Note on a Picard-like Method for Caputo Fuzzy Fractional Differential Equations

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Abstract: A Picard-like approach which has been used to solve a class of Volterra integro-differential equations, is extended in this manuscript to solve fuzzy fractional differential equations. Such technique uses quadrature rules and Picard's iterations in the fuzzy context. In spite of this, it is conceived to become a non-recursive scheme, in terms of operational matrices, in the linear regime. Some properties of the method are thoroughly discussed, and some numerical examples are provided in order to illustrate the effectiveness of the approach.

Keywords: fractional differential equations, Caputo fuzzy derivative, quadrature rules, Picard-like method, convergence

1 Introduction

The use of fractional calculus has been incorporated in many branches of mathematics, engineering and science to provide more accurate deterministic descriptions of physical phenomena [1,10,18]. However, in reality, the presence of uncertainties needs to be considered to achieve a higher level of reliability [16,23]. In this manuscript, the notion of *uncertainty* will be interpreted as *fuzziness* [28]. Fuzziness is an important characteristic to be considered in realistic decision processes [9], in regression analysis [14], efficient data classifiers [26] and, in general, in modeling complex problems in science and engineering [11].

With those considerations in mind, Agarwal et al. [3] introduced the notion of Fuzzy Fractional Differential Equations (FFDEs). To that end, fractional operators such as the Riemann-Liouville and the Caputo derivatives have been adapted to the fuzzy scenario. Various works have investigated these models using the Riemann-Liouville derivatives [7,27]. The analytical solution, its existence and its uniqueness for a class of FFDEs with Caputo derivatives were discussed in [6]. However, the aim of the present work is to solve numerically FFDEs employing Caputo-type derivatives.

Fuzzy fractional initial-value problems under fuzzy fractional derivatives of the Caputo type were solved by means of a modified fractional Euler method in [20]. Also, the shifted Legendre operational matrix of fuzzy fractional derivatives was used in [5] to solve numerically FFDEs. More recently, FFDEs with Caputo derivatives were used in [4] to model the kinetics behavior of the diluted acid hydrolysis in oil palm frond. In that work, the FFDEs were solved numerically by means of a fuzzy operational matrix of generalized Laguerre polynomials. Also, FFDEs with Caputo derivatives were solved by the differential transform method (DTM) in [22]. Moreover, homotopy techniques have been used for solving fuzzy fractional diffusion equations with Caputo derivatives [24], and related initial-value problems [2].

In the present work, the following problem governed by a FFDE is considered for each $x \in I = [0, T]$ and $0 < \beta < 1$:

$$\int {}^{c} D^{\beta} \tilde{y}(x) = F(\tilde{y}) + \tilde{g}(x),$$

(1) subject to $\tilde{y}(0) = \tilde{a}_{0}.$

Here, *T* is a positive number, \tilde{a}_0 is a fuzzy number, $\tilde{y}(x)$ is the unknown fuzzy function, $F(\tilde{y})$ a functional form in \tilde{y} , and $\tilde{g}(x)$ is in general a given fuzzy-valued function. Note that the FFDE of (1) may be seen as the fuzzification

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through the Zadeh's extension principle of the same equation but without fuzzy variables [28].

The purpose of this work is to solve (1) following an approach analogous to that used in [25], which was employed to solve a class of integro-differential equations and recently a class of partial differential equations [15]. More precisely, a numerical scheme combining quadrature rules and a Picard-like recursion is studied here from the perspective of fuzzy fractional derivatives. Some properties are formally discussed. Numerical results obtained through this technique are compared against analytical and numerical solutions available in the literature, obtaining satisfactory results.

This manuscript is sectioned as follows. Some basic definitions are provided in Section 2. Section 3 is devoted to introducing the proposed Picard-like method to solve (1). Some analytical results are derived in Section 4. Meanwhile, Section 5 is devoted to show simulations and numerical comparisons against some results available in the literature in order to illustrate the performance of our technique. Finally, this work closes with a section of concluding remarks.

2 Preliminaries

Throughout, the set U will represent a nonempty and fixed (though arbitrary) closed interval of \mathbb{R} . The term *crisp* will mean 'not fuzzy'.

