



Research article

Semi-analytic solutions of nonlinear multidimensional fractional differential equations

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Abstract: In this paper, the Adomian decomposition method (ADM) and Picard technique are used to solve a class of nonlinear multidimensional fractional differential equations with Caputo-Fabrizio fractional derivative. The main advantage of the Caputo-Fabrizio fractional derivative appears in its non-singular kernel of a convolution type. The sufficient condition that guarantees a unique solution is obtained, the convergence of the series solution is discussed, and the maximum absolute error is estimated. Several numerical problems with an unknown exact solution are solved using the two techniques. A comparative study between the two solutions is presented. A comparative study shows that the time consumed by ADM is much smaller compared with the Picard technique.

Keywords: Adomian decomposition; Picard method; Caputo-Fabrizio; multidimensional; fractional differential equations; bassset problem

1. Introduction

Fractional calculus is a main branch of mathematics that can be considered as the generalisation of integration and differentiation to arbitrary orders. This hypothesis begins with the assumptions of L. Euler (1730) and G. W. Leibniz (1695). Fractional differential equations (FDEs) have lately gained attention and publicity due to their realistic and accurate computations [1–7]. There are various types of fractional derivatives, including Riemann–Liouville, Caputo, Grünwald–Letnikov, Weyl, Marchaud, and Atangana. This topic’s history can be found in [8–11]. Undoubtedly, fractional calculus applies to mathematical models of different phenomena, sometimes more effectively than ordinary calculus [12, 13]. As a result, it can illustrate a wide range of dynamical and engineering models with greater

precision. Applications have been developed and investigated in a variety of scientific and engineering fields over the last few decades, including bioengineering [14], mechanics [15], optics [16], physics [17], mathematical biology, electrical power systems [18–20] and signal processing [21–23].

One of the definitions of fractional derivatives is Caputo-Fabrizio, which adds a new dimension in the study of FDEs. The new derivative's feature is that it has a nonsingular kernel, which is made from a combination of an ordinary derivative with an exponential function, but it has the same supplementary motivating properties with various scales as in the Riemann-Liouville fractional derivatives and Caputo. The Caputo-Fabrizio fractional derivative has been used to solve real-world problems in numerous areas of mathematical modelling for example, numerical solutions for groundwater pollution, the movement of waves on the surface of shallow water modelling [24], RLC circuit modelling [25], and heat transfer modelling [26, 27] were discussed.

Rach (1987), Bellomo and Sarafyan (1987) first compared the Adomian Decomposition method (ADM) [28–32] to the Picard method on a variety of examples. These methods have many benefits: they effectively work with various types of linear and nonlinear equations and also provide an analytic solution for all of these equations with no linearization or discretization. These methods are more realistic compared with other numerical methods as each technique is used to solve a specific type of equations, on the other hand ADM and Picard are useful for many types of equations. In the numerical examples provided, we compare ADM and Picard solutions of multidimensional fractional order equations with Caputo-Fabrizio.

The fractional derivative of Caputo-Fabrizio for the function $x(t)$ is defined as [33]

$${}^{CF}D_0^\alpha x(t) = \frac{B(\alpha)}{1-\alpha} \int_0^t \frac{d}{ds} (x(s)) e^{-\frac{\alpha}{1-\alpha}(t-s)} ds, \quad (1.1)$$

and its corresponding fractional integral is

$${}^{CF}I^\alpha x(t) = \frac{1-\alpha}{B(\alpha)} x(t) + \frac{\alpha}{B(\alpha)} \int_0^t x(s) ds, \quad 0 < \alpha < 1, \quad (1.2)$$

where $x(t)$ be continuous and differentiable on $[0, T]$. Also, in the above definition, the function $B(\alpha) > 0$ is a normalized function which satisfy the condition $B(0) = B(1) = 0$. The relation between the Caputo-Fabrizio fractional derivative and its corresponding integral is given by

$$\left({}^{CF}I_0^\alpha\right)\left({}^{CF}D_0^\alpha f(t)\right) = f(t) - f(a). \quad (1.3)$$

2. Construction of the algorithm

In this section, we will introduce a multidimensional FDE subject to the initial condition. Let $\alpha \in (0, 1]$, $0 < \alpha_1 < \alpha_2 < \dots, \alpha_m < 1$, and m is integer real number,

$$\begin{aligned} {}^{CF}Dx &= f(t, x, {}^{CF}D^{\alpha_1}x, {}^{CF}D^{\alpha_2}x, \dots, {}^{CF}D^{\alpha_m}x), \\ x(0) &= c_0, \end{aligned} \quad (2.1)$$

where $x = x(t)$, $t \in J = [0, T]$, $T \in R^+$, $x \in C(J)$.

