



# A spectral approach to non-linear weakly singular fractional integro-differential equations

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## Abstract

In this work, a class of non-linear weakly singular fractional integro-differential equations is considered, and we first prove existence, uniqueness, and smoothness properties of the solution under certain assumptions on the given data. We propose a numerical method based on spectral Petrov-Galerkin method that handling to the non-smooth behavior of the solution. The most outstanding feature of our approach is to evaluate the approximate solution by means of recurrence relations despite solving complex non-linear algebraic system. Furthermore, the well-known exponential accuracy is established in  $L^2$ -norm, and we provide some examples to illustrate the theoretical results and the performance of the proposed method.

**Keywords** Weakly singular fractional integro-differential equation · Caputo derivative operator · Generalized Jacobi polynomials · Spectral Petrov-Galerkin method · Convergence

**Mathematics Subject Classification** 45E10 · 45J05 · 34K37 · 33C45 · 65D15

## 1 Introduction

The subject of fractional calculus has recently gained significant popularity and importance, due mainly to its memory features and demonstrated applications in numerous seemingly diverse and widespread fields of science and engineering. For further details,

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readers are relegated to the books ([3], [6], [14], [18], [24], [30]) and review papers ([17], [26], [29]).

There is no generally applicable method to find an analytic solution to an arbitrary given fractional-integro differential equation (FIDE). That is why effective numerical methods can help overcome the problems caused by the shortage of analytical methods for the computation of solutions to FIDEs. Various kinds of approximate methods have independently appeared for the numerical solution of FIDEs along with smooth kernel function, such as Quadratic method [15], spline collocation method [23, 25], differential transform method [1], Legendre wavelet method [2], second Chebyshev wavelet method [35], Laguerre collocation method [5], Jacobi collocation method [7], Taylor expansion method [13], Legendre collocation method [19] and hybrid collocation method [16].

Weakly singular FIDEs seem to be investigated less frequently than FIDEs associated with smooth kernel function. For instance, alternative Legendre polynomials method [27], Jacobi Tau method [20], second kind Chebyshev spectral method [22], second kind Chebyshev wavelet method [31], and piecewise polynomial collocation method [34] have introduced for the numerical solution of linear weakly singular FIDEs. Moreover, the most frequently used methods for obtaining the approximate solutions of non-linear weakly singular FIDEs are Legendre wavelet method [32], shifted Jacobi collocation method [4], and hat functions method [21].

In this work we consider a non-linear weakly singular FIDE

$$\begin{cases} D_C^\alpha y(t) = f(t, y(t)) + \lambda \int_0^t (t-s)^{\beta-1} g(t, s, y(s)) ds, & t \in \Lambda = [0, T], \\ y^{(k)}(0) = y_0^{(k)}, \quad k = 0, 1, \dots, [\alpha] - 1, & \alpha > 0, \quad 0 < \beta \leq 1, \end{cases} \quad (1.1)$$

where  $\lambda \in \mathbb{R}$ ,  $\alpha = \frac{a_1}{b_1}$ ,  $\beta = \frac{a_2}{b_2}$  along with  $a_i \geq 1$ ,  $b_i \geq 2$ ,  $i = 1, 2$ , and  $(a_i, b_i)$  are the pairs of co-prime integers.

Here  $[\cdot]$  is the ceiling function, and  $T$  is a finite positive real value.  $f(t, y(t)) : \Lambda \times \mathbb{R} \rightarrow \mathbb{R}$  and  $g(t, s, y(s)) : D \times \mathbb{R} \rightarrow \mathbb{R}$ , with

$$D = \{(t, s) \in \Lambda \times \Lambda : 0 \leq s \leq t \leq T\},$$

are continuous functions, and  $y(t) : \Lambda \rightarrow \mathbb{R}$  is the unknown.  $D_C^\alpha$  is Caputo fractional derivative of order  $\alpha$  defined by

$$D_C^\alpha(\cdot) = I^{[\alpha]-\alpha} \partial_t^{[\alpha]}(\cdot),$$

in which  $I^{[\alpha]-\alpha}$  denotes the Riemann-Liouville fractional integral operator of order  $[\alpha] - \alpha$ , [6, 14, 24].

In this work, a comprehensive investigation regarding the existence, uniqueness, and smoothness properties of (1.1) are provided on the one side, and then an efficient spectral scheme is implemented to (1.1) thanks to the fractional generalized Jacobi functions (FGJFs) introduced in [8]. Needless to direct that spectral methods offer highly accurate approximations for smooth problems. However, some cons are still existed, including the requirement of solving ambiguous and ill-conditioned algebraic

systems and the striking decline in the accuracy of the approximations facing the problems with non-smooth solutions. Contrary to all cons, the numerical strategy in this essay is taken whereby it contributes to both spectral accuracy in attacking non-smooth solution and evaluating approximate solution by means of recurrence relations despite solving non-linear complex algebraic systems.

This paper is organized in the following way. We begin by presenting theorems of existence, uniqueness, and smoothness. This analysis confirms that some derivatives of the solution have probably a discontinuity at the initial point. Subsequently, in Section 3, we first provide the essential concepts and definitions of generalized Jacobi polynomials (GJPs) and FGJFs, and then a state-of-the-art Petrov-Galerkin method is implemented to deal with the solution of (1.1) which satisfies the assumptions of the existence, uniqueness and smoothness theorems. Numerical solvability and practical implementation of the relevant non-linear algebraic system are examined as well. In particular, the error estimate is mightily surveyed in Section 4. Section 5 includes some prototype examples to assess the efficiency and applicability of the introduced method. Section 6 ultimately presents concluding remarks.

## 2 Existence, uniqueness and smoothness

We allocate this section to the existence, uniqueness, and smoothness properties of the solution of (1.1).

$$\text{Let } \psi(t) = \sum_{k=0}^{[\alpha]-1} \frac{t^k}{k!} y_0^{(k)}. \text{ Define the set}$$

$$\Omega_\zeta = \{y \in C([0, T]) : \|y - \psi\|_\Lambda \leq \zeta\},$$

where  $\|\cdot\|_\Lambda = \max_{t \in \Lambda=[0, T]} |z(t)|$  for all  $z \in C(\Lambda)$  and

$$\zeta = \frac{\|f\|_\Lambda T^\alpha}{\Gamma(\alpha + 1)} + \frac{\|g\|_\Lambda \lambda \Gamma(\beta) T^{\alpha+\beta}}{\Gamma(1 + \alpha + \beta)}. \tag{2.1}$$

Through applying the Riemann-Liouville fractional integral operator of order  $\alpha$ , (1.1) is changed into the weakly singular Volterra integral equation

$$y(t) = \psi(t) + I^\alpha(f(t, y(t))) + \lambda I^\alpha \left( \int_0^t (t-s)^{\beta-1} g(t, s, y(s)) ds \right). \tag{2.2}$$

At this stage, we define an operator  $\mathcal{T}_\psi$ , on  $\Omega_\zeta$ , by

$$\begin{aligned} \mathcal{T}_\psi(\varphi)(t) &= \psi(t) + I^\alpha(f(t, \varphi(t))) + \lambda I^\alpha \left( \int_0^t (t-s)^{\beta-1} g(t, s, \varphi(s)) ds \right) \\ &= \psi(t) + \frac{1}{\Gamma(\alpha)} \int_0^t (t-s)^{\alpha-1} f(s, \varphi(s)) ds \\ &\quad + \frac{\lambda}{\Gamma(\alpha)} \int_0^t (t-\tau)^{\alpha-1} \left( \int_0^\tau (\tau-s)^{\beta-1} g(\tau, s, \varphi(s)) ds \right) d\tau. \end{aligned}$$

Using this operator, the equation (1.1) can be rewritten as  $y = \mathcal{T}_\psi(y)$ . Thereby, if the operator  $\mathcal{T}_\psi$  has a unique fixed point on  $\Omega_\zeta$ , (1.1) will possess a unique continuous solution. This result is provided on the next theorem.

To prove theorems of existence, uniqueness, and smoothness, we consider the hypotheses:

- **H1:** Let  $D_1 = \Lambda \times [m_\psi - \zeta, M_\psi + \zeta]$  and  $D_2 = D \times [m_\psi - \zeta, M_\psi + \zeta]$ , where  $m_\psi = \min_{t \in \Lambda} |\psi(t)|$ ,  $M_\psi = \max_{t \in \Lambda} |\psi(t)|$  and  $\zeta$  is defined by (2.1).

The functions  $f(s, y(s)) : D_1 \rightarrow \mathbb{R}$  and  $g(t, s, y(s)) : D_2 \rightarrow \mathbb{R}$  are continuous for all  $s \in \Lambda$  and further the functions  $f$  and  $g$  fulfills Lipschitz conditions with respect to the second and third variables, respectively. Therefore, there exists  $L_1 > 0$  and  $L_2 > 0$  such that

$$\begin{aligned} |f(t, z_1) - f(t, z_2)| &\leq L_1|z_1 - z_2|, & \forall z_1, z_2 \in \Omega_\zeta, \\ |g(t, s, z_1) - g(t, s, z_2)| &\leq L_2|z_1 - z_2|, & \forall z_1, z_2 \in \Omega_\zeta. \end{aligned}$$

- **H2:** We have

$$\begin{aligned} f(t, y(t)) &= \bar{f}(t^{1/b}, y(t)), \\ g(t, s, y(s)) &= \bar{g}(t^{1/b}, s^{1/b}, y(s)), \end{aligned}$$

where  $b$  signifies the least common multiple of  $b_i, i = 1, 2$ , and  $\bar{f}$  and  $\bar{g}$  are analytic functions in the neighborhood of  $(0, y_0^{(0)})$  and  $(0, 0, y_0^{(0)})$ , respectively.

**Theorem 1** (Existence and uniqueness) *Assume that H1 holds. Then, (1.1) possesses a unique continuous solution on  $\Lambda$ .*

**Proof** Suppose that  $y \in \Omega_\zeta$ . It is straightforward prove that  $\mathcal{T}_\psi(y) \in \Omega_\zeta$ .

Let  $\delta > 0$  be a constant such that

$$\max \left( \frac{L_1}{\delta^\alpha} + \frac{L_2 \lambda \Gamma(\beta)}{\delta^{\alpha+\beta}} \right) < 1. \tag{2.3}$$

We introduce a new norm  $\|\cdot\|_\delta$  over the space  $C(\Lambda; \Omega_\zeta)$  as

$$\|\varphi\|_\delta = \left\| \frac{\varphi}{\exp(\delta t)} \right\|_\Lambda.$$

Using standard arguments, it can be readily inferred that  $\Omega_\zeta$  is a closed subset of the Banach space of continuous functions on  $\Lambda$ , associated with the norm  $\|\cdot\|_\delta$ .

