The Efficient Finite Element Methods for Time-Fractional Oldroyd-B Fluid Model Involving Two Caputo Derivatives

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Abstract: In this paper, we consider the numerical schemes for a time-fractional Oldroyd-B fluid model involving the Caputo derivative. We propose two efficient finite element methods by applying the convolution quadrature in time generated by the backward Euler and the second-order backward difference methods. Error estimates in terms of data regularity are established for both the semidiscrete and fully discrete schemes. Numerical examples for two-dimensional problems further confirm the robustness of the schemes with first- and second-order accurate in time.

Keywords: Oldroyd-B fluid model; caputo derivative; finite element method; convolution quadrature; error estimate; data regularity

1 Introduction

In this paper, we investigate efficient numerical schemes for the time-fractional Oldroyd-B fluid model involving the Caputo derivative. Let $\Omega \subset \mathbb{R}^n (n = 1, 2, 3)$ be a bounded convex domain with a smooth boundary $\partial \Omega$, $T > 0$ be a fixed time. The equation is stated as follows.

\[
\left(\partial_t + a_1 cD_{0,t}^{\alpha_1}\right)u(x, t) = \mu \left(1 + a_3 cD_{0,t}^{\gamma_1}\right) \Delta u(x, t) + f(x, t), \quad \text{in } \Omega \times (0, T)
\] (1)

with a homogeneous Dirichlet boundary condition $u(x, t) = 0$ on $\partial \Omega \times (0, T]$ and the initial value conditions $u(x, 0) = v(x)$ and $\partial_t u(x, 0) = b(x)$. Here, the parameters $\mu$, $a_1$ and $a_3$ are some fixed positive constants, the source term $f$ and the initial value conditions $v$ and $b$ are given functions. The fractional orders $\gamma_1 \in (0, 1]$ and $\alpha_1 \in (1, 2]$. The fractional derivative $cD_{0,t}^{\nu}$ with $\nu \in (n-1, n]$ is in Caputo sense defined by

\[
cD_{0,t}^{\nu}u(\cdot, t) = RL D_{0,t}^{-(n-\nu)} \frac{\partial^n}{\partial t^n} u(\cdot, t),
\]

where the Riemann–Liouville (R–L) integral $RL D_{0,t}^{-\nu}$ with $\nu > 0$ is given by

\[
RL D_{0,t}^{-\nu}u(\cdot, t) = \frac{1}{\Gamma(\nu)} \int_0^t (t-s)^{\nu-1} u(\cdot, s) \, ds.
\]

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Here $\frac{\partial^n}{\partial t^n} u(\cdot, t)$ is the $n$-th partial derivative of $u$ with respect to $t$ and $\Gamma(\cdot)$ is the Gamma function.

Over the last few decades, the studies of viscoelastic materials with fractional models have attracted many scholars’ attention [1]. It seems that fractional models can provide another way to preserve the possibility for relatively simple description of the complex behavior of non-Newtonian viscoelastic fluids. The Oldroyd-B fluid can be regarded as one particular subclass of non-Newtonian fluids [2]. One can refer to [3–6] and the references therein for more details about the application. The reason why we prefer to consider the Caputo derivative rather than R–L derivative (see the next section for its definition) in the modeling of abnormal phenomenon in Oldroyd-B fluid is as follows. It is well known that the R–L derivative played an important role in the development of the theory of fractional calculus. However, it seems that the initial conditions for fractional models involving R–L derivative in time is not physically interpretable in applied problems, since they always contain the limit values of the R–L derivative at $t = 0$. So their solutions may not be practically useful due to the unknown physical interpretation for such kinds of initial conditions [7]. This motivates us to use the Caputo derivative instead of R–L derivative in the generalized Oldroyd-B fluid problem, which may lead to a more practical use. By using the method of separation of variables, the analytical solution of Eq. (1) can be derived and expressed in terms of the multivariate Mittag–Leffler function [8]. In this paper, we do not intend to go further for solution theory since it will be off the focus of the current paper. We leave the rigorous analysis of the regularity of the solution to be one of the future possible study.

The numerical schemes for the fractional model (1) or its variants (the Riemann–Liouville derivative version with zero $a_1$ or $a_3$) have been studied extensively in the literatures, [9,10], to name just a few. For more details, see the references therein. Bazhlekova et al. proposed two fully discrete schemes for the Rayleigh–Stokes problem for a generalized second-grade fluid, that is the case $a_1 = 0$ in Riemann–Liouville derivative version [9]. Their numerical schemes are based on the convolution quadrature in time generated by the backward Euler and the second-order backward difference methods. However they only investigated the error estimates for homogeneous problem with $f = 0$. AI-Maskari et al. [10] further studied the finite element methods for time-fractional Oldroyd-B fluid problem with zero initial value condition $b$ by convolution quadrature in time. They derived the optimal error estimates for smooth and nonsmooth data with $v \in \dot{H}^2(\Omega)$ and $v \in L^2(\Omega_1)$ (Section 2 for such definitions), respectively.

Numerical scheme for the case $a_3 = 0$ in Eq. (1) have been studied by some literatures [11–14]. Ren et al. [11] presented a compact alternating direction implicit (ADI) scheme for the two-dimensional time-fractional Cattaneo equation. Zhao et al. [12] proposed a compact Crank–Nicolson scheme and they also considered the two-dimensional case. Chen et al. [13] proposed an efficient ADI Galerkin method for time-fractional partial differential equation with damping. Recently, Li et al. introduced a high-order accurate numerical method for the time-fractional Cattaneo equation. Their numerical scheme is based on the Galkerin–Legendre spectral method in space and the Chebyshev collocation method in time [14]. It is worth noting that Ji et al. [15] proposed a new nanoscale heat transfer model based on the Caputo type fractional dual-phase-lagging heat conduction equation. Their model has the similar structure with the time-fractional Oldroyd-B fluid model, that is the case $\gamma = \alpha - 1$ in Eq. (1). By using the L1 method in time, they developed a finite difference scheme for the time-fractional dual-phase-lagging model.
In order to obtain the desired convergence order, many efficient schemes often rely on some requisite regularity on the solution. Sometimes such assumption on the solution are too strong and make those numerical schemes less efficient or even useless in practical application.

In this paper, we aim to propose the robust and efficient numerical schemes for the fractional model (1) by applying the popular convolution quadrature method. The convolution quadrature method for solving fractional model has received considerable attention [16,17]. The convolution quadrature method developed by Lubich [18] can provide a flexible framework for constructing some reliable numerical methods. So in this paper, we deduce two fully discrete schemes for (1) based on the linear finite element method in space and convolution quadrature in time generated by the backward Euler and the second-order backward difference methods. By the strategy developed in [19], we investigate the error estimates with respect to data regularity (Theorems 4.3 and 4.5).

We find that only smooth data of $v$ fulfill the requirement for the second-order spatial accuracy of the proposed schemes, while the nonsmooth data (that is $v \in L^2(\Omega)$) would deteriorate the spatial accuracy. However, both smooth and nonsmooth data $v$ can ensure the desired temporal convergence orders in the fully discrete schemes. We make some discussions on such situation (Remarks 4.1, 4.3, and 4.5) and numerically confirm our conclusions (Tabs. 7 and 8). To the best of our knowledge, it seems that such results have not been found in the existing literatures yet.

The rest of this paper is organized as follows. In Section 2, we introduce some preliminaries which are need in the numerical study. Two fully discrete schemes using the Galerkin finite element method in space and convolution quadrature in time are developed in Section 3. In Section 4, the error analysis of the schemes are investigated. In Section 5, numerical examples for two-dimensional problems are provided to support our theoretical results. The conclusions are presented in Section 6.

Throughout this paper, the notation $c$ is denoted as a constant which may vary at different occurrences, but is always independent of the mesh size $h$ and the time step size $\tau$.

2 Preliminary

We first list some useful notations in this part.

(1) Some function spaces and the corresponding norms. Denote $(\cdot, \cdot)$ as the inner product on $L^2(\Omega)$ and the $L^2$-norm is $\| \cdot \|$. For $r \geq 0$, let $H^r(\Omega)$ be the subspace of $L^2(\Omega)$ induced by the norm

$$
\| v \|_{H^r(\Omega)} = \| (-\Delta)^{r/2} v \|^2 = \sum_{j=1}^{\infty} \lambda_j^r (v, \phi_j)^2,
$$

where $\{ \lambda_j \}_{j=1}^{\infty}$ and $\{ \phi_j \}_{j=1}^{\infty}$ are the Dirichlet eigenvalues and eigenfunctions of $-\Delta$, respectively. Here, $\{ \phi_j \}_{j=1}^{\infty}$ is an orthonormal basis in $L^2(\Omega)$. Hence, $\| v \|_{H^0(\Omega)} = \| v \|$ is the norm on $L^2(\Omega)$, $\| v \|_{H^1_0(\Omega)} = \| \nabla v \|$ is also the norm on $H^1_0(\Omega)$, and $\| v \|_{H^2(\Omega)} = \| A v \|$ is the equivalent norm in $H^2(\Omega) \cap H^1_0(\Omega)$ [20].
Laplace transform and a related property. The Laplace transform of $f$ with respect to $t$ is denoted as
\[
\hat{f}(z) = L\{f(t); z\} = \int_{0}^{\infty} e^{-zt} f(t) \, dt.
\]
One can readily derive that
\[
L\left\{ c D_{0,t}^{\nu} f(t); z \right\} = z^\nu \hat{f}(z) - \sum_{k=0}^{n-1} z^{\nu-k-1} f^{(k)}(0),
\]
with $n - 1 < \nu \leq n$.