Definition 1.*A* fuzzy number \tilde{u} is defined by a membership function $\mu_u(x) : U \to [0,1]$ which satisfies the following properties:

• \tilde{u} is normal, meaning that $\sup_{x \in U} \mu_u(x) = 1$, • \tilde{u} is convex on U, that is,

$$\mu_u(\alpha x + (1 - \alpha)y) \ge \min(\mu_u(x), \mu_u(y))$$
(2)

for each $x, y \in U$ *and each* $\alpha \in [0, 1]$ *,*

•*ũ* is upper semi-continuous, and

• $[\tilde{u}]_0 = cl(\{x \in U : \mu_u(x) > 0\})$ is compact. Here cl denotes closure in the standard topology of U.

Definition 2.Let $\alpha \ge 0$. The α -cut of the fuzzy number \tilde{u} is the crisp set

$$[\tilde{u}]_{\alpha} = \{ x \in U : \mu_u(x) \ge \alpha \}$$
(3)

when $\alpha > 0$, otherwise $[\tilde{u}]_0$ is given as in Definition 1.

Definition 3.*The* parametric form *of the fuzzy number* \tilde{u} *is a pair of functions* $\underline{u}(\alpha), \overline{u}(\alpha) : U \to \mathbb{R}$ *for each* $\alpha \in [0, 1]$ *, which satisfy the following properties:*

- $1.\underline{u}(\alpha)$ is a bounded, left-continuous, monotonic increasing function,
- $2.\overline{u}(\alpha)$ is a bounded, left-continuous, monotonic decreasing function, and

 $3.\underline{u}(\alpha) \leq \overline{u}(\alpha).$

The notation $[\tilde{u}]_{\alpha} = [\underline{u}(\alpha), \overline{u}(\alpha)]$ is employed if such form is available.

Definition 4.*A fuzzy number* \tilde{u} *is called* triangular *if there exist real numbers* d_C , d_L and d_R , such that

$$[\tilde{u}]_{\alpha} = [d_C + (\alpha - 1)d_L, d_C + (1 - \alpha)d_R], \qquad (4)$$

for each $\alpha \in [0, 1]$. If that is the case then \tilde{u} is identified by the ordered triplet (d_C, d_L, d_R) , and the numbers d_C , d_L and d_R are called the center, the left and the right spreads, respectively.

In the sequel, $\tilde{f}(x)$ will denote a continuous and Lebesgue-integrable fuzzy-valued function on $[a,b] \subset \mathbb{R}$.

Definition 5.*The* fuzzy Riemann-Liouville integral of order β of $\tilde{f}(x)$ is given as

$$J^{\beta}\tilde{f}(x) = \frac{1}{\Gamma(\beta)} \int_0^x (x-s)^{\beta-1} \tilde{f}(s) ds,$$
 (5)

where Γ represents the Gamma function. According to [20], the α -cut representation of $J^{\beta} \tilde{f}(x)$ is provided by

$$[J^{\beta}\tilde{f}(x)]_{\alpha} = [J^{\beta}\underline{f}(x,\alpha), J^{\beta}\overline{f}(x,\alpha)].$$
(6)

Definition 6.Let f(x) be a crisp continuous function, and $0 < \beta < 1$. The Caputo fractional derivative of order β of *f* is defined in [21] as

$$D^{\beta}f(x) = \frac{1}{\Gamma(1-\beta)} \int_{0}^{x} (x-s)^{-\beta} \frac{df}{dx}(s) ds.$$
 (7)

The concept of strong generalized *H*-differentiability studied in [8] was extended in [20] to the context of fractional derivatives. In the latter work, the following definition of differentiability was considered.

Definition 7.Let $0 < \beta < 1$. The fuzzy-valued function $\tilde{f}(x)$ is a Caputo fuzzy fractional differentiable function of order β at $x_0 \in I$ if either

$$\tilde{f}'(x_0) = \lim_{h \to 0^+} \frac{\tilde{f}(x_0 + h) \ominus \tilde{f}(x_0)}{h}$$
$$= \lim_{h \to 0^+} \frac{\tilde{f}(x_0) \ominus \tilde{f}(x_0 - h)}{h}$$
(8)

(9)

$$\tilde{f}'(x_0) = \lim_{h \to 0^+} \frac{\tilde{f}(x_0) \ominus \tilde{f}(x_0)}{-h}$$
$$= \lim \frac{\tilde{f}(x_0 - h) \ominus \tilde{f}(x_0 - h)}{-h}$$

and

or

$${}^{c}D^{\beta}\tilde{f}(x_{0}) = \frac{1}{\Gamma(1-\beta)} \int_{0}^{x_{0}} \tilde{f}'(s)(x_{0}-s)^{-\beta} ds, \quad (10)$$

where ${}^{c}D^{\beta}\tilde{f}(x_{0})$ denotes the Caputo fuzzy fractional derivative of \tilde{f} at x_0 .

