Applied Mathematics & Information Sciences An International Journal

281

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

Stefania Tomasiello<sup>1,\*</sup> and Jorge E. Macías-Díaz<sup>2</sup>

Received: 10 Feb. 2016, Revised: 18 Dec. 2016, Accepted: 21 Dec. 2016 Published online: 1 Jan. 2017

**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  $0 < \beta < 1$ :

$$\int_{a}^{c} D^{\beta} \tilde{y}(x) = F(\tilde{y}) + \tilde{g}(x), \qquad (1)$$
  
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

\* Corresponding author e-mail: stomasiello@unisa.it; stefania.tomasiello@gmail.com

<sup>&</sup>lt;sup>1</sup> Consorzio di Ricerca Sistemi ad Agenti, Dipartimento di Ingegneria dell'Informazione e Elettrica, Matematica Applicata, Università degli Studi di Salerno, Via Giovanni Paolo II 132, 84084 Fisciano, Salerno, Italy

<sup>&</sup>lt;sup>2</sup> Departamento de Matemáticas y Física, Universidad Autónoma de Aguascalientes, Avenida Universidad 940, Ciudad Universitaria, Aguascalientes, Ags. 20131, Mexico

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  $\alpha$ -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  $\beta$  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  $\Gamma$  represents the Gamma function. According to [20], the  $\alpha$ -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  $\beta$  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  $\beta$  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)

-h

(9)

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

 $h \rightarrow 0^+$ 

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  $\alpha$ -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  $J^{\beta}$  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

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

and  $\gamma_0$  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  $\alpha$  and several methods. The FFDE and the parameters used are described in Example 1.

	<b>VALUES OF</b> $\underline{y}(1, \alpha)$						
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	
		VALU	<b>JES OF</b> $\overline{y}(1)$	,α)			
α.	Present	Present	Method	Method	M-4h-J	•	
	7 GCL	15 GCL	of [ <mark>5</mark> ]	of [20]	of [22]	Exact	
0.0	7 GCL 0.589627	15 GCL 0.589661	of [5]	of [20]	of [22]	Exact 0.5897	
0.0	7 GCL 0.589627 0.569973	15 GCL 0.589661 0.570005	of [5] 0.5887 0.5691	0.549 0.534	0.5901 0.5704	Exact 0.5897 0.5700	
0.0 0.1 0.2	7 GCL 0.589627 0.569973 0.550319	15 GCL 0.589661 0.570005 0.550350	of [5] 0.5887 0.5691 0.5495	0.549 0.534 0.513	0.5901 0.5704 0.5507	Exact 0.5897 0.5700 0.5504	
0.0 0.1 0.2 0.3	7 GCL 0.589627 0.569973 0.550319 0.530664	15 GCL 0.589661 0.570005 0.550350 0.530694	of [5] 0.5887 0.5691 0.5495 0.5298	0.549 0.534 0.513 0.494	0.5901 0.5704 0.5507 0.5311	Exact 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	15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039	of [5] 0.5887 0.5691 0.5495 0.5298 0.5102	0.549 0.534 0.513 0.494 0.476	0.5901 0.5704 0.5507 0.5311 0.5114	Exact 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	15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384	of [5] 0.5887 0.5691 0.5495 0.5298 0.5102 0.4906	of [20] 0.549 0.534 0.513 0.494 0.476 0.458	0.5901 0.5704 0.5507 0.5311 0.5114 0.4917	Exact 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	15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384 0.471728	of [5] 0.5887 0.5691 0.5495 0.5298 0.5102 0.4906 0.4710	of [20] 0.549 0.534 0.513 0.494 0.476 0.458 0.439	Method         of [22]           0.5901         0.5704           0.5507         0.5311           0.5114         0.4917           0.4721         0.4721	Exact 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	15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384 0.471728 0.452073	of [5] 0.5887 0.5691 0.5495 0.5298 0.5102 0.4906 0.4710 0.4514	of [20] 0.549 0.534 0.513 0.494 0.476 0.458 0.439 0.421	Method           of [22]           0.5901           0.5704           0.5507           0.5311           0.5114           0.4917           0.4721           0.4524	Exact 0.5897 0.5700 0.5504 0.5307 0.5110 0.4914 0.4717 0.4521	
0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8	<b>7 GCL</b> 0.589627 0.569973 0.550319 0.530664 0.511010 0.491356 0.471702 0.452047 0.432393	15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384 0.471728 0.452073 0.452073 0.432418	of [5] 0.5887 0.5691 0.5298 0.5102 0.4906 0.4710 0.4514 0.4317	of [20] 0.549 0.534 0.513 0.494 0.476 0.458 0.439 0.421 0.403	Method           of [22]           0.5901           0.5704           0.5507           0.5311           0.5114           0.4917           0.4721           0.4524           0.4327	Exact 0.5897 0.5700 0.5504 0.5307 0.5110 0.4914 0.4717 0.4521 0.4324	
0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9	7 GCL 0.589627 0.569973 0.550319 0.530664 0.511010 0.491356 0.471702 0.452047 0.432393 0.412739	15 GCL 0.589661 0.570005 0.550350 0.530694 0.511039 0.491384 0.471728 0.452073 0.432418 0.412762	of [5] 0.5887 0.5691 0.5495 0.5298 0.5102 0.4906 0.4710 0.4514 0.4317 0.4121	of [20] 0.549 0.534 0.513 0.494 0.476 0.458 0.439 0.421 0.403 0.384	Method           of [22]           0.5901           0.5704           0.5507           0.5311           0.5114           0.4917           0.4721           0.4524           0.4327           0.4130	Exact 0.5897 0.5700 0.5504 0.5307 0.5110 0.4914 0.4717 0.4521 0.4324 0.4128	