To facilitate the equation and make it easy for the calculation, we let $x(t) = c_0 + X(t)$ so Eq (2.1) can be written as

$$\begin{aligned} {}^{CF}D^\alpha X &= f(t, c_0 + X, {}^{CF}D^{\alpha_1}X, {}^{CF}D^{\alpha_2}X, \dots, {}^{CF}D^{\alpha_m}X), \\ X(0) &= 0. \end{aligned} \quad (2.2)$$

the algorithm depends on converting the initial condition from a constant c_0 to 0.

Let ${}^{CF}D^\alpha X = y(t)$ then $X = {}^{CF}I^\alpha y$, so we have

$${}^{CF}D^{\alpha_i}X = {}^{CF}I^{\alpha-\alpha_i} {}^{CF}D^\alpha X = {}^{CF}I^{\alpha-\alpha_i}y, \quad i = 1, 2, \dots, m. \quad (2.3)$$

Substituting in Eq (2.2) we obtain

$$y = f(t, c_0 + {}^{CF}I^\alpha y, {}^{CF}I^{\alpha-\alpha_1}y, \dots, {}^{CF}I^{\alpha-\alpha_m}y). \quad (2.4)$$

Assume f satisfies Lipschitz condition with Lipschitz constant L given by,

$$|f(t, y_0, y_1, \dots, y_m) - f(t, z_0, z_1, \dots, z_m)| \leq L \sum_{i=0}^m |y_i - z_i|, \quad (2.5)$$

which implies

$$\begin{aligned} & \left| f(t, c_0 + {}^{CF}I^\alpha y, {}^{CF}I^{\alpha-\alpha_1}y, \dots, {}^{CF}I^{\alpha-\alpha_m}y) \right. \\ & \left. - f(t, c_0 + {}^{CF}I^\alpha z, {}^{CF}I^{\alpha-\alpha_1}z, \dots, {}^{CF}I^{\alpha-\alpha_m}z) \right| \\ & \leq L \sum_{i=0}^m \left| {}^{CF}I^{\alpha-\alpha_i}y - {}^{CF}I^{\alpha-\alpha_i}z \right|. \end{aligned} \quad (2.6)$$

The solution algorithm of Eq (2.4) using ADM is,

$$\begin{aligned} y_0(t) &= a(t) \\ y_{n+1}(t) &= A_n(t), \quad j \geq 0. \end{aligned} \quad (2.7)$$

where $a(t)$ processes all free terms in Eq (2.4) and A_n are the Adomian polynomials of the nonlinear term which takes the form [34]

$$A_n = f(S_n) - \sum_{i=0}^{n-1} A_i, \quad (2.8)$$

where $f(S_n) = \sum_{i=0}^n A_i$. Later, this accelerated formula of Adomian polynomial will be used in convergence analysis and error estimation. The solution of Eq (2.4) can be written in the form,

$$y(t) = \sum_{i=0}^{\infty} y_i(t). \quad (2.9)$$

lastly, the solution of the Eq (2.4) takes the form

$$x(t) = c_0 + X(t) = c_0 + {}^{CF}I^\alpha y(t). \quad (2.10)$$

At which we convert the parameter to the initial form y to x in Eq (2.10), so we have the solution of the original Eq (2.1).

3. Convergence analysis

3.1. Existence and uniqueness

Define a mapping $F : E \rightarrow E$ where $E = (C[J], \|\cdot\|)$ is a Banach space of all continuous functions on J with the norm $\|x\| = \max_{t \in J} x(t)$.

Theorem 3.1. Equation (2.4) has a unique solution whenever $0 < \phi < 1$ where $\phi = L \left(\sum_{i=0}^m \frac{[(\alpha - \alpha_i)(T-1)] + 1}{B(\alpha - \alpha_i)} \right)$.

Proof. First, we define the mapping $F : E \rightarrow E$ as

$$Fy = f(t, c_0 + {}^{CF}I^\alpha y, {}^{CF}I^{\alpha-\alpha_1} y, \dots, {}^{CF}I^{\alpha-\alpha_m} y).$$

Let y and $z \in E$ are two different solutions of Eq (2.4). Then

$$Fy - Fz = f(t, c_0 + {}^{CF}I^\alpha y, {}^{CF}I^{\alpha-\alpha_1} y, \dots, {}^{CF}I^{\alpha-\alpha_m} y) - f(t, c_0 + {}^{CF}I^\alpha z, {}^{CF}I^{\alpha-\alpha_1} z, \dots, {}^{CF}I^{\alpha-\alpha_m} z)$$

