

## RESEARCH ARTICLE

# Numerical simulation of time fractional Cable equations and convergence analysis

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**Funding information**

The work was supported by NSFC Project (11671342, 91430213, 11671339, 11771369), Project of Scientific Research Fund of Hunan Provincial Science and Technology Department (2016JJ3114) and Key Project of Hunan Provincial Department of Education (17A210).

In this article, the numerical solution of time fractional Cable equation is considered. We convert the time fractional Cable equations into equivalent integral equations with singular kernel, then propose a spectral collection method in both time and space discretizations with a spectral expansion of Lagrange interpolation polynomial for this equation. The convergence of the method is rigorously established. Numerical tests are carried out to confirm the theoretical results.

**KEYWORDS**

convergence analysis, Jacobi collocation method, time fractional Cable equation

## 1 | INTRODUCTION

The Cable equation is one of the most fundamental equations for modelling neuronal dynamics. When the ions are undergoing anomalous subdiffusion, the standard or normal models may lead to incorrect or misleading diffusion coefficient values and models, so it is better to use the model incorporate with anomalous diffusion. A recent study on spiny Purkinje cell dendrites that spines trap and release

diffusing molecules resulting in anomalously show molecular diffusion along the dendrite [1]. The diffusive spatial variance  $\langle r^2(t) \rangle$  of an inert tracer was found to evolve as a sublinear power law in time, that is,  $\langle r^2(t) \rangle \sim t^\gamma$  with  $0 < \gamma < 1$ . Henry et al. [2] derived a fractional Cable equation from the fractional Nernst-Planck equations to model anomalous electrodiffusion of ions in spiny dendrites. They subsequently found a fractional Cable equation by treating the neuron and its membrane as two separate materials governed separate fractional Nernst-Planck equations and used a small ionic concentration gradient assumption [3, 4]. The resulting equation involves two fractional temporal Riemann-Liouville derivatives.

In this article, we consider the following time fractional Cable equation

$$u_t(x, t) = {}^R D_t^{1-\mu_1} K u_{xx}(x, t) - \lambda {}^R D_t^{1-\mu_2} u(x, t) + f(x, t), \quad 0 < \mu_2 \leq \mu_1 < 1, \quad (1.1)$$

subject to the initial condition

$$u(x, 0) = \phi(x), \quad a < x < b, \quad (1.2)$$

and boundary conditions

$$u(a, t) = \psi_1(t), \quad u(b, t) = \psi_2(t), \quad 0 < t < T, \quad (1.3)$$

where  $K > 0$  and  $\lambda$  are constants,  ${}^R D_t^\mu$  is defined as Riemann-Liouville fractional derivative of order  $\mu$ ,

$${}^R D_t^\mu P(t) = \frac{1}{\Gamma(1-\mu)} \frac{d}{dt} \int_0^t (t-\tau)^{-\mu} P(\tau) d\tau, \quad \mu \in (0, 1). \quad (1.4)$$

In the order, the well-known Caputo and Riemann-Liouville also have a big problem that their kernel although nonlocal but is singular. This weakness has effect when modeling real world problems. Recently, some newly concept of fractional calculus are proposed to answer some outstanding questions [5–9].

The numerical method to fractional differential equations has been recently discussed by numerous researchers. Some different numerical methods for solving time or space fractional partial differential equations have been proposed [10–14]. Liu and Yang [15], proposed two new implicit numerical methods with convergence analysis for the fractional Cable equation. Ibrahim and Nurdane [16] constructed a new difference scheme based on Crank-Nicholson difference scheme for solution of fractional Cable equation involving Caputo fractional derivative. Bu [17, 18] considered finite element multigrid method for time fractional advection diffusion equations. Recently, we provided Jacobi spectral-collocation method [19, 20] for time-fractional equations. Bhrawy and Zaky [21] reported a spectral collocation method based on shifted Jacobi collocation procedure in conjunction with the shifted Jacobi operational matrix for solving one and two-dimensional variable-order fractional nonlinear Cable equations. Sweilam and Khader [22] introduced numerical study for the fractional Cable equation using weighted average of finite difference methods. Mao and Xiao [23, 24] investigated the numerical solution for a class of fractional diffusion-wave equations based on the sinc-collocation technique. Yang and Xiao [25, 26] discussed the cubic spline collocation method with two parameters for solving the initial value problems of fractional differential equations.

The main purpose of this article is to solve and analyse the fractional Cable problem by Jacobi spectral collocation method. One contribution of this article is that we transform the fractional Cable equation into an integro-differential equation with weakly singular kernel and propose a Jacobi spectral

collocation method in both time and space discretizations with a spectral expansion of Lagrange interpolation polynomial for this equation. Another contribution of this article is that we provide a rigorous error analysis for the proposed method, which shows that the numerical errors decay exponentially in both the infinity norm and the weighted  $L_2$  norms.

## 2 | JACOBI COLLOCATION METHOD FOR THE TIME-FRACTIONAL CABLE EQUATION

First, we give the properties of the fractional derivative and integral, we can transform the time-fractional Cable problem into a Volterra integro-differential equation with a weakly singular kernel equivalently. Therefore, a number of the numerical schemes developed for Volterra-type integro-differential equation with weakly singular kernel can be applied to find numerical solution of fractional differential equations.

First, we define  $\mathbf{J}_t^\mu$  as the Riemann-Liouville fractional integral of order  $\mu$

$$\mathbf{J}_t^\mu P(t) = \frac{1}{\Gamma(\mu)} \int_0^t (t - \tau)^{\mu-1} P(\tau) d\tau, \quad \mu \in (0, 1). \quad (2.1)$$

Letting  $\mathbf{J}_t^{1-\mu_1}$  act on both sides of (1.1), we find

$$\mathbf{J}_t^{1-\mu_1} u_t(x, t) = \mathbf{J}_t^{1-\mu_1} \left( {}^R D_t^{1-\mu_1} K u_{xx}(x, t) - \lambda {}^R D_t^{1-\mu_2} u(x, t) + f(x, t) \right), \quad (2.2)$$

using the following properties

$$\begin{aligned} \mathbf{J}_t^\mu ({}^R D_t^\mu P(t)) &= P(t) - [\mathbf{J}_t^{1-\mu} P(t)]_{t=0} \frac{t^{\mu-1}}{\Gamma(\mu)}, \\ \mathbf{J}_t^{1-\mu} P_t(t) &= {}^R D_t^\mu P(t) - \frac{P(0)t^{-\mu}}{\Gamma(1-\mu)}, \end{aligned} \quad (2.3)$$

(2.2) becomes

$$\begin{aligned} {}^R D_t^\mu u(x, t) - \frac{u(x, 0)t^{-\mu}}{\Gamma(1-\mu)} &= K \left( u_{xx}(x, t) - \frac{t^{-\mu_1}}{\Gamma(1-\mu_1)} \mathbf{J}_t^{\mu_1} u_{xx}(x, t) \Big|_{t=0} \right) \\ &\quad - \lambda \left( \mathbf{J}_t^{\mu_1-\mu_2} u(x, t) - \frac{t^{-\mu_1}}{\Gamma(1-\mu_1)} \mathbf{J}_t^{\mu_1-\mu_2} u(x, t) \Big|_{t=0} \right) \\ &\quad + \mathbf{J}_t^{\mu_1} f(x, t). \end{aligned} \quad (2.4)$$

As  $u(x, t)$  is at least a continuous function w.r.t.  $t$ , then

$$\lim_{t \rightarrow 0} |\mathbf{J}_t^\mu u(x, t)| = \lim_{t \rightarrow 0} \left| \frac{1}{\Gamma(\mu)} \int_0^t (t - \tau)^{\mu-1} u(x, \tau) d\tau \right| \leq \lim_{t \rightarrow 0} M \frac{t^\mu}{\Gamma(1+\mu)} = 0,$$

where  $M = \max_{t \in [0, T]} u(x, t)$  for fixed  $x$ , and  $T$  is any small fixed positive number. Then, we can rewrite (2.4) in the following equivalent form

$${}^R D_t^\mu u(x, t) - \frac{u(x, 0)t^{-\mu}}{\Gamma(1-\mu)} = K u_{xx}(x, t) - \lambda \mathbf{J}_t^{\mu_1-\mu_2} u(x, t) + \mathbf{J}_t^{\mu_1} f(x, t). \quad (2.5)$$

We define  $D_t^\mu$  as Caputo fractional derivative of order  $\mu$ ,

$$D_t^\mu P(t) = \frac{1}{\Gamma(1-\mu)} \int_0^t (t-\tau)^{-\mu} P(\tau) d\tau, \quad \mu \in (0, 1), \quad (2.6)$$

and note that the following properties of Caputo fractional derivative hold

$$\begin{aligned} D_t^\mu P(t) &= {}^R D_t^\mu P(t) - \frac{P(0)t^{-\mu}}{\Gamma(1-\mu)}, \\ \mathbf{J}_t^\mu (D_t^\mu P(t)) &= P(t) - P(0). \end{aligned} \quad (2.7)$$

So, letting  $\mathbf{J}^{\mu_1}$  act on both sides of (2.5), we find the fractional Cable equation can be written in the integrated form with weakly singular kernel  $(t-\tau)^{\mu-1}$  finally

$$\begin{aligned} u(x, t) &= \mathbf{J}_t^{\mu_1} K u_{xx}(x, \tau) - \lambda \mathbf{J}_t^{2\mu_1 - \mu_2} u(x, t) + \mathbf{J}_t^{2\mu_1} f(x, t) + u(x, 0) \\ &= K \frac{1}{\Gamma(\mu_1)} \int_0^t (t-\tau)^{-\gamma_1} u_{xx}(x, \tau) d\tau \\ &\quad - \lambda \frac{1}{\Gamma(2\mu_1 - \mu_2)} \int_0^t (t-\tau)^{-\gamma_2} u(x, \tau) d\tau + \mathbf{J}_t^{2\mu_1} f(x, t) + u(x, 0), \\ &\quad -1 < -\gamma_1 = \mu_1 - 1 < -\gamma_2 = 2\mu_1 - \mu_2 - 1 < 0, \end{aligned} \quad (2.8)$$

subject to the initial condition

$$u(x, 0) = \phi(x), \quad a < x < b, \quad (2.9)$$

and boundary conditions

$$u(a, t) = \psi_1(t), \quad u(b, t) = \psi_2(t), \quad 0 < t < T, \quad (2.10)$$

where  $g(x, t) = \mathbf{J}_t^{2\mu_1} f(x, t)$ .