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  $L_{\infty}$  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  $2\beta$ -times differentiable crisp functions with respect to x. The following bound holds for any 0 < h < 1:

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

$$(33)$$

*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  $h^{\beta}$ .

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

	<b>VALUES OF</b> $\underline{y}(1, \alpha)$							
α	Present 11 GCL	Exact	Error	Error with [5]				
0.0	0.720607	0.7206	$7.60 \times 10^{-6}$	$1.44 \times 10^{-4}$				
0.1	0.722132	0.7221	$3.25 \times 10^{-5}$	$1.44 \times 10^{-4}$				
0.2	0.723657	0.7237	$4.26 \times 10^{-5}$	$1.45 \times 10^{-4}$				
0.3	0.725182	0.7252	$1.76 \times 10^{-5}$	$1.45 \times 10^{-4}$				
0.4	0.726707	0.7267	$7.28 \times 10^{-6}$	$1.46 \times 10^{-4}$				
0.5	0.728232	0.7282	$3.22 \times 10^{-5}$	$1.47 \times 10^{-4}$				
0.6	0.729757	0.7298	$4.29 \times 10^{-5}$	$1.47 \times 10^{-4}$				
0.7	0.731282	0.7313	$1.18 \times 10^{-5}$	$1.48 \times 10^{-4}$				
0.8	0.732807	0.7328	$6.96 \times 10^{-6}$	$1.49 \times 10^{-4}$				
0.9	0.734332	0.7343	$3.19 \times 10^{-5}$	$1.49 \times 10^{-4}$				
1.0	0.735856	0.7359	$4.32 \times 10^{-5}$	$1.50 \times 10^{-4}$				
VALUES OF $\overline{y}(1, \alpha)$								
		VALU	ES OF $\overline{y}(1, \alpha)$					
α	Present 11 GCL	VALUI Exact	<b>ES OF</b> $\overline{y}(1, \alpha)$ <b>Error</b>	Error with [5]				
α	<b>Present</b> <b>11 GCL</b> 0.739669	<b>VALUI</b> <b>Exact</b> 0.7397	<b>ES OF</b> $\overline{y}(1, \alpha)$ <b>Error</b> $3.09 \times 10^{-5}$	Error with [5]				
α 0.0 0.1	Present 11 GCL 0.739669 0.739288	VALUI Exact 0.7397 0.7393	<b>ES OF</b> $\overline{y}(1, \alpha)$ <b>Error</b> $3.09 \times 10^{-5}$ $1.21 \times 10^{-5}$	Error with [5] 1.51×10 <sup>-4</sup> 1.51×10 <sup>-4</sup>				
α 0.0 0.1 0.2	Present 11 GCL 0.739669 0.739288 0.738907	VALUI Exact 0.7397 0.7393 0.7389	ES OF $\overline{y}(1, \alpha)$ Error $3.09 \times 10^{-5}$ $1.21 \times 10^{-5}$ $6.65 \times 10^{-6}$	Error with [5] 1.51×10 <sup>-4</sup> 1.51×10 <sup>-4</sup> 1.51×10 <sup>-4</sup>				
α 0.0 0.1 0.2 0.3	Present 11 GCL 0.739669 0.739288 0.738907 0.738525	VALUI Exact 0.7397 0.7393 0.7389 0.7389	ES OF $\overline{y}(1, \alpha)$ Error $3.09 \times 10^{-5}$ $1.21 \times 10^{-5}$ $6.65 \times 10^{-6}$ $2.54 \times 10^{-5}$	Error with [5] $1.51 \times 10^{-4}$ $1.51 \times 10^{-4}$ $1.51 \times 10^{-4}$ $1.51 \times 10^{-4}$				
α 0.0 0.1 0.2 0.3 0.4	Present 11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144	VALUI Exact 0.7397 0.7393 0.7389 0.7385 0.7381	ES OF $\overline{y}(1,\alpha)$ Error $3.09 \times 10^{-5}$ $1.21 \times 10^{-5}$ $6.65 \times 10^{-6}$ $2.54 \times 10^{-5}$ $4.42 \times 10^{-5}$	$\frac{\text{Error}}{\text{with [5]}}$ $\frac{1.51 \times 10^{-4}}{1.51 \times 10^{-4}}$ $1.51 \times 10^{-4}$ $1.51 \times 10^{-4}$ $1.46 \times 10^{-4}$				
