BOUNDARY VALUE METHODS FOR CAPUTO FRACTIONAL DIFFERENTIAL EQUATIONS*

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Abstract

This paper deals with the numerical computation and analysis for Caputo fractional differential equations (CFDEs). By combining the *p*-order boundary value methods (B-VMs) and the *m*-th Lagrange interpolation, a type of extended BVMs for the CFDEs with γ -order ($0 < \gamma < 1$) Caputo derivatives are derived. The local stability, unique solvability and convergence of the methods are studied. It is proved under the suitable conditions that the convergence order of the numerical solutions can arrive at min { $p, m - \gamma + 1$ }. In the end, by performing several numerical examples, the computational efficiency, accuracy and comparability of the methods are further illustrated.

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Key words: Fractional differential equations, Caputo derivatives, Boundary value methods, Local stability, Unique solvability, Convergence.

1. Introduction

In this paper, we consider the following initial value problems of CFDEs

$$y'(t) = f\left(t, y(t), {}_{t_0}^C D_t^{\gamma} y(t)\right), \quad t \in [t_0, T]; \quad y(t_0) = y_0, \tag{1.1}$$

where $f : [t_0, T] \times \mathbb{R}^d \times \mathbb{R}^d \to \mathbb{R}^d$ is a given sufficiently smooth function, $y_0 \in \mathbb{R}^d$ is an assigned initial value and ${}_{t_0}^C D_t^{\gamma} y(t)$ is the γ -order Caputo derivative of the unknown function y(t) defined by (cf. [30, 32, 36])

$${}_{t_0}^C D_t^{\gamma} y(t) = \frac{1}{\Gamma(1-\gamma)} \int_{t_0}^t \frac{y'(v)}{(t-v)^{\gamma}} dv, \quad 0 < \gamma < 1.$$
(1.2)

The model (1.1) has a wide application in science and technology. For example, in McKee [28] and McKee & Stokes [29], the diffusion of discrete particles in a turbulent fluid is modeled by the so-called Basset equation:

$$y'(t) = f(t, y(t)) + c(t) \int_{t_0}^t \frac{y'(v)}{(t-v)^{\gamma}} dv + g(t), \quad t \in [t_0, T]; \quad y(t_0) = y_0,$$
(1.3)

where f(t, y(t)), c(t) and g(t) are the assigned functions. An extended Basset equation

$$y'(t) = f(t, y(t)) + \frac{1}{\Gamma(1-\gamma)} \int_{t_0}^t \frac{k(t, v, y'(v))}{(t-v)^{\gamma}} dv, \quad t \in [t_0, T]; \quad y(t_0) = y_0,$$
(1.4)

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can be found in Brunner & Tang [7] and Hairer & Maass [14]. Another example is the Babenko's model describing the gas pressure in a fluid (cf. [3]) which is given by

$$\begin{cases} \frac{\partial}{\partial t} \left[V_0 \ g(t/\theta) \ P(t,0) \frac{M}{RT} \right] = F \mathcal{D} \left. \frac{\partial C}{\partial x} \right|_{x=0}, \\ -\sqrt{\mathcal{D}} \left. \frac{\partial C}{\partial x} \right|_{x=0} = {}_0^C D_t^{1/2} [C(t,0) - C(0,x)], \quad t \in [0,\theta], \\ P(t,0) = \kappa C(t,0), \quad P(0,x) = \kappa C(0,x), \quad x \in [0,\infty), \end{cases}$$
(1.5)

where V_0 is the initial gas volume, θ is the time of the gas compression to zero volume, $g(t/\theta)$ is the function reflecting the change of gas volume with g(0) = 1 and g(1) = 0, M, R, \mathcal{D}, F denote the gas molar weight, universal gas constant, diffusion coefficient of gas in the fluid and contact surface between the gas and the fluid, respectively, κ is the Henry's constant, C(t, x) is the gas concentration and P(t, x) is the unknown gas pressure. The gas temperature T is assumed to be constant. From the problem (1.5), we can obtain the following initial-value problem for determining the dimensionless gas pressure $p(t) \equiv p(t, x) = \frac{P(t, x)}{P(0, x)}$ near the contact surface:

$$\frac{d}{dt}(g(t)p(t)) + \lambda_0^C D_t^{1/2}[p(t) - 1] = 0, \quad t \in [0, 1]; \quad p(0) = 1.$$
(1.6)

