Research Article

An Eigenvalue-Eigenvector Method for Solving a System of Fractional Differential Equations with Uncertainty

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A new method is proposed for solving systems of fuzzy fractional differential equations (SFFDEs) with fuzzy initial conditions involving fuzzy Caputo differentiability. For this purpose, three cases are introduced based on the eigenvalue-eigenvector approach; then it is shown that the solution of system of fuzzy fractional differential equations is vector of fuzzy-valued functions. Then the method is validated by solving several examples.

1. Introduction

Recently, a lot of research has been focused on the application of fractional calculus, and such application is in the modelling of many physical and chemical processes as well as in engineering [1–5].

It has been found that the behavior of many physical systems can be properly described by using the fractional order system theory. Fractional derivatives provide an excellent instrument for the description of memory and hereditary properties of various materials and processes. The advantages or the real objects of the fractional order systems are that we have more degrees of freedom in the model and that a “memory” is included in the model. For example, the nonlinear oscillation of earthquake can be modeled with fractional derivatives [6]. In mechanics, fractional calculus plays an important role; for example, it has been successfully employed to model damping forces with memory effects to describe state feedback controllers [7, 8] and dynamics of interfaces between nanoparticles and substrates [9]. Due to its tremendous scope and applications in several disciplines, a considerable attention has been given to exact and numerical solutions of fractional differential equations [10–18]. The analytic results on the existence and uniqueness of solutions to the fractional differential equations have been studied by many authors [19, 20]. From the numerical point of view, several methods have been presented to achieve the goal of highly accurate and reliable solutions for the fractional differential equations. The most commonly used methods are fractional differential transform method [21], operational matrix method [22, 23], finite difference method [24], and Haar wavelets method [25].

On the other hand, fuzzy differential equations have received considerable attention in dealing with various problems. So the development in this field has risen from the theoretical and practical perspectives [26–33].

Recently, Agarwal et al. [34] proposed the concept of solutions for fractional differential equations with uncertainty which was followed by the authors in [35, 36]. They have considered Riemann-Liouville’s differentiability to solve FDEs which is a combination of the Hukuhara difference and the Riemann-Liouville derivative. In [37, 38], the authors considered the generalization of H-differentiability for the fractional case. A lot of research has been devoted to find the accurate and efficient methods for solving fuzzy fractional differential equations (FFDEs). It is well known that the exact
solutions of most of the FFDEs cannot be found easily; therefore, in the recent years, attempts have been made to address this problem [39–41]. It is with this motivation that we introduce in this paper an eigenvalue–eigenvector method for solving fuzzy fractional differential equations (FFDEs).

Also regarding some defects of H-differentiability [42], Bede et al. [43–46] discussed the limitations of the H-differentiability due to the shortcoming of the Hukuhara differentiability. So we motivated our interest to adopt our proposed method based on the generalized differentiability in the sense of fractional order which was introduced by [37,38,41].

In this paper, we intend to investigate the solutions of systems of fractional differential equations with uncertainty which is called system of fuzzy fractional differential equations (SFFDEs). Here, we use the Riemann-Liouville derivative in the fuzzy concept, which applied the concept of the system of fractional derivatives under Caputo’s differentiability by applying the Hukuhara difference, which is denoted as fuzzy Caputo’s derivative. Similar to the deterministic cases, the construction of Caputo’s derivatives is based on the definition of the Riemann-Liouville derivatives in fuzzy cases.

This paper is organized as follows. In Section 2, we review the well-known definitions of fuzzy numbers, and some basic concepts are given. In Section 3, fuzzy Caputo’s derivative is introduced, and the relation between Riemann-Liouville and Caputo’s derivatives and some of its properties is considered. Consequently, the eigenvalue-eigenvector method for solving fuzzy fractional order linear systems with initial values under fuzzy Caputo’s derivative is given, and three cases of eigenvalues are considered in Section 4. The proposed method is illustrated by solving several examples in Section 5 to depict the applicability and validity of the proposed method. Finally, conclusion is drawn in Section 6.

2. Preliminaries

The basic definition of fuzzy numbers is given in [47, 48]. We denote the set of all real numbers by $\mathbb{R}$, and the set of all fuzzy number on $\mathbb{R}$ is indicated by $E$. A fuzzy number is a mapping $u: \mathbb{R} \to [0,1]$ with the following properties:

(a) $u$ is upper semicontinuous,
(b) $u$ is fuzzy convex; that is, $u(\lambda x + (1-\lambda)y) \geq \min\{u(x),u(y)\}$ for all $x, y \in \mathbb{R}$, $\lambda \in [0,1]$,
(c) $u$ is normal; that is, $\exists x_0 \in \mathbb{R}$ for which $u(x_0) = 1$,
(d) $\text{supp} \ u = \{x \in \mathbb{R} \mid u(x) > 0\}$ is the support of the $u$, and its closure $\text{cl}(\text{supp} \ u)$ is compact.

An equivalent parametric definition is also given in [49–51] as follows.

**Definition 1.** A fuzzy number $u$ in parametric form is a pair $(\tilde{u}, \overline{u})$ of functions $u(r), \overline{u}(r)$, $0 \leq r \leq 1$, which satisfy the following requirements:

1. $u(r) \leq \overline{u}(r)$, $0 \leq r \leq 1$.

Moreover, we can also present the $r$-cut representation of fuzzy number as $[u]^r = [(u(r), \overline{u}(r))]$ for all $0 \leq r \leq 1$.

According to Zadeh’s extension principle, operation of addition on $E$ is defined by

$$
(u \oplus v)(x) = \sup_{y \in \mathbb{R}} \min \{u(y), v(x-y)\}, \quad x \in \mathbb{R},
$$

and scalar multiplication of a fuzzy number is given by

$$
(k \odot u)(x) = \begin{cases} u\left(\frac{x}{k}\right), & k > 0, \\ 0, & k = 0, \\ \overline{u}(x), & k < 0 \end{cases}
$$

where $\overline{0} \in \mathbb{E}$.