α 0.0 0.1 0.2 0.3 0.4 0.5	Present 11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763	VALUI Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378	$\frac{\text{Error}}{1.21 \times 10^{-5}}$ $\frac{3.09 \times 10^{-5}}{1.21 \times 10^{-5}}$ $\frac{6.65 \times 10^{-6}}{2.54 \times 10^{-5}}$ $\frac{4.42 \times 10^{-5}}{3.70 \times 10^{-5}}$	$\frac{\text{Error}}{\text{with [5]}}$ 1.51×10 <sup>-4</sup> 1.51×10 <sup>-4</sup> 1.51×10 <sup>-4</sup> 1.51×10 <sup>-4</sup> 1.46×10 <sup>-4</sup> 1.47×10 <sup>-4</sup>				
α 0.0 0.1 0.2 0.3 0.4 0.5 0.6	Present 11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763 0.737382	VALUI Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378 0.7374	$\frac{\text{Error}}{1.21 \times 10^{-5}}$ $\frac{3.09 \times 10^{-5}}{1.21 \times 10^{-5}}$ $\frac{3.65 \times 10^{-6}}{2.54 \times 10^{-5}}$ $\frac{4.42 \times 10^{-5}}{3.70 \times 10^{-5}}$ $\frac{1.83 \times 10^{-5}}{1.83 \times 10^{-5}}$	$\begin{array}{c} \textbf{Error} \\ \textbf{with [5]} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.46 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.47 \times 10^{-4} \end{array}$				
α 0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7	Present 11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763 0.737782 0.737000	VALUI Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378 0.7374 0.7370	$\frac{\text{Error}}{3.09 \times 10^{-5}}$ $\frac{3.09 \times 10^{-5}}{1.21 \times 10^{-5}}$ $6.65 \times 10^{-6}$ $2.54 \times 10^{-5}$ $4.42 \times 10^{-5}$ $3.70 \times 10^{-5}$ $1.83 \times 10^{-5}$ $4.96 \times 10^{-7}$	$\begin{array}{c} \textbf{Error} \\ \textbf{with [5]} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.46 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.48 \times 10^{-4} \end{array}$				
α 0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8	Present 11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763 0.737382 0.737000 0.736619	VALUI Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378 0.7374 0.7370 0.7366	$\frac{\text{Error}}{3.09 \times 10^{-5}}$ $\frac{3.09 \times 10^{-5}}{1.21 \times 10^{-5}}$ $6.65 \times 10^{-6}$ $2.54 \times 10^{-5}$ $4.42 \times 10^{-5}$ $1.83 \times 10^{-5}$ $4.96 \times 10^{-7}$ $1.92 \times 10^{-5}$	$\begin{array}{c} \textbf{Error} \\ \textbf{with [5]} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.48 \times 10^{-4} \\ 1.49 \times 10^{-4} \end{array}$				
α 0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9	Present 11 GCL 0.739669 0.739288 0.738907 0.738525 0.738144 0.737763 0.737763 0.737382 0.7377000 0.736619 0.736238	VALUI Exact 0.7397 0.7393 0.7389 0.7385 0.7381 0.7378 0.7374 0.7370 0.7366 0.7362	$\frac{\text{Error}}{3.09 \times 10^{-5}} \\ \frac{3.09 \times 10^{-5}}{1.21 \times 10^{-5}} \\ 6.65 \times 10^{-6} \\ 2.54 \times 10^{-5} \\ 4.42 \times 10^{-5} \\ 1.83 \times 10^{-5} \\ 4.96 \times 10^{-7} \\ 1.92 \times 10^{-5} \\ 3.80 \times 10^{-5} \\ 3.80 \times 10^{-5} \\ \end{array}$	$\begin{array}{c} \textbf{Error} \\ \textbf{with [5]} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.51 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.47 \times 10^{-4} \\ 1.48 \times 10^{-4} \\ 1.49 \times 10^{-4} \\ 1.49 \times 10^{-4} \end{array}$				