Let y(t) = p(t) - 1, G(t) = g(t)/g'(t) and $\hat{G}(t) = \lambda/g'(t)$. Then (1.6) can be written as a FDE of the form (1.1):

$$G(t)y'(t) + \hat{G}(t) {}_{0}^{C} D_{t}^{1/2} y(t) + y(t) = -1, \ t \in [0,1]; \ y(0) = 0.$$

Besides the above real models, with the semi-discrete method for the spatial variable x, which is also called *method of lines*, the following fractal mobile/immobile transport models (cf. [26,33]):

$$\begin{cases} a_1 \frac{\partial u(x,t)}{\partial t} + a_2 {}_0^C D_t^{\gamma} u(x,t) = a_3 \frac{\partial^2 u(x,t)}{\partial x^2} + f(x,t), \quad (x,t) \in [a,b] \times [t_0,T], \\ u(x,0) = \varphi_0(x), \quad x \in [a,b], \\ u(a,t) = \phi_1(t), \quad u(b,t) = \phi_2(t), \quad t \in [t_0,T] \end{cases}$$

can be transformed into (1.1). A detailed description for this approach refers to Example 6.2. Moreover, some other fractional partial differential equations, such as fractional reaction-subdiffusion equation (cf. [21, 24]), fractional cable equation (cf. [25]) and the equations in references [30, 32, 36], can also be cast into (1.1) by the method of lines.

In contrast to the classical regular Volterra integro-differential equations, the CFDEs have the weakly singular factor $(t - v)^{-\gamma}$ ($0 < \gamma < 1$), which leads to the difficulties to obtain the solutions of the equations. Hence, developing various numerical methods for CFDEs becomes an important issue. In [28,29], for Basset equation (1.3), McKee and Stokes proposed the product integration methods based on backward difference interpolation. Subsequently, for the extended Basset equations (1.4), Brunner and Tang [7] constructed the polynomial spline collocation methods and Hairer and Maass [14] presented the fractional linear multistep methods. As to the other related researches for CFDEs, the readers can find them in [22,26,30,32,36] and the references therein. It should be pointed out that, most of the existed numerical methods for CFDEs are presented for the regularity problems (see e.g. [8,22,27]). However, in general, the solutions of problems (1.1) have the weak singularity at initial point. Hence, it is necessary to consider some computational techniques to treat this issue in order to obtain the expected accuracy of the solution. For this, several approaches have been proposed. For example, in [24,34] the authors used the nonuniform grids to keep errors small near the singularity, and in [9,17] the authors employed the correction terms to restore the theoretical accuracy.

In recent years, due to the fact that BVMs and their block schemes have better stability behavior than the usual linear multistep methods, they have been applied widely to solve various initial and boundary value problems (see e.g. [1, 2, 4-6, 10-12, 15, 16, 18, 19, 38-44]). These researches devoted mainly to the regular equations excepting those in Aceto, Magherini and Novati [1, 2], where the authors extended the generalized Adams methods to solve the following fractional differential equations:

$${}_{t_0}^C D_t^{\gamma} y(t) = f(t, y(t)), \quad t \in [t_0, T], \quad 0 < \gamma < 1; \quad y(t_0) = y_0.$$
(1.7)

It is remarkable that model (1.7) belongs to integral equations because it does not contain any derivative. This shows that model (1.7) is different from CFDEs (1.1). In fact, the research for (1.1) has a greater challenge since it contains both fractional derivative and integer derivative. As we know, up to now, no result has been presented for the BVMs applied to CFDEs (1.1). Hence, in the present paper, we will extend the underlying BVMs to solve (1.1).

The paper is organized as follows. In Section 2, we consider the Lagrange interpolation for γ -order ($0 < \gamma < 1$) Caputo derivatives and investigate its local truncation error. In Section 3, by combining BVMs with the Lagrange interpolation, we derive a class of extended BVMs to solve (1.1). In Section 4, we analyze the local stability and unique solvability of the extended BVMs. In Section 5, under the suitable conditions, we prove that the convergence order of the extended BVMs for (1.1) can arrive at min { $p, m - \gamma + 1$ }, where p and m are the local order of the extended BVMs and the degree of the Lagrange interpolation, respectively. In Section 6, some numerical examples are given to illustrate the computational efficiency, accuracy and comparability of the methods.