The Hausdorff distance between fuzzy numbers is given by $d: E \times E \to \mathbb{R}_+ \cup \{0\}$,

$$
d(u, v) = \sup_{r \in [0,1]} \max \{|u(r) - v(r)|, |\overline{u}(r) - \overline{v}(r)|\},
$$

where $u = (u(r), \overline{u}(r))$ and $v = (v(r), \overline{v}(r)) \in \mathbb{R}$ is utilized in [43]. Then, it is easy to see that $d$ is a metric in $\mathbb{E}$ and has the following properties (see [33]):

1. $d(u + w, v + w) = d(u, v)$, for all $u, v, w \in \mathbb{E}$,
2. $d(ku, kv) = |k|d(u, v)$, for all $k \in \mathbb{R}$, $u, v \in \mathbb{E}$,
3. $d(u + v, w + e) \leq d(u, w) + d(v, e)$, for all $u, v, w, e \in \mathbb{E}$,
4. $(d, \mathbb{E})$ is a complete metric space.

**Definition 2.** Let $x, y \in \mathbb{E}$. If there exists $z \in \mathbb{E}$ such that $x = y + z$, then $z$ is called the H-difference of $x$ and $y$, and it is denoted by $x \ominus y$.

In this paper, the sign “$\ominus$” always stands for H-difference, and also note that $x \ominus y \neq x + (-1)y$.

3. Fuzzy Caputo’s Derivative

In this section, the concept of fuzzy Caputo’s derivatives is considered using the Hukuhara difference. We denote $C^F[a,b]$ as a space of all fuzzy-valued functions which are continuous on $[a,b]$. Also, we denote the space of all Lebesgue integrable fuzzy-value functions on the bounded interval $[a,b] \subset \mathbb{R}$ by $L^F[a,b]$. We denote the space of fuzzy-value functions $f(x)$ which have continuous H-derivative up to order $n-1$ on $[a,b]$ such that $f^{(n-1)}(x) \in AC^F([a,b])$ by $AC^{(n)}([a,b])$.

Now, we define the fuzzy Riemann-Liouville integral of fuzzy-valued function as follows.

**Definition 3.** Let $f(x) \in C^F[a,b] \cap L^F[a,b]$, the fuzzy Riemann-Liouville integral of fuzzy-valued function $f$ is defined as follows:

$$
\left(\mathcal{I}^\beta_{a+} f\right)(x) = \frac{1}{\Gamma(\beta)} \int_a^x \frac{f(t) \, dt}{(x-t)^{1-\beta}}, \quad x > a, \ 0 < \beta < 1.
$$

(4)
Since $f(x; r) = [f(x; r), f(x; r)]$, for all $0 \leq r \leq 1$, then we can indicate the fuzzy Riemann-Liouville integral of fuzzy-valued function $f$ based on the lower and upper functions as follows.

**Theorem 4.** Let $f(x) \in C^\beta[a, b] \cap L^k[a, b]$, the fuzzy Riemann-Liouville integral of fuzzy-valued function $f$ is defined as follows:

$$
\left( I_{a+}^\beta f \right)(x; r) = \left[ \left( I_{a+}^\beta \tilde{f} \right)(x; r), \left( I_{a+}^\beta \tilde{f} \right)(x; r) \right], \quad 0 \leq r \leq 1,
$$

(5)

where

$$
\left( I_{a+}^\beta \tilde{f} \right)(x; r) = \frac{1}{\Gamma(\beta)} \int_a^x \frac{f(t; r) \, dt}{(x-t)^{1-\beta}}, \quad 0 \leq r \leq 1,
$$

(6)

$$
\left( I_{a+}^\beta \tilde{f} \right)(x; r) = \frac{1}{\Gamma(\beta)} \int_a^x \frac{\tilde{f}(t; r) \, dt}{(x-t)^{1-\beta}}, \quad 0 \leq r \leq 1.
$$

Let $f \in C^\beta((0, a)) \cap L^k(0, a)$ be a given function such that $f(t, r) = [f(t; r), \tilde{f}(t; r)]$ for all $t \in (0, a)$ and $0 \leq r \leq 1$. We define the fuzzy fractional Riemann-Liouville derivative of order $0 < \beta < 1$ of $f$,

$$
\text{RL}D^\beta f(t) = \frac{d}{dt} \left[ (t-s)^{-\beta} f(s; r) ds \right],
$$

(7)

and in the parametric form,

$$
\text{RL}D^\beta f(t, r) = \frac{d}{dt} \int_0^t (t-s)^{-\beta} f(s; r) ds,
$$

(8)

provided that the equation defines a fuzzy number $\text{RL}D^\beta f(t) \in E$. In fact,

$$
\text{RL}D^\beta f(t, r) := \left[ \text{RL}D^\beta f(t; r), \text{RL}D^\beta \tilde{f}(t; r) \right].
$$

(9)

Obviously, $\text{RL}D^\beta f(t) = (d/dt)I^{1-\beta} f(t)$ for $t \in (0, a]$.

**Definition 5** (the relation between Riemann-Liouville and Caputo’s operators). Let $f(x) \in C^\alpha(a, b) \cap L^k[a, b]$ be a fuzzy-valued function. One defines the fuzzy fractional Caputo’s derivative of order $0 < \beta < 1$ of $f$,

$$
\left( C^D_{a+}^\beta f \right)(x) = \left( \text{RL}D^\beta_{a+} \left[ f(t) \oplus \sum_{k=0}^{n-1} \frac{t^k}{k!} f^{(k)}(a) \right] \right)(x),
$$

(10)

provided that the equation defines a fuzzy number $C^D_{a+} f(x) \in E$:

$$
\left( C^D_{a+}^\beta f \right)(x; r) = \left[ \left( C^D_{a+}^\beta \tilde{f} \right)(x; r), \left( C^D_{a+}^\beta \tilde{f} \right)(x; r) \right],
$$

(11)

where

$$
\text{RL}D^\beta_{a+} f(t; r) = 1_{(0, a)} f(t; r) - \sum_{k=0}^{n-1} \frac{t^k}{k!} f^{(k)}(a; r),
$$

$$
\text{RL}D^\beta_{a+} \tilde{f}(t; r) = 1_{(0, a)} \tilde{f}(t; r) - \sum_{k=0}^{n-1} \frac{t^k}{k!} \tilde{f}^{(k)}(a; r)
$$

for all $0 \leq r \leq 1$ and $0 < \beta < 1$. Obviously, $C^D_{a+}^\beta f(t) = P^{n-\beta} D^n f(t)$ for $n-1 < \beta < n$, $n \in \mathbb{N}$.