Let $0 \le \alpha \le 1$. Following [20], the α -cut form of the Caputo fuzzy fractional derivative is

$$[{}^{c}D^{\beta}\tilde{f}(x_{0})]_{\alpha} = \begin{cases} [{}^{c}D^{\beta}\underline{f}(x_{0},\alpha), {}^{c}D^{\beta}\overline{f}(x_{0},\alpha)], \text{ for } (8), \\ \\ [{}^{c}D^{\beta}\overline{f}(x_{0},\alpha), {}^{c}D^{\beta}\underline{f}(x_{0},\alpha)], \text{ for } (9). \end{cases}$$
(11)

For briefness, in the following we will refer to the first equation of (11) to illustrate our numerical approach.

Let $\tilde{y}(x)$ be a fuzzy function. Using [12], one may readily check that $[\tilde{y}(x)]_{\alpha} = [y(x,\alpha), \overline{y}(x,\alpha)]$. As shown in [20], problem (1) is equivalent to

$$\begin{cases} {}^{c}D^{\beta}\mathbf{Y}(x,\alpha) = \mathbf{F}(\mathbf{Y}(x,\alpha),\alpha) + \mathbf{G}(x,\alpha), \\ \mathbf{Y}(0,\alpha) = \mathbf{A}(\alpha), \end{cases}$$
(12)

where

$$\mathbf{F}(\mathbf{Y}(x,\alpha),\alpha)^{T} = (\underline{F}(\underline{y}(x,\alpha),\overline{y}(x,\alpha),\alpha), \\ \overline{F}(\underline{y}(x,\alpha),\overline{y}(x,\alpha),\alpha)),$$
(13)
$$\mathbf{Y}(x,\alpha)^{T} = (y(x,\alpha),\overline{y}(x,\alpha)),$$
(14)

$$Y(x,\alpha)^{T} = (\underline{y}(x,\alpha), \overline{y}(x,\alpha)),$$
 (14)

$$\mathbf{G}(x,\alpha)^T = (\underline{g}(x,\alpha), \overline{g}(x,\alpha)) \tag{15}$$

3 Methodology

In this section, the approach employed in [25] is extended in order to solve the FFDE of (12). Letting $\gamma = \frac{1}{\Gamma(\beta)}$, applying the operator J^{β} to both sides of that equation and arguing as in [20], one obtains

$$\mathbf{Y}(x,\alpha) = \mathbf{A}(\alpha) + \gamma \int_0^x [\mathbf{F}(\mathbf{Y}(x,\alpha),\alpha) + \mathbf{G}(x,\alpha)](x-s)^{\beta-1} ds.$$
(16)

Let N be a positive integer and consider a (not necessarily uniform) partition of [0,T] consisting of N points and partition norm equal to h, say,

$$0 = x_1 < x_2 < \dots < x_N = T. \tag{17}$$

Then (16) may be rewritten as

$$\mathbf{Y}(x,\alpha) = \mathbf{A}(\alpha) + \gamma \mathbf{C}(x) \left[\mathbf{F}(\mathbf{Y}(\alpha), \alpha) + \mathbf{Q}(\alpha) \right], \quad (18)$$

in which

$$\mathbf{F}(\mathbf{Y}(\alpha), \alpha)^{T} = (\underline{F}(\mathbf{Y}_{1}, \alpha), \dots, \underline{F}(\mathbf{Y}_{N}, \alpha), \\ \overline{F}(\mathbf{Y}_{1}, \alpha), \dots, \overline{F}(\mathbf{Y}_{N}, \alpha)),$$
(19)
$$\mathbf{Y}^{T} = ((\mathbf{y}, \alpha), \mathbf{y}, \mathbf{y}) = (\mathbf{y}, \alpha)$$
(20)