Let  $\varphi, \hat{\varphi} \in \Omega_\zeta$ . Regarding the Lipschitz assumption on  $f$  and  $g$ , it follows

$$\begin{aligned} \frac{|\mathcal{T}_\psi \varphi(t) - \mathcal{T}_\psi \hat{\varphi}(t)|}{\exp(\delta t)} &\leq \frac{L_1}{\Gamma(\alpha) \exp(\delta t)} \int_0^t (t-s)^{\alpha-1} \exp(\delta s) \frac{|\varphi(s) - \hat{\varphi}(s)|}{\exp(\delta s)} ds \\ &+ \frac{L_2 \lambda}{\Gamma(\alpha) \exp(\delta t)} \int_0^t (t-s)^{\alpha-1} \left( \int_0^s (s-\tau)^{\beta-1} \exp(\delta \tau) \frac{|\varphi(\tau) - \hat{\varphi}(\tau)|}{\exp(\delta \tau)} d\tau \right) ds \end{aligned}$$

$$\begin{aligned}
 &\leq \frac{L_1}{\Gamma(\alpha) \exp(\delta t)} \|\varphi - \hat{\varphi}\|_\delta \int_0^t (t-s)^{\alpha-1} \exp(\delta s) ds \\
 &\quad + \frac{L_2 \lambda}{\Gamma(\alpha) \exp(\delta t)} \|\varphi - \hat{\varphi}\|_\delta \int_0^t (t-s)^{\alpha-1} \left( \int_0^s (s-\tau)^{\beta-1} \exp(\delta \tau) d\tau \right) ds \\
 &\leq \frac{L_1}{\Gamma(\alpha) \delta^\alpha} \|\varphi - \hat{\varphi}\|_\delta \int_0^{\delta t} u^{\alpha-1} \exp(-u) du \\
 &\quad + \frac{L_2 \lambda}{\Gamma(\alpha) \exp(\delta t) \delta^\beta} \|\varphi - \hat{\varphi}\|_\delta \int_0^t (t-s)^{\alpha-1} \exp(\delta s) \left( \int_0^{\delta s} v^{\beta-1} \exp(-v) dv \right) ds.
 \end{aligned}$$

Due to the definition of Gamma function, we have

$$\begin{aligned}
 \frac{|\mathcal{T}_\psi \varphi(t) - \mathcal{T}_\psi \hat{\varphi}(t)|}{\exp(\delta t)} &\leq \frac{L_1}{\delta^\alpha} \|\varphi - \hat{\varphi}\|_\delta + \frac{L_2 \lambda \Gamma(\beta)}{\exp(\delta t) \delta^\beta \Gamma(\alpha)} \int_0^t (t-s)^{\alpha-1} \exp(\delta s) ds \\
 &\leq \frac{L_1}{\delta^\alpha} \|\varphi - \hat{\varphi}\|_\delta + \frac{L_2 \lambda \Gamma(\beta)}{\delta^{\alpha+\beta}} \|\varphi - \hat{\varphi}\|_\delta \\
 &= \left( \frac{L_1}{\delta^\alpha} + \frac{L_2 \lambda \Gamma(\beta)}{\delta^{\alpha+\beta}} \right) \|\varphi - \hat{\varphi}\|_\delta.
 \end{aligned}$$

Then from the above inequality and (2.3), it follows

$$\|\mathcal{T}_\psi(\varphi) - \mathcal{T}_\psi(\hat{\varphi})\|_\delta \leq \left( \frac{L_1}{\delta^\alpha} + \frac{L_2 \lambda \Gamma(\beta)}{\delta^{\alpha+\beta}} \right) \|\varphi - \hat{\varphi}\|_\delta < \|\varphi - \hat{\varphi}\|_\delta.$$

Therefore the operator  $\mathcal{T}_\psi$  is a contraction on  $\Omega_\zeta$ . Finally, by the Banach fixed point principle, the proof of the theorem is complete.

We present the following theorem to dig into the asymptotic behavior of the solution of (1.1) through its series representation near the origin.

**Theorem 2** (Smoothness) *Suppose that H1 and H2 are satisfied. Then the solution of (1.1) has the following series form in the neighborhood of the initial point*

$$y(t) = \psi(t) + \sum_{\mu=\alpha b}^{\infty} \bar{y}_\mu t^{\frac{\mu}{b}}, \tag{2.4}$$

in which  $\bar{y}_\mu$  are known coefficients.

**Proof** Consider the series expansion of  $y(t)$  as

$$y(t) = \sum_{\mu=0}^{\infty} \bar{y}_\mu t^{\frac{\mu}{b}}. \tag{2.5}$$

The unknown coefficients  $\bar{y}_\mu$  are obtained in order that the series (2.5) converges and solves (1.1). We use the series expansions of  $\bar{f}$  and  $\bar{g}$  around  $(0, y_0^{(0)})$  and  $(0, 0, y_0^{(0)})$ ,

respectively, viz.,

$$\begin{aligned}
 f(t, y(t)) &= \bar{f}(t^{1/b}, y(t)) = \sum_{\substack{\mu=0 \\ \theta=0}}^{\infty} f_{\mu,\theta} t^{\frac{\mu}{b}} \left( y(t) - y_0^{(0)} \right)^{\theta}, \\
 g(t, s, y(s)) &= \bar{g}(t^{1/b}, s^{1/b}, y(s)) = \sum_{\substack{\mu,v=0 \\ \theta=0}}^{\infty} g_{\mu,v,\theta} t^{\frac{\mu}{b}} s^{\frac{v}{b}} \left( y(s) - y_0^{(0)} \right)^{\theta}. \tag{2.6}
 \end{aligned}$$

By rearranging, it can be concluded that

$$\left( y(t) - y_0^{(0)} \right)^{\theta} = \left( \sum_{\mu=1}^{\infty} \bar{y}_{\mu} t^{\frac{\mu}{b}} \right)^{\theta} = \sum_{\sigma=0}^{\infty} C_{\sigma}^{\theta} t^{\frac{\sigma}{b}}, \tag{2.7}$$

where

$$C_{\sigma}^{\theta} = \begin{cases} 1, & \theta = 0, \sigma = 0, \\ 0, & \theta = 0, \sigma \geq 1, \\ \sum_{\mu_1+\dots+\mu_{\theta}=\sigma} \bar{y}_{\mu_1} \dots \bar{y}_{\mu_{\theta}}, & \theta \neq 0, \sigma \geq 1. \end{cases}$$

Therefore, in view of the equivalent equation (2.2) and substituting the relation (2.7) into (2.6), we find

$$\begin{aligned}
 \sum_{\mu=0}^{\infty} \bar{y}_{\mu} t^{\frac{\mu}{b}} &= \psi(t) + I^{\alpha} \left( \sum_{\substack{\mu=0 \\ \theta=0}}^{\infty} f_{\mu,\theta} t^{\frac{\mu}{b}} \left( \sum_{\sigma=0}^{\infty} C_{\sigma}^{\theta} t^{\frac{\sigma}{b}} \right) \right) \\
 &+ \lambda I^{\alpha} \left( \int_0^t (t-s)^{\beta-1} \sum_{\substack{\mu,v=0 \\ \theta=0}}^{\infty} g_{\mu,v,\theta} t^{\frac{\mu}{b}} s^{\frac{v}{b}} \left( \sum_{\sigma=0}^{\infty} C_{\sigma}^{\theta} s^{\frac{\sigma}{b}} \right) ds \right).
 \end{aligned}$$

Meanwhile, by considering uniform convergence, the coefficients  $\bar{y}_{\mu}$  satisfy the following equality

$$\begin{aligned}
 \sum_{\mu=0}^{\infty} \bar{y}_{\mu} t^{\frac{\mu}{b}} &= \psi(t) + \varrho_1 \left( \sum_{\substack{\mu=0 \\ \theta,\sigma=0}}^{\infty} f_{\mu,\theta} C_{\sigma}^{\theta} t^{\alpha+\frac{\mu+\sigma}{b}} \right) \\
 &+ \lambda \varrho_2 \left( \sum_{\substack{\mu,v=0 \\ \theta,\sigma=0}}^{\infty} g_{\mu,v,\theta} C_{\sigma}^{\theta} t^{\alpha+\beta+\frac{\mu+v+\sigma}{b}} \right), \tag{2.8}
 \end{aligned}$$

in which

$$q_1 = \frac{\Gamma(\frac{\mu+\sigma}{b} + 1)}{\Gamma(\frac{\mu+\sigma}{b} + \alpha + 1)}, \quad q_2 = \frac{\Gamma(\beta)\Gamma(\frac{v+\sigma}{b} + 1)\Gamma(\frac{\mu+v+\sigma}{b} + \beta + 1)}{\Gamma(\frac{v+\sigma}{b} + \beta + 1)\Gamma(\frac{\mu+v+\sigma}{b} + \alpha + \beta + 1)}.$$

Setting  $\mu = \mu - \sigma - \alpha b$  and  $\mu = \mu - v - \sigma - \alpha b - \beta b$  into the first and second series on the right-hand side of (2.8) respectively deduces

$$\sum_{\mu=0}^{\infty} \bar{y}_{\mu} t^{\frac{\mu}{b}} = \psi(t) + \bar{q}_1 \left( \sum_{\substack{\mu=\sigma+\alpha b \\ \theta,\sigma=0}}^{\infty} f_{\mu-\sigma-\alpha b,\theta} C_{\sigma}^{\theta} t^{\frac{\mu}{b}} \right) + \lambda \bar{q}_2 \left( \sum_{\substack{\mu=v+\sigma+\alpha b+\beta b \\ v,\theta,\sigma=0}}^{\infty} g_{\mu-v-\sigma-\alpha b-\beta b,v,\theta} C_{\sigma}^{\theta} t^{\frac{\mu}{b}} \right), \tag{2.9}$$

where

$$\bar{q}_1 = \frac{\Gamma(\frac{\mu}{b} - \alpha + 1)}{\Gamma(\frac{\mu}{b} + 1)}, \quad \bar{q}_2 = \frac{\Gamma(\beta)\Gamma(\frac{v+\sigma}{b} + 1)\Gamma(\frac{\mu}{b} - \alpha + 1)}{\Gamma(\frac{v+\sigma}{b} + \beta + 1)\Gamma(\frac{\mu}{b} + 1)}.$$

We now arrive at  $\bar{y}_{\mu}$  through making a comparison between the coefficients of  $t^{\frac{\mu}{b}}$  on both sides of (2.9). Evidently, for  $\mu < \alpha b$ , we can write

$$\bar{y}_{\mu} = \begin{cases} \frac{y_0}{(\frac{\mu}{b})!}, & \mu = 0, b, \dots, (\lceil \alpha \rceil - 1)b, \\ 0, & \text{else,} \end{cases}$$

and for  $\mu \geq \alpha b$ , the following recursive relation is derived

$$\bar{y}_{\mu} = \sum_{v,\theta,\sigma=0}^{\infty} C_{\sigma}^{\theta} \left( \bar{q}_1 f_{\mu-\sigma-\alpha b,\theta} + \lambda \bar{q}_2 g_{\mu-v-\sigma-\alpha b-\beta b,v,\theta} \right),$$

such that the coefficients  $f_{\mu-\sigma-\alpha b,\theta}$  and  $g_{\mu-v-\sigma-\alpha b-\beta b,v,\theta}$ , equipped with negative indices, are deemed to be zero, and in the case of  $\mu - \sigma - \alpha b \geq 0$  and  $\mu - v - \sigma - \alpha b - \beta b \geq 0$ , we observe that

$$\begin{aligned} \mu &\geq \sigma + \alpha b > \sigma = \mu_1 + \dots + \mu_{\theta} \geq \mu_i, \\ \mu &\geq v + \sigma + \alpha b + \beta b > \sigma = \mu_1 + \dots + \mu_{\theta} \geq \mu_i, \quad i = 1, 2, \dots, \theta. \end{aligned}$$

In other words,  $\bar{y}_{\mu}$  are obtained recursively, and thereby the series expansion (2.4) is a unique formal solution of (1.1).