**Definition 6** (see [52]). The $n \times n$ linear system is

$$
a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = y_1,
$$

$$
a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = y_2,
$$

$$
\vdots
$$

$$
a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = y_n,
$$

(13)

where the coefficient matrix $A = (a_{ij})$, $1 \leq i, j \leq n$, is a crisp $n \times n$ matrix, and $y_j \in \mathbb{E}$, $1 \leq i \leq n$, is called a fuzzy system of linear equations (FSLEs).

**Definition 7.** A fuzzy number vector $(x_1, x_2, \ldots, x_n)'$ given by $x_i = (\tilde{x}_i(r), \bar{x}_i(r))$, $1 \leq i \leq n$, $0 \leq r \leq 1$, is called a solution of the FSLE if

$$
\sum_{j=1}^n a_{ij}x_j = \sum_{j=1}^n a_{ij}x_j = y_j,
$$

(14)

$$
\sum_{j=1}^n a_{ij}x_j = \sum_{j=1}^n a_{ij}x_j = \bar{y}_j.
$$

(15)

Considering the $i$th equation of the system (13)

$$
a_{i1} \tilde{x}_1 + \cdots + a_{ii} \tilde{x}_i + \cdots + a_{in} \tilde{x}_n = \tilde{y}_i(r),
$$

(16)

we have

$$
\frac{a_{i1} \tilde{x}_1 + \cdots + a_{ii} \tilde{x}_i + \cdots + a_{in} \tilde{x}_n = \tilde{y}_i(r)}{a_{i1} \bar{x}_1 + \cdots + a_{ii} \bar{x}_i + \cdots + a_{in} \bar{x}_n = \bar{y}_i(r)},
$$

$$
1 \leq i \leq n, \quad 0 \leq r \leq 1.
$$

(17)
From (16), we have two crisp $n \times n$ linear systems for all $i$ that can be extended to a $2n \times 2n$ crisp linear system as follows:

$$SX = Y \iff \begin{bmatrix} s_1 & s_2 & \cdots & s_n \\ s_2 & s_1 & \cdots & \vdots \\ \vdots & \ddots & \ddots & \vdots \\ s_n & \cdots & s_2 & s_1 \end{bmatrix} \begin{bmatrix} X \\ Y \end{bmatrix} = \begin{bmatrix} X \\ Y \end{bmatrix},$$

(17)

where $s_{ij}$ are determined as follows:

$$
\begin{align*}
\text{if } a_{ij} \geq 0 & \implies s_{ij} = a_{ij}, s_{i+j,n+j} = a_{ij}, \\
\text{if } a_{ij} < 0 & \implies s_{i+j,n+j} = a_{ij}, s_{i+j,n+j} = a_{ij},
\end{align*}
$$

(18)

and any $s_{ij}$ which is not determined is zero.

In this paper, the following system will be solved:

$$
\begin{align*}
\frac{d^\beta_1 \bar{x}_1}{dt^\beta_1} &= f_1(\bar{x}_1, \bar{x}_2, \ldots, \bar{x}_n) = a_{11} \bar{x}_1 + a_{12} \bar{x}_2 + \cdots + a_{1n} \bar{x}_n, \\
\frac{d^\beta_2 \bar{x}_2}{dt^\beta_2} &= f_2(\bar{x}_1, \bar{x}_2, \ldots, \bar{x}_n) = a_{21} \bar{x}_1 + a_{22} \bar{x}_2 + \cdots + a_{2n} \bar{x}_n, \\
&\hspace{1cm} \vdots \\
\frac{d^\beta_n \bar{x}_n}{dt^\beta_n} &= f_n(\bar{x}_1, \bar{x}_2, \ldots, \bar{x}_n) = a_{n1} \bar{x}_1 + a_{n2} \bar{x}_2 + \cdots + a_{nn} \bar{x}_n.
\end{align*}
$$

(19)

Thus,

$$
\frac{d^\beta}{dt^\beta} \bar{X}(t) = A\bar{X}(t), \quad \bar{X}(0) = \bar{X}_0 \in \mathbb{E}^n,
$$

(20)

where $\bar{X} \in \mathbb{E}^n$, the matrix $A = (a_{ij})$, $1 \leq i, j \leq n$, $\beta = [\beta_1, \beta_2, \ldots, \beta_n]$ indicates the fractional orders, $d^\beta/dt^\beta = [d^{\beta_1}/dt^{\beta_1}, d^{\beta_2}/dt^{\beta_2}, \ldots, d^{\beta_n}/dt^{\beta_n}]$, and $d^\beta_1/dt^\beta_1$ is the fuzzy Caputo’s fractional derivative of order $\beta_i$, where $0 < \beta_i < 1$, for $i = 1, 2, \ldots, n$. To obtain the solution of $(d^\beta/dt^\beta)\bar{X}(t) = A\bar{X}(t)$, the eigenvalue-eigenvector method is used.

### 4. Fuzzy Fractional Order Linear Systems

In this section, we drive the general solution for fuzzy fractional order linear system as follows:

$$
\frac{d^\beta}{dt^\beta} \bar{X}(t) = A\bar{X}(t), \quad 0 < t \leq a,
$$

(21)

where $\bar{X} \in \mathbb{E}^n$, $a > 0$, the coefficient matrix $A = (a_{ij})$, $1 \leq i, j \leq n$, is a crisp $n \times n$ matrix and $\bar{x}_i \in \mathbb{E}$, $1 \leq i \leq n$, and $d^\beta/dt^\beta$ is the fuzzy Caputo’s fractional derivative, where $0 < \beta < 1$. Simply to construct the general solution of the system (21), we proceed by analogy with treatment of homogeneous integer order fuzzy linear systems with the constant coefficient where the exponential function $\text{Exp}(t)$ is replaced by the Mittag-Leffler function $E_\beta(t^\beta)$. Thus, we seek solutions of the form

$$
\bar{X}(t) = \xi E_\beta(\lambda t^\beta),
$$

(22)

where the constant $\lambda$ and the vector $\xi$ are to be determined. Substituting form (22) for $\bar{X}$ in the system (21) gives

$$
\xi \lambda E_\beta(\lambda t^\beta) = A\xi E_\beta(\lambda t^\beta).
$$

(23)

Upon canceling the nonzero factor $E_\beta(\lambda t^\beta)$, we obtain $A\xi = \lambda \xi$ or

$$
(A - \lambda I) \xi = 0,
$$

(24)

where $I$ is the $n \times n$ identity matrix. Therefore, the vector $\bar{X}$ given by (22) is a solution of the system (21) provided that $\lambda$ and the vector $\xi$ are associated eigenvectors of the matrix $A$. In the following Section, three cases for the eigenvalue of matrix $A$ are discussed.