Riemann–Liouville derivative and its relationship with Caputo derivative. For $n - 1 < \nu < n$, the definition of R–L derivative is given by
\[
RL D_{0,t}^{\nu} \varphi(t) = \frac{\partial^n}{\partial t^n} \left( D_{0,t}^{-(n-\nu)} \varphi(t) \right).
\]
The relationships between Caputo and R–L derivatives are as follows.
\[
c D_{0,t}^{\nu} \varphi(t) = RL D_{0,t}^{\nu} \left( \varphi(t) - \varphi(0) \right),
\]
and
\[
c D_{0,t}^{\alpha} \varphi(t) = RL D_{0,t}^{\alpha} \left( \varphi(t) - \varphi(0) - t \varphi'(0) \right),
\]

The resolvent estimate. It is known that the operator $A := -\Delta$ is selfadjoint and positive definite, so the resolvent estimate
\[
\| (z^\nu I + A)^{-1} \| \leq M |z|^{-\nu}, \quad \forall z \in \Sigma_{\theta},
\]
holds \[20\]. Here, the sector $\Sigma_{\theta} = \{ z \in \mathbb{C}, z \neq 0, |\arg z| < \theta \}$ with $\theta \in (\pi/2, \pi)$ and $M$ depends on $\theta$. The notation $\| \cdot \|$ denotes the operator norm from $L^{2}(\Omega) \to L^{2}(\Omega)$, which has the same form with the $L^{2}$-norm.

We also need the notations from the operational calculus \[21\].

Let $K(z)$ be a complex valued or operator valued function that is analytic in a sector $\Sigma_{\theta}$ where $\theta \in (\frac{\pi}{2}, \pi)$. Besides, the function $K(z)$ is bounded by
\[
\|K(z)\| \leq M |z|^{-\lambda}, \quad \forall z \in \Sigma_{\theta},
\]
for some real numbers $\lambda$ and $M$. Then $K(z)$ is the Laplace transform of a distribution $k$ on the real line, which vanishes for $t < 0$, has its singular support empty or concentrated at $t = 0$, and which is an analytic function for $t > 0$ (\[18,21\] for more details). By the inversion Laplace transform for $K(z)$ with $t > 0$, we get
\[
k(t) = \frac{1}{2\pi i} \int_{\Gamma} K(z) e^{zt} \, dz, \quad t > 0.
\]
Here the contour $\Gamma$ lies in $\Sigma_{\theta}$, and parallel to its boundary and oriented with increasing imaginary part. We denote that $\Gamma_{\theta,\delta} = \{ z \in \mathbb{C}: |z| = \delta, |\arg z| \leq \theta \} \cup \{ z \in \mathbb{C}: z = \rho e^{\pm i\theta}, \rho \geq \delta \}$.
We define $K(\partial_t)$ as the operator of convolution with the kernel $k$: $K(\partial_t)g = k * g$. Here $\partial_t$ is the time differentiation and $g$ is suitably smooth. We divide the time interval $[0, T]$ into a uniform grid with a grid point $t_k = k \tau$. Here the time step size $\tau = T/N$ with a positive integer $N$. The convolution quadrature $K(\partial_t)g$ of $K(\partial_t)g$ at $t = t_n$ is given by

$$K(\partial_t)g(t_n) = \sum_{k=0}^n \omega_k(\tau) g(t_{n-k}), \quad (5)$$

where the quadrature weights $\omega_k(\tau)$ are determined by the generating function

$$\sum_{k=0}^\infty \omega_k(\tau) \zeta^k = K\left(\frac{\vartheta(\zeta)}{\tau}\right). \quad (6)$$

Here $\vartheta$ is the quotient of the generating polynomials of linear multistep method [18]. In this paper, we focus on the backward Euler (BE) and the second-order backward difference (SBD) methods, for which $\vartheta(\zeta) = 1 - \zeta$ and $\vartheta(\zeta) = (1 - \zeta) + (1 - \zeta)^2/2$, respectively.

The convolution quadrature has the following error estimates [22].

**Lemma 2.1.** Let $K(z)$ be analytic in $\Sigma_{\theta}$ and (3) hold. Then for $g(t) = ct^{\sigma-1}$, the convolution quadrature based on BE ($p=1$) or SBD ($p=2$) satisfies

$$\| (K(\partial_t) - K(\partial_t)) g(t) \| \leq \begin{cases} c r^{\lambda-1} \tau^\sigma, & 0 < \sigma \leq p, \\ c r^{\lambda-1+\sigma-p} \tau^p, & \sigma \geq p. \end{cases} \quad (7)$$

Finally, we state some basic properties for the functions

$$h(z) = 1/(z + a_1 z^\alpha) \quad \text{and} \quad g(z) = (z + a_1 z^\alpha) / (\mu (1 + a_3 z^\gamma))$$

with $z \in \Sigma_{\theta}$ by the following lemma.

**Lemma 2.2.** Let $\theta \in (\pi/2, \pi/\alpha)$ be fixed. Then, for any $z \in \Sigma_{\theta}$, we have $g(z) \in \Sigma_{\bar{\theta}}$, where $\bar{\theta} = a \theta < \pi$,

$$|g(z)| \leq c \frac{1}{\mu} \left( |z| + a_1 |z|^\alpha \right) \min \left\{ 1, \frac{1}{a_3} |z|^{-\gamma} \right\},$$

and

$$|h(z)| \leq c \min \left\{ |z|^{-1}, \frac{1}{a_1} |z|^{-\alpha} \right\}.$$ 

**Proof.** The property of $g(z)$ has been provided in [10]. The bound $|h(z)|$ can be derived in a similar way with $|g(z)|$, the proof is thus completed.

### 3 Fully Discrete Schemes

In this section, we propose two fully discrete finite element methods by using the standard Galerkin finite element method in space and convolution quadrature in time.
3.1 **Semidiscrete Galerkin Scheme in Space**

A partition of the domain $\Omega$ is denoted by $\mathcal{T}_h$ in which $h$ is the maximal length of the sides of the triangulation $\mathcal{T}_h$. Then the continuous piecewise linear finite element space $V_h$ is given by

$$V_h = \left\{ v_h \in H^1_0(\Omega) : v_h|_T \text{ is a linear function and } \forall T \in \mathcal{T}_h \right\}.$$

The $L^2(\Omega)$ orthogonal projection $P_h : L^2(\Omega) \rightarrow V_h$ is defined by

$$(P_h \varphi, \chi) = (\varphi, \chi), \quad \forall \chi \in V_h,$$

and the Ritz projection $R_h : H^1_0(\Omega) \rightarrow V_h$ is defined by

$$(R_h \varphi, \chi) = (\nabla \varphi, \nabla \chi), \quad \forall \chi \in V_h.$$

From the definitions of $P_h$ and $R_h$, one can see that they are stable in $L^2$ and $H^1_0$, respectively. We remark that these properties would be used frequently and may not be mentioned explicitly in the error estimates. We also need the following approximation properties of these two operators $P_h$ and $R_h$, one may also refer to Theorem 1.1 and Lemma 1.1 in the book [20].

**Lemma 3.1.** The following approximation properties of the operators $P_h$ and $R_h$ hold:

$$||P_h \varphi - \varphi|| + h||\nabla (P_h \varphi - \varphi)|| \leq ch^2||\varphi||_{H^2(\Omega)}, \quad \forall \varphi \in \dot{H}^2(\Omega),$$

$$||R_h \varphi - \varphi|| + h||\nabla (R_h \varphi - \varphi)|| \leq ch^2||\varphi||_{H^2(\Omega)}, \quad \forall \varphi \in \dot{H}^2(\Omega).$$

The semidiscrete Galerkin scheme for (1) reads: Find $u_h(t) = u_h(\cdot, t) \in V_h$ such that

$$\left( \left( \frac{\partial}{\partial t} + a_1 C_D^\alpha \right) u_h(t), \chi \right) + \mu \left( 1 + a_3 C_D^\gamma \right) a\left( u_h(t), \chi \right) = \left( f(t), \chi \right), \quad \forall \chi \in V_h,$$

with the initial value conditions $u_h(0) = v_h \in V_h$ and $\partial_t u_h(0) = b_h \in V_h$. Here $v_h$ and $b_h$ are proper approximations to the functions $v$ and $b$, respectively. The bilinear form $a(u, v)$ is given by $(\nabla u_h, \nabla \chi)$.

By introducing the discrete Laplacian $\Delta_h : V_h \rightarrow V_h$ with the definition:

$$- (\Delta_h \varphi, \chi) = (\nabla \varphi, \nabla \chi), \quad \forall \varphi, \chi \in V_h,$$

and letting $A_h = -\Delta_h$, we can rewrite the semidiscrete scheme (8) as follows.

$$\left( \frac{\partial}{\partial t} + a_1 C_D^\alpha \right) u_h(t) + \mu \left( 1 + a_3 C_D^\gamma \right) A_h u_h(t) = f_h(t), \quad t > 0,$$

where $f_h(t) = P_h f(t)$.

3.2 **Fully Discrete Schemes with Convolution Quadrature in Time**

In this part we apply convolution quadrature based on BE and SBD formulas in time to obtain two fully discrete schemes.