We are now required to verify that such power series is uniformly and absolutely convergent in the neighborhood of zero. In this regard, we apply Lindelof’s majorant

approach [6]. Deem the weakly singular Volterra integral equation

$$Y(t) = \bar{\psi}(t) + I^\alpha F(t, y(t)) + \lambda I^\alpha \left( \int_0^t (t-s)^{\beta-1} G(t, s, Y(s)) ds \right),$$

where  $\bar{\psi}(t) = \sum_{k=0}^{[\alpha]-1} \frac{t^k}{k!} |y_0^{(k)}|$

$$F(t, y(t)) = \bar{F}(t^{1/b}, Y(t)) = \sum_{\substack{\mu=0 \\ \theta=0}}^{\infty} |f_{\mu,\theta}| t^{\frac{\mu}{b}} \left( Y(t) - |y_0^{(0)}| \right)^\theta,$$

$$G(t, s, Y(s)) = \bar{G}(t^{1/b}, s^{1/b}, Y(s)) = \sum_{\substack{\mu,v=0 \\ \theta=0}}^{\infty} |g_{\mu,v,\theta}| t^{\frac{\mu}{b}} s^{\frac{v}{b}} \left( Y(t) - |y_0^{(0)}| \right)^\theta.$$

It is clear that all coefficients of  $Y$  are positive, and it is a majorant for  $y$ . The series expansion  $Y$  may be calculated in precisely the same manner as above. Currently, for some  $\omega > 0$  given in the sequel, we prove that the series  $Y(t)$  converges absolutely over  $[0, \omega]$ . For this purpose, the key is to justify that the finite partial sum of the formal solution  $Y(t)$  i.e.

$$S_{L+1}(t) = \bar{\psi}(t) + \sum_{\mu=\alpha b}^{L+1} \bar{Y}_\mu t^{\frac{\mu}{b}},$$

can be uniformly bounded on  $[0, \omega]$ . Clearly, we have the inequality below

$$S_{L+1}(t) \leq \bar{\psi}(t) + I^\alpha F(t, S_L(t)) + \lambda I^\alpha \left( \int_0^t (t-s)^{\beta-1} G(t, s, S_L(s)) ds \right), \tag{2.10}$$

in accordance with the recursive evaluation of the coefficients. Albeit, all the coefficients  $\bar{Y}_\mu$  with  $\frac{\mu}{b} \leq \frac{(L+1)}{b}$  are removed from both sides by expanding the right-hand side of (2.10), there still exists some extra terms of higher-order in the right-hand side. Considering

$$\begin{aligned} D^1 &= \sum_{k=0}^{[\alpha]-1} \frac{T^k}{k!} |y_0^{(k)}|, \\ D^2 &= \max_{(t,z) \in \Lambda \times [0, 3D^1]} \frac{|F(t, z)|}{\Gamma(\alpha + 1)}, \\ D^3 &= \max_{(t,s,z) \in \Lambda \times \Lambda \times [0, 3D^1]} \frac{\Gamma(\beta) |G(t, s, z)|}{\Gamma(\alpha + \beta + 1)}, \end{aligned}$$



we define

$$\omega = \min \left\{ T, \left[ \frac{D^1}{D^2} \right]^{\frac{1}{\alpha}}, \left[ \frac{D^1}{D^3} \right]^{\frac{1}{\alpha+\beta}} \right\}.$$

At this step, the aim is to show that  $|S_L(t)| \leq 3D^1, t \in [0, \omega]$  by means of mathematical induction on  $L$ . The result is valid for  $L = 0$  because

$$S_0(t) = |y_0^{(0)}| \leq D^1.$$

We regard that it is valid for  $L$ , and proceed to  $L + 1$  as follows

$$\begin{aligned} |S_{L+1}(t)| &= S_{L+1}(t) \\ &\leq \bar{\psi}(t) + I^\alpha F(t, S_L(t)) + \lambda I^\alpha \left( \int_0^t (t-s)^{\beta-1} G(t, s, S_L(s)) ds \right) \\ &\leq \sum_{k=0}^{[\alpha]-1} \frac{\omega^k}{k!} |y_0^{(k)}| + \max_{s \in [0, t]} |F(s, S_L(s))| \frac{t^\alpha}{\Gamma(\alpha + 1)} \\ &\quad + \max_{s \in [0, t]} |G(t, s, S_L(s))| \frac{\Gamma(\beta) t^{\alpha+\beta}}{\Gamma(\alpha + \beta + 1)} \\ &\leq D^1 + \max_{(s, z) \in [0, \omega] \times [0, 3D^1]} |F(s, z)| \frac{\omega^\alpha}{\Gamma(\alpha + 1)} \\ &\quad + \max_{(t, s, z) \in [0, \omega] \times [0, \omega] \times [0, 3D^1]} |G(t, s, z)| \frac{\Gamma(\beta) \omega^{\alpha+\beta}}{\Gamma(\alpha + \beta + 1)} \\ &\leq D^1 + \omega^\alpha D^2 + \omega^{\alpha+\beta} D^3 \leq 3D^1. \end{aligned}$$

This establishes that  $S_{L+1}(t)$  is uniformly bounded on  $[0, \omega]$ . Since all the coefficients  $S_{L+1}(t)$  are positive, it is monotone as well. As a result, in according to the power series structure of  $Y(t)$ , the majorant  $Y(t)$  converges absolutely over  $[0, \omega]$ , and it is uniformly convergent on the compact subsets of  $[0, \omega]$ . Ultimately, Lindelof’s majorant theorem concludes that the series expansion  $y(t)$  inherits the same features. That is why the above exchange of integration and series was legal.

Theorem 2 tells us that  $\partial_t^{[\alpha]} y(t)$  may not be continuous at the initial point. In consequence, the accuracy of the classical spectral methods might be threatened by this difficulty. That is why constructing an exponentially accurate or high-order spectral method is a kind of challenging task. Next section includes such approach to approximate the solution of (1.1) which satisfies the assumptions of Theorems 1 and 2.

### 3 Numerical approach

In this section, the goal is firstly to direct some critical properties of Jacobi polynomials, GJPs, and FGJFs, and present an effective strategy based on an advanced operational Petrov-Galerkin method to approximate the solution of (1.1).

#### 3.1 Jacobi polynomials

The orthogonal relation of classical Jacobi polynomials  $J_n^{(\rho,\eta)}(s)$ ,  $\rho, \eta > -1$  is as follows [28]

$$\int_I J_m^{(\rho,\eta)}(s) J_n^{(\rho,\eta)}(s) w^{(\rho,\eta)}(s) ds = \lambda_n^{(\rho,\eta)} \delta_{mn}, \quad m, n \geq 0,$$

in which  $w^{(\rho,\eta)}(s) = (1 - s)^\rho (1 + s)^\eta$ ,

$$\lambda_n^{(\rho,\eta)} = \|J_n^{(\rho,\eta)}\|_{w^{(\rho,\eta)}}^2 = \frac{2^{\rho+\eta+1} \Gamma(n + \rho + 1) \Gamma(n + \eta + 1)}{(2n + \rho + \eta + 1) n! \Gamma(n + \rho + \eta + 1)},$$

and  $\delta_{mn}$  directs the Kronecker delta function. Jacobi polynomials satisfy the Rodrigues formula below

$$w^{(\rho,\eta)}(s) J_n^{(\rho,\eta)}(s) = \frac{(-1)^n}{2^n n!} \frac{d^n}{ds^n} \{ (1 - s)^{n+\rho} (1 + s)^{n+\eta} \}.$$

#### 3.2 Generalized Jacobi polynomials

Let us first define  $\hat{\rho}, \hat{\eta}$  and  $\tilde{\rho}, \tilde{\eta}$  from  $\rho, \eta$  as follows

$$\hat{\rho} = \begin{cases} -\rho, & \rho \leq -1, \\ 0, & \rho > -1, \end{cases} \quad \tilde{\rho} = \begin{cases} -\rho, & \rho \leq -1, \\ \rho, & \rho > -1, \end{cases}$$

likewise for  $\hat{\eta}$  and  $\tilde{\eta}$ .

For each  $\rho, \eta \in \mathbb{Z}$ , the GJPs eliminating the restriction  $\rho, \eta > -1$  are defined by [12]

$$\bar{J}_n^{(\rho,\eta)}(s) = (1 - s)^{\hat{\rho}} (1 + s)^{\hat{\eta}} J_{\tilde{n}}^{(\tilde{\rho},\tilde{\eta})}(s), \quad \tilde{n} = n - \kappa_{\rho,\eta} \geq 0, \quad \kappa_{\rho,\eta} = \hat{\rho} + \hat{\eta}.$$

The GJPs are mutually  $L^2_{w^{(\rho,\eta)}}(I)$ -orthogonal, i.e.,

$$\int_I \bar{J}_m^{(\rho,\eta)}(s) \bar{J}_n^{(\rho,\eta)}(s) w^{(\rho,\eta)}(s) ds = \lambda_{\tilde{n}}^{(\tilde{\rho},\tilde{\eta})} \delta_{mn}, \quad m, n \geq \kappa_{\rho,\eta}.$$

The following relations indicate outstanding properties of the GJPs

$$\partial_s^i \bar{J}_n^{(\rho,\eta)}(1) = 0, \quad i = 0, 1, \dots, \hat{\rho} - 1, \quad \rho \leq -1, \quad \eta > -1,$$

$$\partial_s^j \bar{J}_n^{(\rho, \eta)}(-1) = 0, \quad j = 0, 1, \dots, \hat{\eta} - 1, \quad \rho > -1, \eta \leq -1,$$

and for  $(\rho, \eta) \leq -1$ , we have

$$\begin{aligned} \partial_s^i \bar{J}_n^{(\rho, \eta)}(1) &= 0, & i &= 0, 1, \dots, \hat{\rho} - 1, \\ \partial_s^j \bar{J}_n^{(\rho, \eta)}(-1) &= 0, & j &= 0, 1, \dots, \hat{\eta} - 1. \end{aligned}$$

This feature invokes the GJPs toward appropriate basis functions for the Galerkin and Petrov-Galerkin approximations of smooth solutions of functional differential equations associated with the following boundary conditions

$$\begin{aligned} \partial_s^i P(1) &= 0, & i &= 0, 1, \dots, \hat{\rho} - 1, \\ \partial_s^j P(-1) &= 0, & j &= 0, 1, \dots, \hat{\eta} - 1. \end{aligned}$$

### 3.3 The fractional generalized Jacobi functions

Faghieh and Mokhtary [8] introduced the fractional generalized Jacobi functions which have outstanding approximate properties to the functions owning singularity at boundaries.