In this article, we consider the numerical solution of integral equations with singular kernel (2.8), which is equivalent to the time fractional Cable equation (1.1). The numerical treatment of the Equation 2.8 is not simple, mainly due to the fact that the solutions of (2.8) usually have a weak singularity near  $t = 0^+$ . Recently, we provided Jacobi spectral-collocation method [27, 28] or spectral Petrov-Galerkin method [29, 30] and convergence analysis for integro-differential equations. We propose a spectral collocation method in both temporal and spatial discretizations with a spectral expansion of Jacobi interpolation polynomial for this equation. The main advantage of the present scheme is that it gives very accurate convergence by choosing less number of grid points and the problem can be solved up to big time, and the storage requirement due to the time memory effect can be considerably reduced. At the same time, the technique is simple and easy to apply to multidimensional problems.

Let  $\omega^{\alpha, \beta}(x) = (1-x)^\alpha (1+x)^\beta$  be a weight function in the usual sense, for  $\alpha, \beta > -1$ . The set of Jacobi polynomials  $\{J_n^{\alpha, \beta}(x)\}_{n=0}^\infty$  forms a complete  $L_{\omega^{\alpha, \beta}}^2(-1, 1)$ —orthogonal system, where  $L_{\omega^{\alpha, \beta}}^2(-1, 1)$  is a weighted space defined by

$$L_{\omega^{\alpha, \beta}}^2(-1, 1) = \{v : v \text{ is measurable and } \|v\|_{\omega^{\alpha, \beta}} < \infty\},$$

equipped with the inner product

$$(u, v)_{\omega^{\alpha, \beta}} = \int_{-1}^1 u(x)v(x)\omega^{\alpha, \beta}(x)dx \quad \forall u, v \in L^2_{\omega^{\alpha, \beta}}(-1, 1),$$

and the norm

$$\|v\|_{\omega^{\alpha, \beta}} = (v, v)_{\omega^{\alpha, \beta}}^{\frac{1}{2}}.$$

For a given  $N \geq 0$ , we denote by  $\{\theta_k\}_{k=0}^N$  the Legendre points, and by  $\{\omega_k\}_{k=0}^N$  the corresponding Legendre weights (i.e., Jacobi weights  $\{\omega_k^{0,0}\}_{k=0}^N$ ). Then, the Legendre-Gauss integration formula is

$$\int_{-1}^1 f(x)dx \approx \sum_{k=0}^N f(\theta_k)\omega_k. \quad (2.11)$$

Similarly, we denote by  $\{\tilde{\theta}_k\}_{k=0}^N$  the Jacobi-Gauss points, and by  $\{\omega_k^{\alpha, \beta}\}_{k=0}^N$  the corresponding Jacobi weights. Then, the Jacobi-Gauss integration formula is

$$\int_{-1}^1 f(x)\omega^{\alpha, \beta}(x)dx \approx \sum_{k=0}^N f(\tilde{\theta}_k)\omega_k^{\alpha, \beta}. \quad (2.12)$$

For a given positive integer  $N$ , we denote the collocation points by  $\{x_i^{\alpha, \beta}\}_{i=0}^N$ , which is the set of  $(N+1)$  Jacobi-Gauss points, corresponding to the weight  $\omega^{\alpha, \beta}(x)$ . Let  $\mathcal{P}_N$  denote the space of all polynomials of degree not exceeding  $N$ . For any  $v \in C[-1, 1]$ , we can define the Lagrange interpolating polynomial  $I_N^{\alpha, \beta} v \in \mathcal{P}_N$ , satisfying

$$I_N^{\alpha, \beta} v(x_i^{\alpha, \beta}) = v(x_i^{\alpha, \beta}), \quad 0 \leq i \leq N.$$

The Lagrange interpolating polynomial can be written in the form

$$I_N^{\alpha, \beta} v(x) = \sum_{i=0}^N v(x_i^{\alpha, \beta})F_i(x), \quad 0 \leq i \leq N,$$

where  $F_i(x)$  is the Lagrange interpolation basis function associated with  $\{x_i^{\alpha, \beta}\}_{i=0}^N$ .

To apply the theory of orthogonal polynomials, we make following variable change

$$\begin{aligned} x &= \frac{b-a}{2}(1 + \bar{x}) + a, \quad \bar{x} = \frac{2(x-a)}{b-a} - 1, \quad \bar{x} \in [-1, 1], \\ t &= \frac{T}{2}(1 + \bar{t}), \quad \bar{t} = \frac{2t}{T} - 1, \quad \bar{t} \in [-1, 1], \\ \tau &= \frac{T}{2}(1 + s), \quad s = \frac{2\tau}{T} - 1, \quad s \in [-1, y], \end{aligned}$$

then the singular problems (2.8) can be rewritten as

$$\bar{u}(\bar{x}, \bar{t}) = \frac{K}{\Gamma(\mu_1)} \left(\frac{T}{2}\right)^{\mu_1} \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} \bar{u}_{\bar{x}\bar{x}}(\bar{x}, s) ds$$

$$\begin{aligned}
& - \frac{\lambda}{\Gamma(2\mu_1 - \mu_2)} \left(\frac{T}{2}\right)^{2\mu_1 - \mu_2} \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} \bar{u}(\bar{x}, s) ds \\
& + \bar{g}(\bar{x}, \bar{t}) + \bar{\phi}(\bar{x}),
\end{aligned} \tag{2.13}$$

with boundary conditions

$$\bar{u}(-1, \bar{t}) = \bar{\psi}_1(\bar{t}), \quad \bar{u}(1, \bar{t}) = \bar{\psi}_2(\bar{t}), \quad -1 < \bar{t} < 1, \tag{2.14}$$

where

$$\begin{aligned}
\bar{u}(\bar{x}, \bar{t}) &= u\left(\frac{b-a}{2}(1+\bar{x})+a, \frac{T}{2}(1+\bar{t})\right), \\
\bar{u}_{\bar{x}\bar{x}}(\bar{x}, s) &= \frac{4}{(b-a)^2} u\left(\frac{b-a}{2}(1+\bar{x})+a, \frac{T}{2}(1+s)\right), \\
\bar{g}(\bar{x}, \bar{t}) &= g\left(\frac{b-a}{2}(1+\bar{x})+a, \frac{T}{2}(1+\bar{t})\right), \\
\bar{\phi}(\bar{x}) &= \phi\left(\frac{b-a}{2}(1+\bar{x})+a\right), \quad \bar{\psi}_i(\bar{t}) = \psi_i\left(\frac{T}{2}(1+\bar{t})\right), \quad i = 1, 2.
\end{aligned}$$

## 2.1 | Spectral collocation method in time

For the collocation methods in time, (2.13) holds at Jacobi collocation points  $\{\bar{t}_j\}_{j=0}^M$  with Jacobi weight functions  $\omega^{\bar{\alpha}, \bar{\beta}}(\bar{t}) = (1-\bar{t})^{\bar{\alpha}}(1+\bar{t})^{\bar{\beta}}$  on  $[-1, 1]$ , namely,

$$\begin{aligned}
\bar{u}(\bar{x}, \bar{t}_j) &= \zeta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_1} \bar{u}_{\bar{x}\bar{x}}(\bar{x}, s) ds + \eta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_2} \bar{u}(\bar{x}, s) ds \\
&+ \bar{g}(\bar{x}, \bar{t}_j) + \bar{\phi}(\bar{x})
\end{aligned} \tag{2.15}$$

where  $\zeta = \frac{K}{\Gamma(\mu_1)} \left(\frac{T}{2}\right)^{\mu_1}$ ,  $\eta = -\frac{\lambda}{\Gamma(2\mu_1 - \mu_2)} \left(\frac{T}{2}\right)^{2\mu_1 - \mu_2}$ .

To use some appropriate quadrature rule, we will transfer the integral interval  $[-1, \bar{t}_j]$  to a fixed interval  $[-1, 1]$  and  $j$ , we make a simple linear transformation

$$s = s_j(\theta) = \frac{1 + \bar{t}_j}{2} \theta + \frac{\bar{t}_j - 1}{2}, \quad \theta \in [-1, 1], \tag{2.16}$$

we can rewrite (2.15) as follows:

$$\begin{aligned}
\bar{u}(\bar{x}, \bar{t}_j) &= \zeta_j \int_{-1}^1 (1-\theta)^{-\gamma_1} \bar{u}_{\bar{x}\bar{x}}(\bar{x}, s_j(\theta)) d\theta + \eta_j \int_{-1}^1 (1-\theta)^{-\gamma_2} \bar{u}(\bar{x}, s_j(\theta)) d\theta \\
&+ \bar{g}(\bar{x}, \bar{t}_j) + \bar{\phi}(\bar{x}),
\end{aligned} \tag{2.17}$$

with  $\zeta_j = \zeta \left(\frac{1+\bar{t}_j}{2}\right)^{1-\gamma_1}$ ,  $\eta_j = \eta \left(\frac{1+\bar{t}_j}{2}\right)^{1-\gamma_2}$ .