$$\mathbf{Y}_{i} = (\underline{y}(x_{i}, \alpha), y(x_{i}, \alpha)), \qquad (20)$$
$$\mathbf{A}(\alpha)^{T} = (\underline{a}_{0}(\alpha), \overline{a}_{0}(\alpha)), \qquad (21)$$

$$\mathbf{Q}^{T}(\boldsymbol{\alpha}) = \left(\underline{g}(x_{1},\boldsymbol{\alpha}),\ldots,\underline{g}(x_{N},\boldsymbol{\alpha}), \frac{\overline{g}(x_{1},\boldsymbol{\alpha}),\ldots,\overline{g}(x_{N},\boldsymbol{\alpha})}{\overline{g}(x_{1},\boldsymbol{\alpha}),\ldots,\overline{g}(x_{N},\boldsymbol{\alpha})}\right),$$
(22)

and

$$\mathbf{C}(x) = \begin{pmatrix} C_1^{\beta}(x) \dots C_N^{\beta}(x) & 0 \dots & 0\\ 0 & \dots & 0 & C_1^{\beta}(x) \dots & C_N^{\beta}(x) \end{pmatrix}.$$
 (23)

Here, for each $i = 1, 2, \ldots, N$,

$$C_i^{\beta}(x) = \int_0^x l_i(s)(x-s)^{\beta-1} ds,$$
 (24)

with $l_i(s)$ being the *i*th Lagrange polynomial. By successive approximations, the solution $\mathbf{Y}(x, \alpha)$ is

$$\mathbf{Y}(x,\alpha) = \sum_{k=0}^{\infty} \mathbf{Y}_k(x,\alpha),$$
(25)

where each $\mathbf{Y}_{k}^{T}(x, \alpha)$ has to be determined recursively using the formulas

$$\mathbf{Y}_0(x,\alpha) = \mathbf{A}(\alpha) + \gamma \mathbf{C}(x)\mathbf{Q}(\alpha), \qquad (26)$$

$$\mathbf{Y}_{k+1}(x,\alpha) = \gamma \mathbf{C}(x) \mathbf{F}(\mathbf{Y}_k(\alpha),\alpha).$$
(27)

On the other hand, truncating (25) at the *p*th term yields

$$\mathbf{Y}^{[p]}(x,\alpha) = \mathbf{Y}_0(x,\alpha) + \gamma \mathbf{C}(x) \sum_{k=1}^{p-1} \mathbf{F}(\mathbf{Y}_k(\alpha),\alpha).$$
(28)

Note that (27) reduces to $\mathbf{Y}_{k+1}(x, \alpha) = \gamma_0 \mathbf{C}(x) \mathbf{Y}_k(\alpha)$ when $F(\tilde{y})$ is a linear function, where

$$\mathbf{Y}_{k}(\boldsymbol{\alpha})^{T} = (\underbrace{\mathbf{y}_{k}(x_{1},\boldsymbol{\alpha}),\ldots,\underbrace{\mathbf{y}_{k}(x_{N},\boldsymbol{\alpha}),}_{\overline{\mathbf{y}_{k}}(x_{1},\boldsymbol{\alpha}),\ldots,\overline{\mathbf{y}_{k}}(x_{N},\boldsymbol{\alpha}))}_{(29)}$$

and γ_0 is a constant. Let **D** be the matrix whose entries are $D_{ij} = C_j(x_i)$. Since $\mathbf{Y}_k(\alpha) = \gamma_0 \mathbf{D} \mathbf{Y}_{k-1}(\alpha) = \gamma_0^k \mathbf{D}^k \mathbf{Y}_0(\alpha)$, the truncation of (25) after p terms becomes

$$\mathbf{Y}^{[p]}(x,\alpha) = \mathbf{Y}_0(x,\alpha) + \gamma_0 \mathbf{C}(x) \sum_{k=0}^{p-1} \gamma_0^k \mathbf{D}^k \mathbf{Y}_0(\alpha).$$
(30)

4 Properties

The present section summarizes the main properties of the method of Section 3. For the remainder, the spectral radius of the matrix **D** will be represented by $\rho(\mathbf{D})$, and the symbol I will denote the identity matrix of size 2N.