Suppose that  $\gamma \in (0, 1]$  and  $t \in \Lambda$ , the essence of the FGJFs  $\bar{J}_n^{(\rho, \eta, \gamma)}(t)$  comes from the GJPs by means of the coordinate transformation  $s = 2\left(\frac{t}{T}\right)^\gamma - 1$  as [8]

$$\bar{J}_n^{(\rho, \eta, \gamma)}(t) = \bar{J}_n^{(\rho, \eta)}\left(2\left(\frac{t}{T}\right)^\gamma - 1\right) = \frac{2^{\hat{\rho} + \hat{\eta}}}{T^{\gamma(\hat{\rho} + \hat{\eta})}} \left(T^\gamma - t^\gamma\right)^{\hat{\rho}} t^{\gamma \hat{\eta}} J_{\hat{n}}^{(\hat{\rho}, \hat{\eta}, \gamma)}(t), \tag{3.1}$$

in which  $\rho, \eta \in \mathbb{Z}, n \geq \kappa_{\rho, \eta}$  and

$$J_{\hat{n}}^{(\hat{\rho}, \hat{\eta}, \gamma)}(t) = J_{\hat{n}}^{(\hat{\rho}, \hat{\eta})}\left(2\left(\frac{t}{T}\right)^\gamma - 1\right),$$

indicates the fractional Jacobi function (FJF) satisfying the explicit formula [11, 33]

$$J_{\hat{n}}^{(\hat{\rho}, \hat{\eta}, \gamma)}(t) = \sum_{j=0}^{\hat{n}} \frac{1}{T^{\gamma j}} \Upsilon_j^{(\hat{\rho}, \hat{\eta}, \hat{n})} t^{\gamma j} = \text{Span}\{1, t^\gamma, \dots, t^{\hat{n}\gamma}\},$$

where

$$\Upsilon_j^{(\hat{\rho}, \hat{\eta}, \hat{n})} = \frac{(-1)^{\hat{n}-j} \Gamma(\hat{n} + \hat{\eta} + 1) \Gamma(\hat{n} + \hat{\rho} + \hat{\eta} + j + 1)}{\Gamma(\hat{\eta} + j + 1) j! \Gamma(\hat{n} + \hat{\rho} + \hat{\eta} + 1) (\hat{n} - j)!}.$$

It can be inferred that the FJFs are  $L^2_{w^{(\tilde{\rho}, \tilde{\eta}, \gamma)}}(\Lambda)$ -orthogonal along with the weight function  $w^{(\tilde{\rho}, \tilde{\eta}, \gamma)}(t) = \frac{\gamma}{T\gamma(\tilde{\rho} + \tilde{\eta} + 1) - 1} \left( T^\gamma - t^\gamma \right)^{\tilde{\rho}} t^{\gamma(\tilde{\eta} + 1) - 1}$ , i.e., we have

$$\int_{\Lambda} J_{\tilde{m}}^{(\tilde{\rho}, \tilde{\eta}, \gamma)}(t) J_{\tilde{n}}^{(\tilde{\rho}, \tilde{\eta}, \gamma)}(t) w^{(\tilde{\rho}, \tilde{\eta}, \gamma)}(t) dt = \lambda_{\tilde{n}}^{(\tilde{\rho}, \tilde{\eta}, \gamma)} \delta_{\tilde{m}\tilde{n}},$$

in which

$$\lambda_{\tilde{n}}^{(\tilde{\rho}, \tilde{\eta}, \gamma)} = \|J_{\tilde{n}}^{(\tilde{\rho}, \tilde{\eta}, \gamma)}\|_{w^{(\tilde{\rho}, \tilde{\eta}, \gamma)}}^2 = \frac{T}{2\tilde{\rho} + \tilde{\eta} + 1} \lambda_{\tilde{n}}^{(\tilde{\rho}, \tilde{\eta})}.$$

The FGJFs are mutually orthogonal along with the following orthogonal relation

$$\int_{\Lambda} \bar{J}_m^{(\alpha, \beta, \gamma)}(t) \bar{J}_n^{(\alpha, \beta, \gamma)}(t) w^{(\alpha, \beta, \gamma)}(t) dt = \frac{T}{2\alpha + \beta + 1} \lambda_{\tilde{n}}^{(\tilde{\alpha}, \tilde{\beta})} \delta_{mn}.$$

The following relations hold

$$\begin{aligned} \partial_t^i \bar{J}_n^{(\rho, \eta, \gamma)}(T) &= 0, & i &= 0, 1, \dots, \hat{\rho} - 1, & \rho &\leq -1, \eta > -1, \\ \partial_t^j \bar{J}_n^{(\rho, \eta, \gamma)}(0) &= 0, & j &= 0, 1, \dots, \lceil \gamma \hat{\eta} \rceil - 1, & \rho &> -1, \eta \leq -1, \end{aligned}$$

and for  $\rho, \eta \leq -1$ , we have

$$\begin{aligned} \partial_t^i \bar{J}_n^{(\rho, \eta, \gamma)}(T) &= 0, & i &= 0, 1, \dots, \hat{\rho} - 1, \\ \partial_t^j \bar{J}_n^{(\rho, \eta, \gamma)}(0) &= 0, & j &= 0, 1, \dots, \lceil \gamma \hat{\eta} \rceil - 1, \end{aligned}$$

which is one of the remarkable features of the FGJFs.

### 3.4 Operational Petrov-Galerkin method

This section offers a highly accurate operational Petrov-Galerkin method based on the FGJFs to approximate the solution of (1.1). Solvability analysis of the relevant non-linear algebraic system is also provided through a sequence of matrix operations. Inserting  $\rho = 0, \eta = -\alpha b$  and  $\gamma = \frac{1}{b}$  in (3.1), the FGJFs  $\{\bar{J}_k^{(0, -\alpha b, \gamma)}(t)\}_{k \geq \alpha b}$

$$\bar{J}_k^{(0, -\alpha b, \gamma)}(t) = \frac{2^{\alpha b}}{T^\alpha} t^\alpha J_{k-\alpha b}^{(0, \alpha b, \gamma)}(t) = \bar{J}_k^{(0, -\alpha b, \gamma)}(t) = \text{Span}\{t^\alpha, t^{\alpha+\gamma}, \dots, t^{k\gamma}\}, \tag{3.2}$$

satisfy the initial conditions of the equation (1.1). Thereby they are capable of serving as basis functions to obtain the Petrov-Galerkin approximation of (1.1).

Based on (3.2), we arrive at  $\underline{J}^{(0, -\alpha b, \gamma)} = \underline{J} T_t$ , where

$$\underline{J}^{(0, -\alpha b, \gamma)}(t) = [\bar{J}_{\alpha b}^{(0, -\alpha b, \gamma)}(t), \bar{J}_{\alpha b+1}^{(0, -\alpha b, \gamma)}(t), \dots, \bar{J}_N^{(0, -\alpha b, \gamma)}(t), \dots]',$$

indicating the vector of FGJFs with degree  $(\bar{J}_k^{(0,-\alpha b,\gamma)}(t)) \leq k\gamma$ ,  $J$  is a lower-triangular matrix of order infinity, and  $\underline{T}_t = [t^\alpha, t^{\alpha+\gamma}, \dots, t^{N\gamma}, \dots]'$ . At this stage, the Petrov-Galerkin approximation  $y_N(t)$  of the exact solution  $y(t)$  is stated as

$$y_N(t) = \sum_{k=0}^{\infty} c_k \bar{J}_{k+\alpha b}^{(0,-\alpha b,\gamma)}(t) = \underline{c} \underline{J}^{(0,-\alpha b,\gamma)}(t) = \underline{c} J \underline{T}_t = \underline{c} J \Psi \hat{\underline{T}}_t, \tag{3.3}$$

where  $\underline{c} = [c_0, c_1, \dots, c_{\hat{N}}, 0, \dots]$ ,  $\hat{\underline{T}}_t = [1, t^\gamma, \dots, t^{\hat{N}\gamma}, \dots]'$ ,  $\hat{N} = N - \alpha b$ , and

$$\Psi = \begin{bmatrix} \overbrace{0 \dots 0}^{\alpha b} & 1 & 0 & \dots \\ \vdots & 0 & 1 & 0 & \dots \\ \vdots & \vdots & 0 & 1 & 0 & \dots \\ \dots & \dots & \ddots & \ddots & \ddots & \ddots \end{bmatrix}. \tag{3.4}$$

Meanwhile, the relation (2.6) can be restated as

$$\begin{aligned} f(t, y(t)) &= \sum_{\substack{\mu=0 \\ \theta=0}}^{\infty} f_{\mu,\theta} t^{\mu\gamma} y^\theta(t), \\ g(t, s, y(s)) &= \sum_{\substack{\mu,v=0 \\ \theta=0}}^{\infty} g_{\mu,v,\theta} t^{\mu\gamma} s^{v\gamma} y^\theta(s). \end{aligned} \tag{3.5}$$

Inserting the relation (3.3) into (1.1) and using (3.5), we have

$$\underline{c} J D_C^\alpha \underline{T}_t = \sum_{\substack{\mu=0 \\ \theta=0}}^{\infty} f_{\mu,\theta} t^{\mu\gamma} y_N^\theta(t) + \lambda \sum_{\substack{\mu,v=0 \\ \theta=0}}^{\infty} g_{\mu,v,\theta} t^{\mu\gamma} \int_0^t (t-s)^{\beta-1} s^{v\gamma} y_N^\theta(s) ds. \tag{3.6}$$

Hence, we first compute  $D_C^\alpha \underline{T}_t$ . In this respect, using the relation [6, 14, 24]

$$D_C^\alpha t^\tau = \frac{\Gamma(\tau + 1)}{\Gamma(\tau - \alpha + 1)} t^{\tau-\alpha}, \quad \tau > 0,$$

we can write

$$D_C^\alpha \underline{T}_t = \left[ D_C^\alpha t^{\alpha+\gamma i} \right]_{i \geq 0} = \left[ \frac{\Gamma(\alpha + \gamma i + 1)}{\Gamma(\gamma i + 1)} t^{\gamma i} \right]_{i \geq 0} = \chi \hat{\underline{T}}_t, \tag{3.7}$$

in which

$$\chi = \begin{bmatrix} \Gamma(\alpha + 1) & 0 & 0 & \dots \\ 0 & \frac{\Gamma(\alpha+\gamma+1)}{\Gamma(\gamma+1)} & 0 & \dots \\ 0 & 0 & \frac{\Gamma(\alpha+2\gamma+1)}{\Gamma(2\gamma+1)} & \dots \\ \vdots & \vdots & 0 & \ddots \end{bmatrix}.$$

The key of this strategy is to attain a matrix form for the right-hand side of (3.6) vigorously. In this sense, we first substitute an appropriate matrix representation for  $y_N^\theta(t)$  through the following lemma.