which implies that

$$\begin{aligned} |Fy - Fz| &= \left| f(t, c_0 + {}^{CF}I^\alpha y, {}^{CF}I^{\alpha-\alpha_1} y, \dots, {}^{CF}I^{\alpha-\alpha_m} y) \right. \\ &\quad \left. - f(t, c_0 + {}^{CF}I^\alpha z, {}^{CF}I^{\alpha-\alpha_1} z, \dots, {}^{CF}I^{\alpha-\alpha_m} z) \right| \\ &\leq L \sum_{i=0}^m \left| {}^{CF}I^{\alpha-\alpha_i} y - {}^{CF}I^{\alpha-\alpha_i} z \right| \\ &\leq L \sum_{i=0}^m \left| \frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} (y - z) + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} \int_0^t (y - z) ds \right| \\ \|Fy - Fz\| &\leq L \sum_{i=0}^m \frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} \max_{t \in J} |y - z| + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} \max_{t \in J} |y - z| \int_0^t ds \\ &\leq L \sum_{i=0}^m \frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} \|y - z\| + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} \|y - z\| T \\ &\leq L \|y - z\| \left(\sum_{i=0}^m \frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} T \right) \\ &\leq L \|y - z\| \left(\sum_{i=0}^m \frac{[(\alpha - \alpha_i)(T-1)] + 1}{B(\alpha - \alpha_i)} \right) \\ &\leq \phi \|y - z\|. \end{aligned}$$

under the condition $0 < \phi < 1$, the mapping F is contraction and hence there exists a unique solution $y \in C[J]$ for the problem Eq (2.4) and this completes the proof.

3.2. Proof of convergence

Theorem 3.2. The series solution of the problem Eq (2.4) converges if $|y_1(t)| < c$ and c is finite.

Proof. Define a sequence $\{S_p\}$ such that $S_p = \sum_{i=0}^p y_i(t)$ is the sequence of partial sums from the series solution $\sum_{i=0}^{\infty} y_i(t)$, we have

$$f(t, c_0 + {}^{CF}I^\alpha y, {}^{CF}I^{\alpha-\alpha_1} y, \dots, {}^{CF}I^{\alpha-\alpha_m} y) = \sum_{i=0}^{\infty} A_i,$$

So

$$f(t, c_0 + {}^{CF}I^\alpha S_p, {}^{CF}I^{\alpha-\alpha_1} S_p, \dots, {}^{CF}I^{\alpha-\alpha_m} S_p) = \sum_{i=0}^p A_i,$$

From Eq (2.7) we have

$$\sum_{i=0}^{\infty} y_i(t) = a(t) + \sum_{i=0}^{\infty} A_{i-1}$$

let S_p, S_q be two arbitrary sums with $p \geq q$. Now, we are going to prove that $\{S_p\}$ is a Cauchy sequence in this Banach space. We have

$$\begin{aligned} S_p &= \sum_{i=0}^p y_i(t) = a(t) + \sum_{i=0}^p A_{i-1}, \\ S_q &= \sum_{i=0}^q y_i(t) = a(t) + \sum_{i=0}^q A_{i-1}. \end{aligned}$$

$$\begin{aligned} S_p - S_q &= \sum_{i=0}^p A_{i-1} - \sum_{i=0}^q A_{i-1} = \sum_{i=q+1}^p A_{i-1} = \sum_{i=q}^{p-1} A_{i-1} \\ &= f(t, c_0 + {}^{CF}I^\alpha S_{p-1}, {}^{CF}I^{\alpha-\alpha_1} S_{p-1}, \dots, {}^{CF}I^{\alpha-\alpha_m} S_{p-1}) - \\ &\quad f(t, c_0 + {}^{CF}I^\alpha S_{q-1}, {}^{CF}I^{\alpha-\alpha_1} S_{q-1}, \dots, {}^{CF}I^{\alpha-\alpha_m} S_{q-1}) \end{aligned}$$

$$\begin{aligned} |S_p - S_q| &= \left| f(t, c_0 + {}^{CF}I^\alpha S_{p-1}, {}^{CF}I^{\alpha-\alpha_1} S_{p-1}, \dots, {}^{CF}I^{\alpha-\alpha_m} S_{p-1}) - \right. \\ &\quad \left. f(t, c_0 + {}^{CF}I^\alpha S_{q-1}, {}^{CF}I^{\alpha-\alpha_1} S_{q-1}, \dots, {}^{CF}I^{\alpha-\alpha_m} S_{q-1}) \right| \\ &\leq L \sum_{i=0}^m \left| {}^{CF}I^{\alpha-\alpha_i} S_{p-1} - {}^{CF}I^{\alpha-\alpha_i} S_{q-1} \right| \\ &\leq L \sum_{i=0}^m \left| \frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} (S_{p-1} - S_{q-1}) + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} \int_0^t (S_{p-1} - S_{q-1}) ds \right| \\ &\leq L \sum_{i=0}^m \frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} |S_{p-1} - S_{q-1}| + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} \int_0^t |S_{p-1} - S_{q-1}| ds \end{aligned}$$

