrated through tables. It is seen that the obtained errors range from 10^{-6} to 10^{-12} at the upper bound and from 10^{-6} to 10^{-11} at the lower bound. The efficiency of the proposed methodology is also presented in the theoretical analysis. From this, it can be noticed that the He–Mohand algorithm is a convergent scheme. The behavior of interest rate, investment demand, and price index is analyzed in two-dimensional and three-dimensional plots at both upper and lower bounds. The effect of time and fractional parameters on the system profile with regard to r -cut is also studied. It is estimated that as r -cut approaches to 1, correspondingly solution becomes less fuzzy and eventually changes into a crisp form at $r=1$. It is also observed that the smaller value saving rate, elasticity of demands, and per-investment cost has a significant effect on the chaotic behavior of the system. In conclusion, the modeled fuzzy-fractional financial chaotic system has the potential in helping the analyst to better comprehend the predictions and risk assessments of financial systems. Moreover, the proposed methodology can be efficiently utilized to tackle various research areas of the financial market such as risk analysis, portfolio administration, and decision-making procedures in fractional and fuzzy environments in the future.

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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