**Lemma 1** Assuming  $\underline{c} = \underline{c}J = [\underline{c}_0, \underline{c}_1, \dots, \underline{c}_{\hat{N}}, 0, \dots]$ , the following relation holds

$$y_N^\theta(t) = \underline{c} \Psi Q^{\theta-1} \hat{T}_t, \quad \theta > 0,$$

where  $Q$  indicates the following upper-triangular matrix of order infinity

$$Q = \begin{bmatrix} \underline{c}\Psi_0 & \underline{c}\Psi_1 & \underline{c}\Psi_2 & \dots \\ 0 & \underline{c}\Psi_0 & \underline{c}\Psi_1 & \dots \\ 0 & 0 & \underline{c}\Psi_0 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix},$$

with  $\Psi_r = \{\Psi_{m,r}\}_{m=0}^\infty, r = 0, 1, \dots$ .

**Proof** The mathematical induction on  $\theta$  is utilized to prove this lemma. For  $\theta = 1$ , the result is obviously valid. We consider that it is valid for  $\theta$ , and proceed to  $\theta + 1$  as follows

$$\begin{aligned} y_N^{\theta+1}(t) &= y_N^\theta(t) \times y_N(t) = (\underline{c} \Psi Q^{\theta-1} \hat{T}_t) \times (\underline{c} \Psi \hat{T}_t) \\ &= \underline{c} \Psi Q^{\theta-1} (\hat{T}_t \times (\underline{c} \Psi \hat{T}_t)). \end{aligned} \tag{3.8}$$

Next, it suffices to demonstrate that

$$\hat{T}_t \times (\underline{c} \Psi \hat{T}_t) = Q \hat{T}_t. \tag{3.9}$$

For this purpose, we can derive

$$\begin{aligned} \hat{T}_t \times (\underline{c} \Psi \hat{T}_t) &= \hat{T}_t \times \left( \sum_{h=0}^\infty \sum_{k=0}^\infty \underline{c}_k \Psi_{k,h} t^{h\gamma} \right) = \left[ \sum_{h=0}^\infty \sum_{k=0}^\infty \underline{c}_k \Psi_{k,h} t^{(h+m)\gamma} \right]_{m=0}^\infty \\ &= \left[ \sum_{r=m}^\infty \left( \sum_{k=0}^\infty \underline{c}_k \Psi_{k,r-m} t^{r\gamma} \right) \right]_{m=0}^\infty, \end{aligned}$$

whereby it yields (3.9). Trivially, by inserting (3.9) into (3.8) the desired result can be achieved.

Now, it is time to employ Lemma 1 to convert the first term of the right-hand side of (3.6) into the following vector-matrix form

$$\begin{aligned}
 \sum_{\substack{\mu=0 \\ \theta=0}}^{\infty} f_{\mu,\theta} t^{\mu\gamma} y_N^\theta(t) &= \sum_{\mu=0}^{\infty} f_{\mu,0} t^{\mu\gamma} + \sum_{\substack{\mu=0 \\ \theta=1}}^{\infty} f_{\mu,\theta} t^{\mu\gamma} y_N^\theta(t) \\
 &= \underline{f}_0 \hat{T}_t + \sum_{\substack{\mu=0 \\ \theta=1}}^{\infty} f_{\mu,\theta} t^{\mu\gamma} \underline{c} \Psi Q^{\theta-1} \hat{T}_t \\
 &= \underline{f}_0 \hat{T}_t + \sum_{\theta=1}^{\infty} \underline{c} \Psi Q^{\theta-1} \left( \sum_{\mu=0}^{\infty} f_{\mu,\theta} \left[ t^{(\mu+m)\gamma} \right]_{m=0}^{\infty} \right) \\
 &= \left( \underline{f}_0 + \sum_{\theta=1}^{\infty} \underline{c} \Psi Q^{\theta-1} F_\theta \right) \hat{T}_t, \tag{3.10}
 \end{aligned}$$

in which  $\underline{f}_0 = [f_{0,0}, f_{1,0}, \dots, f_{N,0}, \dots]$  and  $F_\theta$  directs an upper-triangular matrix of order infinity with the following components

$$[F_\theta]_{i,j=0}^{\infty} = \begin{cases} 0, & i > j, \\ f_{j-i,\theta}, & i \leq j. \end{cases}$$

Applying the same strategy to the second term of (3.6) concludes

$$\begin{aligned}
 &\sum_{\substack{\mu,v=0 \\ \theta=0}}^{\infty} g_{\mu,v,\theta} t^{\mu\gamma} \int_0^t (t-s)^{\beta-1} s^{v\gamma} y_N^\theta(s) ds \\
 &= \sum_{\mu,v=0}^{\infty} g_{\mu,v,0} t^{\mu\gamma} \int_0^t (t-s)^{\beta-1} s^{v\gamma} ds \\
 &\quad + \sum_{\substack{\mu,v=0 \\ \theta=1}}^{\infty} g_{\mu,v,\theta} t^{\mu\gamma} \int_0^t (t-s)^{\beta-1} s^{v\gamma} \underline{c} \Psi Q^{\theta-1} \hat{T}_s ds \\
 &= \sum_{\mu,v=0}^{\infty} g_{\mu,v,0} t^{\mu\gamma} \int_0^t (t-s)^{\beta-1} s^{v\gamma} ds + \sum_{\theta=1}^{\infty} \underline{c} \Psi Q^{\theta-1} \\
 &\quad \left( \sum_{\mu,v=0}^{\infty} g_{\mu,v,\theta} t^{\mu\gamma} \left[ \int_0^t (t-s)^{\beta-1} s^{(v+m)\gamma} ds \right]_{m=0}^{\infty} \right). \tag{3.11}
 \end{aligned}$$

Evidently, we have [6]

$$\left[ \int_0^t (t-s)^{\beta-1} s^{(v+m)\gamma} ds \right]_{m=0}^\infty = \left[ \mathcal{A}_v^m t^{(v+m)\gamma+\beta} \right]_{m=0}^\infty,$$

$$\int_0^t (t-s)^{\beta-1} s^{v\gamma} ds = \mathcal{A}_v^0 t^{v\gamma+\beta},$$

in which  $\mathcal{A}_v^m = \frac{\Gamma(\beta)\Gamma((v+m)\gamma+1)}{\Gamma((v+m)\gamma+\beta+1)}$ . Therefore, the equality (3.11) can be rewritten as

$$\sum_{\substack{\mu, v=0 \\ \theta=0}}^\infty g_{\mu, v, \theta} t^{\mu\gamma} \int_0^t (t-s)^{\beta-1} s^{v\gamma} y_N^\theta(s) ds = \sum_{\mu, v=0}^\infty g_{\mu, v, 0} \mathcal{A}_v^0 t^{(\mu+v)\gamma+\beta}$$

$$+ \sum_{\theta=1}^\infty \underline{c} \Psi Q^{\theta-1} \left( \sum_{\mu, v=0}^\infty g_{\mu, v, \theta} \left[ \mathcal{A}_v^m t^{(\mu+v+m)\gamma+\beta} \right]_{m=0}^\infty \right)$$

$$= \left( \underline{K} + \sum_{\theta=1}^\infty \underline{c} \Psi Q^{\theta-1} H_\theta \right) \hat{T}_t, \tag{3.12}$$

along with the infinite-order row vector  $\underline{K}$  as

$$[\underline{K}]_{j=0}^\infty = \begin{cases} 0, & j < \beta b, \\ \sum_{v=0}^{j-\beta b} g_{v, j-\beta b-v, 0} \mathcal{A}_{j-\beta b-v}^0, & j \geq \beta b, \end{cases}$$

$$= \underbrace{[0 \dots 0]_{\beta b}}_{\beta b}, [\underline{K}]_{\beta b}, [\underline{K}]_{\beta b+1}, \dots, [\underline{K}]_{\hat{N}}, \dots,$$

and the infinite-order upper-triangular matrix  $H_\theta$  with the components

$$[H_\theta]_{i, j=0}^\infty = \begin{cases} 0, & i \geq j - \beta b + 1, \\ \sum_{v=0}^{j-i-\beta b} g_{v, j-i-\beta b-v, \theta} \mathcal{A}_{j-i-\beta b-v}^i, & i < j - \beta b + 1. \end{cases}$$

At this stage, we employ the relations (3.7), (3.10) and (3.12) in (3.6), and thereby we derive

$$\underline{c} \chi \hat{T}_t = \left( \underline{f}_0 + \sum_{\theta=1}^\infty \underline{c} \Psi Q^{\theta-1} F_\theta \right) \hat{T}_t + \lambda \left( \underline{K} + \sum_{\theta=1}^\infty \underline{c} \Psi Q^{\theta-1} H_\theta \right) \hat{T}_t. \tag{3.13}$$

Due to the orthogonality of  $\{J_k^{(0, \alpha b, \gamma)}\}_{k \geq 0}$ , we project (3.13) onto  $\{J_k^{(0, \alpha b, \gamma)}\}_{k=0}^{\hat{N}}$ . Ultimately, after some simple manipulations, the algebraic form of the Petrov-Galerkin



discretization is obtained as

$$\underline{c}^{\hat{N}} \chi^{\hat{N}} = \underline{f}_0^{\hat{N}} + \lambda \underline{K}^{\hat{N}} + \sum_{\theta=1}^{\infty} \underline{c}^{\hat{N}} \Psi^{\hat{N}} (Q^{\theta-1})^{\hat{N}} (F_{\theta}^{\hat{N}} + \lambda H_{\theta}^{\hat{N}}). \tag{3.14}$$

Here, the sign  $\hat{N}$  above the matrices and vectors signifies respectively the principle sub-matrices and sub-vectors of order  $\hat{N} + 1$ . Needless to mention, we are able to get the unknown vector through solving the system of  $\hat{N} + 1$  non-linear algebraic equations (3.14). The next section presents an outstanding strategy to overcome this system regardless of its complexity.