Using the relationship between Caputo and R–L derivatives, we derive from the semidiscrete scheme (9) that

$$\partial_t u_h(t) + a_{1RL} D_{0,t}^\alpha (u_h(t) - v_h - tb_h) + \mu A_h u_h(t) + \mu a_{3RL} D_{0,t}^\gamma A_h (u_h(t) - v_h) = f_h(t),$$

that is

$$\left( \frac{\partial}{\partial t} + a_{1RL} D_{0,t}^\alpha \right) u_h(t) + \mu \left( 1 + a_{3RL} D_{0,t}^\gamma \right) A_h u_h(t) = a_{1RL} D_{0,t}^\alpha (v_h + tb_h) + \mu a_{3RL} D_{0,t}^\gamma A_h v_h + f_h(t).$$

$(10)$
Applying \( K(\partial_t) = \left( \partial_t + a_1 RL D_{0,t}^\eta + \mu \left( 1 + a_3 RL D_{0,t}^\gamma \right) A_h \right)^{-1} \) on both sides of the semidiscrete scheme (10), we obtain
\[
u_h(t) = K(\partial_t) \left( a_1 RL D_{0,t}^\eta(v_h + tb_h) + \mu a_3 RL D_{0,t}^\gamma A_h v_h + f_h(t) \right),
\]
with \( t > 0 \). Combining the convolution quadrature with the associativity of convolution, we have the approximation of \( \nu_h(t_n) \) at \( t_n = n\tau \) with \( U^n_h \) by
\[
U^n_h = K(\partial_t) \left( a_1 \partial_t^\eta (v_h + tb_h) + \mu a_3 \partial_t^\gamma A_h v_h + F^n_h \right),
\]
where \( n = 1, 2, \ldots, N \), \( U^n_h = v_h \), and \( F^n_h = P_h f(t_n) \). Here \( K(\partial_t), \partial_t^\eta \), and \( \partial_t^\gamma \) are the convolution quadratures with \( \vartheta(\zeta) = 1 - \zeta \) for BE scheme, or \( \vartheta(\zeta) = \frac{3}{2} - 2\zeta + \frac{1}{2}\zeta^2 \) for SBD scheme.

Thus, the BE scheme for (12) is stated as: Find \( U^n_h \) for \( n \geq 1 \) such that
\[
\left( \partial_t + a_1 \partial_t^\eta \right) U^n_h + \mu \left( 1 + a_3 \partial_t^\gamma \right) A_h U^n_h = a_1 \partial_t^\eta (v_h + tb_h) + \mu a_3 \partial_t^\gamma A_h v_h + F^n_h.
\]
\[ \text{(13)} \]

**Remark 3.1.** Actually, the term \( \partial_t^\eta tb_h \) in Eq. (13) may also be replaced with the exact expression \( RL D_{0,t}^\eta tb_h \), that is \( RL D_{0,t}^\eta tb_h = t^{1-\alpha}b_h/\Gamma(2-\alpha) \).

Next, we present a robust numerical scheme which can maintain the second-order accuracy for the scheme (12). Since
\[
K(\partial_t) = \left( a_1 RL D_{0,t}^\eta + \mu a_3 RL D_{0,t}^\gamma A_h \right)^{-1} - K(\partial_t) \left( a_1 RL D_{0,t}^\eta + \mu a_3 RL D_{0,t}^\gamma A_h \right)^{-1} (\partial_t + \mu A_h),
\]
the semidiscrete scheme (11) can be rewritten as
\[
u_h(t) = v_h + K(\partial_t) \left( -\left( \partial_t + \partial_t RL D_{0,t}^{-1}(\mu A_h) \right) v_h + a_1 RL D_{0,t}^\eta tb_h + \partial_t RL D_{0,t}^{-1} f_h(0) + \bar{f}_h(t) \right),
\]
where \( t > 0 \) and \( \bar{f}_h(t) = f_h(t) - f_h(0) \). By noting that \( -\partial_t v_h = 0 \), one has
\[
U^n_h = v_h + K(\partial_t) \left( -\partial_t RL D_{0,t}^{-1}(\mu A_h) v_h + a_1 \partial_t^\eta tb_h + \partial_t RL D_{0,t}^{-1} f_h(0) + \bar{F}_h \right). \]
\[ \text{(15)} \]

Letting \( \tau_1 = (0, 3/2, 1, 1, \ldots) \), and noting \( \tau_1 = \partial_t RL D_{0,t}^{-1} \), we obtain the SBD scheme as follows: Find \( U^n_h \) for \( n \geq 1 \) such that
\[
\left( \partial_t + a_1 \partial_t^\eta + \mu \left( 1 + a_3 \partial_t^\gamma \right) A_h \right) \left( U^n_h - v_h \right) = -\tau_1 (\mu A_h) v_h + a_1 \partial_t^\eta tb_h + 1\tau f_h(0) + \bar{F}_h^n.
\]
\[ \text{(16)} \]
that is,
\[
\partial_t U^n_h + a_1 \partial_t^\eta U^n_h + \mu \left( 1 + a_3 \partial_t^\gamma \right) A_h U^n_h = \partial_t U^n_h + a_1 \partial_t^\eta U^n_h + \mu a_3 \partial_t^\gamma A_h U^n_h + a_1 \partial_t^\gamma tb_h + \bar{F}_h^n,
\]
with \( \bar{F}_h^n = F_h^n - \frac{1}{2} (\mu A_h) U^n_h + \frac{1}{2} F^n_h \) when \( n = 1 \), and \( \bar{F}_h^n = F_h^n \) for \( n \geq 2 \).

**4 Error Estimates**

In this section, we derive the error estimates for the proposed numerical schemes by using the strategy developed in [19] which is originated in [21]. For the convenience of discussion, we always assume \( a_1 = 1 \) when necessary.
4.1 Error Estimates for the Semidiscrete Scheme

We first consider the homogeneous problem with \( f = 0 \) and derive the corresponding integral representation of the solution.

An application of the Laplace transform to (1) yields
\[
zh(z) - v + a_1 (z^a \hat{u}(z) - z^{a-1}v - z^{a-2}b) + \mu A (\hat{u}(z) + a_3 (z^{\gamma} \hat{u}(z) - z^{\gamma-1}v)) = 0.
\]

So, \( \hat{u}(z) = \hat{E}(z) \left( \left( 1 + a_1 z^{a-1} + \mu a_3 z^{\gamma-1} \right) v + a_1 z^{a-2}b \right) \),
\[
(17)
\]
where
\( \hat{E}(z) = h(z) g(z) (g(z) I + A)^{-1} \),
and the functions \( h(z) \) and \( g(z) \) are given by Eq. (7).

So the solution \( u(t) \) can be represented by
\[
u(t) = \frac{1}{2\pi i} \int_{\Gamma_{0,\delta}} e^{zt} \hat{E}(z) \left( \left( 1 + a_1 z^{a-1} + \mu a_3 z^{\gamma-1} \right) v + a_1 z^{a-2}b \right) dz.
\]
\[
(18)
\]
Similarly, the solution \( u_h(t) \) to Eq. (9) can be represented by the following:
\[
u_h(t) = \frac{1}{2\pi i} \int_{\Gamma_{0,\delta}} e^{zt} \hat{E}_h(z) \left( \left( 1 + a_1 z^{a-1} + \mu a_3 z^{\gamma-1} \right) v_h + a_1 z^{a-2}b_h \right) dz,
\]
\[
(19)
\]
where
\( \hat{E}_h(z) = h(z) g(z) (g(z) I + A_h)^{-1} \).

From the structures of \( \hat{E}(z) \) and \( \hat{E}_h(z) \), we let \( F_h(z) = (g(z)I + A_h)^{-1} - (g(z)I + A_h)^{-1}P_h \) for notation convenience. The operator \( F_h(z) \) has the following property which plays a key role in the error estimate [9].

**Lemma 4.1.** The following estimate holds:
\[
\|F_h(z) \psi\| + h \|\nabla F_h(z) \psi\| \leq ch^2 \|\psi\|.
\]

Now we are ready to present the error estimates for the semidiscrete problem.