2. The Lagrange Interpolation for Caputo Derivatives

Let *m* and *N* be two assigned positive integers, $y(t) \in C^{(m+1)}([t_0, T])$, $t_n = t_0 + nh$ (n = 0, 1, ..., N) and $h = \frac{T-t_0}{N}$. Define the following sets:

$$\mathcal{A}_{ij} = \{a \mid a \in [0, j], a \neq i, a \in \mathbb{Z}\}, \quad \mathcal{B}_{ij} = \{b \mid b \in [-j, 0], b \neq i - j, b \in \mathbb{Z}\}, \\ \mathcal{C}^m_{ij} = \{c \mid c \in [j - m, j], c \neq j - m + i, c \in \mathbb{Z}\}.$$

For any given positive integer s with s < m, we introduce the following notations:

- $\mu_{j,i}^{m,s}$ is the sum of products of all the different combinations of m-s elements in \mathcal{A}_{ij} ;
- $\nu_{i\,i}^{m,s}$ is the sum of products of all the different combinations of m-s elements in \mathcal{B}_{ij} ;
- $\sigma_{j,i}^{m,s}$ is the sum of products of all the different combinations of m-s elements in \mathcal{C}_{ij}^m .

When s = m, we set $\mu_{j,i}^{m,s} = \nu_{j,i}^{m,s} = \sigma_{j,i}^{m,s} = 1$ for all i, j.

In order to construct a class of numerical methods for (1.1) in the subsequent section, we first consider an approximation to the Caputo derivative at t_n . The approximation is divided into the following two cases:

(i) When $t \in [t_{j-m}, t_j]$ $(m < j \le n, m < n \le N), y(t)$ can be approximated by the Lagrange interpolation polynomial of degree m:

$$L_{m,j}(t) = \sum_{i=0}^{m} y(t_{j-i}) \prod_{l=0, l \neq i}^{m} \frac{t - t_{j-l}}{t_{j-i} - t_{j-l}}$$

and thus

$$\begin{split} & \frac{1}{\Gamma(1-\gamma)} \int_{t_{j-m}}^{t_{j}} \frac{y'(v)}{(t_{n}-v)^{\gamma}} dv \approx \frac{1}{\Gamma(1-\gamma)} \int_{t_{j-m}}^{t_{j}} \frac{L'_{m,j}(v)}{(t_{n}-v)^{\gamma}} dv \\ &= \frac{1}{\Gamma(1-\gamma)} \sum_{i=0}^{m} \frac{(-1)^{i} y(t_{j-i})}{i!(m-i)!h^{m}} \int_{t_{j-m}}^{t_{j}} (t_{n}-v)^{-\gamma} \left[\prod_{l=0,l\neq i}^{m} (v-t_{j-l}) \right]' dv \\ &= \frac{1}{\Gamma(1-\gamma)} \left\{ \sum_{i=0}^{m} \frac{(-1)^{i+1} y(t_{j-i})}{i!(m-i)!h^{m}} \left[\sum_{s=1}^{m} \frac{(t_{n}-v)^{s-\gamma}}{\prod_{l=1}^{s} (l-\gamma)} \left(\prod_{l=0,l\neq i}^{m} (v-t_{j-l}) \right)^{(s)} \right] \right\} \Big|_{t_{j-m}}^{t_{j}} \\ &= \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^{m} \omega_{i,j,n}^{m} y(t_{j-i}), \end{split}$$

where

$$\omega_{i,j,n}^{m} = \frac{(-1)^{i+1}}{i!(m-i)!} \sum_{s=1}^{m} \left[\frac{s!}{\prod\limits_{l=1}^{s} (l-\gamma)} (\mu_{m,i}^{m,s} (n-j)^{s-\gamma} - \nu_{m,i}^{m,s} (n-j+m)^{s-\gamma}) \right], \quad 0 \le i \le m.$$