Lemma 1. If $F(\tilde{y})$ is linear and $\rho(\mathbf{D}) < \frac{1}{|\mathbf{y}_0|}$ then the solution of (25) is approximated by

$$\mathbf{Y}(x,\alpha) = \mathbf{Y}_0(x,\alpha) + \gamma_0 \mathbf{C}(x)(\mathbf{I} - \gamma_0 \mathbf{D})^{-1} \mathbf{Y}_0(\alpha).$$
(31)

Proof. The proof follows as that of Theorem 1 in [25], considering a geometric series approximation of (30) with matrices and taking then the limit when $p \to \infty$. \Box

Table 1: Values of $\underline{y}(1,\alpha)$ and $\overline{y}(1,\alpha)$, for various values of α and several methods. The FFDE and the parameters used are described in Example 1.

		VALU	JES OF $\underline{y}(1)$	$(, \alpha)$		
α	Present 7 GCL	Present 15 GCL	Method of [5]	Method of [20]	Method of [22]	Exact
0.0	0.196543	0.196554	0.1962	0.183	0.1967	0.1966
0.1	0.216197	0.216209	0.2159	0.201	0.2164	0.2162
0.2	0.235851	0.235864	0.2355	0.219	0.2360	0.2359
0.3	0.255505	0.255520	0.2551	0.238	0.2557	0.2555
0.4	0.275160	0.275175	0.2747	0.256	0.2754	0.2752
0.5	0.294814	0.294831	0.2944	0.274	.2950	0.2948
0.6	0.314468	0.314486	0.3140	0.293	0.3147	0.3145
0.7	0.334122	0.334141	0.3336	0.311	0.3344	0.3341
0.8	0.353777	0.353797	0.3532	0.329	.3540	0.3538
0.9	0.373431	0.373452	0.3729	0.348	0.3737	0.3735
1.0	0.393085	0.393107	0.3925	0.366	0.3934	0.3931
VALUES OF $\overline{y}(1, \alpha)$						
		VALU	JES OF $\overline{y}(1)$	Ι,α)		
α	Present 7 GCL	VALU Present 15 GCL	JES OF y(1 Method of [5]	.(α) Method of [20]	Method of [22]	Exact
α 0.0		Present	Method	Method		Exact 0.5897
	7 GCL	Present 15 GCL	Method of [5]	Method of [20]	of [22]	
0.0	7 GCL 0.589627	Present 15 GCL 0.589661	Method of [5] 0.5887	Method of [20] 0.549	of [22]	0.5897
0.0 0.1	7 GCL 0.589627 0.569973	Present 15 GCL 0.589661 0.570005	Method of [5] 0.5887 0.5691	Method of [20] 0.549 0.534	of [22] 0.5901 0.5704	0.5897
0.0 0.1 0.2	7 GCL 0.589627 0.569973 0.550319	Present 15 GCL 0.589661 0.570005 0.550350	Method of [5] 0.5887 0.5691 0.5495	Method of [20] 0.549 0.534 0.513	of [22] 0.5901 0.5704 0.5507	0.5897 0.5700 0.5504
0.0 0.1 0.2 0.3	7 GCL 0.589627 0.569973 0.550319 0.530664	Present 15 GCL 0.589661 0.570005 0.550350 0.530694	Method of [5] 0.5887 0.5691 0.5495 0.5298	Method of [20] 0.549 0.534 0.513 0.494	of [22] 0.5901 0.5704 0.5507 0.5311	0.5897 0.5700 0.5504 0.5307
0.0 0.1 0.2 0.3 0.4	7 GCL 0.589627 0.569973 0.550319 0.530664 0.511010	Present 15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039	Method of [5] 0.5887 0.5691 0.5495 0.5298 0.5102	Method of [20] 0.549 0.534 0.513 0.494 0.476	of [22] 0.5901 0.5704 0.5507 0.5311 0.5114	0.5897 0.5700 0.5504 0.5307 0.5110
0.0 0.1 0.2 0.3 0.4 0.5	7 GCL 0.589627 0.569973 0.550319 0.530664 0.511010 0.491356	Present 15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384	Method of [5] 0.5887 0.5691 0.5495 0.5298 0.5102 0.4906	Method of [20] 0.549 0.534 0.513 0.494 0.476 0.458	of [22] 0.5901 0.5704 0.5507 0.5311 0.5114 0.4917	0.5897 0.5700 0.5504 0.5307 0.5110 0.4914
0.0 0.1 0.2 0.3 0.4 0.5 0.6	7 GCL 0.589627 0.569973 0.550319 0.530664 0.511010 0.491356 0.471702	Present 15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384 0.471728	Method of [5] 0.5887 0.5691 0.5495 0.5298 0.5102 0.4906 0.4710	Method of [20] 0.549 0.534 0.513 0.494 0.476 0.458 0.439	of [22] 0.5901 0.5704 0.5507 0.5311 0.5114 0.4917 0.4721	0.5897 0.5700 0.5504 0.5307 0.5110 0.4914 0.4717
0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7	7 GCL 0.589627 0.569973 0.550319 0.530664 0.511010 0.491356 0.471702 0.452047	Present 15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384 0.471728 0.47273	Method of [5] 0.5887 0.5691 0.5495 0.5298 0.5102 0.4906 0.4710 0.4514	Method of [20] 0.549 0.534 0.513 0.494 0.476 0.458 0.439 0.421	of [22] 0.5901 0.5704 0.5507 0.5311 0.5114 0.4917 0.4721 0.4524	0.5897 0.5700 0.5504 0.5307 0.5110 0.4914 0.4717 0.4521