### 3.5 Solvability analysis

It is noteworthy that the system of non-linear algebraic equations (3.14) involves high computational costs to be solved, specifically for large degrees of approximation, and it can undoubtedly result in inaccurate solution due to its complexity. In order to cope with this barrier, we aim to provide a productive and well-conditioned implementation that gives the unknown of (3.14) by means of some recurrence relations instead of solving a complex non-linear algebraic system. For this purpose, applying Lemma 1 enables us to write

$$Q = \begin{bmatrix} \underline{c}\Psi_0 & \underline{c}\Psi_1 & \underline{c}\Psi_2 & \dots \\ 0 & \underline{c}\Psi_0 & \underline{c}\Psi_1 & \dots \\ 0 & 0 & \underline{c}\Psi_0 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix} = \begin{bmatrix} \overbrace{0 \dots 0}^{ab} & \underline{c}_0 & \underline{c}_1 & \underline{c}_2 & \dots \\ \vdots & 0 & \underline{c}_0 & \underline{c}_1 & \dots \\ \vdots & \vdots & 0 & \underline{c}_0 & \dots \\ \dots & \dots & \ddots & \ddots & \ddots \end{bmatrix}.$$

Through simple calculations, we observe that  $(Q^{\theta-1})^{\hat{N}}$  has the following upper-triangular structure

$$(Q^{\theta-1})^{\hat{N}} = \begin{bmatrix} \overbrace{0 \dots 0}^{(\theta-1)ab} & (\underline{c}_0)^{\theta-1} & (\theta-1) (\underline{c}_0)^{\theta-2} \underline{c}_1 & \dots & \dots \\ \vdots & 0 & (\underline{c}_0)^{\theta-1} & (\theta-1) (\underline{c}_0)^{\theta-2} \underline{c}_1 & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \\ 0 & 0 & \ddots & 0 & (\underline{c}_0)^{\theta-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} \overbrace{0 \dots 0}^{(\theta-1)\alpha b} & Q_{0,0}^{\theta-1} & Q_{0,1}^{\theta-1} & Q_{0,2}^{\theta-1} & \dots \\ \vdots & 0 & Q_{0,0}^{\theta-1} & Q_{0,1}^{\theta-1} & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \\ 0 & 0 & \ddots & 0 & Q_{0,0}^{\theta-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix},$$

where  $Q_{0,r}^{\theta-1}$ ,  $r = 0, 1, \dots, N - \theta\alpha b$ , are non-linear functions of the elements  $c_0, c_1, \dots, c_r$ .

In addition, from (3.4) and (3.15), the following upper-triangular matrix of order  $\hat{N} + 1$  can be derived

$$\Psi^{\hat{N}}(Q^{\theta-1})^{\hat{N}} = \begin{bmatrix} \overbrace{0 \dots 0}^{\theta\alpha b} & Q_{0,0}^{\theta-1} & Q_{0,1}^{\theta-1} & Q_{0,2}^{\theta-1} & \dots \\ \vdots & 0 & Q_{0,0}^{\theta-1} & Q_{0,1}^{\theta-1} & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \\ 0 & 0 & \ddots & 0 & Q_{0,0}^{\theta-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

Consequently, one can be checked that the matrix  $\Pi$  defined by

$$\Pi = \Psi^{\hat{N}}(Q^{\theta-1})^{\hat{N}}\mathcal{B}, \quad \mathcal{B} = F_{\theta}^{\hat{N}} + \lambda H_{\theta}^{\hat{N}},$$

has an upper-triangular structure with the components below

$$[\Pi]_{i,j=0}^{\hat{N}} = \begin{cases} 0, & i \geq j - \theta\alpha b + 1, \\ \sum_{r=0}^{j-i-\theta\alpha b} Q_{0,r}^{\theta-1} [\mathcal{B}]_{i+r+\theta\alpha b,j}, & i < j - \theta\alpha b + 1. \end{cases}$$

Afterward, considering the structure of  $\Pi$ , we have

$$\underline{c}^{\hat{N}} \Pi = \underline{c}^{\hat{N}} \begin{bmatrix} \overbrace{0 \dots 0}^{\theta \alpha b} & [\Pi]_{0, \theta \alpha b} & [\Pi]_{0, \theta \alpha b + 1} & \dots & [\Pi]_{0, \hat{N}} \\ \vdots & 0 & [\Pi]_{1, \theta \alpha b + 1} & \dots & [\Pi]_{1, \hat{N}} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \ddots & 0 & [\Pi]_{\hat{N} - \theta \alpha b, \hat{N}} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

Now, due to the above relation, one can deduce

$$\begin{aligned} \left[ \underline{c}^{\hat{N}} \Pi \right]_{j=0}^{\hat{N}} &= \begin{cases} 0, & j < \theta \alpha b, \\ \sum_{i=0}^{j - \theta \alpha b} c_i [\Pi]_{i, j}, & j \geq \theta \alpha b, \end{cases} \\ &= \left[ \overbrace{0 \dots 0}^{\theta \alpha b}, \mathbf{Z}_0^\theta, \mathbf{Z}_1^\theta, \dots, \mathbf{Z}_{\hat{N} - \theta \alpha b}^\theta \right], \end{aligned}$$

where  $\mathbf{Z}_r^\theta, r = 0, 1, \dots, \hat{N} - \theta \alpha b$ , are non-linear functions of the elements  $\underline{c}_0, \underline{c}_1, \dots, \underline{c}_r$ .

Eventually we attain

$$\sum_{\theta=1}^{\infty} \left[ \overbrace{0 \dots 0}^{\theta \alpha b}, \mathbf{Z}_0^\theta, \mathbf{Z}_1^\theta, \dots, \mathbf{Z}_{\hat{N} - \theta \alpha b}^\theta \right] = \left[ \overbrace{0 \dots 0}^{\alpha b}, \tilde{\mathbf{Z}}_0, \tilde{\mathbf{Z}}_1, \dots, \tilde{\mathbf{Z}}_{\hat{N} - \alpha b} \right], \quad (3.15)$$

in which  $\tilde{\mathbf{Z}}_r, r = 0, 1, \dots, \hat{N} - \alpha b$  are non-linear functions in terms of the components  $\underline{c}_0, \underline{c}_1, \dots, \underline{c}_r$ . We substitute (3.15) into (3.14) and compute the unknown elements of the unknown vector  $\underline{c}$  through the recurrence relations below

$$\begin{aligned} \underline{c}_0 &= \frac{1}{\Gamma(\alpha + 1)} (f_{0,0} + \lambda[\underline{K}]_0), \\ &\vdots \\ \underline{c}_{\alpha b - 1} &= \frac{\Gamma(\gamma(\alpha b - 1) + 1)}{\Gamma(\alpha + \gamma(\alpha b - 1) + 1)} (f_{\alpha b - 1, 0} + \lambda[\underline{K}]_{\alpha b - 1}), \\ \underline{c}_{\alpha b} &= \frac{\Gamma(\gamma(\alpha b) + 1)}{\Gamma(\alpha + \gamma(\alpha b) + 1)} \left( f_{\alpha b, 0} + \lambda[\underline{K}]_{\alpha b} + \tilde{\mathbf{Z}}_0 \right), \\ &\vdots \\ \underline{c}_{\hat{N}} &= \frac{\Gamma(\gamma \hat{N} + 1)}{\Gamma(\alpha + \gamma \hat{N} + 1)} \left( f_{\hat{N}, 0} + \lambda[\underline{K}]_{\hat{N}} + \tilde{\mathbf{Z}}_{\hat{N} - \alpha b} \right). \end{aligned}$$

Ultimately, solving the lower-triangular system  $\underline{c}^{\hat{N}} = \underline{c}^{\hat{N}} J^{\hat{N}}$  fuels the main unknown  $\underline{c}^{\hat{N}}$  whereby we can obtain the Petrov-Galerkin approximation (3.3).

### 4 Error estimate

In this section we obtain a convergence result of the method, introduced in previous section, through employing an appropriate error bound in  $L^2$ -norm.

Let us introduce  $\Pi_N^{(0,\alpha b,\gamma)}$  as the  $L^2_{w^{(0,\alpha b,\gamma)}}(\Lambda)$ -orthogonal projection relevant to the fractional Jacobi space

$$\mathcal{F}_N^{(0,\alpha b,\gamma)} = \text{Span}\{J_k^{(0,\alpha b,\gamma)} : k = 0, 1, \dots, N\}.$$

Meanwhile, for  $p \in L^2_{w^{(0,\alpha b,\gamma)}}(\Lambda)$ , we can write

$$\left( p - \Pi_N^{(0,\alpha b,\gamma)} p, \phi \right)_{w^{(0,\alpha b,\gamma)}} = 0, \quad \forall \phi \in \mathcal{F}_N^{(0,\alpha b,\gamma)}.$$

To obtain an upper bound of truncation error  $\Pi_N^{(0,\alpha b,\gamma)} p - p$ , we first define the space

$$H_{w^{(\rho,\eta)}}^m(I) = \{P : \|P\|_{m,w^{(\rho,\eta)}} < \infty, m \in \mathbb{N}\},$$

along with

$$\|P\|_{m,w^{(\rho,\eta)}}^2 = \sum_{l=0}^m \|\partial_s^l P\|_{w^{\rho+l,\eta+l}}^2, \quad |P|_{m,w^{(\rho,\eta)}} = \|\partial_s^m P\|_{w^{\rho+m,\eta+m}},$$

considered as the norm and semi-norm.

In this step, if we assume that the coordinate transformation  $s = 2(\frac{t}{T})^\gamma - 1$  associate the function  $p(t)$  with  $P(s)$ , their derivatives will be connected as follows

$$\begin{aligned} D_t p &:= \partial_s P(s) = \partial_s t \partial_t p, \\ D_t^2 p &:= \partial_s^2 P(s) = \partial_s t \partial_t (D_t p), \\ &\vdots \\ D_t^n p &:= \partial_s^n P(s) = \partial_s t \partial_t (\partial_s t \partial_t (\dots (\partial_s t \partial_t p) \dots)), \end{aligned}$$

in which  $\partial_s t = \frac{T}{2^\gamma} (\frac{t}{T})^{1-\gamma}$ . Moreover, it can be deduced that

$$\begin{aligned} \|P(s)\|_{w^{(\rho,\eta)}}^2 &= \int_I |P(s)|^2 w^{(\rho,\eta)}(s) ds = d^{(\rho,\eta)} \int_\Lambda |p(t)|^2 w^{(\rho,\eta,\gamma)}(t) dt \\ &= d^{(\rho,\eta)} \|p(t)\|_{w^{(\rho,\eta,\gamma)}}^2, \end{aligned}$$

$$\begin{aligned} \|\partial_s^m P(s)\|_{w^{(\rho,\eta)}}^2 &= \int_I |\partial_s^m P(s)|^2 w^{(\rho,\eta)}(s) ds = d^{(\rho,\eta)} \int_\Lambda |D_t^m p(t)|^2 w^{(\rho,\eta,\gamma)}(t) dt \\ &= d^{(\rho,\eta)} \|D_t^m p(t)\|_{w^{(\rho,\eta,\gamma)}}^2, \end{aligned}$$

where  $d^{(\rho,\eta)} = \frac{2^{\rho+\eta+1}}{T}$ .