**Theorem 4.1.** Let \( u \) be the solution of problem (1) with \( v \in H^2(\Omega) \) and \( f = 0 \), and set \( u_h \) be the solution of Eq. (9) with \( v_h = R_h v \). Then, the following estimates hold:

(a) If \( b \in \dot{H}^2(\Omega) \) and \( b_h = R_h b \), then
\[
\|e_h(t)\| + h \|\nabla e_h(t)\| \leq ch^2 \left( (t^{\alpha-1} + 1) \|v\|_{H^2(\Omega)} + t \|b\|_{H^2(\Omega)} \right).
\]

(b) If \( b \in L^2(\Omega) \) and \( b_h = P_h b \), then
\[
\|e_h(t)\| + h \|\nabla e_h(t)\| \leq ch^2 \left( (t^{\alpha-1} + 1) \|v\|_{H^2(\Omega)} + t^{\gamma-2} \|b\|_1 \right).
\]
Proof. For \( v, b \in \dot{H}^2(\Omega) \), the error \( e_h(t) := u(t) - u_h(t) \) can be represented by

\[
e_h(t) = \frac{1}{2\pi i} \int_{\Gamma_{\alpha,\beta}} e^{zt} \left( 1 + a_1z^{\alpha - 1} \right) \left( \tilde{E}(z) - \hat{E}_h(z) R_h \right)vdz
\]

\[+ \frac{1}{2\pi i} \int_{\Gamma_{\alpha,\beta}} e^{zt} \left( \mu a_3z^{\gamma - 1} \right) \left( \tilde{E}(z) A - \hat{E}_h(z) A_h R_h \right)vdz
\]

\[+ \frac{1}{2\pi i} \int_{\Gamma_{\alpha,\beta}} e^{zt} a_1z^{\alpha - 2} \left( \tilde{E}(z) - \hat{E}_h(z) R_h \right) b)dz =: I + II + III.
\] (20)

For the first term \( I \), by the identity

\[h(z) g(z) (g(z) I + A)^{-1} = h(z) I - h(z) (g(z) I + A)^{-1} A,
\]

and noting that \( A_hR_h = P_hA \), we can obtain

\[I = \frac{1}{2\pi i} \left( \int_{\Gamma_{\alpha,\beta}} e^{zt} \left( 1 + a_1z^{\alpha - 1} \right) (-h(z)) F_h(z) Avdz + \int_{\Gamma_{\alpha,\beta}} e^{zt} \left( 1 + a_1z^{\alpha - 1} \right) h(z) (v - R_h v) dz \right).
\]

By Lemmas 3.1 and 4.1, we obtain

\[\|I\| \leq c h^2 \|v\|_{H^2(\Omega)} \int_{\Gamma_{\alpha,\beta}} e^{Re(z)t} |h(z)| \left( 1 + a_1 |z|^{\alpha - 1} \right) |dz|
\]

\[\leq c h^2 \|v\|_{H^2(\Omega)} \int_{\Gamma_{\alpha,\beta}} e^{Re(z)t} \min \left\{ |z|^{-1}, \frac{1}{a_1} |z|^{-\alpha} \right\} \left( 1 + a_1 |z|^{\alpha - 1} \right) |dz|
\]

\[\leq c h^2 \|v\|_{H^2(\Omega)} \min \left\{ 1, \frac{1}{a_1} t^{\alpha - 1} \right\} \left( 1 + a_1 t^{-\alpha + 1} \right)
\]

\[\leq c \left( t^{\alpha - 1} + 1 \right) h^2 \|v\|_{H^2(\Omega)}.
\]

Note that the last two terms \( II \) and \( III \) can be recast as

\[II = \frac{1}{2\pi i} \int_{\Gamma_{\alpha,\beta}} e^{zt} \left( \mu a_3z^{\gamma - 1} \right) h(z) g(z) F_h(z) Avdz,
\]

\[III = \frac{1}{2\pi i} \int_{\Gamma_{\alpha,\beta}} e^{zt} a_1z^{\alpha - 2} h(z) (-F_h(z) Ab + (b - R_h b)) dz.
\]

Thus, for the second term \( II \), by Lemma 4.1, we have

\[\|II\| \leq c h^2 \|v\|_{H^2(\Omega)} \int_{\Gamma_{\alpha,\beta}} e^{Re(z)t} \left( \mu a_3 |z|^{\gamma - 1} \right) |h(z) g(z)| |dz|
\]

\[\leq c h^2 \|v\|_{H^2(\Omega)} \int_{\Gamma_{\alpha,\beta}} e^{Re(z)t} \left( a_3 |z|^{\gamma - 1} \right) \min \left\{ 1, \frac{1}{a_3} |z|^{-\gamma} \right\} |dz|
\]

\[\leq c h^2 \|v\|_{H^2(\Omega)} \left( a_3 t^{-\gamma + 1} \right) \min \left\{ t^{-1}, \frac{1}{a_3} t^{\gamma - 1} \right\} \leq c h^2 \|v\|_{H^2(\Omega)},
\]
For the last term III, a use of Lemma 4.1 again yields
\[ \|\text{III}\| \leq ch^2 \|b\|_{H^2(\Omega)} \int_{\Gamma_{\theta,1}} e^{R(z)t} a_1 |z|^{\alpha-2} |h(z)| |dz| \]
\[ \leq ch^2 \|b\|_{H^2(\Omega)} a_1 t^{-\alpha+2} \min\left\{ 1, \frac{t^{\alpha-1}}{t^{\gamma}} \right\} \leq ch^2 t \|b\|_{H^2(\Omega)}. \]

Putting the bounds of I, II, and III together, we complete the proof for the case (a) with \(L^2\)-estimates.

For the second case \(b \in L^2(\Omega)\), by the error estimates in Eq. (20) for the first case (a), one just need to bound the third term III in Eq. (20) with \(P_h\) replacing \(R_h\). By Lemma 4.1, we derive that
\[ \|\text{III}\| \leq ch^2 \|v\| \int_{\Gamma_{\theta,1}} e^{R(z)t} \left( a_1 |z|^{\alpha-2} \right) |h(z)g(z)||dz| \]
\[ \leq ch^2 \|v\| \int_{\Gamma_{\theta,1}} e^{R(z)t} \left( a_1 |z|^{\alpha-2} \right) \min\left\{ 1, \frac{1}{a_3 t^{\gamma-1}} \right\} |dz| \]
\[ \leq ch^2 \|v\| \left( a_1 t^{-\alpha+2} \right) \min\left\{ t^{1}, \frac{1}{a_3 t^{\gamma-1}} \right\} \leq ct^{\gamma-\alpha+1} h^2 \|v\|. \]

So the \(L^2\)-estimates are proved completed. A similar argument yields the \(H^1\)-estimates for both cases (a) and (b).

**Remark 4.1.** The result in Theorem 4.1 may not be improved to the case: \(v, b \in L^2(\Omega)\). As a matter of fact, by the error representation in Eq. (20) (let \(v_h = P_h v\)), one may derive that
\[ \|\text{II}\| \leq \left\| \frac{1}{2\pi i} \int_{\Gamma_{\theta,\delta}} e^{zt} \left( \mu a_3 z^{\gamma-1} \right) h(z) g(z) (v - v_h) (-g(z)F_h(z)v) |dz| \right\| \]
\[ \leq c \|v - v_h\| \int_{\Gamma_{\theta,1}} e^{R(z)t} \left( \mu a_3 |z|^{\gamma-1} \right) |h(z)g(z)| |dz| \]
\[ + ch^2 \|v\| \int_{\Gamma_{\theta,1}} e^{R(z)t} \left( \mu a_3 |z|^{\gamma-1} \right) |h(z)g(z)||g(z)||dz|. \]

By Lemma 3.1, we conclude that the above error estimate is not \(O(h^2)\) for the case \(v \in L^2(\Omega)\) due to the appearance of the term \(\|v - v_h\|\).

Now we consider the inhomogeneous problem with \(f \neq 0\). By applying the Laplace transform, the solution \(u(t)\) can be represented by
\[ u(t) = \int_0^t G(t-s)f(s) \, ds, \quad (21) \]
where
\[ G(t) = \frac{1}{2\pi i} \int_{\Gamma_{\theta,\delta}} e^{zt} \hat{E}(z) \, dz. \]
By a similar argument, the solution $u_h(t)$ to Eq. (9) is represented by

$$u_h(t) = \int_0^t G_h(t-s) P_h f(s) \, ds,$$

(22)

where

$$G_h(t) = \frac{1}{2\pi i} \int_{\Gamma_{\theta,\delta}} e^z \hat{E}_h(z) \, dz.$$

Subtracting Eq. (21) from Eq. (22), we get

$$e_h(t) := u_h(t) - u(t) = \int_0^t (G_h(t-s) P_h - G(t-s)) f(s) \, ds = \int_0^t \tilde{G}_h(t-s) f(s) \, ds.$$

(23)

The operator $\tilde{G}_h$ has the following error estimate.

**Lemma 4.2.** Let $\phi \in L^2(\Omega)$, then there holds

$$\|\tilde{G}_h(t)\| + h\|\nabla \tilde{G}_h(t)\| \leq c h^2 t^{\gamma-1}.$$

**Proof.** By Lemma 4.1, we readily obtain

$$\|\tilde{G}_h(t)\| = \left\| \frac{1}{2\pi i} \int_{\Gamma_{\theta,\delta}} e^z \left( \hat{E}_h(z) P_h - \hat{E}(z) \right) \, dz \right\|$$

$$\leq c h^2 \int_{\Gamma_{\theta,\delta}} e^{Re(z)/t} |h(z) g(z)| \cdot |dz| \leq c h^2 t^{\gamma-1}.$$

A similar argument also yields

$$h\|\nabla \tilde{G}_h(t)\| \leq c h^2 t^{\gamma-1}.$$

The proof is thus completed.

We are in position to state the error estimate for the inhomogeneous problem.

**Theorem 4.2.** Let $u$ and $u_h$ be the solutions of problems (1) with $f \in L^p(0,T;\Omega)$ and $v = b = 0$, and (9) with $v_h = b_h = 0$ and $f_h = P_h f$, respectively. Then, for $t > 0$,

$$\|e_h(t)\| + h\|\nabla e_h(t)\| \leq c h^2 t^{\gamma-1/p} \|f\|_{L^p(0,T;L^2(\Omega))}.$$

**Proof.** Using Lemma 4.2 and Hölder’s inequality for the error $e_h(t)$ in Eq. (23), we can readily derive the desired result. One may refer to Theorem 3.4 in [10] for similar discussions. Thus the proof is completed.