$$\begin{aligned}
\|S_p - S_q\| &\leq L \sum_{i=0}^m \frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} \max_{t \in J} |S_{p-1} - S_{q-1}| + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} \max_{t \in J} |S_{p-1} - S_{q-1}| \int_0^t ds \\
&\leq L \|S_p - S_q\| \sum_{i=0}^m \left(\frac{1 - (\alpha - \alpha_i)}{B(\alpha - \alpha_i)} + \frac{\alpha - \alpha_i}{B(\alpha - \alpha_i)} T \right) \\
&\leq L \|S_p - S_q\| \left(\sum_{i=0}^m \frac{[(\alpha - \alpha_i)(T - 1)] + 1}{B(\alpha - \alpha_i)} \right) \\
&\leq \phi \|S_p - S_q\|
\end{aligned}$$

let $p = q + 1$ then,

$$\|S_{q+1} - S_q\| \leq \phi \|S_q - S_{q-1}\| \leq \phi^2 \|S_{q-1} - S_{q-2}\| \leq \dots \leq \phi^q \|S_1 - S_0\|$$

From the triangle inequality we have

$$\begin{aligned}
\|S_p - S_q\| &\leq \|S_{q+1} - S_q\| + \|S_{q+2} - S_{q+1}\| + \dots + \|S_p - S_{p-1}\| \\
&\leq [\phi^q + \phi^{q+1} + \dots + \phi^{p-1}] \|S_1 - S_0\| \\
&\leq \phi^q [1 + \phi + \dots + \phi^{p-q+1}] \|S_1 - S_0\| \\
&\leq \phi^q \left[\frac{1 - \phi^{p-q}}{1 - \phi} \right] \|y_1(t)\|
\end{aligned}$$

Since $0 < \phi < 1, p \geq q$ then $(1 - \phi^{p-q}) \leq 1$. Consequently

$$\|S_p - S_q\| \leq \frac{\phi^q}{1 - \phi} \|y_1(t)\| \leq \frac{\phi^q}{1 - \phi} \max_{t \in J} |y_1(t)| \quad (3.1)$$

but $|y_1(t)| < \infty$ and as $q \rightarrow \infty$ then, $\|S_p - S_q\| \rightarrow 0$ and hence, $\{S_p\}$ is a Cauchy sequence in this Banach space then the proof is complete.

3.3. Error estimate

Theorem 3.3. *The maximum absolute truncated error Eq (2.4) is estimated to be*
 $\max_{t \in J} |y(t) - \sum_{i=0}^q y_i(t)| \leq \frac{\phi^q}{1 - \phi} \max_{t \in J} |y_1(t)|$

Proof. From the convergence theorem inequality (Eq 3.1) we have

$$\|S_p - S_q\| \leq \frac{\phi^q}{1 - \phi} \max_{t \in J} |y_1(t)|$$

but, $S_p = \sum_{i=0}^p y_i(t)$ as $p \rightarrow \infty$ then, $S_p \rightarrow y(t)$ so,

$$\|y(t) - S_q\| \leq \frac{\phi^q}{1 - \phi} \max_{t \in J} |y_1(t)|$$

so, the maximum absolute truncated error in the interval J is,

$$\max_{t \in J} \left| y(t) - \sum_{i=0}^q y_i(t) \right| \leq \frac{\phi^q}{1 - \phi} \max_{t \in J} |y_1(t)| \quad (3.2)$$

and this completes the proof.

4. Numerical examples

In this part, we introduce several numerical examples with unknown exact solution and we will use inequality (Eq 3.2) to estimate the maximum absolute truncated error.

Example 4.1. *Application of linear FDE*

$${}^{CF}Dx(t) + 2a{}^{CF}D^{1/2}x(t) + bx(t) = 0, \quad x(0) = 1. \quad (4.1)$$

A Basset problem in fluid dynamics is a classical problem which is used to study the unsteady movement of an accelerating particle in a viscous fluid under the action of the gravity [36]

Set

$$X(t) = x(t) - 1$$

Equation (4.1) will be

$${}^{CF}DX(t) + 2a{}^{CF}D^{1/2}X(t) + bX(t) = 0, \quad X(0) = 0. \quad (4.2)$$

Applying Eq (2.3) to Eq (4.2), and using initial condition, also we take $a = 1$, $b = 1/2$,

$$y = -\frac{1}{2} - 2I^{1/2}y - \frac{1}{2}Iy \quad (4.3)$$

Applying ADM to Eq (4.3), we find the solution algorithm become

$$\begin{aligned} y_0(t) &= -\frac{1}{2}, \\ y_i(t) &= -2 {}^{CF}I^{1/2}y_{i-1} - \frac{1}{2} {}^{CF}I y_{i-1}, \quad i \geq 1. \end{aligned} \quad (4.4)$$

Applying Picard solution to Eq (4.2), we find the solution algorithm become

$$\begin{aligned} y_0(t) &= -\frac{1}{2}, \\ y_i(t) &= -\frac{1}{2} - 2 {}^{CF}I^{1/2}y_{i-1} - \frac{1}{2} {}^{CF}I y_{i-1}, \quad i \geq 1. \end{aligned} \quad (4.5)$$

From Eq (4.4), the solution using ADM is given by $y(t) = \lim_{q \rightarrow \infty} \sum_{i=0}^q y_i(t)$ while from Eq (4.5), the solution using Picard technique is given by $y(t) = \lim_{i \rightarrow \infty} y_i(t)$. Lately, the solution of the original problem Eq (4.2), is

$$x(t) = 1 + {}^{CF}I y(t).$$

One the same processor ($q = 20$), the time consumed using ADM is 0.037 seconds, while the time consumed using Picard is 7.955 seconds.