Finally, we define the transformed space

$$H_{w^{(\rho,\eta,\gamma)}}^m(\Lambda) = \{p : \|p\|_{m,w^{(\rho,\eta,\gamma)}} < \infty\},$$

associated with the norm and semi-norm

$$\begin{aligned} \|p\|_{m,w^{(\rho,\eta,\gamma)}}^2 &= \sum_{l=0}^m d^{(\rho+l,\eta+l)} \|D_t^l p\|_{w^{(\rho+l,\eta+l,\gamma)}}^2, \\ |p|_{m,w^{(\rho,\eta,\gamma)}} &= \sqrt{d^{(\rho+m,\eta+m)}} \|D_t^m p\|_{w^{(\rho+m,\eta+m,\gamma)}}, \end{aligned}$$

and from Theorem 3.2 of [9], the following estimation holds

$$\|\Pi_N^{(0,\alpha b,\gamma)} p - p\|_{w^{(0,\alpha b,\gamma)}} \leq CN^{-m} |p|_{m,w^{(0,\alpha b,\gamma)}}, \quad m \geq 0. \tag{4.1}$$

It is time to exhibit the convergence theorem directing the proper error bound for  $y(t) - y_N(t)$  in  $L^2$ -norm.

**Theorem 3** (Convergence) *Let  $y_N(t)$  given by (3.3) be the approximate solution of (1.1). If we have*

1.  $\int_0^t (t-s)^{\beta-1} g(t,s,y(s)) ds \in H_{w^{(0,\alpha b,\gamma)}}^\varepsilon(\Lambda)$  such that

$$D_t^{\varepsilon+1} \left( \int_0^t (t-s)^{\beta-1} g(t,s,y(s)) ds \right) \in C(\Lambda), \quad \varepsilon \geq 0.$$

2.  $f \in H_{w^{(0,\alpha b,\gamma)}}^\varsigma(\Lambda)$  such that  $D_t^{\varsigma+1} f \in C(\Lambda)$ ,  $\varsigma \geq 0$ .

Then the following upper bound holds for sufficiently large values of  $N$

$$\|e_N(t)\| \leq C \left( \hat{N}^{-\varepsilon} \left| \int_0^t (t-s)^{\beta-1} g(t,s,y(s)) ds \right|_{\varepsilon,w^{(0,\alpha b,\gamma)}} + \hat{N}^{-\varsigma} |f|_{\varsigma,w^{(0,\alpha b,\gamma)}} \right),$$

where  $e_N(t) = y(t) - y_N(t)$  dictates the error function, and  $C > 0$  denotes a generic constant independent of  $N$ .

**Proof** Based on the devised numerical approach in Section 3, we get the following operator equation

$$\Pi_{\hat{N}}^{(0,\alpha b,\gamma)} \left( D_C^\alpha y_N(t) - f(t, y_N(t)) - \lambda \int_0^t (t-s)^{\beta-1} g(t,s,y_N(s)) ds \right) = 0,$$

and equivalently we have

$$D_C^\alpha y_N(t) = \Pi_{\hat{N}}^{(0,\alpha b,\gamma)}(f(t, y_N(t))) + \lambda \Pi_{\hat{N}}^{(0,\alpha b,\gamma)}\left(\int_0^t (t-s)^{\beta-1} g(t, s, y_N(s)) ds\right), \tag{4.2}$$

since we have  $D_C^\alpha y_N(t) \in \text{Span}\{J_0^{(0,\alpha b,\gamma)}, J_1^{(0,\alpha b,\gamma)}, \dots, J_N^{(0,\alpha b,\gamma)}\}$ . Subtracting (4.2) from (1.1) yields

$$D_C^\alpha e_N(t) = f(t, y(t)) - \Pi_{\hat{N}}^{(0,\alpha b,\gamma)}(f(t, y_N(t))) + \lambda \left(\int_0^t (t-s)^{\beta-1} g(t, s, y(s)) ds - \Pi_{\hat{N}}^{(0,\alpha b,\gamma)}\left(\int_0^t (t-s)^{\beta-1} g(t, s, y_N(s)) ds\right)\right),$$

which can be restated as

$$D_C^\alpha e_N(t) = (f - \bar{f}) + \lambda \int_0^t (t-s)^{\beta-1} (g(t, s, y(s)) - g(t, s, y_N(s))) ds + e_{\Pi_{\hat{N}}^{(0,\alpha b,\gamma)}}\left(\bar{f} + \lambda \int_0^t (t-s)^{\beta-1} g(t, s, y_N(s)) ds\right), \tag{4.3}$$

where  $e_{\Pi_{\hat{N}}^{(0,\alpha b,\gamma)}}(z) = z - \Pi_{\hat{N}}^{(0,\alpha b,\gamma)}(z)$  and  $\bar{f} = f(t, y_N(t))$ . We enforce the Riemann-Liouville fractional integral operator of order  $\alpha$  on (4.3) and utilize the relation  $I^\alpha D_C^\alpha e_N(t) = e_N(t)$  whereby we deduce

$$e_N(t) = I^\alpha (f - \bar{f}) + \lambda \int_0^t (t-s)^{\alpha-1} \left(\int_0^s (s-\tau)^{\beta-1} (g(s, \tau, y(\tau)) - g(s, \tau, y_N(\tau))) d\tau\right) ds + R,$$

along with

$$R = I^\alpha e_{\Pi_{\hat{N}}^{(0,\alpha b,\gamma)}}\left(\bar{f} + \lambda \int_0^t (t-s)^{\beta-1} g(t, s, y_N(s)) ds\right).$$

Due to the Lipschitz assumption on  $f$  and  $g$ , we arrive at

$$\begin{aligned} |e_N(t)| &\leq L_1 I^\alpha |e_N(t)| + \frac{\lambda L_2}{\Gamma(\alpha)} \int_0^t (t-s)^{\alpha-1} \left(\int_0^s (s-\tau)^{\beta-1} |e_N(\tau)| d\tau\right) ds + |R| \\ &= L_1 I^\alpha |e_N(t)| + \lambda L_2 \Gamma(\beta) I^{\alpha+\beta} |e_N(t)| + |R|. \end{aligned} \tag{4.4}$$

In addition to this, we can write

$$\begin{aligned} L_1 I^\alpha |e_N| + \lambda L_2 \Gamma(\beta) I^{\alpha+\beta} |e_N| &\leq \max(L_1, \lambda L_2 \Gamma(\beta)) (I^\alpha |e_N| + I^{\alpha+\beta} |e_N|) \\ &= \max(L_1, \lambda L_2 \Gamma(\beta)) \left(\int_0^t (t-s)^{\alpha-1} \left(\frac{1}{\Gamma(\alpha)} + \frac{1}{\Gamma(\alpha+\beta)} (t-s)^\beta\right) |e_N(s)| ds\right) \\ &\leq \aleph \int_0^t (t-s)^{\alpha-1} |e_N(s)| ds, \end{aligned} \tag{4.5}$$

in which

$$\aleph = \max(L_1, \lambda L_2 \Gamma(\beta)) \left( \frac{1}{\Gamma(\alpha)} + \frac{T^\beta}{\Gamma(\alpha + \beta)} \right).$$

Substituting (4.5) into (4.4) and employing Gronwall’s inequality, i.e., Lemma 6 of [10], it can be concluded that

$$\|e_N(t)\| \leq C \|R\|.$$

We rewrite the above inequality in the following sense

$$\|e_N(t)\| \leq C \|e_{\Pi_{\hat{N}}^{(0,\alpha b,\gamma)}}\left(\bar{f} + \lambda \int_0^t (t-s)^{\beta-1} g(t,s,y_N(s)) ds\right)\|_{w^{(0,\alpha b,\gamma)}}, \tag{4.6}$$

due to the Cauchy-Schwarz inequality and some manipulations.

Currently, the suitable upper bounds are sought for each term of the right-hand side of (4.6). In this regard, utilizing the estimation (4.1) and the first-order Taylor formula give

$$\begin{aligned} & \|e_{\Pi_{\hat{N}}^{(0,\alpha b,\gamma)}}\left(\lambda \int_0^t (t-s)^{\beta-1} g(t,s,y_N(s)) ds\right)\|_{w^{(0,\alpha b,\gamma)}} \\ & \leq C \hat{N}^{-\varepsilon} \left| \int_0^t (t-s)^{\beta-1} g(t,s,y_N(s)) ds \right|_{\varepsilon,w^{(0,\alpha b,\gamma)}} \\ & \leq C \hat{N}^{-\varepsilon} \left( \left| \int_0^t (t-s)^{\beta-1} g(t,s,y(s)) ds \right|_{\varepsilon,w^{(0,\alpha b,\gamma)}} \right. \\ & \quad \left. + C_1 \|g(t,s,y(s)) - g(t,s,y_N(s))\| \right) \\ & \leq C \hat{N}^{-\varepsilon} \left( \left| \int_0^t (t-s)^{\beta-1} g(t,s,y(s)) ds \right|_{\varepsilon,w^{(0,\alpha b,\gamma)}} + C_1 L_2 \|e_N\| \right), \tag{4.7} \end{aligned}$$

due to the Lipschitz assumption on  $g$ . Here,  $C_1 > 0$  is a generic constant independent of  $N$ . Proceeding the same way as (4.7), we derive

$$\|e_{\Pi_{\hat{N}}^{(0,\alpha b,\gamma)}}(\bar{f})\|_{w^{(0,\alpha b,\gamma)}} \leq C \hat{N}^{-\varepsilon} |\bar{f}|_{\varepsilon,w^{(0,\alpha b,\gamma)}} \leq C \hat{N}^{-\varepsilon} \left( |f|_{\varepsilon,w^{(0,\alpha b,\gamma)}} + C_2 L_2 \|e_N\| \right), \tag{4.8}$$

where  $C_2 > 0$  denotes a generic constant independent of  $N$ .

Inserting (4.7) and (4.8) into (4.6) concludes

$$\begin{aligned} & \|e_N(t)\| - C(\hat{N}^{-\varepsilon} C_1 L_2 + \hat{N}^{-\varepsilon} C_2 L_2) \|e_N(t)\| \\ & \leq C \left( \hat{N}^{-\varepsilon} \left| \int_0^t (t-s)^{\beta-1} g(t,s,y(s)) ds \right|_{\varepsilon,w^{(0,\alpha b,\gamma)}} + \hat{N}^{-\varepsilon} |f|_{\varepsilon,w^{(0,\alpha b,\gamma)}} \right). \end{aligned}$$

Evidently, the desired result can be attained for sufficiently large values of  $N$ .