4.2 **Error Estimates for the Fully Discrete Schemes**

In this part, we derive the $L^2$-error estimates for the fully discrete schemes (13) and (16).
4.2.1 Error Estimates for the BE Method

We first consider the homogeneous problem with \( f = 0 \).

Since the spatial error \( u(t) - u_h(t) \) has been discussed in the previous part, we focus on the temporal error \( u_h(t_0) - U_h^n \). We state the error estimates as below.

**Lemma 4.3.** Let \( u_h \) and \( U_h^n \) be the solutions of Eq. (9) and Eq. (13), respectively. Particularly, \( v \in \dot{H}^2(\Omega), f = 0, \) and \( v_h = R_h v \). Then the following error estimates hold:

(a) If \( b \in \dot{H}^2(\Omega) \) and \( b_h = R_h b \), then
\[
\| u_h(t_0) - U_h^n \| \leq c \tau \left( \left( t_n^{-1} + a_3 t_n^{a-\gamma-1} \right) \| v \|_{\dot{H}^2(\Omega)} + \| b \|_{\dot{H}^2(\Omega)} \right).
\]

(b) If \( b \in L^2(\Omega) \) and \( b_h = P_h b \), then
\[
\| u_h(t_0) - U_h^n \| \leq c \tau \left( \left( t_n^{-1} + a_3 t_n^{a-\gamma-1} \right) \| v \|_{\dot{H}^2(\Omega)} + \| b \| \right).
\]

**Proof.** By Eqs. (11) and (12), we have
\[
u_h(t_0) - U_h^n = (F_1 (\partial_t) - F_1 (\partial_x)) v_h + (F_2 (\partial_t) - F_2 (\partial_x)) t b_h = I + II,
\]
where
\[
F_1 (z) = \hat{E}_h (z) \left( a_1 z^\alpha + \mu a_3 z^\gamma A_h \right)
\]
and \( F_2 (z) = \hat{E}_h (z) (a_1 z^\alpha) \).

Recall that \( \hat{E}_h (z) = h (z) g (z) (g (z) I + A_h)^{-1} \), then we have
\[
\| F_1 (z) \| \leq c \left( |z|^\alpha + \mu a_3 |z|^\gamma A_h \right) \min \left\{ |z|^{-1}, |z|^{-\alpha} \right\},
\]
and
\[
\| F_2 (z) \| \leq c |z|^\alpha \min \left\{ |z|^{-1}, |z|^{-\alpha} \right\}.
\]

So it follows from Lemma 2.1 (with \( \sigma = 1, p = 1, \) and the suitable chosen parameters \( \lambda \)) that
\[
\| I \| \leq c \tau \left( \min \left\{ t_n^\alpha, t_n^{-1} \right\} \| R_h v \| + \mu a_3 \min \left\{ t_n^{\gamma'}, t_n^{\alpha-\gamma-1} \right\} \| A_h R_h v \| \right)
\]
\[
\leq c \tau \left( t_n^{-1} \| R_h v \| + a_3 t_n^{a-\gamma-1} \| A v \| \right) \leq c \tau \left( t_n^{-1} + a_3 t_n^{a-\gamma-1} \right) \| A v \|.
\]

where the estimate for the term \( R_h v \) in the last inequality has used the following two facts: \( H_0^1 (\Omega) \)-stability of \( R_h \) and the equivalence between \( \| \nabla v \| \) and \( \| v \|_1 \) in \( H_0^1 (\Omega) \).

For \( b \in L^2(\Omega) \) and \( b_h = P_h b \), by the bound of \( \| F_2 (z) \| \) and Lemma 2.1 (with \( \sigma = 2, p = 1, \) and some suitable parameters \( \lambda \)), the second term \( II \) can be derived that
\[
\| II \| \leq \| (F_2 (\partial_t) - F_2 (\partial_x)) t \| b_h \| \leq c \tau \min \left\{ t_n^{1-\alpha}, 1 \right\} \| b \| \leq c \tau \| b \|.
\]

This together the bound \( \| I \| \) yields the case (b). The case (a) can be derived similarly, so the proof is completed.

**Remark 4.2.** When proving the error estimate in terms of smooth data in \( \dot{H}^2 (\Omega) \), we only can obtain the terms with \( \| v_h \| \) and \( \| b_h \| \) instead of those with \( \| A_h v_h \| \) and \( \| A_h b_h \| \), and then...
bound them by \( \|Av\| \) and \( \|Ab\| \) in view of \( v_h = R_h v \) and \( b_h = R_h b \). One may wonder whether this result can be improved to \( \|A_h v_h\| \) and \( \|A_h b_h\| \) where we can naturally have \( \|Av\| \) and \( \|Ab\| \) by using the identity \( A_h R_h = P_h A \) and the \( L^2 \)-stability of \( P_h \). We find that such situation may not happen unless the term \( \partial_t u \) in Eq. (1) vanishes.

Remark 4.3. By the equality

\[
\hat{E}_h(z) A_h = h(z) g(z) (g(z) I + A_h)^{-1} A_h = h(z) g(z) \left( I - g(z) (g(z) I + A_h)^{-1} \right),
\]

one can improve the above result with respect to the initial data \( v \in L^2(\Omega) \) and \( v_h = P_h v \), that is

\[
\|U_h^n - u_h(t_n)\| \leq c(\tau t_n^{-1}\|v\| + \|b\|).
\]

Indeed, one may consider the term \( \mu a_3 \hat{E}_h(z) z^\gamma A_h =: F_{12}(z) \) in Eq. (24). By noting that

\[
\|F_{12}(z)\| \leq c|z|^\gamma |h(z) g(z)| \leq c|z|^\gamma \min \left\{ 1, \frac{1}{a_3} |z|^{-\gamma} \right\}, \quad \forall z \in \Sigma_\theta,
\]
a use of Lemma 2.1 (with \( \sigma = 1, p = 1 \), and the suitable chosen parameters \( \lambda \)) yields

\[
\| (F_{12}(\partial_t) - F_{12}(\partial_t)) v_h \| \leq c \min \left\{ t_n^{-\gamma - 1}, \frac{1}{a_3} t_n^{-\gamma - 1} \right\} \|v\| \leq c \tau t_n^{-1} \|v\|.
\]

In view of Remark 4.3, we readily obtain the desired results.

We now summarize the results for the error estimates of the fully discrete BE scheme (13) in the next theorem.

Theorem 4.3. Let \( u \) and \( U_h^n \) be the solutions of (1) and (13), respectively. Particularly, \( v \in H^2(\Omega) \), \( f = 0 \), and \( v_h = R_h v \). Then the following error estimates hold:

(a) If \( b \in H^2(\Omega) \) and \( b_h = R_h b \), then

\[
\|U_h^n - u(t_n)\| \leq c \left( \left( t_n^{-1} + a_3 t_n^\alpha - \gamma - 1 \right) \tau + h^2 \right) \|v\|_{H^2(\Omega)} + \left( \tau + h^2 \right) \|b\|_{H^2(\Omega)}.
\]

(b) If \( b \in L^2(\Omega) \) and \( b_h = P_h b \), then

\[
\|U_h^n - u(t_n)\| \leq c \left( \left( t_n^{-1} + a_3 t_n^\alpha - \gamma - 1 \right) \tau + h^2 \right) \|v\|_{H^2(\Omega)} + \left( \tau + h^2 t_n^\gamma \right) \|b\|.
\]

Next we state the error estimates for the inhomogeneous problem with \( f \neq 0 \) but \( v = b = 0 \).

Theorem 4.4. Let \( u \) be the solution of the problem (1) with \( v = b = 0 \) and \( f \in L^\infty(0,T;L^2(\Omega)) \), and set \( U_h^n \) be the solution of (13) with \( f_h = P_h f \). If \( \int_0^T (t_n - s)^{a-1} \|f'(s)\| ds < \infty \) with \( t > 0 \), then we have

\[
\|U_h^n - u(t_n)\| \leq c \left( \tau t_n^{\alpha - 1} \|f'(0)\| + \tau \int_0^{t_n} (t_n - s)^{a-1} \|f'(s)\| ds + h^2 t_n^{\gamma} \|f\|_{L^\infty(0,T;L^2(\Omega))} \right).
\]
Proof. It suffices to bound $\|U^m_h - u_h(t_n)\|$. From (11) and (12), we have

$$U^m_h - u_h(t_n) = (F(\partial_t) - F(\partial_t))f_h,$$

where $F(z) = \tilde{E}_h(z)$.

By the Taylor expansion $f_h(t) = f_h(0) + 1 * f'_h$, we have

$$U^m_h - u_h(t_n) = (K(\partial_t) - K(\partial_t))f_h(0) + ((K(\partial_t) - K(\partial_t))1) * f'_h = I + II.$$

For the first term $I$, noting that $\|F(z)\| \leq c|\lambda(z)|$, we apply Lemma 2.2 and Lemma 2.1 (with $\sigma = 1, p = 1$, and some suitable parameters $\lambda$) to derive that

$$\|I\| = \|(K(\partial_t) - K(\partial_t))f_h(0)\|$$

$$\leq c\tau \min\left\{1, t^\alpha_n^{-1}\right\} \|f_h(0)\| \leq c\tau t^\alpha_n^{-1}\|f(0)\|.$$  

The second term $II$ has the following estimate:

$$\|II\| \leq \int_0^{t_n} \|(K(\partial_t) - K(\partial_t))1\) (t_n - s) f'_h(s)\| ds$$

$$\leq c\tau \int_0^{t_n} (t_n - s)^\alpha - 1 \|f'(s)\| ds.$$  

Combining the above estimates $I$ and $II$ with Theorem 4.2, we complete the proof.