### **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  $\alpha$  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  $\alpha$ , 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.

## Acknowledgement

The authors would like to thank the anonymous reviewers and the editor in charge of handling this manuscript for all their invaluable comments.

## References

- T. Abdeljawad. On conformable fractional calculus. *Journal* of Computational and Applied Mathematics, 279:57–66, 2015.
- [2] O. Abu-Arqub, A. El-Ajou, S. Momani, and N. Shawagfeh. Analytical solutions of fuzzy initial value problems by ham.
- [3] R. P Agarwal, V Lakshmikantham, and J. J Nieto. On the concept of solution for fractional differential equations with uncertainty. *Nonlinear Analysis: Theory, Methods & Applications*, 72(6):2859–2862, 2010.
- [4] A. Ahmadian, S. Salahshour, D. Baleanu, H. Amirkhani, and R. Yunus. Tau method for the numerical solution of a fuzzy fractional kinetic model and its application to the oil palm frond as a promising source of xylose. *Journal of Computational Physics*, 294:562–584, 2015.
- [5] A. Ahmadian, M. Suleiman, and S. Salahshour. An operational matrix based on Legendre polynomials for solving fuzzy fractional-order differential equations. *Abstract and Applied Analysis*, 2013(1):Article ID 505903, 2013.
- [6] T Allahviranloo, A Armand, and Z Gouyandeh. Fuzzy fractional differential equations under generalized fuzzy caputo derivative. *Journal of Intelligent & Fuzzy Systems*, 26:1481–1490, 2014.
- [7] T. Allahviranloo, S. Salahshour, and S. Abbasbandy. Explicit solutions of fractional differential equations with uncertainty. *Soft Computing*, 16(2):297–302, 2012.
- [8] B. Bede and S. G Gal. Generalizations of the differentiability of fuzzy-number-valued functions with applications to fuzzy differential equations. *Fuzzy Sets and Systems*, 151(3):581–599, 2005.
- [9] R. E Bellman and L. A. Zadeh. Decision-making in a fuzzy environment. *Management Science*, 17(4):B–141, 1970.
- [10] B Bonilla, M Rivero, L Rodríguez-Germá, and J J Trujillo. Fractional differential equations as alternative models to nonlinear differential equations. *Applied Mathematics and Computation*, 187(1):79–88, 2007.
- [11] J. J Buckley and L. J Jowers. *Simulating continuous fuzzy systems*. Springer-Verlag, Berlin Heidelberg, 2006.
- [12] J. J. Buckley and Y. Qu. On using  $\alpha$ -cuts to evaluate fuzzy equations. *Fuzzy Sets and Systems*, 38(3):309–312, 1990.
- [13] S. Corveleyn, E. Rosseel, and S. Vandewalle. Iterative solvers for a spectral Galerkin approach to elliptic partial differential equations with fuzzy coefficients. *SIAM Journal* on Scientific Computing, 35(5):S420–S444, 2013.
- [14] J de Andrés Sánchez. Calculating insurance claim reserves with fuzzy regression. *Fuzzy Sets and Systems*, 157(23):3091–3108, 2006.
- [15] Massimo de Falco, Matteo Gaeta, Vincenzo Loia, Luigi Rarità, and Stefania Tomasiello. Differential quadraturebased numerical solutions of a fluid dynamic model for supply chains. *Communications in Mathematical Sciences*, 14(5):1467–1476, 2016.