(ii) When $t \in [t_0, t_j]$ $(1 \le j \le m, j \le n \le N)$, y(t) can be approximated by $L_{m,m}(t)$ and thus

$$\frac{1}{\Gamma(1-\gamma)} \int_{t_0}^{t_j} \frac{y'(v)}{(t_n-v)^{\gamma}} dv \approx \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^m \varpi_{i,j,n}^m y(t_{m-i}),$$

where

$$\varpi_{i,j,n}^{m} = \frac{(-1)^{i+1}}{i!(m-i)!} \sum_{s=1}^{m} \left[\frac{s!}{\prod\limits_{l=1}^{s} (l-\gamma)} (\sigma_{j,i}^{m,s} (n-j)^{s-\gamma} - \nu_{m,i}^{m,s} n^{s-\gamma}) \right], \quad 0 \le i \le m.$$

Let R_m^n be the truncation error of Lagrange interpolation for the Caputo derivative at t_n . Then, when $1 \le n \le m$, we have that

and, when $m < n \le N$, we have by selecting a positive integer r with $m(r+1) \ge n$ that

$${}_{t_0}^C D_t^{\gamma} y(t_n) = \frac{1}{\Gamma(1-\gamma)} \int_{t_0}^{t_n} \frac{y'(v)}{(t_n-v)^{\gamma}} \, dv$$

$$= \frac{1}{\Gamma(1-\gamma)} \int_{t_0}^{t_{n-rm}} \frac{L'_{m,m}(v)}{(t_n-v)^{\gamma}} dv + \frac{1}{\Gamma(1-\gamma)} \sum_{j=0}^{r-1} \int_{t_{n-(r-j-1)m}}^{t_{n-(r-j-1)m}} \frac{L'_{m,n-(r-j-1)m}(v)}{(t_n-v)^{\gamma}} dv + R_m^n$$

$$= \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \Lambda_{n,m}^n \Omega_{n,m} \mathcal{Y}_n + R_m^n,$$
(2.2)

where $\mathcal{Y}_n = (y(t_0), y(t_1), \cdots, y(t_n))^T$ and

$$\Omega_{n,m} = \begin{bmatrix} \varpi_{m,1,n}^{m} & \cdots & \cdots & \varpi_{0,1,n}^{m} & 0 & \cdots & 0 \\ \vdots & \cdots & \vdots & \vdots & \cdots & \vdots \\ \varpi_{m,m-1,n}^{m} & \cdots & \cdots & \varpi_{0,m-1,n}^{m} & 0 & \cdots & 0 \\ \varpi_{m,m,n}^{m} & \cdots & \cdots & \varpi_{0,m,n}^{m} & 0 & \cdots & 0 \\ 0 & \omega_{m,m+1,n}^{m} & \cdots & \cdots & \omega_{0,m+1,n}^{m} & \ddots & \vdots \\ \vdots & \ddots & \ddots & & \ddots & 0 \\ 0 & \cdots & 0 & \omega_{m,n,n}^{m} & \cdots & \cdots & \omega_{0,n,n}^{m} \end{bmatrix}_{n \times (n+1)}$$

and $\Lambda^n_{n,m}$ denotes the n-th row of matrix

$$\Lambda_{n,m} = \begin{bmatrix} 1 & & & & & \\ 0 & 1 & & & & \\ \vdots & \ddots & \ddots & & & \\ 0 & & \ddots & \ddots & & & \\ 1 & \ddots & & \ddots & \ddots & & \\ 0 & 1 & \ddots & \ddots & \ddots & \ddots & \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \\ 0 & & \ddots & \ddots & \ddots & \ddots & \ddots & \\ 1 & 0 & & \ddots & \ddots & \ddots & \ddots & \ddots & \\ \vdots & \ddots & \\ \vdots & \cdots & 1 & 0 & \cdots & 0 & 1 & 0 & \cdots & 0 & 1 \end{bmatrix}_{n \times n}$$

in which the first column of $\Lambda_{n,m}$ is the vector $(1, 0, \dots, 0, 1, 0, \dots, 0, 1, \dots)^T$. Write

$$\theta_{i,n}^m = \begin{cases} \varpi_{m-i,n,n}^m, & 0 \le i \le m, \ 1 \le n \le m, \\ (i+1)\text{-th element of vector } \Lambda_{n,m}^n \Omega_{n,m}, & 0 \le i \le n, \ m < n \le N. \end{cases}$$