4.2.2 Error Estimate for the SBD Method

We now present the error estimates for the homogeneous problem in the next lemma.

**Lemma 4.4.** Let $u_h$ and $U^m_h$ be the solutions of Eqs. (9) and (16), respectively. Particularly, $v \in \dot{H}^2(\Omega), f = 0$, and $v_h = R_h v, b_h = P_h b$. Then the following error estimates hold:

(a) If $b \in \dot{H}^2(\Omega)$ and $b_h = R_h b$, then

$$\|u_h(t_n) - U^m_h\| \leq c\tau^2 \left(t^\alpha_n - 2\|v\|_{\dot{H}^2(\Omega)} + t^\alpha_n^{-1}\|b\|_{\dot{H}^2(\Omega)}\right).$$

(b) If $b \in L^2(\Omega)$ and $b_h = P_h b$, then

$$\|u_h(t_n) - U^m_h\| \leq c\tau^2 \left(t^\alpha_n - 2\|v\|_{\dot{H}^2(\Omega)} + t^\alpha_n^{-1}\|b\|\right).$$

**Proof.** Let $F_1(z) = \tilde{E}_h(z) z$ and $F_2(z) = \tilde{E}_h(z) z^\alpha$. From Eqs. (14) and (15), one has

$$u_h(t_n) - U^m_h = -(F_1(\partial_t) - F_1(\partial_t))_{RL} D_{0,t}^{-1}(\mu A_h) v_h + (F_2(\partial_t) - F_2(\partial_t)) a_1 t b_h.$$  

From Lemma 2.2, we have

$$\|F_1(z)\| = \|\tilde{E}_h(z) z\| \leq c|\lambda(z)| \min\left\{|z|^{-1}, \frac{1}{a_1} |z|^{-\alpha}\right\}, \ \forall z \in \Sigma_\theta,$$

and

$$\|F_2(z)\| = \|\tilde{E}_h(z) z^\alpha\| \leq c|\lambda(z)|^\alpha \min\left\{|z|^{-1}, \frac{1}{a_1} |z|^{-\alpha}\right\}, \ \forall z \in \Sigma_\theta.$$
So for the case \( b \in L^2 \) and \( b_h = P_h b \), by Lemma 2.1 (with \( \sigma = 2, p = 2 \), and the suitable chosen parameters \( \lambda \)), we derive that

\[
\| u_h (t_n) - U^n_h \| \leq \| -(F_1 (\partial_\tau) - F_1 (\partial_\tau))_{RL} D_{0,t}^{-1} (\mu A_h) v_h \| + \| (F_2 (\partial_\tau) - F_2 (\partial_\tau)) a_1 t h b \|
\]

\[
\leq c t^2 \left( \min \left\{ t_n^{-1}, \alpha_n^{-2} \right\} \| v \|_{H^2(\Omega)} + \min \left\{ t_n^{-\alpha}, t_n^{-1} \right\} \| b \| \right)
\]

\[
\leq c t^2 \left( \alpha_n^{-2} \| v \|_{H^2(\Omega)} + t_n^{-1} \| b \| \right).
\]

The error estimates for the case \( b \in \dot{H}^2 (\Omega) \) and \( b_h = R_h v \) is similar with that in (b), so the proof is completed.

**Remark 4.4.** It seems that the error estimates for the smooth data \( b \in \dot{H}^2 (\Omega) \) in the proof of Lemma 4.4 may not be sharp due to the appearance of the prefactor \( t_n^{-1} \) in \( \| b \|_{H^2(\Omega)} \). Nevertheless, this is enough for the analysis and use of our schemes in this paper. See Remark 4.2 for some similar discussions.

**Remark 4.5.** Similar to the discussion of Remark 4.3, one may obtain the improved result of Lemma 4.4 with respect to the initial data \( v \in L^2 (\Omega) \) and \( v_h = P_h v \), that is

\[
\| U^n_h - u_h (t_n) \| \leq c t^2 \left( \alpha_n^{-2} \| v \| + t_n^{-1} \| b \| \right).
\]

It seems that the convergent order may maintain the first/second-order temporal accuracy for the semidiscrete scheme (14) by using the BE/SBD scheme, even though the initial data is nonsmooth, i.e., \( v \in L^2 (\Omega) \), while one need to impose the condition \( \alpha_3 = 0 \) in order to obtain second-order spatial accuracy in the spatial error estimate of the semidiscrete scheme, see Remark 4.1 for more details.

Now we summarize the error estimates of the SBD scheme (16) for the homogeneous problem.

**Theorem 4.5.** Let \( u \) and \( U^n_h \) be the solutions of (1) with \( v \in \dot{H}^2 (\Omega), b \in L^2 (\Omega), f = 0 \), and (16) with \( v_h = R_h v \), respectively. Then the following error estimates hold:

(a) If \( b \in \dot{H}^2 (\Omega) \) and \( b_h = R_h b \), then

\[
\| U^n_h - u (t_n) \| \leq c \left( t^2 \alpha_n^{-2} \| v \|_{\dot{H}^2(\Omega)} + c \left( t^2 \alpha_n^{-1} \| b \|_{\dot{H}^2(\Omega)} \right) \right);
\]

(b) If \( b \in L^2 (\Omega) \), and \( b_h = P_h b \), then

\[
\| U^n_h - u (t_n) \| \leq c \left( \alpha_n^{-2} \| v \|_{\dot{H}^2(\Omega)} + c \left( t^2 \alpha_n^{-1} \| b \|_{\dot{H}^2(\Omega)} \right) \right).
\]

Finally, we present the following error estimate for the inhomogeneous problem with \( v = b = 0 \).

**Theorem 4.6.** Let \( u \) be the solution of the problem (1) with \( v = b = 0 \) and \( f \in L^{\infty} (0, T; L^2 (\Omega)) \), and let \( U^n_h \) be the solution of (16) with \( v_h = b_h = 0 \). If \( \int_0^t (t - s)^{\alpha - 1} \| f''(s) \| \, ds < \infty \) for \( t > 0 \), then we have

\[
\| u (t_n) - U^n_h \| \leq c \left( t^2 \alpha_n^{-2} \| f \|_{L^{\infty} (0, T; L^2 (\Omega))} + t^2 \alpha_n^{-1} \| f' (0) \| + t^2 \alpha_n^{-1} \| f'' (0) \| + \tau^2 \int_0^{t_n} (t_n - s)^{\alpha - 1} \| f''(s) \| \, ds \right).
\]
Proof. Similar to the proof of Theorem 4.4, we just need to bound \( e^n_h = u_h(t_n) - U^n_h \). Using the Taylor expansion of \( f_h(t) \) at \( t = 0: f_h(t) = f_h(0) + tf_h'(0) + t * f_h'' \), we rewrite the expressions of \( u_h(t_n) \) and \( U^n_h \) in (14) and (15) respectively as

\[
\begin{align*}
    u_h(t_n) &= F(\partial_t) \left( \partial_{1RL} D_{0,t}^{-1} f_h(0) + tf_h'(0) + t * f_h'' \right), \\
    U^n_h &= F(\partial_{1}) \left( \partial_{RL} D_{0,t}^{-1} f_h(0) + tf_h'(0) + t * f_h'' \right),
\end{align*}
\]

where \( F(z) = \hat{E}_h(z) \). So,

\[
\| e^n_h \| \leq \left\| (F(\partial_t) \partial_t - F(\partial_{1}) \partial_{RL}) D_{0,t}^{-1} f_h(0) \right\|
\]

\[
+ \left\| (F(\partial_t) - F(\partial_{1})) t f_h'(0) \right\| + \left\| ((F(\partial_t) - F(\partial_{1})) t) * f_h'' \right\|
\]

\[
:= I + II + III.
\]

Noting that for \( \forall z \in \Sigma_{\theta} \), one has

\[
\| F(z) x \| = \| \hat{E}_h(z) x \| \leq c |z| \min \left\{ |z|^{-\alpha}, \frac{1}{a_1} |z|^{-\alpha} \right\},
\]

and

\[
\| F(z) \| = \| \hat{E}_h(z) \| \leq c \min \left\{ |z|^{-\alpha}, \frac{1}{a_1} |z|^{-\alpha} \right\}.
\]

So for the first two terms \( I \) and \( II \), by Lemma 2.1 (with \( \sigma = 2, p = 2 \), and the suitable chosen parameters \( \lambda \)), we obtain

\[
\| I + II \| \leq c \tau^2 \left( \min \left\{ t_n^{-1}, t_n^\alpha \right\} \| f(0) \| \right. + \left. + \min \left\{ 1, t_n^{\alpha-1} \right\} \| f'(0) \| \right)
\]

\[
\leq c \tau^2 (t_n^{-2} \| f(0) \| + t_n^{-\alpha-1} \| f'(0) \| )
\]

For the third term \( III \), we similarly have

\[
\| III \| \leq \int_0^{t_n} \left\| ((F(\partial_t) - F(\partial_{1})) t (t-s) f_h'(s)) \right\| ds
\]

\[
\leq c \tau^2 \int_0^{t_n} (t_n-s)^{\alpha-1} \| f''(s) \| ds.
\]

Putting the bounds together we thus finish the proof.