Figure 1 gives a comparison between ADM and Picard solution of Ex. 4.1.

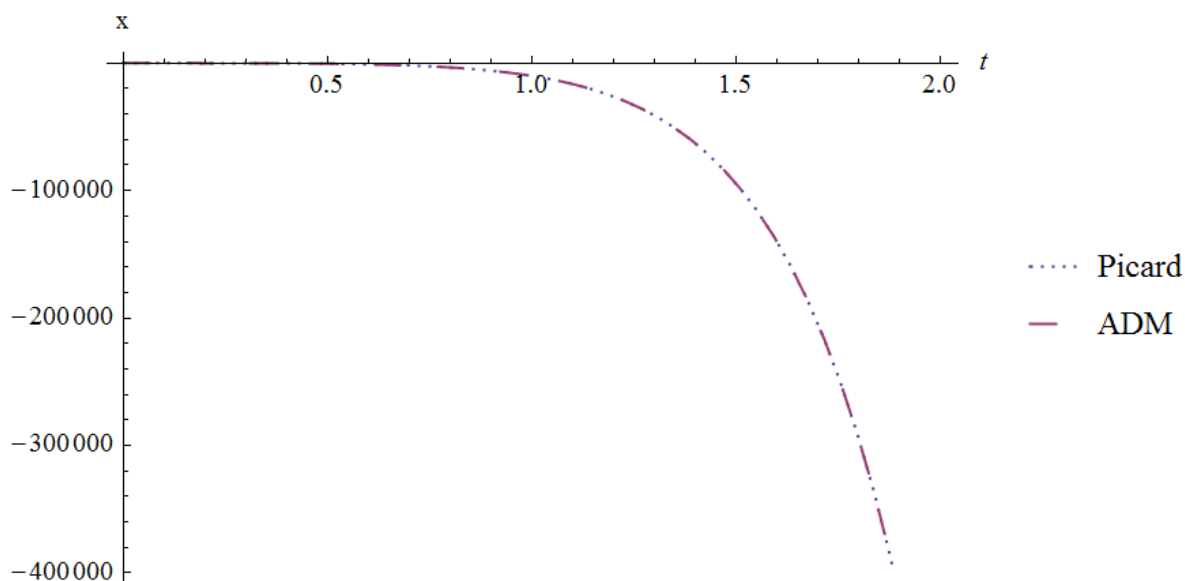


Figure 1. ADM and Picard solution of Ex. 4.1.

Example 4.2. Consider the following nonlinear FDE [35]

$${}^{CF}D^{1/2}x = \frac{8t^{3/2}}{3\sqrt{\pi}} - \frac{t^{7/4}}{4\Gamma\left(\frac{11}{4}\right)} - \frac{t^4}{4} + \frac{1}{8} {}^{CF}D^{1/4}x + \frac{1}{4}x^2, \quad (4.6)$$

$$x(0) = 0.$$

Applying Eq (2.3) to Eq (4.6), and using initial condition,

$$y = \frac{8t^{3/2}}{3\sqrt{\pi}} - \frac{t^{7/4}}{4\Gamma\left(\frac{11}{4}\right)} - \frac{t^4}{4} + \frac{1}{8} {}^{CF}I^{1/4}y + \frac{1}{4} \left({}^{CF}I^{1/2}y\right)^2. \quad (4.7)$$

Applying ADM to Eq (4.7), we find the solution algorithm will be become

$$y_0(t) = \frac{8t^{3/2}}{3\sqrt{\pi}} - \frac{t^{7/4}}{4\Gamma\left(\frac{11}{4}\right)} - \frac{t^4}{4},$$

$$y_i(t) = \frac{1}{8} {}^{CF}I^{1/4}y_{i-1} + \frac{1}{4} (A_{i-1}), \quad i \geq 1. \quad (4.8)$$

at which A_i are Adomian polynomial of the nonlinear term $\left({}^{CF}I^{1/2}y\right)^2$.