- [16] R. M. Jafelice, L. C. Barros, and R. C. Bassanezi. Study of the dynamics of HIV under treatment considering fuzzy delay. *Computational and Applied Mathematics*, 33(1):45– 61, 2014.
- [17] S. K. Khattri. From Lobatto quadrature to the Euler constant e. *PRIMUS*, 20(6):488–497, 2010.
- [18] A Kilbas, A. Anatolii, H. M. Srivastava, and J. J Trujillo. *Theory and applications of fractional differential equations*, volume 204. Elsevier Science Limited, 2006.
- [19] J. E. Macías-Díaz and S. Tomasiello. A differential quadrature-based approach à la Picard for systems of partial differential equations associated with fuzzy differential equations. *Journal of Computational and Applied Mathematics*, page doi:10.1016/j.cam.2015.08.009, 2015.
- [20] M. Mazandarani and A. V. Kamyad. Modified fractional Euler method for solving fuzzy fractional initial value problem. *Communications in Nonlinear Science and Numerical Simulation*, 18(1):12–21, 2013.
- [21] I Podlubny. Fractional differential equations: an introduction to fractional derivatives, fractional differential equations, to methods of their solution and some of their applications, volume 198. Academic press, 1998.
- [22] A. Rivaz, O. S. Fard, and Bidgoli T. A. Solving fuzzy fractional differential equations by a generalized differential transform method. *SeMA*, 73:149–170, 2016.
- [23] SM Sadatrasoul and R Ezzati. Numerical solution of two-dimensional nonlinear Hammerstein fuzzy integral equations based on optimal fuzzy quadrature formula. *Journal of Computational and Applied Mathematics*, 292:430–446, 2016.
- [24] S. Tapaswini and S. Chakraverty. Non-probabilistic solutions of uncertain fractional order diffusion equations. *Fundamenta Informaticae*, 133:19–34, 2014.
- [25] S Tomasiello. Some remarks on a new DQ-based method for solving a class of Volterra integro-differential equations. *Applied Mathematics and Computation*, 219(1):399–407, 2012.
- [26] G Y. Tütüncü and N. Kayaalp. An aggregated fuzzy naive Bayes data classifier. *Journal of Computational and Applied Mathematics*, 286:17–27, 2015.
- [27] H. Wang and Y. Liu. Existence results for fuzzy integral equations of fractional order. *International Journal of Mathematical Analysis*, 5(17):811–818, 2011.
- [28] L. A Zadeh. Fuzzy sets. Information and Control, 8(3):338– 353, 1965.

Stefania Tomasiello, Ph.D. in computer science (University of Salerno, Italy), is currently a senior researcher and project manager at CO.RI.SA. (Research Consortium on Agent Systems), University of Salerno, Italy. She was an adjunct professor of Fundamentals of Computer Science, Human-Computer Interaction, Computational Methods and Finite Element Analysis at the University of Basilicata, Italy. She is an expert evaluator (ex-ante and ex-post) of research projects joining academia and industry for the Italian Ministry of Economic Development and the Italian Ministry of University and Research. Her research interests lie in scientific and soft computing, fuzzy mathematics. She is member of the editorial board of International Journal of System Assurance Engineering and Management (Springer) and formerly Applied Mathematics (Scientific and Academic Publishing). She recently joined the editorial board of CAAI Transactions on Intelligence Technology (Elsevier).

Jorge E. Macías-Díaz received the PhD degree in Mathematics from the Tulane University of New Orleans, and the PhD in Physics from the University of New Orleans. His vast research interests range from nonlinear partial differential equations, to structure-preserving numerical methods, computer simulation in sciences and engineering (with emphasis on nonlinear physics, biology and chemistry), and module theory. Jorge belongs to the faculty of Mathematics and Physics at the Universidad Autónoma de Aguascalientes, Mexico, where he holds a full professorship and teaches Real, Complex and Functional Analysis. He has published several research articles in international journals of mathematical and physical sciences. Jorge is referee and editor of various journals on numerical mathematics, and a member of the Mexican Academy of Sciences.