Then Eqs. (2.1) and (2.2) can be rewritten as

$${}^{C}_{t_{0}}D^{\gamma}_{t}y(t_{n}) = \begin{cases} \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^{m} \theta^{m}_{i,n}y(t_{i}) + R^{n}_{m}, & 1 \le n \le m, \\ \\ \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^{n} \theta^{m}_{i,n}y(t_{i}) + R^{n}_{m}, & m < n \le N. \end{cases}$$
(2.3)

In order to analyze truncation error \mathbb{R}_m^n in (2.3), the following lemma is needed.

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Lemma 2.1 (see e.g. [35]). Suppose that $y(t) \in C^{(m+1)}([t_0,T])$. Then the interpolation $L_{m,j}(t)$ satisfies for all $t \in (t_{j-m},t_j)$ $(m \leq j \leq N)$ that

$$\|y(t) - L_{m,j}(t)\|_{\infty} \le M_{m+1}^j h^{m+1}, \tag{2.4}$$

$$\|y'(t) - L'_{m,j}(t)\|_{\infty} \le M^{j}_{m+1}h^{m},$$
(2.5)

where $M_{m+1}^{j} = \max_{t \in [t_{j-m}, t_{j}]} \|y^{(m+1)}(t)\|_{\infty}.$

We are now in a position to give the estimation of the truncation error \mathbb{R}_m^n in (2.3).

Theorem 2.1. Let $y(t) \in C^{(m+1)}([t_0,T])$. Then, there exists a constant $c_0 > 0$ such that

$$|R_m^n||_{\infty} \le c_0 h^{m-\gamma+1}, \quad 1 \le n \le N, \quad 0 < \gamma < 1.$$

Proof. In the following, we will perform the proof in two cases. When $1 \le n \le m$, it follows from (2.1) and Lemma 2.1 that

$$\begin{split} \|R_{m}^{n}\|_{\infty} &= \left\|\frac{1}{\Gamma(1-\gamma)} \left[\int_{t_{0}}^{t_{n}} \frac{y'(v)}{(t_{n}-v)^{\gamma}} \, dv - \int_{t_{0}}^{t_{n}} \frac{L'_{m,m}(v)}{(t_{n}-v)^{\gamma}} \, dv\right]\right\|_{\infty} \\ &= \left\|\frac{1}{\Gamma(1-\gamma)} \left[\frac{y(v) - L_{m,m}(v)}{(t_{n}-v)^{\gamma}}\right]_{t_{0}}^{t_{n-1}} - \gamma \int_{t_{0}}^{t_{n-1}} \frac{y(v) - L_{m,m}(v)}{(t_{n}-v)^{\gamma+1}} \, dv + \int_{t_{n-1}}^{t_{n}} \frac{[y(v) - L_{m,m}(v)]'}{(t_{n}-v)^{\gamma}} \, dv\right]\right\|_{\infty} \\ &= \left\|\frac{-\gamma}{\Gamma(1-\gamma)} \int_{t_{0}}^{t_{n-1}} \frac{y(v) - L_{m,m}(v)}{(t_{n}-v)^{\gamma+1}} \, dv + \frac{1}{\Gamma(1-\gamma)} \int_{t_{n-1}}^{t_{n}} \frac{[y(v) - L_{m,m}(v)]'}{(t_{n}-v)^{\gamma}} \, dv\right\|_{\infty} \\ &\leq \frac{\gamma M_{m+1}^{m}}{\Gamma(1-\gamma)} h^{m+1} \left|\int_{t_{0}}^{t_{n-1}} (t_{n}-v)^{-\gamma-1} \, dv\right| + \frac{M_{m+1}^{m}}{\Gamma(1-\gamma)} h^{m} \left|\int_{t_{n-1}}^{t_{n}} (t_{n}-v)^{-\gamma} \, dv\right| \\ &\leq \frac{M_{m+1}^{m}}{\Gamma(1-\gamma)} \left(1 + \frac{1}{1-\gamma}\right) h^{m-\gamma+1}. \end{split}$$