### 4.1. Real and Distinct Eigenvalues.

In this case, suppose that $\lambda_i$, for $i = 1, \ldots, n$, are the real eigenvalues of matrix $A$. Therefore, the solution of (20) is as follows:

$$
\bar{X}(t) = \sum_{i=1}^{n} \zeta_i v_i(t),
$$

(25)

where $\zeta_i$ are fuzzy numbers, $v_i(t) = E_\beta(\lambda_i t^\beta) \xi_i$ for $i = 1, \ldots, n$ and $\lambda_i$ and $\xi_i$ are the real eigenvalues and eigenvector of matrix $A$, respectively.

By setting initial values $t = 0$, in (25),

$$
\bar{X}(a) = \sum_{i=1}^{n} \zeta_i E_\beta(\lambda_i a^\beta) \xi_i = \bar{\alpha}
$$

(26)

is obtained, where $\xi_i = [\xi_{i1}, \xi_{i2}, \ldots, \xi_{in}]^T$ and $\bar{\alpha} = [\bar{\alpha}_1, \bar{\alpha}_2, \ldots, \bar{\alpha}_n]^T$. Thus, $\zeta_i$, $i = 1, \ldots, n$. From the following, fuzzy systems are obtained:

$$
\begin{align*}
\zeta_1 E_\beta \lambda_1 a^\beta \xi_{11} + \zeta_2 E_\beta \lambda_2 a^\beta \xi_{12} + \cdots + \zeta_n E_\beta \lambda_n a^\beta \xi_{1n} &= \bar{\alpha}_1, \\
\zeta_1 E_\beta \lambda_1 a^\beta \xi_{21} + \zeta_2 E_\beta \lambda_2 a^\beta \xi_{22} + \cdots + \zeta_n E_\beta \lambda_n a^\beta \xi_{2n} &= \bar{\alpha}_2, \\
&\hspace{1cm} \vdots \\
\zeta_1 E_\beta \lambda_1 a^\beta \xi_{n1} + \zeta_2 E_\beta \lambda_2 a^\beta \xi_{n2} + \cdots + \zeta_n E_\beta \lambda_n a^\beta \xi_{nn} &= \bar{\alpha}_n.
\end{align*}
$$

(27)
The parametric form of (27) is as follows:

\[
\begin{align*}
(\xi_1 (r), \bar{c}_1 (r)) E_\beta \lambda_1 a^\beta \xi_1 1 + (\xi_2 (r), \bar{c}_2 (r)) E_\beta \lambda_2 a^\beta \xi_2 1 \\
+ \cdots + (\xi_n (r), \bar{c}_n (r)) E_\beta \lambda_n a^\beta \xi_n 1 & = (\xi_1 (r), \bar{c}_1 (r)), \\
(\xi_1 (r), \bar{c}_1 (r)) E_\beta \lambda_1 a^\beta \xi_1 2 + (\xi_2 (r), \bar{c}_2 (r)) E_\beta \lambda_2 a^\beta \xi_2 2 \\
+ \cdots + (\xi_n (r), \bar{c}_n (r)) E_\beta \lambda_n a^\beta \xi_n 2 & = (\xi_1 (r), \bar{c}_1 (r)), \\
\vdots & \\
(\xi_1 (r), \bar{c}_1 (r)) E_\beta \lambda_1 a^\beta \xi_1 n + (\xi_2 (r), \bar{c}_2 (r)) E_\beta \lambda_2 a^\beta \xi_2 n \\
+ \cdots + (\xi_n (r), \bar{c}_n (r)) E_\beta \lambda_n a^\beta \xi_n n & = (\xi_1 (r), \bar{c}_1 (r)).
\end{align*}
\]

(28)

Now similar to (17), there is a 2n×2n crisp system. Therefore, \( \bar{c}_i = (\xi_i (r), \bar{c}_i (r)), i = 1, 2, \ldots, n \) are obtained from (28) and are set in (25). Finally the solution of (20) will be obtained from \( \bar{X}(t) = [\bar{x}_1, \bar{x}_2, \ldots, \bar{x}_n] \).

**Theorem 8.** The solution of fuzzy system (20) with real eigenvalues is a fuzzy number (25).

**Proof.** It is shown that \( \bar{X}(t) = \sum_{i=1}^{n} c_i E(\lambda, t^\beta) \xi_i \) for \( i = 1, 2, \ldots, n \) and \( 0 < \beta < 1 \) is the solution of \( (d^\beta/dt^\beta) \bar{X} = \lambda \bar{X} \).