The generalized Taylor's formula under the Caputo fractional derivative was established in [20]. That formula will be a useful tool in the next results. In the sequel, the exact solution of the FFDE in (12) will be represented by $\mathbf{Y}(x,\alpha) = [\underline{y}(x,t,\alpha), \overline{y}(x,\alpha)]$. The norm of L_{∞} will be employed next in view of the continuity of the functions involved (see [13]). So, in what follows,

$$e_{\infty}(x_j, \alpha) = \left\| \mathbf{Y}^{[p]}(x_j, \alpha) - \mathbf{Y}(x_j, \alpha) \right\|_{\infty}.$$
 (32)

Theorem 1.Let $0 < \beta < 1$ and $0 \le \alpha \le 1$. Let $\underline{F}(\mathbf{Y}(x,\alpha),\alpha)$, $\overline{F}(\mathbf{Y}(x,\alpha),\alpha)$, $\underline{g}(\alpha)$ and $\overline{g}(\alpha)$ be continuous 2β -times differentiable crisp functions with respect to x. The following bound holds for any 0 < h < 1:

$$e_{\infty}(x_{j}, \alpha) \leq \left\| \gamma \mathbf{C}((j-1)h) \sum_{k=1}^{p-1} \mathbf{F}(\mathbf{Y}_{k}(\alpha), \alpha) - \frac{(j-1)^{\beta}h^{\beta}}{\Gamma(\beta+1)} \left[\mathbf{F}(\mathbf{Y}(0, \alpha), \alpha) + \mathbf{G}(0, \alpha) \right] \right\|_{\infty} + \mathcal{O}(h^{2\beta}),$$
(33)

Proof. The truncation of the generalized Taylor expansion of $\mathbf{Y}(x, \alpha)$ around x = 0 at the first two terms is

$$\mathbf{Y}(x,\alpha) = \mathbf{A}(\alpha) + \frac{x^{\beta}}{\Gamma(\beta+1)} D^{\beta} \mathbf{Y}(0,\alpha) + D^{2\beta} \mathbf{Y}(\chi,\alpha) \frac{x^{2\beta}}{\Gamma(2\beta+1)},$$
(34)



Fig. 1: Absolute errors with respect to the exact solution for the calculated values of (a) $\underline{y}(1,\alpha)$ and (b) $\overline{y}(1,\alpha)$ for $\alpha \in [0,1]$, using the present method (solid), the method used in [5] (dashed), and that employed in [20] (dotted). The FFDE and the parameters used are described in Example 1.