Next, using a  $(L+1)$ -point Gauss quadrature formula, the first integration term in (2.17) can be approximated by

$$\int_{-1}^1 (1-\theta)^{-\gamma_1} \bar{u}_{\bar{x}\bar{x}}(\bar{x}, s_j(\theta)) d\theta \approx \sum_{k=0}^L \bar{u}_{\bar{x}\bar{x}}(\bar{x}, s_j(\theta_k)) \omega_k^{-\gamma_1, 0}, \tag{2.18}$$

where  $L \geq M$ , and  $\{\theta_k\}_{k=0}^L$  is the set of Jacobi-Gauss points corresponding to the weights  $\omega^{-\gamma_1,0}(\theta) = (1-\theta)^{-\gamma_1}(1+\theta)^0$  and  $\omega_k^{-\gamma_1,0} = \omega^{-\gamma_1,0}(\theta_k)$  on  $[-1, 1]$ . Similarly, the second integration term in (2.17) can be approximated by

$$\int_{-1}^1 (1-\theta)^{-\gamma_2} \bar{u}(\bar{x}, s_j(\theta)) d\theta \approx \sum_{k=0}^P \bar{u}(\bar{x}, s_j(\vartheta_k)) \omega_k^{-\gamma_2,0}, \quad (2.19)$$

where  $P \geq M$ , and  $\{\vartheta_k\}_{k=0}^P$  is the set of Jacobi-Gauss points corresponding to the weights  $\omega^{-\gamma_2,0}(\vartheta) = (1-\vartheta)^{-\gamma_2}(1+\vartheta)^0$  and  $\omega_k^{-\gamma_2,0} = \omega^{-\gamma_2,0}(\vartheta_k)$  on  $[-1, 1]$ .

We first use  $\bar{u}^j(\bar{x})$  to indicate the approximate value for  $\bar{u}(\bar{x}, \bar{t}_j)$ , and

$$\bar{u}(\bar{x}, \bar{t}) \approx \bar{u}^M(\bar{x}, \bar{t}) = \sum_{j=0}^M \bar{u}^j(\bar{x}) F_j(\bar{t}), \quad (2.20)$$

where  $F_j(\bar{t})$  is the  $j$ th Lagrange interpolation polynomial associated with the collocation points  $\{\bar{t}_j\}_{j=0}^M$ . Combining the above equations, the semidiscretized problem (2.17) can be approximated by

$$\begin{aligned} \bar{u}^j(\bar{x}) &= \zeta_j \sum_{k=0}^L \bar{u}_{\bar{x}\bar{x}}(\bar{x}, s_j(\theta_k)) \omega_k^{-\gamma_1,0} + \eta_j \sum_{k=0}^P \bar{u}(\bar{x}, s_j(\vartheta_k)) \omega_k^{-\gamma_2,0} \\ &\quad + \bar{g}(\bar{x}, \bar{t}_j) + \bar{\phi}(\bar{x}). \end{aligned} \quad (2.21)$$

## 2.2 | Spectral collocation method in space

For Jacobi collocation methods in space, (2.21) holds at the Jacobi-Gauss collocation points  $\{\bar{x}_i\}_{i=0}^N$  with Jacobi weight functions  $\omega^{\tilde{\alpha},\tilde{\beta}}(\bar{x}) = (1-\bar{x})^{\tilde{\alpha}}(1+\bar{x})^{\tilde{\beta}}$  on  $[-1, 1]$ ,

$$\begin{aligned} \bar{u}^j(\bar{x}_i) &= \zeta_j \sum_{k=0}^L \bar{u}_{\bar{x}\bar{x}}(\bar{x}_i, s_j(\theta_k)) \omega_k^{-\gamma_1,0} + \eta_j \sum_{k=0}^P \bar{u}(\bar{x}_i, s_j(\vartheta_k)) \omega_k^{-\gamma_2,0} \\ &\quad + \bar{g}(\bar{x}_i, \bar{t}_j) + \bar{\phi}(\bar{x}_i). \end{aligned} \quad (2.22)$$

Use  $\bar{u}_i^j$  to approximate the function  $\bar{u}^j(\bar{x}_i)$ , and

$$\bar{u}^j(\bar{x}) \approx \sum_{i=0}^N \bar{u}_i^j H_i(\bar{x}), \quad \bar{u}(\bar{x}, \bar{t}) \approx \bar{u}_N^M(\bar{x}, \bar{t}) = \sum_{i=0}^N \sum_{j=0}^M \bar{u}_i^j H_i(\bar{x}) F_j(\bar{t}), \quad (2.23)$$

where  $H_i(\bar{x})$  is the  $i$ th Lagrange interpolation polynomial associated with the collocation points  $\{\bar{x}_i\}_{i=0}^N$ . The discretized problem in space (2.22) can be approximated by

$$\begin{aligned} \bar{u}_i^j &= \zeta_j \sum_{k=0}^L \bar{u}_{N\bar{x}\bar{x}}^M(\bar{x}_i, s_j(\theta_k)) \omega_k^{-\gamma_1,0} + \eta_j \sum_{k=0}^P \bar{u}_N^M(\bar{x}_i, s_j(\vartheta_k)) \omega_k^{-\gamma_2,0} \\ &\quad + \bar{g}(\bar{x}_i, \bar{t}_j) + \bar{\phi}(\bar{x}_i). \end{aligned} \quad (2.24)$$

Then the solution of the full-discrete problem (2.13) using collocation spectral method is to seek  $\bar{u}_N^M(\bar{x}, \bar{t})$  of the form (2.23) such that  $\bar{u}_i^j$  satisfies the above collocation Equations 2.24 for  $1 \leq i \leq N-1$ ,  $0 \leq j \leq M$ .

For completeness sake, the implementation is briefly described here. To simplify the computation, we rewrite the above collocation Equations 2.24 into the following:

$$\begin{aligned}
 \bar{u}_i^j &= \zeta_j \sum_{n=0}^N \sum_{m=0}^M \bar{u}_n^m \bar{H}_n''(\bar{x}_i) \sum_{k=0}^L F_m(s_j(\theta_k)) \omega_k^{-\gamma_1, 0} \\
 &\quad + \eta_j \sum_{n=0}^N \sum_{m=0}^M \bar{u}_n^m \bar{H}_n(\bar{x}_i) \sum_{k=0}^P F_m(s_j(\vartheta_k)) \omega_k^{-\gamma_2, 0} + \bar{g}(\bar{x}_i, \bar{t}_j) + \bar{\phi}(\bar{x}_i) \\
 &= \sum_{n=0}^N \sum_{m=0}^M b_{i,n} c_{j,m} \bar{u}_n^m + \bar{g}_i^j + \bar{\phi}_i \\
 &= \sum_{n=1}^{N-1} \sum_{m=0}^M b_{i,n} c_{j,m} \bar{u}_n^m + \sum_{m=0}^M (b_{i,0} \bar{u}_0^m + b_{i,N} \bar{u}_N^m) c_{j,m} + \bar{g}_i^j + \bar{\phi}_i, \\
 i &= 1, \dots, N-1, \quad j = 0, \dots, M,
 \end{aligned} \tag{2.25}$$

where

$$\begin{aligned}
 b_{i,n} &= H_n''(\bar{x}_i) + H_n(\bar{x}_i), \quad \bar{g}_i^j = \bar{g}(\bar{x}_i, \bar{t}_j), \quad \bar{\phi}_i = \bar{\phi}(\bar{x}_i) \\
 c_{j,m} &= \zeta_j \sum_{k=0}^L F_m(s_j(\theta_k)) \omega_k^{-\gamma_1, 0} + \eta_j \sum_{k=0}^P F_m(s_j(\vartheta_k)) \omega_k^{-\gamma_2, 0}.
 \end{aligned}$$

The Dirichlet boundary conditions (2.10) are directly applied in (2.25) and give numerical solutions on boundary in the following way

$$\bar{u}_0^m = \psi_1 \left( \frac{T(1 + \bar{t}_m)}{2} \right), \quad \bar{u}_N^m = \psi_2 \left( \frac{T(1 + \bar{t}_m)}{2} \right), \quad m = 0, \dots, M. \tag{2.26}$$

Let us set

$$\begin{aligned}
 \bar{d}_i^j &= \sum_{m=0}^M (b_{i,0} \bar{u}_0^m + b_{i,N} \bar{u}_N^m) c_{j,m} + \bar{g}_i^j, \\
 \bar{\mathbf{U}}_N^M &= [\bar{u}_1^0, \dots, \bar{u}_{N-1}^0, \bar{u}_1^1, \dots, \bar{u}_{N-1}^1, \dots, \bar{u}_1^M, \dots, \bar{u}_{N-1}^M]^T, \\
 \bar{\mathbf{F}}_N^M &= [\bar{\phi}_1, \dots, \bar{\phi}_{N-1}, \bar{\phi}_1, \dots, \bar{\phi}_{N-1}, \dots, \bar{\phi}_1, \dots, \bar{\phi}_{N-1}]^T, \\
 \bar{\mathbf{D}}_N^M &= [\bar{d}_1^0, \dots, \bar{d}_{N-1}^0, \bar{d}_1^1, \dots, \bar{d}_{N-1}^1, \dots, \bar{d}_1^M, \dots, \bar{d}_{N-1}^M]^T, \\
 \mathbf{B} &= (b_{i,n}), \quad i = 1, \dots, N-1, \quad n = 1, \dots, N-1, \\
 \mathbf{A} &= [c_{j,m} \mathbf{B}], \quad j = 0, \dots, M, \quad m = 0, \dots, M, \\
 \mathbf{A} &\text{ is a matrix of } (N-1) \times (M+1) \text{ by } (N-1) \times (M+1).
 \end{aligned} \tag{2.27}$$

Thus, the numerical scheme (2.10) leads to a system of equation of the form

$$\bar{\mathbf{U}}_N^M = \mathbf{A} \bar{\mathbf{U}}_N^M + \bar{\mathbf{D}}_N^M + \bar{\mathbf{F}}_N^M. \tag{2.28}$$



### 3 | CONVERGENCE ANALYSIS

In this section, we will carry out error estimations for the solution of the semidiscretized full-discretized problems. First, we will provide some elementary lemmas, which are important for the derivation of the main results. Let  $I := (-1, 1)$ .