Let \( \bar{C} = [\bar{c}_1, \bar{c}_2, \ldots, \bar{c}_n] \) with \( \bar{c}_i \) which are the fuzzy numbers and \( [\bar{C}]_r = \Pi_{i=1}^{n} [\bar{c}_i]_r \), then

\[
\begin{align*}
X (t; r) & = \min \left\{ \sum_{i=1}^{n} c_i E(\lambda, t^\beta) \xi_i \mid c_i \in [\bar{C}]_r \right\} \\
& = \sum_{i=1}^{n} c_i E(\lambda, t^\beta) \xi_i = \sum_{i=1}^{n} c_i E(\lambda, t^\beta) \xi_i, \\
\bar{X} (t; r) & = \max \left\{ \sum_{i=1}^{n} c_i E(\lambda, t^\beta) \xi_i \mid c_i \in [\bar{C}]_r \right\} \\
& = \sum_{i=1}^{n} c_i E(\lambda, t^\beta) \xi_i = \sum_{i=1}^{n} c_i E(\lambda, t^\beta) \xi_i.
\end{align*}
\]

(29)

With differentiation of (29), we are obtained:

\[
\begin{align*}
\frac{d^\beta}{dt^\beta} X (t; r) & = \sum_{i=1}^{n} \lambda_i c_i E(\lambda, t^\beta) \xi_i, \quad 0 \leq r \leq 1, \\
\frac{d^\beta}{dt^\beta} \bar{X} (t; r) & = \sum_{i=1}^{n} \lambda_i c_i E(\lambda, t^\beta) \xi_i, \quad 0 \leq r \leq 1.
\end{align*}
\]

(30)

Since \( \lambda_i \) is an eigenvalue and \( \xi_i \) is its corresponding eigenvalue of matrix \( A \), then \( A \xi_i = \lambda \xi_i, i = 1, \ldots, n \). Therefore,

\[
\frac{d^\beta}{dt^\beta} X (t; r) = \sum_{i=1}^{n} \lambda_i c_i E(\lambda, t^\beta) \xi_i = \sum_{i=1}^{n} \lambda_i \xi_i E(\lambda, t^\beta) = \lambda \bar{X} (t; r),
\]

(31)

\[
\frac{d^\beta}{dt^\beta} \bar{X} (t; r) = \sum_{i=1}^{n} \lambda_i c_i E(\lambda, t^\beta) \xi_i = \sum_{i=1}^{n} \lambda_i \xi_i E(\lambda, t^\beta) = \lambda \bar{X} (t; r).
\]

(32)

Such that

\[
\left( \frac{d^\beta}{dt^\beta} X (t; r) ; \frac{d^\beta}{dt^\beta} \bar{X} (t; r) \right) = \lambda \left( X (t; r) ; \bar{X} (t; r) \right),
\]

(33)

\[
0 < \beta < 1, \quad 0 \leq r \leq 1.
\]

This means that \( (d^\beta/dt^\beta) \bar{X}(t) = \lambda \bar{X}(t) \). From (7), (10), and \( (d^\beta/dt^\beta) \bar{X}(t) = [(d^\beta/dt^\beta) \bar{x}_1, (d^\beta/dt^\beta) \bar{x}_2, \ldots, (d^\beta/dt^\beta) \bar{x}_n]^t \), it is clear that \( \bar{X}(t) \) is a fuzzy number vector.

4.2. Complex Eigenvalues. In this case, suppose that some eigenvalues of \( \lambda_i \), for \( i = 1, 2, \ldots, k \), are complex numbers. Since the entries of matrix \( A \) are real, therefore characteristic polynomial has real coefficients; therefore, complex roots are in conjugate pairs.

**Lemma 9.** Let the entries of matrix \( A \) be real and \( \lambda \) an eigenvalue of matrix \( A \), where \( \lambda = \mu + iv, v \neq 0 \), and \( \xi = a + ib \) are the corresponding eigenvectors of \( \lambda \), then \( u_r (t) = \text{Re}[E_\beta(\mu + iv)t^\beta)] \) and \( \bar{u}_r (t) = \text{Im}[E_\beta(\mu + iv)t^\beta] \) are solutions.

Therefore, from the above-mentioned lemma, the solution of each pair of conjugate complex eigenvalues \( \lambda = \mu + iv \) is as follows:

\[
\bar{w}(t) = \bar{c}_r \text{Re} \left[ \xi E_\beta (\mu + iv t^\beta) \right] + \bar{c}_i \text{Im} \left[ \xi E_\beta (\mu + iv t^\beta) \right],
\]

(33)

where \( \bar{c}_r \) is the corresponding eigenvector of eigenvalue \( \lambda \). Hence, the solution of (33) is as follows:

\[
\bar{X}(t) = \sum_{i=k+1}^{k} \bar{w}_i (t) + \sum_{i=1}^{k} \bar{v}_i (t),
\]

(34)

where \( \bar{w}_i (t) = \bar{c}_r \text{Re}[\xi E_\beta (\lambda, t^\beta)] + \bar{c}_i \text{Im}[\xi E_\beta (\lambda, t^\beta)] \) from each pair of conjugate complex eigenvalues and \( \bar{v}_i (t) = \bar{c}_r \xi E_\beta (\lambda, t^\beta) \) from real eigenvalues are obtained. Then by setting initial values \( t = a \) in (34) and by solving a fuzzy system similar to (28), fuzzy coefficients are obtained. By setting fuzzy coefficient in (34), \( \bar{X}(t) \) is obtained; finally the solution of (20) will be obtained from \( \bar{X}(t) = [x_1, x_2, \ldots, x_n]^t \).
Theorem 10. The solution of fuzzy system (20) with complex eigenvalues is a fuzzy number (34).

Proof. It is shown that $\bar{X}(t) = \sum_{i=1}^{k/2} \bar{w}_i(t) + i \sum_{i=k/2}^{n} \bar{v}_i(t)$ is the solution of $(d^\beta/dt^\beta) \bar{X} = A \bar{X}$.

\begin{align*}
\bar{X}(t; r) &= \min \left\{ \sum_{i=1}^{k/2} \bar{c}_1 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) + c_2 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) \right\} \\
&= \sum_{i=1}^{k/2} \bar{c}_1 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) + c_2 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) + \sum_{i=k/2+1}^{n} c_i \xi_i E_\beta (\lambda_i t^\theta), \quad 0 < \beta < 1, \ 0 \leq r \leq 1.
\end{align*}

(35)

With differentiation of the above equations we obtain the following

\begin{align*}
\frac{d^\beta}{dt^\beta} \bar{X}(t; r) &= \sum_{i=1}^{k/2} \bar{c}_1 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) + c_2 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) \\
&+ \sum_{i=k/2+1}^{n} c_i \lambda_i \xi_i E_\beta (\lambda_i t^\theta), \quad 0 \leq r \leq 1.
\end{align*}