for some $\chi \in [0, T]$. Using (12), the following inequality results for each node x_i and each $\alpha \in [0, 1]$:

$$e_{\infty}(x_{j},\alpha) \leq \left\| \gamma \mathbf{C}(x_{j}) \sum_{k=1}^{p-1} \mathbf{F}(\mathbf{Y}_{k}(\alpha),\alpha) - \frac{x_{j}^{\beta}}{\Gamma(\beta+1)} [\mathbf{F}(\mathbf{Y}(0,\alpha),\alpha) + \mathbf{G}(0,\alpha)] \right\|_{\infty} + \mathcal{O}(x_{j}^{2\beta}),$$
(35)

For each j = 1, ..., N, note that $x_j = x_1 + (j-1)h$ or $x_j \le x_1 + (j-1)h$ in the case of uniform and nonuniform partitions, respectively. The conclusion of the theorem readily follows now. \Box

Theorem 1 implies that the error tends to zero when $h \rightarrow 0$, and that the approximation error has order h^{β} .

Table 2: Values and errors of $\underline{y}(1,\alpha)$ and $\overline{y}(1,\alpha)$, for various values of α and several methods. The FFDE and the parameters used are described in Example 3.

		VALUI	ES OF $\underline{y}(1, \alpha)$			
α	Present 11 GCL	Exact	Error	Error with [5]		
0.0	0.720607	0.7206	7.60×10^{-6}	1.44×10^{-4}		
0.1	0.722132	0.7221	3.25×10^{-5}	1.44×10^{-4}		
0.2	0.723657	0.7237	4.26×10^{-5}	1.45×10^{-4}		
0.3	0.725182	0.7252	1.76×10^{-5}	1.45×10^{-4}		
0.4	0.726707	0.7267	7.28×10^{-6}	1.46×10^{-4}		
0.5	0.728232	0.7282	3.22×10^{-5}	1.47×10^{-4}		
0.6	0.729757	0.7298	4.29×10^{-5}	1.47×10^{-4}		
0.7	0.731282	0.7313	1.18×10^{-5}	1.48×10^{-4}		
0.8	0.732807	0.7328	6.96×10^{-6}	1.49×10^{-4}		
0.9	0.734332	0.7343	3.19×10^{-5}	1.49×10^{-4}		
1.0	0.735856	0.7359	4.32×10^{-5}	1.50×10^{-4}		
VALUES OF $\overline{y}(1, \alpha)$						
		ALU	1501 y(1, a)			
α	Present 11 GCL	Exact	Error	Error with [<mark>5</mark>]		
α 0.0			,	with [5]		
	11 GCL	Exact	Error			
0.0	11 GCL 0.739669	Exact 0.7397	Error 3.09×10 ⁻⁵	with [5]		
0.0 0.1	11 GCL 0.739669 0.739288	Exact 0.7397 0.7393	Error 3.09×10^{-5} 1.21×10^{-5}	with [5] 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴		
0.0 0.1 0.2	11 GCL 0.739669 0.739288 0.738907	Exact 0.7397 0.7393 0.7389	Error 3.09×10^{-5} 1.21×10^{-5} 6.65×10^{-6}	with [5] 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴		
0.0 0.1 0.2 0.3	11 GCL 0.739669 0.739288 0.738907 0.738525	Exact 0.7397 0.7393 0.7389 0.7385	Error 3.09×10^{-5} 1.21×10^{-5} 6.65×10^{-6} 2.54×10^{-5}	with [5] 1.51×10^{-4} 1.51×10^{-4} 1.51×10^{-4} 1.51×10^{-4} 1.46×10^{-4}		
0.0 0.1 0.2 0.3 0.4	11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144	Exact 0.7397 0.7393 0.7389 0.7385 0.7381	Error 3.09×10^{-5} 1.21×10^{-5} 6.65×10^{-6} 2.54×10^{-5} 4.42×10^{-5}	with [5] 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴		
0.0 0.1 0.2 0.3 0.4 0.5	11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763	Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378	$\frac{\text{Error}}{1.21 \times 10^{-5}}$ 6.65×10^{-6} 2.54×10^{-5} 4.42×10^{-5} 3.70×10^{-5}	with [5] 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.51×10 ⁻⁴ 1.46×10 ⁻⁴ 1.47×10 ⁻⁴		
0.0 0.1 0.2 0.3 0.4 0.5 0.6	11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763 0.737382	Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378 0.7374	$\begin{array}{c} \textbf{Error} \\ \hline 3.09 \times 10^{-5} \\ 1.21 \times 10^{-5} \\ 6.65 \times 10^{-6} \\ 2.54 \times 10^{-5} \\ 4.42 \times 10^{-5} \\ 3.70 \times 10^{-5} \\ 1.83 \times 10^{-5} \end{array}$	$\begin{array}{c} \text{with [5]} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.46 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.48 \times 10^{-4} \\ 1.49 \times 10^{-4} \end{array}$		
0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7	11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763 0.737382 0.737000	Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378 0.7374 0.7370	$\frac{\text{Error}}{3.09 \times 10^{-5}}$ 1.21×10^{-5} 6.65×10^{-6} 2.54×10^{-5} 4.42×10^{-5} 3.70×10^{-5} 1.83×10^{-5} 4.96×10^{-7}	with [5] 1.51×10^{-4} 1.51×10^{-4} 1.51×10^{-4} 1.51×10^{-4} 1.51×10^{-4} 1.46×10^{-4} 1.47×10^{-4} 1.48×10^{-4}		