**Lemma 3.1** (see [31]) *Assume that an  $(N+1)$ -point Gauss quadrature formula relative to the Jacobi weight is used to integrate the product  $u\varphi$ , where  $u \in H_{\omega^{\alpha,\beta}}^{m,N}$  with  $I$  for some  $m \geq 1$  and  $\varphi \in \mathcal{P}_N$ . Then there exists a constant  $C$  independent of  $N$  such that*

$$\left| \int_{-1}^1 u(x)\varphi(x)dx - (u, \varphi)_N \right| \leq CN^{-m} |u|_{H_{\omega^{\alpha,\beta}}^{m,N}(I)} \|\varphi\|_{L_{\omega^{\alpha,\beta}}^2(I)}, \quad (3.1)$$

where

$$|u|_{H_{\omega^{\alpha,\beta}}^{m,N}(I)} = \left( \sum_{j=\min(m,N+1)}^m \|u^{(j)}\|_{L_{\omega^{\alpha,\beta}}^2(I)}^2 \right)^{1/2},$$

$$(u, \varphi)_N = \sum_{j=0}^N u(x_j)\varphi(x_j)\omega_j. \quad (3.2)$$

**Lemma 3.2** (see [31]) *Assume that  $u \in H_{\omega^{\alpha,\beta}}^{m,N}(I)$  and denote by  $I_N^{\alpha,\beta}u$  its interpolation polynomial associated with the  $(N+1)$  Jacobi-Gauss points  $\{x_j\}_{j=0}^N$ , namely,*

$$I_N^{\alpha,\beta}u = \sum_{i=0}^N u(x_i)F_i(x).$$

Then the following estimates hold:

$$\|u - I_N^{\alpha,\beta}u\|_{L_{\omega^{\alpha,\beta}}^2(I)} \leq CN^{-m} |u|_{H_{\omega^{\alpha,\beta}}^{m,N}(I)}, \quad (3.3a)$$

$$\|u - I_N^{\alpha,\beta}u\|_{L^\infty(I)} \leq \begin{cases} CN^{\frac{1}{2}-m} |u|_{H_{\omega^c}^{m,N}(I)}, & -1 \leq \alpha, \beta < -\frac{1}{2}, \\ CN^{1+\varepsilon-m} \log N |u|_{H_{\omega^c}^{m,N}(I)}, & \text{otherwise,} \end{cases} \quad (3.3b)$$

where  $\varepsilon = \max(\alpha, \beta)$  and  $\omega^c = \omega^{-\frac{1}{2}, -\frac{1}{2}}$  denotes the Chebyshev weight function.

**Lemma 3.3** (see [32]) *For every bounded function  $v$ , there exists a constant  $C$ , independent of  $v$ , such that*

$$\sup_N \left\| \sum_{j=0}^N v(x_j)F_j(x) \right\|_{L_{\omega^{\alpha,\beta}}^2(I)} \leq C \max_{x \in [-1,1]} |v(x)|,$$

where  $F_j(x)$ ,  $j = 0, 1, \dots, N$ , are the Lagrange interpolation basis functions associated with the Jacobi collocation points  $\{x_j\}_{j=0}^N$

**Lemma 3.4** (see [33]) Assume that  $\{F_j(x)\}_{j=0}^N$  are the  $N$ -th degree Lagrange basis polynomials associated with the Gauss points of the Jacobi polynomials. Then,

$$\|I_N^{\alpha,\beta}\|_{L^\infty(I)} \leq \max_{x \in [-1,1]} \sum_{j=0}^N |F_j(x)| = \begin{cases} \mathcal{O}(\log N), & -1 < \alpha, \beta \leq -\frac{1}{2}, \\ \mathcal{O}(N^{\varepsilon+\frac{1}{2}}), & \text{otherwise.} \end{cases} \quad (3.4)$$

**Lemma 3.5** (Gronwall inequality, see [34] Lemma 7.1.1) Suppose  $L \geq 0$ ,  $0 < \mu < 1$ , and  $u$  and  $v$  are a non-negative, locally integrable functions defined on  $[-1, 1]$  satisfying

$$u(x) \leq v(x) + L \int_{-1}^x (x - \tau)^{-\mu} u(\tau) d\tau.$$

Then there exists a constant  $C = C(\mu)$  such that

$$u(x) \leq v(x) + CL \int_{-1}^x (x - \tau)^{-\mu} v(\tau) d\tau, \quad \text{for } -1 \leq x < 1.$$

If a nonnegative integrable function  $E(x)$  satisfies

$$E(x) \leq L \int_{-1}^x E(s) ds + J(x), \quad -1 < x \leq 1,$$

where  $J(x)$  is an integrable function, then

$$\begin{aligned} \|E\|_{L^\infty(-1,1)} &\leq C \|J\|_{L^\infty(-1,1)}, \\ \|E\|_{L_{\omega^{\alpha,\beta}}^p(-1,1)} &\leq C \|J\|_{L_{\omega^{\alpha,\beta}}^p(-1,1)}, \quad p \geq 1. \end{aligned} \quad (3.5)$$

**Lemma 3.6** (see [35, 36]) For a nonnegative integer  $r$  and  $\kappa \in (0, 1)$ , there exists a constant  $C_{r,\kappa} > 0$  such that for any function  $v \in C^{r,\kappa}([-1, 1])$ , there exists a polynomial function  $\mathcal{T}_N v \in \mathcal{P}_N$  such that

$$\|v - \mathcal{T}_N v\|_{L^\infty(I)} \leq C_{r,\kappa} N^{-(r+\kappa)} \|v\|_{r,\kappa}, \quad (3.6)$$

where  $\|\cdot\|_{r,\kappa}$  is the standard norm in  $C^{r,\kappa}([-1, 1])$ ,  $\mathcal{T}_N$  is a linear operator from  $C^{r,\kappa}([-1, 1])$  into  $\mathcal{P}_N$ , as stated in [35, 36].

**Lemma 3.7** (see [37]) Let  $\kappa \in (0, 1)$  and let  $\mathcal{M}$  be defined by

$$(\mathcal{M}v)(x) = \int_{-1}^x (x - \tau)^{-\mu} K(x, \tau) v(\tau) d\tau.$$

Then, for any function  $v \in C([-1, 1])$ , there exists a positive constant  $C$  such that

$$\frac{|\mathcal{M}v(x') - \mathcal{M}v(x'')|}{|x' - x''|} \leq C \max_{x \in [-1,1]} |v(x)|,$$

under the assumption that  $0 < \kappa < 1 - \mu$ , for any  $x', x'' \in [-1, 1]$  and  $x' \neq x''$ . This implies that

$$\|\mathcal{M}v\|_{0,\kappa} \leq C \max_{x \in [-1,1]} |v(x)|, \quad 0 < \kappa < 1 - \mu.$$

Now, we devote to providing a convergence analysis for the numerical scheme. The goal is to show that the rate of convergence is exponential, that is, that spectral accuracy can be obtained for the proposed approximations.

**Theorem 3.1** *Let  $\bar{u}(\bar{x}, \bar{t})$  be the solution of the continuous problem (2.13),  $\bar{u}^M(\bar{x}, \bar{t}) = \sum_{m=0}^M \bar{u}^m(\bar{x}) F_m(\bar{t})$  be the time-discrete solution of (2.21),  $\bar{\varepsilon} = \max(\bar{\alpha}, \bar{\beta})$ . If  $\bar{u}(\bar{x}, \bar{t})$ ,  $\bar{u}_{\bar{x}}(\bar{x}, \bar{t})$  and  $\bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t}) \in H_{\omega_{\bar{\alpha}, \bar{\beta}}}^{m, M}(I)$  for  $t, m > 1$ , then for  $M$  sufficiently large,*

$$\|\bar{u}(\bar{x}, \bar{t}) - \bar{u}^M(\bar{x}, \bar{t})\|_{L^\infty} \leq \begin{cases} CM^{\frac{1}{2}-m} Q_1(\bar{x}), & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{1+\bar{\varepsilon}-m} \log M Q_1(\bar{x}), & -\frac{1}{2} \leq \bar{\varepsilon} < \frac{1}{2} - \mu, \end{cases} \quad (3.7a)$$

$$\|\bar{u}(\bar{x}, \bar{t}) - \bar{u}^M(\bar{x}, \bar{t})\|_{L^2_{\omega_{\bar{\alpha}, \bar{\beta}}}} \leq \begin{cases} CM^{-m} \left( Q_2(\bar{x}) + M^{\frac{1}{2}-\kappa_1} Q_1(\bar{x}) \right), & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{-m} \log M \left( Q_2(\bar{x}) + M^{1+\bar{\varepsilon}-\kappa_1} Q_1(\bar{x}) \right), & -\frac{1}{2} \leq \bar{\varepsilon} < \frac{1}{2} - \gamma_1, \end{cases} \quad (3.7b)$$

for  $\kappa_1 \in (0, 1 - \gamma_1)$ , where  $C$  is a constant independent of  $M$ ,

$$\begin{aligned} Q_1(x) &= |\bar{u}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}} + |\bar{u}_{\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}} + |\bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}}, \\ Q_2(x) &= |\bar{u}(\bar{x}, \bar{t})|_{H_{\omega_{\bar{\alpha}, \bar{\beta}}}^{m, M}} + |\bar{u}_{\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega_{\bar{\alpha}, \bar{\beta}}}^{m, M}} + |\bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega_{\bar{\alpha}, \bar{\beta}}}^{m, M}}. \end{aligned}$$