(36)

\begin{align*}
\frac{d^\beta}{dt^\beta} \bar{X}(t; r) &= \sum_{i=1}^{k/2} \bar{c}_1 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) + c_2 \left( \xi_i E_\beta (\lambda_i t^\theta) \right) \\
&+ \sum_{i=k/2+1}^{n} c_i \lambda_i \xi_i E_\beta (\lambda_i t^\theta), \quad 0 \leq r \leq 1.
\end{align*}

Since $\lambda_i$ is an eigenvalue and $\xi_i$ is its corresponding eigenvalue of matrix $A$, then $A \xi_i = \lambda_i \xi_i$, $i = 1, \ldots, n$. Therefore,

\begin{align*}
\frac{d^\beta}{dt^\beta} \bar{X}(t; r) &= \sum_{i=1}^{k/2} \bar{c}_1 \left( A \xi_i E_\beta (\lambda_i t^\theta) \right) + c_2 \left( A \xi_i E_\beta (\lambda_i t^\theta) \right) \\
&+ \sum_{i=k/2+1}^{n} c_i A \xi_i E_\beta (\lambda_i t^\theta).
\end{align*}

(37)

Then,

\begin{align*}
\left( \frac{d^\beta}{dt^\beta} \bar{X}(t; r) ; \frac{d^\beta}{dt^\beta} \bar{X}(t; r) \right) &= A \left( \bar{X}(t; r) ; \bar{X}(t; r) \right), \quad 0 < \beta < 1, \ 0 \leq r \leq 1.
\end{align*}

(38)
This means that \( (d^\beta/dt^\beta)\bar{X}(t) = A\bar{X}(t) \). From (7), (10), and \( (d^\beta/dt^\beta)\bar{X}(t) = [(d^\beta/dt^\beta)\bar{x}_1, (d^\beta/dt^\beta)\bar{x}_2, \ldots, (d^\beta/dt^\beta)\bar{x}_n]^T \), it is clear that \( \bar{X}(t) \) is a fuzzy number vector.

4.3. Multiple Eigenvalues. In this case, suppose that some eigenvalues of matrix \( A \) are multiple. Suppose that \( \lambda_0 \) is an eigenvalue of matrix \( A \) with multiplicity \( k \), and the corresponding eigenvectors of eigenvalue \( \lambda_0 \) are \( \xi_1, \xi_2, \ldots, \xi_k \), if all \( \xi_i \) are linearly independent, then

\[
\bar{X}(t) = \bar{c}_1\xi_1 E_\beta(\lambda_0 t^\beta) + \bar{c}_2\xi_2 E_\beta(\lambda_0 t^\beta) + \cdots + \bar{c}_k\xi_k E_\beta(\lambda_0 t^\beta).
\]

(39)

If \( \xi \) and \( m \) are linearly independent vectors, that is, \( m < k \), then the following lemma is brought.

**Lemma 11.** Let \( \lambda_0 \) be an eigenvalue of matrix \( A \) with multiple \( k > 1 \), and let the numbers of \( \xi_i \) which are linearly independent be less than \( m_0 \), therefore at least one non-zero vector exists such that

\[
(A - \lambda I) \xi \neq 0, \quad (A - \lambda I)^2 \xi = 0. \tag{40}
\]

If \( \xi \) is satisfied in (40), the solution is as follows:

\[
v'(t) = E_\beta^{(1)}(\lambda_0 t^\beta) t^\beta (A - \lambda_0 I) \xi + \xi E_\beta(\lambda_0 t^\beta), \tag{41}
\]

based on the properties of the Mittag-Leffler type functions, where

\[
E_\beta^{(1)}(z) = \frac{d}{dt} E_\beta(z). \tag{42}
\]

In general, if matrix \( A \) has a repeated eigenvalue \( \lambda_0 \) of multiplicity \( k \) with \( m \) linearly independent eigenvectors, where \( m < k \), then the following

\[
x_1(t) = \xi_1 E_\beta(\lambda_0 t^\beta),
\]

\[
x_2(t) = \xi_1 t^\beta E_\beta^{(1)}(\lambda_0 t^\beta) + \xi_2 E_\beta(\lambda_0 t^\beta),
\]

\[
\vdots
\]

\[
x_{k-m}(t) = \xi_1 t^\beta(k-m-1) E_\beta^{(k-m-1)}(\lambda_0 t^\beta) + \xi_2 t^\beta(k-m-2) E_\beta^{(k-m-2)}(\lambda_0 t^\beta) + \cdots + \xi_{k-m} E_\beta(\lambda_0 t^\beta),
\]

are \( k - m \) linearly independent solutions of the system (20).

Hence, with the above-mentioned lemma, the solution of (20) is as follows:

\[
\bar{X}(t) = \sum_{i=1}^{i=k} \bar{c}_i v_i(t) + \sum_{i=k+1}^{i=n} \bar{c}_i t^\beta v_i(t), \tag{44}
\]

where \( v_i(t) = E_\beta^{(k)}(\lambda_i t^\beta) t^\beta (A - \lambda_i I) \xi_i + \xi_i E_\beta(\lambda_i t^\beta) \) for \( \lambda_i \) which are satisfied in Lemma 11 and \( v_i(t) = \xi_i E_\beta(\lambda_i t^\beta) \) for real eigenvalues are obtained. Then by setting initial values \( t = a \) in (44) and by solving a fuzzy system similar to (28), fuzzy coefficient is obtained. By setting fuzzy coefficient in (44), \( \bar{X}(t) \) is obtained, and finally the solution of (20) will be obtained from \( \bar{X}(t) = [\bar{x}_1, \bar{x}_2, \ldots, \bar{x}_n]^T \).