Applying Picard solution to Eq (4.7), we find the the solution algorithm become

$$y_0(t) = \frac{8t^{3/2}}{3\sqrt{\pi}} - \frac{t^{7/4}}{4\Gamma\left(\frac{11}{4}\right)} - \frac{t^4}{4},$$

$$y_i(t) = y_0(t) + \frac{1}{8} {}^{CF}I^{1/4}y_{i-1} + \frac{1}{4} \left({}^{CF}I^{1/2}y_{i-1}\right)^2, \quad i \geq 1. \quad (4.9)$$

From Eq (4.8), the solution using ADM is given by $y(t) = \lim_{q \rightarrow \infty} y_i(t)$ while from Eq (4.9), the solution using Picard technique is given by $y(t) = \lim_{i \rightarrow \infty} y_i(t)$. Finally, the solution of the original problem Eq (4.7), is.

$$x(t) = {}^{CF}I^{1/2}y.$$

On the same processor ($q = 2$), the time consumed using ADM is 65.13 seconds, while the time consumed using Picard is 544.787 seconds.

Table 1 showed the maximum absolute truncated error of of ADM solution (using Theorem 3.3) at different values of m (when $t = 0.5$; $N = 2$):

Table 1. Max. absolute error.

q	max. absolute error
2	0.114548
5	0.099186
10	0.004363

Figure 2 gives a comparison between ADM and Picard solution of Ex. 4.2.

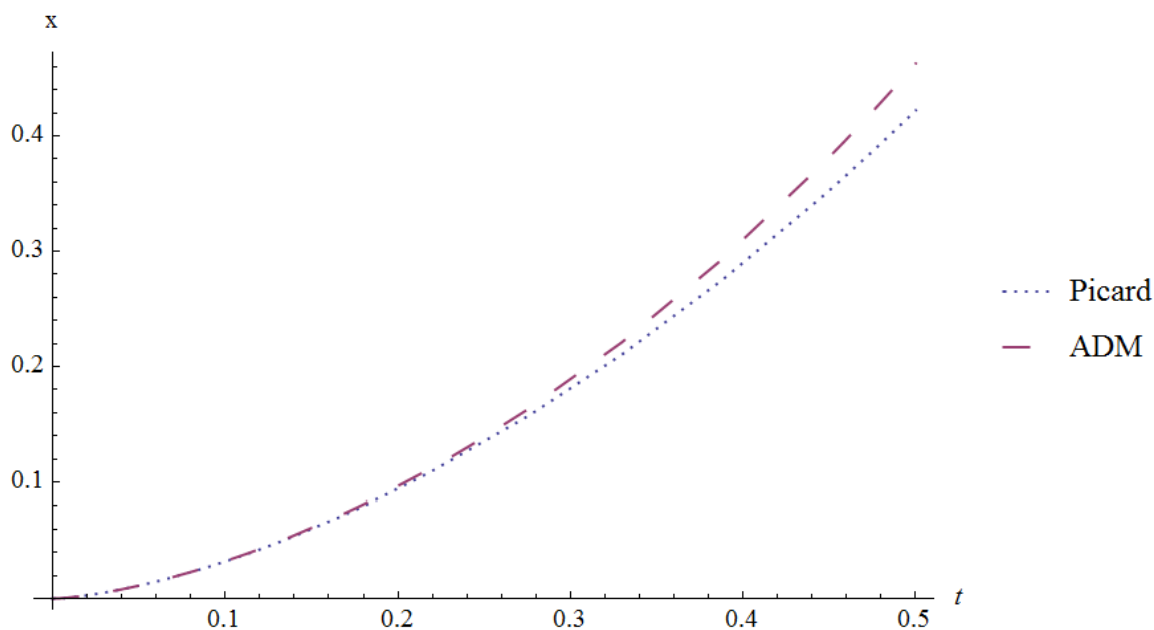


Figure 2. ADM and Picard solution of Ex. 4.2.

Example 4.3. Consider the following nonlinear FDE [35]

$$\begin{aligned} {}^{CF}D^\alpha x &= 3t^2 - \frac{128}{125\pi}t^5 + \frac{1}{10} \left({}^{CF}D^{1/2}x \right)^2, \\ x(0) &= 0. \end{aligned} \quad (4.10)$$

Applying Eq (2.3) to Eq (4.10), and using initial condition,

$$y = 3t^2 - \frac{128}{125\pi}t^5 + \frac{1}{10} \left({}^{CF}I^{1/2}y \right)^2 \quad (4.11)$$

Applying ADM to Eq (4.11), we find the solution algorithm become

$$\begin{aligned} y_0(t) &= 3t^2 - \frac{128}{125\pi}t^5, \\ y_i(t) &= \frac{1}{10} (A_{i-1}), \quad i \geq 1 \end{aligned} \quad (4.12)$$

at which A_i are Adomian polynomial of the nonlinear term $\left({}^{CF}I^{1/2}y \right)^2$.