*Proof* Using the boundary conditions, we have

$$\begin{aligned} \bar{u}(\bar{x}, \bar{t}_j) &= \int_{-1}^{\bar{x}} \bar{u}_\tau(\tau, \bar{t}_j) d\tau + \bar{\psi}_1(\bar{t}_j), \\ \bar{u}_{\bar{x}}(\bar{x}, \bar{t}_j) &= \int_{\xi}^{\bar{x}} \bar{u}_{\tau\tau}(\tau, \bar{t}_j) d\tau + \frac{\bar{\psi}_2(\bar{t}_j) - \bar{\psi}_1(\bar{t}_j)}{2}, \end{aligned} \quad (3.8)$$

where  $\partial_{\bar{x}} \bar{u}(\bar{x}, \bar{t})|_{\bar{x}=\xi} = \frac{\bar{\psi}_2(\bar{t}) - \bar{\psi}_1(\bar{t})}{2}$ . Consequently,  $\bar{u}^M(\bar{x}, \bar{t})$  satisfies the following collocation equations for  $0 \leq j \leq M$ ,

$$\begin{aligned} \bar{u}^j(x) &= \int_{-1}^{\bar{x}} \bar{u}_\tau^M(\tau, \bar{t}_j) d\tau + \bar{\psi}_1(\bar{t}_j), \\ \bar{u}_{\bar{x}}^M(x, \bar{t}_j) &= \int_{\xi}^{\bar{x}} \bar{u}_{\tau\tau}^M(\tau, \bar{t}_j) d\tau + \frac{\bar{\psi}_2(\bar{t}_j) - \bar{\psi}_1(\bar{t}_j)}{2}. \end{aligned} \quad (3.9)$$

As  $L \geq M, P \geq M$   $\bar{u}^M(\bar{x}, \bar{t}) = \sum_{m=0}^M \bar{u}^m(\bar{x}) F_m(\bar{t}) \in \mathcal{P}_M(\bar{t})$ , we have

$$\begin{aligned} \int_{-1}^1 (1-\theta)^{-\gamma_1} \bar{u}_{\bar{x}\bar{x}}^M(\bar{x}, s_j(\theta)) d\theta &= \sum_{k=0}^L \bar{u}_{\bar{x}\bar{x}}^M(\bar{x}, s_j(\theta_k)) \omega_k^{-\gamma_1, 0}, \\ \int_{-1}^1 (1-\theta)^{-\gamma_2} \bar{u}^M(\bar{x}, s_j(\theta)) d\theta &= \sum_{k=0}^P \bar{u}^M(\bar{x}, s_j(\vartheta_k)) \omega_k^{-\gamma_2, 0}. \end{aligned}$$

Subtracting (2.17) from (2.21), subtracting (3.8) from (3.9), yields

$$\begin{aligned}
 \bar{u}(\bar{x}, \bar{t}_j) - \bar{u}^j(\bar{x}) &= \zeta_j \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_1} (\bar{u}_{\bar{x}\bar{x}}(\bar{x}, s) - \bar{u}_{\bar{x}\bar{x}}^M(\bar{x}, s)) ds \\
 &\quad + \eta_j \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_2} (\bar{u}(\bar{x}, s) - \bar{u}^M(\bar{x}, s)) ds \\
 \bar{u}(\bar{x}, \bar{t}_j) - \bar{u}^j(\bar{x}) &= \int_{-1}^{\bar{x}} \bar{u}_\tau(\tau, \bar{t}_j) d\tau - \int_{-1}^{\bar{x}} \bar{u}_\tau^M(\tau, \bar{t}_j) d\tau, \\
 \bar{u}_{\bar{x}}(\bar{x}, \bar{t}_j) - \bar{u}_{\bar{x}}^j(\bar{x}) &= \int_{\xi}^{\bar{x}} \bar{u}_{\tau\tau}(\tau, \bar{t}_j) d\tau - \int_{\xi}^{\bar{x}} \bar{u}_{\tau\tau}^M(\tau, \bar{t}_j) d\tau.
 \end{aligned} \tag{3.10}$$

Multiplying by  $F_j(\bar{t})$  both sides of (3.10) and summing from 0 to  $M$  for  $j$  yield

$$\begin{aligned}
 I_M^{\bar{\alpha}, \bar{\beta}} \bar{u}(\bar{x}, \bar{t}) - \bar{u}^M(\bar{x}, \bar{t}) &= I_M^{\bar{\alpha}, \bar{\beta}} \zeta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} (\bar{u}_{\bar{x}\bar{x}}(\bar{x}, s) - \bar{u}_{\bar{x}\bar{x}}^M(\bar{x}, s)) ds \\
 &\quad + I_M^{\bar{\alpha}, \bar{\beta}} \eta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} (\bar{u}(\bar{x}, s) - \bar{u}^M(\bar{x}, s)) ds, \\
 I_M^{\bar{\alpha}, \bar{\beta}} \bar{u}(\bar{x}, \bar{t}) - \bar{u}^M(\bar{x}, \bar{t}) &= I_M^{\bar{\alpha}, \bar{\beta}} \int_{-1}^{\bar{x}} \bar{u}_\tau(\tau, \bar{t}) d\tau - \int_{-1}^{\bar{x}} \bar{u}_\tau^M(\tau, \bar{t}) d\tau, \\
 I_M^{\bar{\alpha}, \bar{\beta}} \bar{u}_{\bar{x}}(\bar{x}, \bar{t}) - \bar{u}_{\bar{x}}^M(\bar{x}, \bar{t}) &= I_M^{\bar{\alpha}, \bar{\beta}} \int_{\xi}^{\bar{x}} \bar{u}_{\tau\tau}(\tau, \bar{t}) d\tau - \int_{\xi}^{\bar{x}} \bar{u}_{\tau\tau}^M(\tau, \bar{t}) d\tau.
 \end{aligned} \tag{3.11}$$

Let

$$\begin{aligned}
 E(\bar{x}, \bar{t}) &= \bar{u}(\bar{x}, \bar{t}) - \bar{u}^M(\bar{x}, \bar{t}), \quad E_{\bar{x}}(\bar{x}, \bar{t}) = \bar{u}_{\bar{x}}(\bar{x}, \bar{t}) - \bar{u}_{\bar{x}}^M(\bar{x}, \bar{t}), \\
 E_{\bar{x}\bar{x}}(\bar{x}, \bar{t}) &= \bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t}) - \bar{u}_{\bar{x}\bar{x}}^M(\bar{x}, \bar{t}).
 \end{aligned}$$

It follows from (3.11) that

$$\begin{aligned}
 E(\bar{x}, \bar{t}) &= \zeta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} E_{\bar{x}\bar{x}}(\bar{x}, s) ds + \eta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} E(\bar{x}, s) ds \\
 &\quad + K_1(\bar{x}, \bar{t}) + K_2(\bar{x}, \bar{t}) + K_3(\bar{x}, \bar{t}), \\
 E(\bar{x}, \bar{t}) &= \int_{-1}^{\bar{x}} E_\tau(\tau, \bar{t}) d\tau + K_1(\bar{x}, \bar{t}) + K_4(\bar{x}, \bar{t}), \\
 E_{\bar{x}}(\bar{x}, \bar{t}) &= \int_{\xi}^{\bar{x}} E_{\tau\tau}(\tau, \bar{t}) d\tau + K_5(\bar{x}, \bar{t}) + K_6(\bar{x}, \bar{t}),
 \end{aligned} \tag{3.12}$$

where

$$\begin{aligned}
 K_1(\bar{x}, \bar{t}) &= \bar{u}(\bar{x}, \bar{t}) - I_M^{\bar{\alpha}, \bar{\beta}} \bar{u}(\bar{x}, \bar{t}), \\
 K_2(\bar{x}, \bar{t}) &= I_M^{\bar{\alpha}, \bar{\beta}} \zeta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} E_{\bar{x}\bar{x}}(\bar{x}, s) ds - \zeta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} E_{\bar{x}\bar{x}}(\bar{x}, s) ds, \\
 K_3(\bar{x}, \bar{t}) &= I_M^{\bar{\alpha}, \bar{\beta}} \eta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} E(\bar{x}, s) ds - \eta(\bar{t}) \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} E(\bar{x}, s) ds, \\
 K_4(\bar{x}, \bar{t}) &= I_M^{\bar{\alpha}, \bar{\beta}} \int_{-1}^{\bar{x}} E_\tau(\tau, \bar{t}) d\tau - \int_{-1}^{\bar{x}} E_\tau(\tau, \bar{t}) d\tau,
 \end{aligned}$$

$$K_5(\bar{x}, \bar{t}) = \bar{u}_{\bar{x}}(\bar{x}, \bar{t}) - \bar{I}_M^{\bar{\alpha}, \bar{\beta}} \bar{u}_{\bar{x}}(\bar{x}, \bar{t}),$$

$$K_6(\bar{x}, \bar{t}) = \bar{I}_M^{\bar{\alpha}, \bar{\beta}} \int_{\bar{\xi}}^{\bar{x}} \bar{u}_{\tau\tau}^M(\tau, \bar{t}) d\tau - \int_{\bar{\xi}}^{\bar{x}} \bar{u}_{\tau\tau}^M(\tau, \bar{t}) d\tau.$$