**Theorem 12.** The solution of fuzzy system (20) with multiple eigenvalues is a fuzzy number (44).

**Proof.** It is shown that \( \bar{X}(t) = \sum_{i=1}^{i=k} \bar{c}_i v_i(t) + \sum_{i=k+1}^{i=n} \bar{c}_i t^\beta v_i(t) \) is the solution of \((d^\beta/dt^\beta)\bar{X} = A\bar{X} \). Let \( \mathcal{C} = \{\bar{c}_1, \bar{c}_2, \ldots, \bar{c}_n\} \) with \( \bar{c}_1 \) which are the fuzzy numbers and \( |\mathcal{C}_r| = \Gamma_{m_0}^{n-1} |\mathcal{C}_r| \), then

\[
\bar{X}(t; r) = \min \left\{ \sum_{i=1}^{i=k} \bar{c}_i E_\beta^{(k)}(\lambda_i t^\beta) (t^\beta (A - \lambda_i I) \xi_i) + \xi_i E_\beta(\lambda_i t^\beta) + \sum_{i=k+1}^{i=n} \bar{c}_i E_\beta(\lambda_i t^\beta) \mid c_i \in [\mathcal{C}]_r \right\}
\]

\[
= \sum_{i=1}^{i=k} \bar{c}_i E_\beta^{(k)}(\lambda_i t^\beta) (t^\beta (A - \lambda_i I) \xi_i) + \xi_i E_\beta(\lambda_i t^\beta) + \sum_{i=k+1}^{i=n} \bar{c}_i E_\beta(\lambda_i t^\beta)
\]

\[
= \sum_{i=1}^{i=k} \bar{c}_i E_\beta^{(k)}(\lambda_i t^\beta) (t^\beta (A - \lambda_i I) \xi_i) + \xi_i E_\beta(\lambda_i t^\beta) + \sum_{i=k+1}^{i=n} \bar{c}_i E_\beta(\lambda_i t^\beta)
\]

\[
= \sum_{i=1}^{i=k} \bar{c}_i E_\beta^{(k)}(\lambda_i t^\beta) (t^\beta (A - \lambda_i I) \xi_i) + \xi_i E_\beta(\lambda_i t^\beta) + \sum_{i=k+1}^{i=n} \bar{c}_i E_\beta(\lambda_i t^\beta).
\]

(45)

With differentiation of the above equations, we obtain
\[
\frac{d^\beta X(t; r)}{dt^\beta} = \sum_{i=1}^{k} \left[ \lambda_i E_{\beta}^k \left( \lambda \right) \left( \begin{array}{c} (A - \lambda_i I) \xi_i + \xi_i \lambda_i E_{\beta} \left( \lambda \right) \end{array} \right) \right]
\]
\[
\frac{d^\beta X(t; r)}{dt^\beta} = \sum_{i=1}^{k} \left[ \lambda_i E_{\beta}^k \left( \lambda \right) \left( \begin{array}{c} (A - \lambda_i I) \xi_i + \xi_i \lambda_i E_{\beta} \left( \lambda \right) \end{array} \right) \right]
\]
\[
= \left( \begin{array}{c} 0.75 + 0.25r \\ 1.5 + 0.5r \\ 3.75 + 0.25r \\ 1.25 - 0.25r \\ 2.5 - 0.5r \\ 4.25 - 0.25r \end{array} \right)
\]

Since \( \lambda \) is an eigenvalue and \( \xi \) is its corresponding eigenvalue of matrix \( A \), then \( A \xi = \lambda \xi \). Therefore,
\[
\left( \begin{array}{c} \frac{d^\beta X(t; r)}{dt^\beta} \\ \frac{d^\beta X(t; r)}{dt^\beta} \end{array} \right) = A \left( \begin{array}{c} \frac{d^\beta X(t; r)}{dt^\beta} \\ \frac{d^\beta X(t; r)}{dt^\beta} \end{array} \right),
\]
\[
0 < \beta < 1, \quad 0 \leq r \leq 1.
\]

This means that \( \frac{d^\beta}{dt^\beta} \bar{X}(t) \) and \( \bar{X}(t) \) are fuzzy number vectors. From (7), (10), (47) follows:
\[
\frac{d^\beta}{dt^\beta} \bar{X}(t) = A \bar{X}(t), \quad \bar{X}(t) = \left( \begin{array}{c} + \frac{2}{3} E_{\beta} \left( t^\beta \right) + \frac{1}{5} E_{\beta} \left( -2t^\beta \right) \\ \left( \frac{1}{3} - r \right) E_{\beta} \left( t^\beta \right) + 4 \left( \frac{1}{3} - \frac{2}{3} \right) E_{\beta} \left( -2t^\beta \right) \end{array} \right).
\]

Example 2. Consider that the system is with initial value
\[
\left( \begin{array}{c} \frac{d^\beta \bar{x}}{dt^\beta} \\ \frac{d^\beta \bar{y}}{dt^\beta} \\ \frac{d^\beta \bar{z}}{dt^\beta} \end{array} \right) = A \left( \begin{array}{c} \bar{x} \\ \bar{y} \\ \bar{z} \end{array} \right), \quad A = \left( \begin{array}{c} 0 \ 0 \ 0 \\ 0 \ 0 \ 1 \\ 0 \ 5 \ 3 \ 1 \end{array} \right)
\]

where \( 0 < \beta < 1 \).

The eigenvalues of the matrix \( A \) are \( \lambda_1 = 1 \), \( \lambda_2 = -1 + 2i \), and \( \lambda_3 = -1 - 2i \), and their corresponding eigenvectors are \( \bar{x}_1 = [1, 1, 1]^t \), \( \bar{x}_2 = [1, 4, 1]^t \), and \( \bar{x}_3 = [1, -4, 1]^t \), respectively. Therefore, the general solution of the system (48) is
\[
\bar{X}(t) = \bar{c}_1 \left( \begin{array}{c} 1 \\ 1 \\ 1 \end{array} \right) E_{\beta} \left( t^\beta \right) + \bar{c}_2 \left( \begin{array}{c} 1/4 \\ 1/4 \\ 1/4 \end{array} \right) E_{\beta} \left( -2t^\beta \right)
\]

where \( \bar{c}_1 \) and \( \bar{c}_2 \) are fuzzy coefficients. This is obtained by setting the initial values in parametric form, then the following system is obtained:
\[
\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 4 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 4 \end{bmatrix} \begin{bmatrix} \bar{c}_1 \\ \bar{c}_2 \\ \bar{c}_1 \\ \bar{c}_2 \end{bmatrix} = \begin{bmatrix} r + 1 \\ 2 + r \\ 3 - r \\ 5 - r \end{bmatrix}
\]

where \( \bar{c}_1 = (r + 2/3; 7/3 - r) \) and \( \bar{c}_2 = (1/3; 2/3) \) are obtained from the above system and are set in (25). Therefore, the solution of SFFDEs is as follows:

Example 2. Consider that the system is with initial value

\[
\left( \begin{array}{c} \frac{d^\beta \bar{x}}{dt^\beta} \\ \frac{d^\beta \bar{y}}{dt^\beta} \\ \frac{d^\beta \bar{z}}{dt^\beta} \end{array} \right)
\]

where \( 0 < \beta < 1 \).