Then applying Picard solution to Eq (4.11), we find the solution algorithm become

$$\begin{aligned} y_0(t) &= 3t^2 - \frac{128}{125\pi}t^5, \\ y_i(t) &= y_0(t) + \frac{1}{10} \left({}^{CF}I^{1/2}y_{i-1} \right)^2, \quad i \geq 1. \end{aligned} \quad (4.13)$$

From Eq (4.12), the solution using ADM is given by $y(t) = \lim_{q \rightarrow \infty} y_i(t)$ while from Eq (4.13), the solution is $y(t) = \lim_{i \rightarrow \infty} y_i(t)$. Finally, the solution of the original problem Eq (4.11), is

$$x(t) = {}^{CF}Iy(t).$$

One the same processor ($q = 4$), the time consumed using ADM is 2.09 seconds, while the time consumed using Picard is 44.725 seconds.

Table 2 showed the maximum absolute truncated error of of ADM solution (using Theorem 3.3) at different values of m (when $t = 0:5$; $N = 4$):

Table 2. Max. absolute error.

q	max. absolute error
2	0.00222433
5	0.0000326908
10	$2.88273 \cdot 10^{-8}$

Figure 3 gives a comparison between ADM and Picard solution of Ex. 4.3 with $\alpha = 1$.

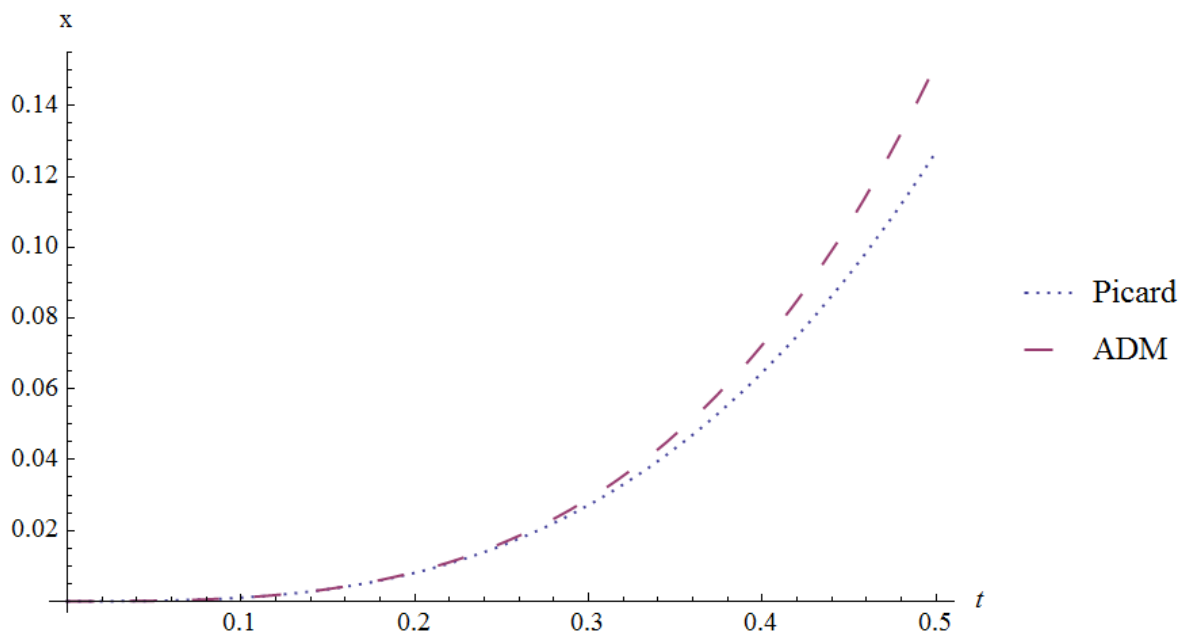


Figure 3. ADM and Picard solution where of Ex. 4.3.

Example 4.4. Consider the following nonlinear FDE [35]

$$\begin{aligned} {}^{CF}D^\alpha x &= t^2 + \frac{1}{2} {}^{CF}D^{\alpha_1} x + \frac{1}{4} {}^{CF}D^{\alpha_2} x + \frac{1}{6} {}^{CF}D^{\alpha_3} x + \frac{1}{8} x^4, \\ x(0) &= 0. \end{aligned} \quad (4.14)$$

Applying Eq (2.3) to Eq (4.10), and using initial condition,

$$y = t^2 + \frac{1}{2} ({}^{CF}I^{\alpha-\alpha_1}y) + \frac{1}{4} ({}^{CF}I^{\alpha-\alpha_2}y) + \frac{1}{6} ({}^{CF}I^{\alpha-\alpha_3}y) + \frac{1}{8} ({}^{CF}I^\alpha y)^4, \quad (4.15)$$

Applying ADM to Eq (4.15), we find the solution algorithm become

$$\begin{aligned} y_0(t) &= t^2, \\ y_i(t) &= \frac{1}{2} ({}^{CF}I^{\alpha-\alpha_1}y) + \frac{1}{4} ({}^{CF}I^{\alpha-\alpha_2}y) + \frac{1}{6} ({}^{CF}I^{\alpha-\alpha_3}y) + \frac{1}{8} A_{i-1}, \quad i \geq 1 \end{aligned} \quad (4.16)$$

where A_i are Adomian polynomial of the nonlinear term $({}^{CF}I^\alpha y)^4$.