5 Numerical Experiments

In this part, we demonstrate numerical examples to verify the theoretical results derived in previous section. In all numerical tests, we fix coefficient \( \mu = a_1 = a_3 = 1 \), unless otherwise specified. The spatial and temporal convergence rates are computed separately on the domain \( \Omega = (0,1)^2 \) and at a fixed final time \( T \). The \( L^2 \)-norm errors are measured by \( \| U^n_h - u_h(t_n) \|_{L^2(\Omega)} \) at \( t_n = T \).
Example 5.1. (The solution is known). Consider problem (1) with the following source term:

\[ f = \sin(\pi x) \sin(\pi y)(pt^{p-1} + 2\pi^2(1 + t^p) + g(p, \alpha) + 2\pi^2 g(p, \gamma)) \]

where \( g(x, y) = \frac{\Gamma(1+x)}{\Gamma(1+x-y)} t^{x-y}. \)

The corresponding solution to (1) is \( u(x, t) = \sin(\pi x) \sin(\pi y)(1 + t). \) Taking \( p = 2.5 \) and \( T = 1, \) we compute the numerical solution of (1) by using the BE (13) and SBD scheme (16), respectively. The convergence rate \( O(\tau^2 + h^2) \) in the \( L^2 \)-norm can be observed from the numerical results in Tabs. 1 and 2, which are in line with the theoretical results.

Example 5.2. (The solution is unknown). Consider the problem (1) with the following data:

(a) \( v = x(1-x)y(1-y), \) \( b = 0, \) and \( f = 0, \)
(b) \( v = 0, \) \( b = x(1-x)y(1-y), \) and \( f = 0, \)
(c) \( v = 0, b = \chi_{[0,1/2] \times (0,1)}(x, y), \) and \( f = 0, \)
(d) \( v = 0, \) \( b = 0, \) and \( f = (1 + t^{1.2})\chi_{[0,1/2] \times (0,1)}(x, y). \)
(e) \( v = \chi_{[0,1/2] \times (0,1)}(x, y), b = 0, \) and \( f = 0. \)

We compute a reference solution in each case on very refined mesh due to the exact solutions of (1) with the above data are difficult to obtain. The numerical results for cases (a)–(d) with \( T = 0.1 \) are demonstrated on Tabs. 3–6, respectively. We observe the \( O(\tau) \) and \( O(\tau^2) \) rates for the temporal error for the BE and SBD schemes of (1), respectively. These observations agree well with the convergence theory.

| Scheme | \( N \) | \((\alpha, \gamma) = (1.9, 0.8)\) | \((\alpha, \gamma) = (1.5, 0.4)\) | \((\alpha, \gamma) = (1.2, 0.1)\) |
|--------|--------|----------------|----------------|----------------|
|        | \( L^2 \) error | Rate | \( L^2 \) error | Rate | \( L^2 \) error | Rate |
| BE     | 4 | 8.98E-02 | – | 4.63E-02 | – | 1.95E-02 | – |
|        | 8 | 4.57E-02 | 0.97 | 2.35E-02 | 0.98 | 9.94E-03 | 0.97 |
|        | 16 | 2.31E-02 | 0.99 | 1.19E-02 | 0.99 | 5.01E-03 | 0.99 |
|        | 32 | 1.16E-02 | 0.99 | 5.95E-03 | 1.00 | 2.51E-03 | 1.00 |
| SBD    | 4 | 1.61E-02 | – | 8.32E-03 | – | 2.83E-03 | – |
|        | 8 | 4.07E-03 | 1.99 | 2.13E-03 | 1.97 | 7.08E-04 | 2.00 |
|        | 16 | 1.00E-03 | 2.02 | 5.34E-04 | 2.00 | 1.74E-04 | 2.03 |
|        | 32 | 2.42E-04 | 2.05 | 1.29E-04 | 2.05 | 3.87E-05 | 2.16 |

We also numerically examine the errors for the nonsmooth initial data \( v \in L^2(\Omega) \) by using the BE and SBD schemes for (1), in which the error estimates are not available in this paper, see Tabs. 7 and 8. We observe that the temporal ones are \( O(\tau) \) and \( O(\tau^2) \) with three different combination of \( \alpha \) and \( \gamma, \) which verify the discussions in Remarks 4.3 and 4.5. From Tab. 8, one can see that the spatial errors are affected by the parameter \( a_3. \) The smaller value of \( a_3, \) the smaller the spatial errors. This further numerically confirms the conclusion in Remark 4.1.
### Table 2: The $L^2$ norm errors in space for Example 5.1 with $N = T/2048$

| Scheme | $1/h$  | $(\alpha, \gamma) = (1.9, 0.8)$ | Rate | $(\alpha, \gamma) = (1.5, 0.4)$ | Rate | $(\alpha, \gamma) = (1.2, 0.1)$ | Rate |
|--------|--------|---------------------------------|------|---------------------------------|------|---------------------------------|------|
|        | $L^2$ error | L^2 error | Rate | L^2 error | Rate | L^2 error | Rate |
| BE     | 4      | 1.19E-01 - | 1.16E-01 - | 1.13E-01 - | | | |
|        | 8      | 3.09E-02 1.94 | 3.01E-02 1.94 | 2.94E-02 1.95 | | | |
|        | 16     | 7.68E-03 2.01 | 7.52E-03 2.00 | 7.38E-03 1.99 | | | |
|        | 32     | 1.80E-03 2.09 | 1.82E-03 2.05 | 1.82E-03 2.02 | | | |
|        | 4      | 1.19E-01 - | 1.16E-01 - | 1.13E-01 - | | | |
| SBD    | 8      | 3.11E-02 1.94 | 3.01E-02 1.94 | 2.94E-02 1.94 | | | |
|        | 16     | 7.85E-03 1.98 | 7.61E-03 1.99 | 7.41E-03 1.99 | | | |
|        | 32     | 1.96E-03 2.00 | 1.90E-03 2.00 | 1.85E-03 2.00 | | | |

### Table 3: The $L^2$ norm errors in time for case (a) in Example 5.2 with $h = 1/10$

| Scheme | $N$  | $(\alpha, \gamma) = (1.9, 0.8)$ | Rate | $(\alpha, \gamma) = (1.5, 0.4)$ | Rate | $(\alpha, \gamma) = (1.2, 0.1)$ | Rate |
|--------|------|---------------------------------|------|---------------------------------|------|---------------------------------|------|
|        | $L^2$ error | L^2 error | Rate | L^2 error | Rate | L^2 error | Rate |
| BE     | 16   | 9.68E-06 - | 1.45E-04 - | 2.79E-04 - | | | |
|        | 32   | 5.03E-06 0.94 | 7.39E-05 0.97 | 1.42E-04 0.97 | | | |
|        | 64   | 2.57E-06 0.97 | 3.73E-05 0.98 | 7.16E-05 0.99 | | | |
|        | 128  | 1.30E-06 0.99 | 1.88E-05 0.99 | 3.60E-05 0.99 | | | |
| SBD    | 16   | 3.88E-06 - | 4.02E-06 - | 6.63E-06 - | | | |
|        | 32   | 9.72E-07 2.00 | 8.88E-07 2.18 | 1.80E-06 1.89 | | | |
|        | 64   | 2.43E-07 2.00 | 2.08E-07 2.09 | 4.64E-07 1.95 | | | |
|        | 128  | 6.07E-08 2.00 | 5.04E-08 2.05 | 1.18E-07 1.98 | | | |

### Table 4: The $L^2$ norm errors in time for case (b) in Example 5.2 with $h = 1/10$

| Scheme | $N$  | $(\alpha, \gamma) = (1.9, 0.8)$ | Rate | $(\alpha, \gamma) = (1.5, 0.4)$ | Rate | $(\alpha, \gamma) = (1.2, 0.1)$ | Rate |
|--------|------|---------------------------------|------|---------------------------------|------|---------------------------------|------|
|        | $L^2$ error | L^2 error | Rate | L^2 error | Rate | L^2 error | Rate |
| BE     | 16   | 4.08E-05 - | 3.11E-05 - | 2.13E-05 - | | | |
|        | 32   | 2.07E-05 0.97 | 1.58E-05 0.98 | 1.08E-05 0.98 | | | |
|        | 64   | 1.05E-05 0.99 | 7.97E-06 0.99 | 5.43E-06 0.99 | | | |
|        | 128  | 5.26E-06 0.99 | 4.00E-06 0.99 | 2.73E-06 1.00 | | | |
| SBD    | 16   | 1.57E-06 - | 1.74E-06 - | 1.21E-06 - | | | |
|        | 32   | 4.11E-07 1.94 | 4.44E-07 1.97 | 3.06E-07 1.98 | | | |
|        | 64   | 1.05E-07 1.97 | 1.12E-07 1.99 | 7.67E-08 1.99 | | | |
|        | 128  | 2.63E-08 1.99 | 2.80E-08 2.00 | 1.92E-08 2.00 | | | |
**Table 5:** The $L^2$ norm errors in time for case (c) in Example 5.2 with $h = 1/10$

| Scheme | $N$ | $\alpha, \gamma = (1.9, 0.8)$ | Rate | $\alpha, \gamma = (1.5, 0.4)$ | Rate | $\alpha, \gamma = (1.2, 0.1)$ | Rate |
|--------|-----|-------------------------------|------|-------------------------------|------|-------------------------------|------|
|        | $L^2$ error | Rate | $L^2$ error | Rate | $L^2$ error | Rate |
| BE     | 16   | 5.00E-04 | – | 3.80E-04 | – | 2.60E-04 | – |
|        | 32   | 2.54E-04 | 0.98 | 1.93E-04 | 0.98 | 1.32E-04 | 0.98 |
|        | 64   | 1.28E-04 | 0.99 | 9.72E-05 | 0.99 | 6.62E-05 | 0.99 |
|        | 128  | 6.43E-05 | 0.99 | 4.88E-05 | 0.99 | 3.32E-05 | 1.00 |
| SBD    | 16   | 3.02E-05 | – | 2.71E-05 | – | 1.73E-05 | – |
|        | 32   | 7.35E-06 | 2.04 | 6.59E-06 | 2.04 | 4.24E-06 | 2.03 |
|        | 64   | 1.82E-06 | 2.02 | 1.63E-06 | 2.02 | 1.05E-06 | 2.01 |
|        | 128  | 4.51E-07 | 2.01 | 4.05E-07 | 2.01 | 2.61E-07 | 2.01 |