5 Simulations

The present section shows comparisons of known exact solutions of (1) against approximations obtained via (28) when (31) holds. Throughout, Gauss-Tchebyshev-Lobatto (GCL) grid points [17] are used, and all fuzzy numbers are triangular with T = 1. The exact solution of the first example was taken from [20], and the solutions of the last two are borrowed from [5].

*Example 1.*Let $\beta = 0.75$, $d_c = 1$, $d_L = 0.5$, $d_R = 0.5$, $F(\tilde{y}) = -\tilde{y}(x)$, and fix $g \equiv 0$. Table 1 shows the solutions y and \overline{y} at x = 1 using the method proposed in the present paper (with different number of GCL points), and other methods available in the literature. In turn, Figure 1 shows the absolute errors of the present method and the methods of the literature with respect to the exact solutions. Clearly, the present methodology yields more accurate results. \Box

*Example 2.*Let $\beta = 0.75$, $d_c = 0$, $d_L = 1$, $d_R = 1$, and consider

$$F(\tilde{y}(x)) = -\tilde{y}(x), \tag{36}$$

$$g(x) = \frac{2x^{2-\beta}}{\Gamma(3-\beta)} - \frac{x^{1-\beta}}{\Gamma(2-\beta)} + x^2 - x.$$
 (37)

The method introduced in this manuscript is used to calculate $\overline{y}(1, \alpha)$ for various values of α using 20 GCL



Fig. 2: Absolute errors with respect to the exact solution for the calculated values of $\overline{y}(1, \alpha)$ for $\alpha \in [0, 1]$, using the present method (dashed) and the method in [5] (dotted). The FFDE and the parameters used are described in Example 2.

points. Figure 2 shows the absolute error using the present method (dashed line) and the technique reported in [5] (dotted line). The graph shows that the approximation reported in this manuscript produces more accurate results for values of $\alpha \in [0.05, 0.6]$. In any case, the order of the error is 10^{-5} . The simulations corresponding to <u>y</u> have been omitted in view that they are qualitatively similar to those of \overline{y} . \Box

*Example 3.*Fix $\beta = 0.85$, $d_c = 1$, $d_L = 0.04$, $d_R = 0.01$. Let $F(\tilde{y}(x)) = -\tilde{y}(x)$ and $g(x) = \sin x$. Table 2 provides the approximate solution \underline{y} and \overline{y} at x = 1 for several values of α , calculated using the present method. The results show that the absolute error using the present technique with 11 GCL points is substantially smaller than the error obtained using the technique reported in [5]. \Box

6 Conclusions

A Picard-like numerical scheme, which was previously employed to solve a class of Volterra integro-differential equations, has been extended in this manuscript to solve problems involving Caputo fuzzy fractional differential equations.

The proposed approach presents two main advantages:

- -it is able to reproduce exactly the initial condition;
- -in the linear regime, it becomes a non-recursive scheme in terms of known operational matrices.

For the general nonlinear scenario, some formal considerations on the error have been discussed. Some numerical examples illustrate the effectiveness of the proposed approach. However, ongoing research is focusing on extending the method towards piecewise solutions for certain classes of problems, similarly to the approach proposed in [19] for fuzzy partial differential equations.

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