Then applying Picard solution to Eq (4.15), we find the solution algorithm become

$$\begin{aligned} y_0(t) &= t^2, \\ y_i(t) &= t^2 + \frac{1}{2} ({}^{CF}I^{\alpha-\alpha_1}y_{i-1}) + \frac{1}{4} ({}^{CF}I^{\alpha-\alpha_2}y_{i-1}) + \frac{1}{6} ({}^{CF}I^{\alpha-\alpha_3}y_{i-1}) + \frac{1}{8} ({}^{CF}I^\alpha y_{i-1})^4 \quad i \geq 1. \end{aligned} \quad (4.17)$$

From Eq (4.16), the solution using ADM is given by $y(t) = \lim_{q \rightarrow \infty} \sum_{i=0}^q y_i(t)$ while from Eq (4.17), the solution using Picard technique is $y(t) = \lim_{i \rightarrow \infty} y_i(t)$. Finally, the solution of the original problem Eq (4.14), is

$$x(t) = {}^{CF}I^\alpha y(t).$$

On the same processor ($q = 3$), the time consumed using ADM is 0.437 seconds, while the time consumed using Picard is (16.816) seconds. Figure 4 shows a comparison between ADM and Picard solution of Ex. 4.4 at $\alpha = 0.7$, $\alpha_1 = 0.1$, $\alpha_2 = 0.3$, $\alpha_3 = 0.5$.

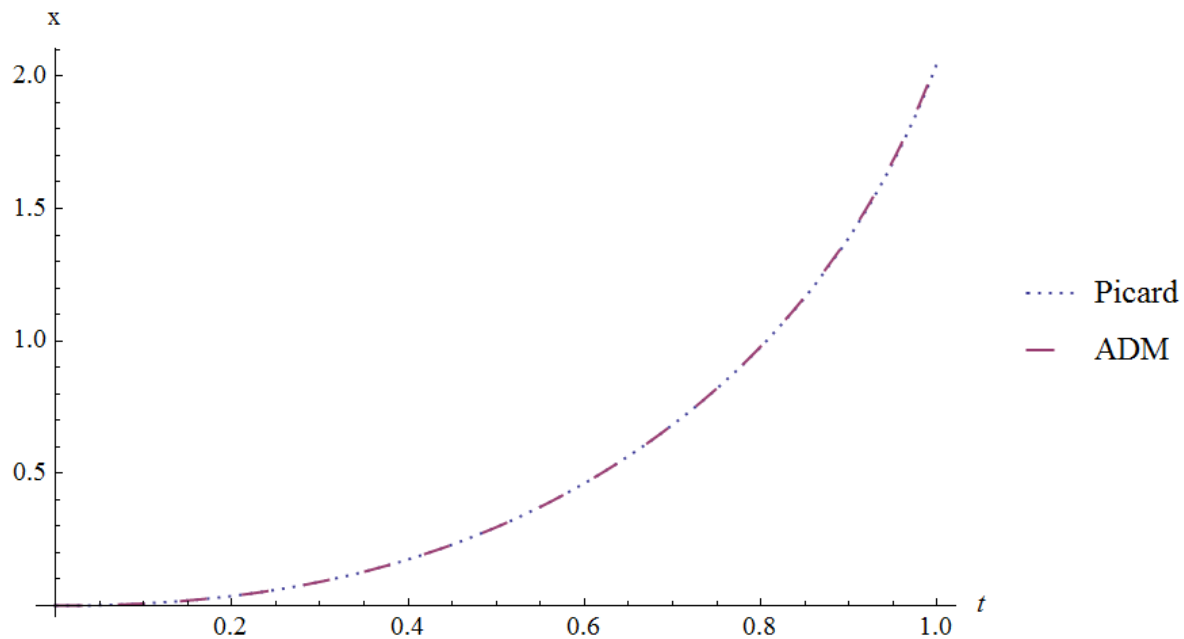


Figure 4. ADM and Picard solution where of Ex. 4.4.

5. Conclusions

The Caputo-Fabrizio fractional derivative has a nonsingular kernel, and consequently, this definition is appropriate in solving nonlinear multidimensional FDE [37, 38]. Since the selected numerical problems have an unknown exact solution, the formula (3.2) can be used to estimate the maximum absolute truncated error. By comparing the time taken on the same processor (i7-2670QM), it was found that the time consumed by ADM is much smaller compared with the Picard technique. Furthermore, Picard gives a more accurate solution than ADM at the same interval with the same number of terms.

Conflict of interest

The authors declare there is no conflict of interest.

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