**Table 6:** The $L^2$ norm errors in time for case (d) in Example 5.2 with $h = 1/10$

| Scheme | $N$ | $\alpha, \gamma = (1.9, 0.8)$ | Rate | $\alpha, \gamma = (1.5, 0.4)$ | Rate | $\alpha, \gamma = (1.2, 0.1)$ | Rate |
|--------|-----|-------------------------------|------|-------------------------------|------|-------------------------------|------|
|        | $L^2$ error | Rate | $L^2$ error | Rate | $L^2$ error | Rate |
| BE     | 16   | 1.51E-05 | – | 9.13E-05 | – | 1.65E-04 | – |
|        | 32   | 7.77E-06 | 0.96 | 4.65E-05 | 0.97 | 8.35E-05 | 0.98 |
|        | 64   | 3.95E-06 | 0.98 | 2.35E-05 | 0.99 | 4.21E-05 | 0.99 |
|        | 128  | 1.99E-06 | 0.99 | 1.18E-05 | 0.99 | 2.11E-05 | 0.99 |
| SBD    | 16   | 2.13E-06 | – | 4.11E-06 | – | 7.55E-06 | – |
|        | 32   | 5.30E-07 | 2.00 | 9.87E-07 | 2.04 | 1.84E-06 | 2.04 |
|        | 64   | 1.32E-07 | 2.00 | 2.42E-07 | 2.03 | 4.55E-07 | 2.01 |
|        | 128  | 3.29E-08 | 2.00 | 5.98E-08 | 2.01 | 1.13E-07 | 2.01 |

**Table 7:** The $L^2$ norm errors in time for case (e) in Example 5.2 with $h = 1/10$

| Scheme | $N$ | $\alpha, \gamma = (1.9, 0.8)$ | Rate | $\alpha, \gamma = (1.5, 0.4)$ | Rate | $\alpha, \gamma = (1.2, 0.1)$ | Rate |
|--------|-----|-------------------------------|------|-------------------------------|------|-------------------------------|------|
|        | $L^2$ error | Rate | $L^2$ error | Rate | $L^2$ error | Rate |
| BE     | 16   | 8.25E-04 | – | 3.17E-03 | – | 4.03E-03 | – |
|        | 32   | 4.22E-04 | 0.97 | 1.60E-03 | 0.98 | 2.02E-03 | 1.00 |
|        | 64   | 2.14E-04 | 0.98 | 8.05E-04 | 0.99 | 1.01E-03 | 1.00 |
|        | 128  | 1.07E-04 | 0.99 | 4.04E-04 | 1.00 | 5.07E-04 | 1.00 |
| SBD    | 16   | 5.30E-05 | – | 1.95E-04 | – | 3.49E-04 | – |
|        | 32   | 1.29E-05 | 2.04 | 4.79E-05 | 2.03 | 8.19E-05 | 2.09 |
|        | 64   | 3.18E-05 | 2.02 | 1.19E-05 | 2.02 | 1.99E-05 | 2.04 |
|        | 128  | 7.91E-07 | 2.01 | 2.95E-06 | 2.01 | 4.91E-06 | 2.02 |
Table 8: The $L^2$ norm errors in space for case (e) in Example 5.2 with $N = 100$, $\alpha = 1.5$, and $\gamma = 0.4$

| Scheme | $1/h$ | $a_3 = 1$ | $a_3 = 0.01$ | $a_3 = 1E-8$ |
|--------|-------|-----------|---------------|---------------|
|        | $L^2$ error | Rate | $L^2$ error | Rate | $L^2$ error | Rate |
| BE     | 8     | 8.18E-02 | –   | 1.97E-02 | –   | 2.16E-02 | –   |
|        | 16    | 5.71E-02 | 0.52 | 6.64E-03 | 1.57 | 7.03E-03 | 1.62 |
|        | 32    | 4.02E-02 | 0.51 | 2.00E-03 | 1.73 | 1.89E-03 | 1.90 |
|        | 64    | 2.84E-02 | 0.50 | 8.27E-04 | 1.27 | 4.76E-04 | 1.99 |
| SBD    | 8     | 8.18E-02 | –   | 2.04E-02 | –   | 2.24E-02 | –   |
|        | 16    | 5.71E-02 | 0.52 | 7.00E-03 | 1.54 | 7.45E-03 | 1.59 |
|        | 32    | 4.02E-02 | 0.51 | 2.09E-03 | 1.74 | 2.02E-03 | 1.88 |
|        | 64    | 2.84E-02 | 0.50 | 8.40E-04 | 1.32 | 5.10E-04 | 1.98 |

Figure 1: Comparison behavior of $u(x, y, 0.01)$, $u(x, y, 0.1)$ and $u(x, y, 1)$ for (1) with fixed $(\alpha, \gamma) = (1.9, 0.5)$ and different $a_3$, computed by SBD scheme with $h = 0.05$, $N = 200$.
Example 5.3. In this example, we perform numerical simulation on the model (1) with the following data: \( f = 0, b = 0, \) and
\[
v = \frac{1}{2\pi \sigma^2} \exp\left(-\frac{x^2 + y^2}{2\sigma^2}\right),
\]
with the parameter \( \sigma \) denotes the impact level of the initial values. Here, we let \( \sigma = 0.1. \)

The computational domain is on a circle \( \Omega = (x, y)|x^2 + y^2 = 1, \) in which we are more interested. The numerical results are demonstrated in Fig. 1.

For fixed \( a_3 = 0.01 \) in Fig. 1, the peak of the solution is located at the origin at the beginning. However as time goes by, say, \( t = 0.1, \) the peak of the solution to the Eq. (1) is no longer at the origin, but spreads around the origin. This phenomenon becomes more apparent with increasing time and exhibits a period-like evolution. Compared to the case \( a_3 = 0.01, \) the evolution of the solution in case \( a_3 = 0.1 \) becomes slower, see the right-hand column in Fig. 1 for the details. From Fig. 1, we can also observe that the solution of the model is not significantly different for various parameters \( a_3 = 0.01 \) and \( 0.1 \) at the beginning \( t = 0.01. \) As time goes by, the effect of the value of parameter \( a_3 \) on the model becomes more and more obvious. It seems that the smaller the parameter \( a_3, \) the more the fractional model (1) behaves like the time-fractional diffusion-wave equation with damping [13]. This may further validates the claims in [9] that the fractional derivative acting on the Laplacian in the model (1) may be used to capture the viscoelastic behavior of the flow.

6 Conclusions

In this paper we have proposed two efficient and robust fully discrete schemes for the time-fractional Oldroyd-B fluid equation which involves the Caputo derivative in time. These two numerical schemes are based on Galerkin finite element method in space and convolution quadrature in time. The error estimates for the schemes are investigated with respect to data regularity. Numerical results are provided and they are all well agree with the theoretical results.

We find that the spatial errors of the proposed schemes would be deteriorated if the data \( v \in L^2(\Omega) \) due to the appearance of the term \( cD_{0,t}^\lambda \) acting on the Laplacian in (1), while the temporal errors are not affected. It seems that such situation does not exist in other similar fractional model which involves the R–L derivative in time [9,10]. So this makes the numerical investigation of fractional model considered in this paper more difficult. Nevertheless, we make some comments on such situation and numerically confirm our conclusions, see Remarks 4.1, 4.3, and 4.5, and Tabs. 7 and 8.

Our strategy in this paper can be extended to solve other types of fractional models, for example, the non-Newtonian fluid model [2]
\[
(\partial_t + a_1 c D_{0,t}^\beta) u(x, t) + \kappa \left(1 + a_2 c D_{0,t}^\beta\right) u(x, t) = \mu \left(1 + a_3 c D_{0,t}^\beta\right) \Delta u(x, t) + f(x, t),
\]
which can be regarded as a more general case of (1). It seems that the numerical study is trivial so we omit the details here. It is noteworthy that our numerical schemes are not very computationally efficient for long time problems due to the nonlocal nature of Caputo derivative and the high dimensionality of the space. Recently, some authors have proposed fast algorithms to resolve such issue, [23–25], just to name a few. So combining the numerical schemes here with a fast algorithm is an interesting thing to do and worthy of further study.
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