Furtherly, we have

$$|E(\bar{x}, \bar{t})| \leq C_1 \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} |E(\bar{x}, s)| ds + C_2 \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} |E(\bar{x}, s)| ds$$

$$+ \sum_{i=1,5,7,8,9} |K_i(\bar{x}, \bar{t})|, \quad (3.13)$$

where

$$K_7(\bar{x}, \bar{t}) = \bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t}) - \bar{I}_M^{\bar{\alpha}, \bar{\beta}} \bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t}),$$

$$K_8(\bar{x}, \bar{t}) = \bar{I}_M^{\bar{\alpha}, \bar{\beta}} C_1 \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} |E(\bar{x}, s)| ds - C_1 \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_1} |E(\bar{x}, s)| ds,$$

$$K_9(\bar{x}, \bar{t}) = \bar{I}_M^{\bar{\alpha}, \bar{\beta}} C_2 \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} |E(\bar{x}, s)| ds - C_2 \int_{-1}^{\bar{t}} (\bar{t} - s)^{-\gamma_2} |E(\bar{x}, s)| ds.$$

Using the Gronwall inequality in Lemma 3.5, we deduce that

$$\|E(\bar{x}, \bar{t})\|_{L^\infty(I)} \leq \sum_{i=1,5,7,8,9} |K_i(\bar{x}, \bar{t})|_{L^\infty(I)}. \quad (3.14)$$

and

$$\|E(\bar{x}, \bar{t})\|_{L^2_{\bar{\alpha}, \bar{\beta}}(I)} \leq \sum_{i=1,5,7,8,9} |K_i(\bar{x}, \bar{t})|_{L^2_{\bar{\alpha}, \bar{\beta}}}. \quad (3.15)$$

Using the  $L^\infty$  and weighted  $L^2_{\omega^{\alpha, \beta}}$  error bounds for the interpolation polynomials (see Lemma 3.2) gives

$$\|K_1(\bar{x}, \bar{t})\|_{L^\infty} \leq \begin{cases} CM^{\frac{1}{2}-m} |\bar{u}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}}, & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{1+\bar{\epsilon}-m} \log M |\bar{u}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}}, & \text{otherwise,} \end{cases} \quad (3.16)$$

$$\|K_5(\bar{x}, \bar{t})\|_{L^\infty} \leq \begin{cases} CM^{\frac{1}{2}-m} |\bar{u}_{\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}}, & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{1+\bar{\epsilon}-m} \log M |\bar{u}_{\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}}, & \text{otherwise,} \end{cases} \quad (3.17)$$

$$\|K_7(\bar{x}, \bar{t})\|_{L^\infty} \leq \begin{cases} CM^{\frac{1}{2}-m} |\bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}}, & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{1+\bar{\epsilon}-m} \log M |\bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^c}^{m, M}}, & \text{otherwise,} \end{cases} \quad (3.18)$$

$$\|K_1(\bar{x}, \bar{t})\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \leq CM^{-m} |\bar{u}(\bar{x}, \bar{t})|_{H_{\omega^{\bar{\alpha}, \bar{\beta}}}^{m, M}}. \quad (3.19)$$

$$\|K_5(\bar{x}, \bar{t})\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \leq CM^{-m} |\bar{u}_{\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^{\bar{\alpha}, \bar{\beta}}}^{m, M}}. \quad (3.20)$$

$$\|K_7(\bar{x}, \bar{t})\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \leq CM^{-m} |\bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t})|_{H_{\omega^{\bar{\alpha}, \bar{\beta}}}^{m, M}}. \quad (3.21)$$

Finally, it follows from Lemma 3.6, Lemma 3.7, and Lemma 3.4 that

$$\begin{aligned}
\|K_8(\bar{x}, \bar{t})\|_{L^\infty} &= \|(I_M^{\bar{\alpha}, \bar{\beta}} - I)\mathcal{M}|E(\bar{x}, s)|\|_{L^\infty} \\
&= \|(I_M^{\bar{\alpha}, \bar{\beta}} - I)(\mathcal{M}|E(\bar{x}, s)| - \mathcal{T}_M\mathcal{M}|E(\bar{x}, s)|)\|_{L^\infty} \\
&\leq \left(1 + \|I_M^{\bar{\alpha}, \bar{\beta}}\|_{L^\infty(I)}\right) CM^{-\kappa_1} \|\mathcal{M}|E(\bar{x}, s)|\|_{0, \kappa_1}, \quad \kappa_1 \in (0, 1 - \gamma_1) \\
&\leq \begin{cases} CM^{-\kappa_1} \log M \| |E(\bar{x}, s)| \|_{L^\infty}, & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{\frac{1}{2} + \bar{\varepsilon} - \kappa_1} \| |E(\bar{x}, s)| \|_{L^\infty}, & -\frac{1}{2} \leq \bar{\varepsilon} < \frac{1}{2} - \gamma_1, \end{cases} \quad (3.22) \\
\|K_8(\bar{x}, \bar{t})\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} &= \|(I_M^{-\mu, -\mu} - I)\mathcal{M}|E(\bar{x}, s)|\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \\
&= \|(I_M^{-\mu, -\mu} - I)(\mathcal{M}|E(\bar{x}, s)| - \mathcal{T}_M|E(\bar{x}, s)|)\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \\
&\leq \|I_M^{\bar{\alpha}, \bar{\beta}}(\mathcal{M}|E(\bar{x}, s)| - \mathcal{T}_M|E(\bar{x}, s)|)\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} + \|\mathcal{M}|E(\bar{x}, s)| - \mathcal{T}_M|E(\bar{x}, s)|\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \\
&\leq C \|\mathcal{M}|E(\bar{x}, s)| - \mathcal{T}_M|E(\bar{x}, s)|\|_{L^\infty} \\
&\leq CM^{-\kappa_1} \|\mathcal{M}|E(\bar{x}, s)|\|_{0, \kappa_1} \\
&\leq CM^{-\kappa_1} \| |E(\bar{x}, s)| \|_{L^\infty}. \quad (3.23)
\end{aligned}$$

Similarly, we have

$$\|K_9(\bar{x}, \bar{t})\|_{L^\infty} \leq \begin{cases} CM^{-\kappa_2} \log M \| |E(\bar{x}, s)| \|_{L^\infty}, & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{\frac{1}{2} + \bar{\varepsilon} - \kappa_2} \| |E(\bar{x}, s)| \|_{L^\infty}, & -\frac{1}{2} \leq \bar{\varepsilon} < \frac{1}{2} - \gamma_2, \end{cases} \quad (3.24)$$

$$\|K_9(\bar{x}, \bar{t})\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \leq CM^{-\kappa_2} \| |E(\bar{x}, s)| \|_{L^\infty}, \quad \kappa_2 \in (0, 1 - \gamma_2). \quad (3.25)$$

Noting  $\gamma_1 > \gamma_2$ , estimates (3.16), (3.17), (3.18), (3.22), and (3.24) lead to (3.7a), estimates (3.19), (3.20), (3.21), (3.23), and (3.25), lead to (3.7b). ■

**Theorem 3.2** *Let  $\bar{u}^M(\bar{x}, \bar{t})$  be the time-discrete solution of (2.21) and  $\bar{u}_N^M(\bar{x}, \bar{t})$  the solution of the full-discrete problem (2.24) with boundary conditions (2.14). If  $\bar{u}^M(\bar{x}, \bar{t})$ ,  $\bar{u}_x^M(\bar{x}, \bar{t})$  and  $\bar{u}_{xx}^M(\bar{x}, \bar{t}) \in H_{\omega^{\bar{\alpha}, \bar{\beta}}}^{n, N}$  for  $n > 1$ ,  $\tilde{\varepsilon} = \max(\tilde{\alpha}, \tilde{\beta})$ ,  $\bar{\varepsilon} = \max(\bar{\alpha}, \bar{\beta})$ , then the following error estimate holds*

$$\begin{aligned}
\|\bar{u}^M(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t})\|_{L^\infty} &\leq \begin{cases} CN^{\frac{1}{2} - n} \log MV_1, & -1 \leq \tilde{\alpha}, \tilde{\beta}, \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CN^{1 + \bar{\varepsilon} - n} \log NM^{1 + \bar{\varepsilon}} V_1, & \text{otherwise,} \end{cases} \\
\|\bar{u}^M(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t})\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} &\leq CN^{-n} \log MV_2, \quad (3.26)
\end{aligned}$$

where

$$\begin{aligned}
V_1 &= \max_{\bar{t} \in [-1, 1]} \left( |\bar{u}^M(\bar{x}, \bar{t})|_{H_{\omega^c}^{n, N}} + |\bar{u}_{xx}^M(\bar{x}, \bar{t})|_{H_{\omega^c}^{n, N}} \right), \\
V_2 &= \max_{\bar{t} \in [-1, 1]} \left( |\bar{u}^M(\bar{x}, \bar{t})|_{H_{\omega^{\bar{\alpha}, \bar{\beta}}}^{n, N}} + |\bar{u}_{xx}^M(\bar{x}, \bar{t})|_{H_{\omega^{\bar{\alpha}, \bar{\beta}}}^{n, N}} \right),
\end{aligned}$$

$C$  is independent of  $N$  and  $M$

*Proof* By subtracting (2.22) from (2.24), we obtain

$$\begin{aligned}
 \bar{u}^j(\bar{x}_i) - \bar{u}_i^j &= \zeta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_1} (\bar{u}_{\bar{x}\bar{x}}^M(\bar{x}_i, s) - \bar{u}_{N\bar{x}\bar{x}}^M(\bar{x}_i, s)) ds \\
 &\quad + \eta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_2} (\bar{u}^M(\bar{x}_i, s) - \bar{u}_N^M(\bar{x}_i, s)) ds \\
 &= \zeta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_1} \sum_{m=0}^M \left( \bar{u}_{\bar{x}\bar{x}}^m(\bar{x}_i) - \sum_{n=0}^N \bar{u}_n^m H_n''(\bar{x}_i) \right) F_m(s) ds \\
 &\quad + \eta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_2} \sum_{m=0}^M \left( \bar{u}^m(\bar{x}_i) - \sum_{n=0}^N \bar{u}_n^m H_n(\bar{x}_i) \right) F_m(s) ds \\
 &= \sum_{m=0}^M \left[ C_1^{jm} \left( \bar{u}_{\bar{x}\bar{x}}^m(\bar{x}_i) - \sum_{n=0}^N \bar{u}_n^m H_n''(\bar{x}_i) \right) + C_2^{jm} (\bar{u}^m(\bar{x}_i) - \bar{u}_i^m) \right], \quad (3.27)
 \end{aligned}$$

where

$$C_1^{jm} = \zeta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_1} F_m(s) ds, \quad C_2^{jm} = \eta \int_{-1}^{\bar{t}_j} (\bar{t}_j - s)^{-\gamma_2} F_m(s) ds.$$