### 5 Numerical results

This section is devoted to confirming the effectiveness and productivity of the proposed implementation through demonstrating the numerical experiments derived from solving some non-linear weakly singular FIEDs. In this respect, this section is organized in the following sense

- To assess the computational capability of the introduced strategy, we illustrate some essential properties including numerical errors  $\mathbf{e}(N)$ , convergence order  $CO(N)$ , and CPU-time elapsed. We evaluate the numerical errors through the  $L^2(\Lambda)$ -norm by

$$\mathbf{e}(N) = \|e_N(t)\|, \quad \|e_N(t)\|^2 \simeq \frac{T}{2} \sum_{k=0}^N \left( e_N(T/2(x_k^{(0,0)} + 1)) \right)^2 w_k^{(0,0)},$$

where  $\{x_k^{(0,0)}\}_{k=0}^N$  and  $\{w_k^{(0,0)}\}_{k=0}^N$  are the Legendre-Gauss quadrature nodes and corresponding weights respectively [28]. In addition, convergence order is defined as

$$CO(N) = \left| \log_2 \frac{\mathbf{e}(N)}{\mathbf{e}(\frac{N}{2})} \right|.$$

- The stability of the method is also examined via approximating the highly oscillatory solution of a test problem associated with the long domain  $\Lambda$  and large values of  $N$ .
- The predominance of the suggested approach is assessed by comparing our results to those obtained by a modification of hat functions (MHFs) introduced in [21].

The computation is conducted by means of Mathematica v11.2, running in a computer system with an Intel (R) Core (TM) i5-4210U CPU @ 2.40 GHz.

**Example 1** Let us consider the non-linear weakly singular FIDEs

$$\begin{cases} D_C^{\frac{3}{2}} y(t) = f(t) + \int_0^t (t-s)^{-\frac{3}{4}} g(t, s, y(s)) ds, & t \in [0, 1], \\ y(0) = 0. \end{cases}$$

The source function  $f(t)$  is chosen in a way that the exact solution is

$$y(t) = E_{\frac{3}{2}}(t^{\frac{3}{2}}) - 1,$$

and

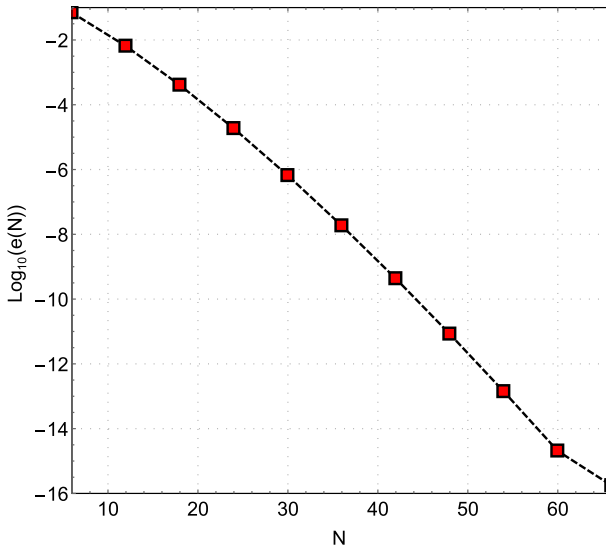
$$g(t, s, y(s)) = \frac{1}{2} s J_0(t^{\frac{7}{4}}) \sin(y(s)) + s^{\frac{5}{2}} y^4(s) + t^{\frac{1}{2}} s^{\frac{1}{4}},$$

where  $E_c(t)$  denotes Mittag-Leffler function, and for integer number  $d$ ,  $J_d(t)$  is known as Bessel function.



**Table 1** The numerical consequences of Example 1 for different degrees of approximation  $N$

$N$	$e(N)$	$CO(N)$	$CPU - time(sec)$
8	6.93E-2	–	1.51
16	6.44E-3	3.43	18.92
32	6.51E-7	13.27	113.13
64	2.04E-15	28.25	555.19



**Fig. 1** Semi-log depiction of the numerical errors versus  $N$  for Example 1

It can easily be checked that the forcing function  $f(t)$  has the following asymptotic behavior

$$f(t) = 1 - 3.71t + 0.75t^{\frac{3}{2}} - 1.02t^{\frac{11}{4}} + 0.17t^3 + \dots,$$

and we have  $y(t) = O(t^{\frac{3}{2}})$  which coincide with the result of Theorem 2.

We assess this problem by means of the proposed implementation and report the results in Table 1 and Fig. 1. From Table 1, it is reasoned that the approximate solutions are highly accurate, it is because the numerical errors are declined regularly in the short CPU-time used particularly for the large degrees of approximation  $N$ . In addition, the semi-log depiction of the numerical errors demonstrated in Fig. 1 confirms the well-known exponential accuracy predicting in Theorem 3 caused by the linear variations of the semi-log depiction of errors versus  $N$  (notice that we have  $\varepsilon, \zeta = \infty$  in Theorem 3).

**Example 2** Let us give the following highly oscillatory non-linear weakly singular FIDEs

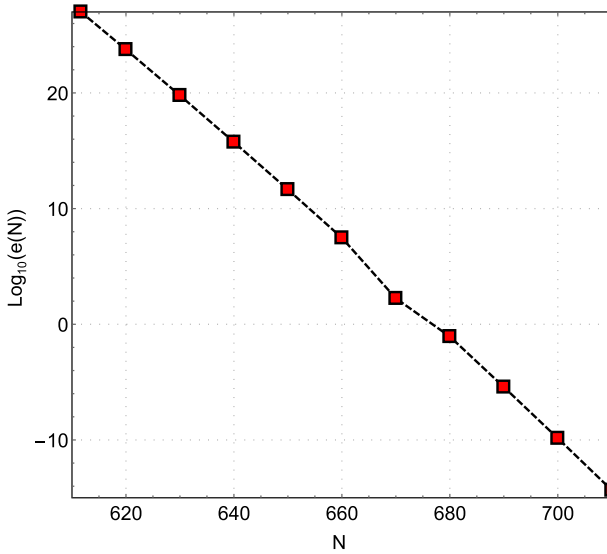


Fig. 2 Semi-log depiction of the numerical errors versus  $N$  for Example 2

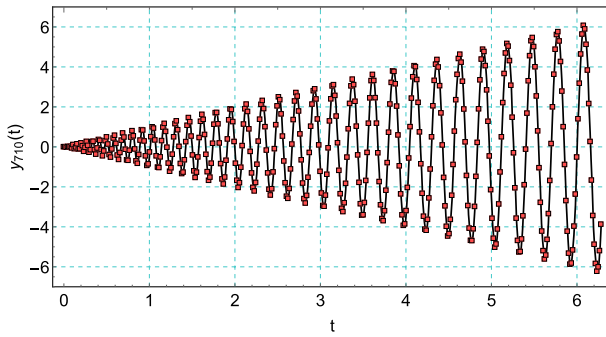
$$\begin{cases} D_C^{\frac{1}{2}}y(t) = f(t) + \int_0^t (t-s)^{-\frac{1}{2}}g(t,s,y(s))ds, & t \in [0, 2\pi], \\ y(0) = 0. \end{cases}$$

The forcing function  $f(t)$  is selected in a way that

$$g(t,s,y(s)) = t^2s^{\frac{3}{2}}y^2(s) + 5_2F_2(\{\frac{1}{4}, \frac{3}{4}\}; \{\frac{1}{5}, \frac{2}{5}\}; -\frac{t}{2}),$$

and  $y(t) = t \sin 100t^{\frac{1}{2}}$ .

${}_x F_\xi(\{a_1, \dots, a_\xi\}; \{b_1, \dots, b_\xi\}; t)$  denotes the generalized hypergeometric function. We apply the introduced scheme to this problem and the derived consequences are illustrated in Figs. 2 and 3. Needless to mention, the highly oscillatory behavior of the solution may cause instability in approximation, particularly for large degrees of approximation  $N$ . Regardless of this fact, the numerical results demonstrate that our scheme, however, produces highly accurate approximate solutions. Indeed, From Fig. 2, it can be concluded that the method is on the path of convergence for  $N > 680$ , and the effective computational performance of our strategy let the numerical errors decline regularly specifically for large degrees of approximation  $N$ . Furthermore, the well-known exponential accuracy is confirmed caused by the linear variations of semi-log depiction of errors versus  $N$ .



**Fig. 3** Graphs of the exact solution (solid line) and the approximate solution (red squares) of Example 2 for  $N = 710$

**Table 2** The absolute errors at some selected grid points of Example 3 with different values of  $n$  utilizing the suggested method in [21]

$n$	<i>Error</i>
2	7.09E-2
4	1.40E-2
8	3.31E-3
16	8.85E-4
32	2.46E-4
64	6.92E-5
128	2.42E-5
256	8.57E-6
512	3.03E-6
1024	1.07E-6

**Example 3** [21] Consider the following non-linear weakly singular FIDEs

$$\begin{cases} D_C^{\frac{2}{3}}y(t) = f(t, y(t)) + \int_0^t (t-s)^{-\frac{1}{2}}y^2(s)ds, & t \in [0, 1], \\ y(0) = 0, \end{cases} \tag{5.1}$$

where

$$f(t, y(t)) = \frac{3\Gamma(\frac{1}{2})}{4\Gamma(\frac{11}{6})}t^{\frac{5}{6}} - t^{\frac{5}{2}} - \frac{32}{35}t^{\frac{7}{2}} + ty(t).$$

The exact solution of this problem is  $y(t) = t^{\frac{3}{2}}$ . Considering  $\gamma = \frac{1}{6}$ , this problem is solved via the implemented scheme, and the exact solution is obtained with the degree of approximation 9, in machine precision. Also, in [21], the approximate solution of (5.1) is computed by means of a modification of hat functions (MHFs), and the absolute errors at some selected grid points derived in [21] for various values of  $n$  are listed in Table 2. In this method,  $n$  directs the number of uniform sub-intervals. We refer the reader to [21] for more details about this method. Comparison results justify the dominance of our suggested method over the presented scheme in [21].

## 6 Conclusions

A comprehensive survey of the existence, uniqueness, and smoothness properties of the solution of (1.1) was presented, and in particular, was demonstrated that the solution has a singularity at the origin. Taking into account the smoothness of the solution we proposed a creative strategy based on the spectral Petrov-Galerkin method to solve (1.1) numerically. This strategy offered some recurrence relations for deriving the approximate solution rather than solving a non-linear complex algebraic system. Finally, our implementation drove us to verify the spectral accuracy of the proposed method through the convergence theorem and approximating some illustrative examples. This strategy enables us to attack a vast majority of non-linear fractional functional equations, which would possibly motivate us to do research on it in the future.

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## Declarations

**Conflict of interest** The authors declare that they have no conflict of interest.

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