The eigenvalues of the matrix \( A \) are \( \lambda_1 = 1 \), \( \lambda_2 = -1 + 2i \), and \( \lambda_3 = -1 - 2i \), and their corresponding eigenvectors are \( \xi_1 = [1, 1, 1]^t \), \( \xi_2 = [(−3/25) + (4/25)i, (−1/5) - (2/5)i, 1]^t \), and \( \xi_3 = [(−3/25) - (4/25)i, (−1/5) + (2/5)i, 1]^t \), which are eigenvalues and eigenvectors of matrix \( A \), respectively, and by setting the initial values in parametric form, the following system is obtained:
\[ \tilde{c}_1 = (1.553571428 + 0.071428r, \\
1.696428572 - 0.0714285714r), \]
\[ \tilde{c}_2 = (2.196428572 + 0.1785714286r, \\
2.553571428 - 0.1785714286r), \]
\[ \tilde{c}_3 = (-3.107142858 + 0.9821428567r, \\
-1.142857142 - 0.9821428527r), \]
\[ (53) \]

which are obtained from the above system and are set in (25). Therefore, the solution of SFFDEs is as follows:

\[
\begin{align*}
\Xi(t;r) &= (1.553571428 + 0.071428r) E_\beta(t^\beta) \\
&+ \frac{(2.196428572 + 0.1785714286r)}{2i} \left( \begin{array}{c}
\frac{-3}{25} \\
-\frac{2}{5}
\end{array} \right) \\
&\times \left( E_\beta((-1 + 2i)t^\beta) + E_\beta((-1 - 2i)t^\beta) \right) \\
&+ \frac{(2.553571428 - 0.1785714286r)}{2} \left( \begin{array}{c}
\frac{4}{25} \\
-\frac{2}{5}
\end{array} \right) \\
&\times \left( E_\beta((-1 + 2i)t^\beta) - E_\beta((-1 - 2i)t^\beta) \right) \\
&+ \frac{(3.107142858 + 0.9821428567r)}{2} \left( \begin{array}{c}
\frac{-3}{25} \\
-\frac{2}{5}
\end{array} \right) \\
&\times \left( E_\beta((-1 + 2i)t^\beta) + E_\beta((-1 - 2i)t^\beta) \right) \\
&+ \frac{(3.107142858 + 0.9821428567r)}{2i} \left( \begin{array}{c}
\frac{-3}{25} \\
-\frac{2}{5}
\end{array} \right) \\
&\times \left( E_\beta((-1 + 2i)t^\beta) - E_\beta((-1 - 2i)t^\beta) \right) \\
&\times \left( E_\beta((-1 + 2i)t^\beta) + E_\beta((-1 - 2i)t^\beta) \right).
\end{align*}
\]
\[ (54) \]

**Example 3.** Consider the system

\[
\begin{pmatrix}
\frac{d^\beta \tilde{x}}{dt^\beta} \\
\frac{d^\beta \tilde{y}}{dt^\beta} \\
\frac{d^\beta \tilde{z}}{dt^\beta}
\end{pmatrix} = B \begin{pmatrix}
\tilde{x} \\
\tilde{y} \\
\tilde{z}
\end{pmatrix}, \\
B = \begin{pmatrix}
1 & 1 & 1 \\
2 & 1 & -1 \\
0 & -1 & 1
\end{pmatrix},
\]
\[ (55) \]

where \(0 < \beta < 1\).

The eigenvalues of the matrix \(A\) are \(\lambda_1 = 1\), \(\lambda_2 = \lambda_3 = 2\), and their corresponding eigenvectors are \(\xi_1 = [-3, 4, 2]^T\), \(\xi_2 = \xi_3 = [0, 1, -1]^T\), respectively. Therefore, the general solution of the system (55) is

\[
\Xi(t) = \tilde{c}_1 \begin{pmatrix}
\frac{-3}{4} \\
0 \\
0
\end{pmatrix} E_\beta(-t^\beta) + \tilde{c}_2 \begin{pmatrix}
0 \\
1 \\
1
\end{pmatrix} E_\beta(2t^\beta) \\
+ \tilde{c}_3 \begin{pmatrix}
0 \\
1 \\
-1
\end{pmatrix} t E_\beta(2t^\beta) + \begin{pmatrix}
0 \\
1 \\
1
\end{pmatrix} E_\beta(2t^\beta),
\]
\[ (56) \]

where \(\tilde{c}_1\), \(\tilde{c}_2\), and \(\tilde{c}_3\) are fuzzy coefficients. In particular, if we take \(\beta = 1\), then the general solution (56) can be written as

\[
\Xi(t) = \tilde{c}_1 \begin{pmatrix}
\frac{-3}{4} \\
0 \\
0
\end{pmatrix} \exp(-t) + \tilde{c}_2 \begin{pmatrix}
0 \\
1 \\
1
\end{pmatrix} \exp(2t) \\
+ \tilde{c}_3 \begin{pmatrix}
0 \\
1 \\
-1
\end{pmatrix} t \exp(2t) + \begin{pmatrix}
0 \\
1 \\
1
\end{pmatrix} \exp(2t).
\]
\[ (57) \]
6. Conclusion

In this paper, we investigated an analytical method (eigenvalue-eigenvector) for solving a system of fuzzy fractional differential equation under fuzzy Caputo’s derivative. To this end, we exploited generalized H-differentiability and derived the solutions based on this concept. To illustrate the effectiveness of the proposed method, several examples were solved. From Section 5, one can conclude that the solution of the system of fuzzy fractional differential is a fuzzy number.

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