Multiplying by  $H_i(x)$  both sides of (3.27) and summing from 0 to  $N$  for  $i$ , and noting that  $\sum_{n=0}^N \bar{u}_n^m H_n''(\bar{x}) \in \mathcal{P}_N(\bar{x})$ , then  $I_N^{\tilde{\alpha}, \tilde{\beta}} \sum_{n=0}^N \bar{u}_n^m H_n''(\bar{x}_i) = \sum_{n=0}^N \bar{u}_n^m H_n''(\bar{x})$ , yield

$$\begin{aligned}
 I_N^{\tilde{\alpha}, \tilde{\beta}} \bar{u}^j(\bar{x}) - \sum_{i=0}^N \bar{u}_i^j H_i(\bar{x}) \\
 = \sum_{m=0}^M \left[ C_1^{jm} \left( I_N^{\tilde{\alpha}, \tilde{\beta}} \bar{u}_{\bar{x}\bar{x}}^m(\bar{x}) - \sum_{n=0}^N \bar{u}_n^m H_n''(\bar{x}) \right) + C_2^{jm} \left( I_N^{\tilde{\alpha}, \tilde{\beta}} \bar{u}^m(\bar{x}) - \sum_{n=0}^N \bar{u}_n^m H_n(\bar{x}) \right) \right]. \quad (3.28)
 \end{aligned}$$

Let

$$\begin{aligned}
 e(\bar{x}, \bar{t}_j) &= e^j(\bar{x}) = \bar{u}^j(\bar{x}) - \sum_{i=0}^N \bar{u}_i^j H_i(\bar{x}), \\
 e_{\bar{x}}(\bar{x}, \bar{t}_j) &= e_{\bar{x}}^j(\bar{x}) = \bar{u}_{\bar{x}}^j(\bar{x}) - \sum_{i=0}^N \bar{u}_i^j H_i'(\bar{x}) \\
 e_{\bar{x}\bar{x}}(\bar{x}, \bar{t}_j) &= e_{\bar{x}\bar{x}}^j(\bar{x}) = \bar{u}_{\bar{x}\bar{x}}^j(\bar{x}) - \sum_{i=0}^N \bar{u}_i^j H_i''(\bar{x}).
 \end{aligned}$$

Thus, we have

$$\begin{aligned}
 |e^j| &\leq C(|J_1(\bar{x})| + |J_2(\bar{x})|) \\
 J_1(x) &= \bar{u}^j(\bar{x}) - I_N^{\tilde{\alpha}, \tilde{\beta}} \bar{u}^j(\bar{x}), \quad J_2(x) = \bar{u}_{\bar{x}\bar{x}}^j(\bar{x}) - I_N^{\tilde{\alpha}, \tilde{\beta}} \bar{u}_{\bar{x}\bar{x}}^j(\bar{x}). \quad (3.29)
 \end{aligned}$$

Similar to the proof of Theorem 3.1, using the  $L^\infty$  and  $L_{\omega^{\alpha,\beta}}^2$  error bounds for the interpolation polynomials in Lemma 3.2, gives

$$\|J_1(\bar{x})\|_{L^\infty} \leq \begin{cases} CN^{\frac{1}{2}-n} |\bar{u}^j(\bar{x})|_{H_{\omega^c}^{n,N}}, & -1 \leq \tilde{\alpha}, \tilde{\beta} < -\frac{1}{2}, \\ CN^{1+\varepsilon_x-n} \log N |\bar{u}^j(\bar{x})|_{H_{\omega^c}^{n,N}}, & \text{otherwise,} \end{cases} \quad (3.30)$$

$$\|J_2(\bar{x})\|_{L^\infty} \leq \begin{cases} CN^{\frac{1}{2}-n} |\bar{u}_{\bar{x}\bar{x}}^j(\bar{x})|_{H_{\omega^c}^{n,N}}, & -1 \leq \tilde{\alpha}, \tilde{\beta} < -\frac{1}{2}, \\ CN^{1+\varepsilon_x-n} \log N |\bar{u}_{\bar{x}\bar{x}}^j(\bar{x})|_{H_{\omega^c}^{n,N}}, & \text{otherwise,} \end{cases}$$

$$\begin{aligned} \|J_1(x)\|_{L_{\tilde{\alpha},\tilde{\beta}}^2} &\leq CN^{-n} |\bar{u}^j(\bar{x})|_{H_{\omega^{\tilde{\alpha},\tilde{\beta}}}^{n,N}}, \\ \|J_2(x)\|_{L_{\omega^{\tilde{\alpha},\tilde{\beta}}}^2} &\leq CN^{-n} |\bar{u}_{\bar{x}\bar{x}}^j(\bar{x})|_{H_{\omega^{\tilde{\alpha},\tilde{\beta}}}^{n,N}}. \end{aligned} \quad (3.31)$$

We note that

$$\begin{aligned} \bar{u}^M(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t}) &= \sum_{j=0}^M \left( \bar{u}^j(\bar{x}) - \sum_{i=0}^N \bar{u}_i^j H_i(\bar{x}) \right) F_j(\bar{t}) \\ &= \sum_{j=0}^M e(\bar{x}, \bar{t}_j) F_j(\bar{t}) = I_M^{\bar{\alpha}, \bar{\beta}} e(\bar{x}, \bar{t}). \end{aligned}$$

Using Lemma 3.4, we have

$$\|\bar{u}^M(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t})\|_{L^\infty} \leq \begin{cases} C \log M \max_{\bar{t} \in [-1,1]} |e(\bar{x}, \bar{t})|, & -1 \leq \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{1+\bar{\varepsilon}} \max_{\bar{t} \in [-1,1]} |e(\bar{x}, \bar{t})|, & \text{otherwise.} \end{cases} \quad (3.32)$$

By Lemma 3.3, we obtain

$$\|\bar{u}^M(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t})\|_{L_{\omega^{\bar{\alpha}, \bar{\beta}}}^2} \leq C \log M \max_{\bar{t} \in [-1,1]} |e(\bar{x}, \bar{t})|. \quad (3.33)$$

Then using (3.29)–(3.33) and above inequality, we will obtain estimate (3.26), the proof is completed.  $\blacksquare$

For the full-discrete error, we state the following result.

**Theorem 3.3** *Let  $\bar{u}(\bar{x}, \bar{t})$  be the solution of the continuous problem (2.13) and let  $\bar{u}_N^M(\bar{x}, \bar{t})$  be the solution of the full-discrete problem (2.24) with the initial condition (1.2) and boundary conditions (2.10). If  $\bar{u}(\bar{x}, \bar{t})$ ,  $\bar{u}_{\bar{x}}(\bar{x}, \bar{t})$  and  $\bar{u}_{\bar{x}\bar{x}}(\bar{x}, \bar{t}) \in H_{\omega^{\tilde{\alpha}, \tilde{\beta}}}^{n,N} \otimes H_{\omega^{\bar{\alpha}, \bar{\beta}}}^{m,M}$  for  $\bar{x}$  and  $\bar{t}$ , with  $n, m > 1$ ,  $\tilde{\varepsilon} = \max(\tilde{\alpha}, \tilde{\beta})$ ,  $\bar{\varepsilon} = \max(\bar{\alpha}, \bar{\beta})$ , then for  $M$  and  $N$  sufficiently large, the following error estimate holds,*

$$\|\bar{u}(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t})\|_{L^\infty} \leq \begin{cases} CM^{\frac{1}{2}-m} Q_1 + CN^{\frac{1}{2}-n} \log M V_1, & -1 \leq \tilde{\alpha}, \tilde{\beta}, \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{1+\bar{\varepsilon}-m} \log M Q_1 + CN^{1+\tilde{\varepsilon}-n} \log N M^{1+\varepsilon_y} V_1, & -\frac{1}{2} \leq \bar{\varepsilon} < \mu - \frac{1}{2}, \end{cases}$$



$$\|\bar{u}(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t})\|_{L^2_{\omega^{\bar{\alpha}, \bar{\beta}}}} \leq \begin{cases} CM^{-m} \left( Q_2 + M^{\frac{1}{2}-\kappa} Q_1 \right) + CN^{-n} \log MV_2, \\ \quad -1 \leq \bar{\alpha}, \bar{\beta}, \bar{\alpha}, \bar{\beta} < -\frac{1}{2}, \\ CM^{-m} \log M \left( Q_2 + M^{1+\bar{\varepsilon}-\kappa} Q_1 \right) + CN^{-n} \log MV_2, \\ \quad -\frac{1}{2} \leq \bar{\varepsilon} < \mu - \frac{1}{2}, \end{cases} \quad (3.34)$$

for  $\kappa \in (0, 1 - \gamma_1)$

*Proof* Using the triangle inequality

$$|\bar{u}(x, y) - \bar{u}_N^M(x, y)| \leq |\bar{u}(\bar{x}, \bar{t}) - \bar{u}^M(\bar{x}, \bar{t})| + |\bar{u}^M(\bar{x}, \bar{t}) - \bar{u}_N^M(\bar{x}, \bar{t})|,$$

Theorem 3.1 and Theorem 3.2, leads to estimation (3.34). ■

## 4 | NUMERICAL EXPERIMENTS

In this section, we present numerical results obtained by the proposed space-time spectral method to support our theoretical statements. The main purpose is to check the convergence behavior of the numerical solution with respect to the time polynomial degree  $M$  and space polynomial degree  $N$  used in the calculation.

**Example 4.1** Consider the following initial and boundary problems of the fractional cable equation

$$u_t(x, t) = {}^R D_t^{1-\mu_1} u_{xx}(x, t) - {}^R D_t^{1-\mu_2} u(x, t),$$

with the initial and boundary condition

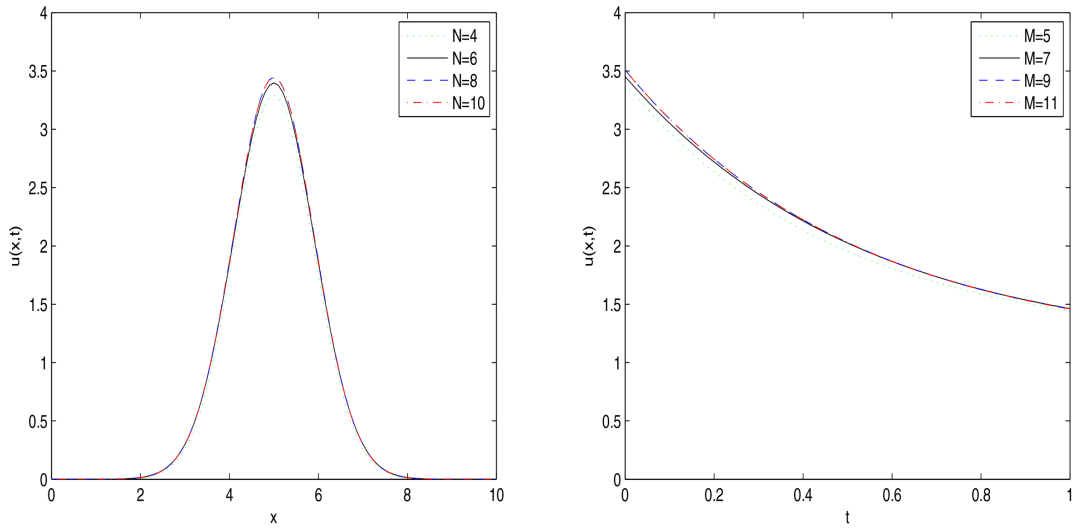
$$\begin{aligned} u(x, 0) &= 10\delta(x - 5), \quad 0 \leq x \leq 10, \\ u(0, t) &= u(10, t) = 0, \quad 0 \leq t \leq 0.5, \end{aligned}$$

where  $\delta(x)$  is the Dirac delta function. This problem does not exist an analytical solution.

In order to test the effectiveness of our method, we set  $\mu_1 = \mu_2 = 0.5$ . First, we fix the number of time collocation points  $M = 36$ , and investigate the behavior of the numerical solutions for  $N = 4, 6, 8, 10$  with  $t = 0.5$ . From Figure 1, we can see that the numerical solution of  $N = 8$  is mostly overlapped with that of  $N = 10$ . We also study the mesh refinement in time with fixing space points  $N = 38$ , it shows similar results. Thus, we obtain that the numerical solution is convergent with the collocation points increased.

**Example 4.2** Consider the following initial-boundary value problem of the fractional Cable equation

$$u_t(x, t) = {}^R D_t^{1-\mu_1} u_{xx}(x, t) - {}^R D_t^{1-\mu_2} u(x, t) + f(x, t),$$



**FIGURE 1** The behavior of numerical solutions for different  $N$  at  $t = 0.5$  (left) and different  $M$  (right) at  $x = 5$  with  $\mu_1 = \mu_2 = 0.5$  [Color figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

with the initial and boundary condition

$$\begin{aligned} u(x, 0) &= 0, \quad 0 \leq x \leq 1, \\ u(0, t) &= u(1, t) = 0, \quad 0 \leq t \leq 1, \end{aligned}$$

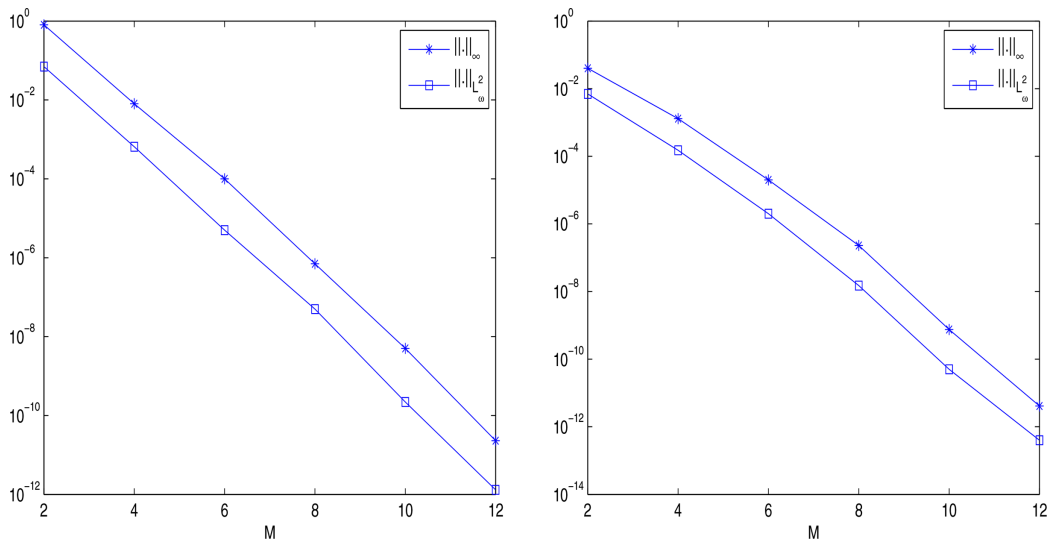
where

$$f(x, t) = \left( 2t + \frac{2\pi^2 t^{\mu_1+1}}{\Gamma(2 + \mu_1)} + \frac{2t^{\mu_2+1}}{\Gamma(2 + \mu_2)} \right) \sin(\pi x).$$

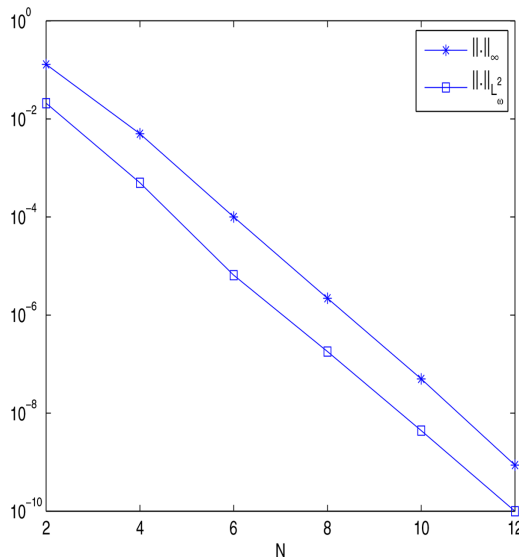
The exact solution is

$$u(x, t) = t^2 \sin(\pi x).$$

We first investigate the temporal errors. In this test, we fix the number of space collocation points  $M = 24$ , a value large enough such that the space discretization error is negligible as compared with the time error. In Figures 2, we plot the errors in the  $L^\infty$  norm and weighted  $L^2$ -norm in semilog scale varying with various of the number of time collocation points  $N$  and various values of the fractional derivatives  $\mu_1$  and  $\mu_2$ . Similarly, we investigate the spatial error by letting the number of space collocation points  $M$  vary and fixing the number of time collocation points  $N = 24$  to avoid contamination of the temporal errors. In Figure 3, we plot the errors as functions of  $M$  for  $\alpha = \beta = 0.5$ . It is clear that the spectral convergence is achieved both of spatial and temporal errors. This indicates that the convergence in space and time of the time-space collocation spectral method is exponential, as in these semilog representation one observes that the error variations are essentially linear versus the degrees of polynomial. This is the so-called spectral accuracy, which can be expected for smooth solutions.



**FIGURE 2** The errors as a function of time collocation points  $M$  for  $\mu_1 = 0.2, \mu_2 = 0.8$  (left) and  $\mu_1 = 0.9, \mu_2 = 0.1$  (right) with  $N = 24$  [Color figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]



**FIGURE 3** The errors as a function of space collocation points  $N$  for  $\mu_1 = 0.5, \mu_2 = 0.5$  with  $M = 24$  [Color figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

## 5 | CONCLUSION

In this article, we consider the numerical solution of the time fractional Cable equation. First, we convert the time fractional Cable equation into an equivalent integral equation with singular kernel, then we propose a spectral collocation method in both time and space discretizations with a spectral expansion of Jacobi interpolation polynomial for this equation. The convergence of the method is rigorously established. Numerical tests are performed to confirm the theoretical results. The main advantage of the present scheme is that it gives very accurate convergence by choosing less number

of grid points and the problem can be solved up to big time, and the storage requirement due to the time memory effect can be considerably reduced. At the same time, the technique is simple and easy to apply to multidimensional problems.

## ACKNOWLEDGMENTS

The work was supported by NSFC Project (11671342, 91430213, 11671339, 11771369), Project of Scientific Research Fund of Hunan Provincial Science and Technology Department (2016JJ3114) and Key Project of Hunan Provincial Department of Education (17A210).

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**How to cite this article:** Yang Y, Huang Y, Zhou Y. Numerical simulation of time fractional Cable equations and convergence analysis. *Numer. Methods Partial Differential Eq.* 2017;00:1–21. <https://doi.org/10.1002/num.22225>