When $m < n \le N$, by (2.2) and Lemma 2.1 we have that

$$\begin{split} \|R_{m}^{n}\|_{\infty} &= \left\| \frac{1}{\Gamma(1-\gamma)} \left[\int_{t_{0}}^{t_{n}} \frac{y'(v)}{(t_{n}-v)^{\gamma}} dv - \int_{t_{0}}^{t_{n-rm}} \frac{L'_{m,m}(v)}{(t_{n}-v)^{\gamma}} dv - \sum_{j=0}^{r-1} \int_{t_{n-(r-j)m}}^{t_{n-(r-j-1)m}} \frac{L'_{m,m-(r-j-1)m}(v)}{(t_{n}-v)^{\gamma}} dv \right] \right\|_{\infty} \\ &= \left\| \frac{1}{\Gamma(1-\gamma)} \left[\frac{y(v) - L_{m,m}(v)}{(t_{n}-v)^{\gamma}} \right|_{t_{0}}^{t_{n-rm}} - \gamma \int_{t_{0}}^{t_{n-rm}} \frac{y(v) - L_{m,m}(v)}{(t_{n}-v)^{\gamma+1}} dv + \sum_{j=0}^{r-2} \frac{y(v) - L_{m,n-(r-j-1)m}(v)}{(t_{n}-v)^{\gamma}} \right|_{t_{n-(r-j)m}}^{t_{n-(r-j-1)m}} \\ &- \gamma \sum_{j=0}^{r-2} \int_{t_{n-(r-j)m}}^{t_{n-(r-j-1)m}} \frac{y(v) - L_{m,n-(r-j-1)m}(v)}{(t_{n}-v)^{\gamma+1}} dv + \int_{t_{n-m}}^{t_{n}} \frac{[y(v) - L_{m,n-(v-j-1)m}(v)]}{(t_{n}-v)^{\gamma}} dv \right] \right\|_{\infty} \\ &= \left\| \frac{-\gamma}{\Gamma(1-\gamma)} \int_{t_{0}}^{t_{n-rm}} \frac{y(v) - L_{m,m}(v)}{(t_{n}-v)^{\gamma+1}} dv + \frac{-\gamma}{\Gamma(1-\gamma)} \sum_{j=0}^{r-2} \int_{t_{n-(r-j-1)m}}^{t_{n-(r-j-1)m}} \frac{y(v) - L_{m,n-(r-j-1)m}(v)}{(t_{n}-v)^{\gamma+1}} dv \\ &+ \frac{1}{\Gamma(1-\gamma)} \int_{t_{n-m}}^{t_{n}} \frac{[y(v) - L_{m,n}(v)]'}{(t_{n}-v)^{\gamma}} dv \right\|_{\infty} \\ &\leq \frac{\gamma M_{m+1}^{m}}{\Gamma(1-\gamma)} h^{m+1} \left\| \int_{t_{0}}^{t_{n-rm}} (t_{n}-v)^{-\gamma-1} dv \right\| \\ &+ \frac{\gamma}{\Gamma(1-\gamma)} h^{m+1} \left\| \sum_{j=0}^{r-2} M_{m+1}^{n-(r-j-1)m} \int_{t_{n-(r-j)m}}^{t_{n-(r-j-1)m}} (t_{n}-v)^{-\gamma-1} dv \right\| + \frac{M_{m+1}^{m}}{\Gamma(1-\gamma)} h^{m} \left\| \int_{t_{n-m}}^{t_{n}} (t_{n}-v)^{-\gamma} dv \right\| \end{aligned}$$

		m=1		m=2	m=2		3
γ	h	$\operatorname{err}_1(h)$	\bar{p}_1	$\operatorname{err}_2(h)$	\bar{p}_2	$\operatorname{err}_3(h)$	\bar{p}_3
0.25	1/4	7.0241e-2	_	2.6755e-2	_	1.5071e-2	_
	1/8	2.4071e-2	1.5450	4.1355e-3	2.6937	1.1982e-3	3.6528
	1/16	7.7592e-3	1.6333	6.2175e-4	2.7336	9.4202e-5	3.6690
	1/32	2.4595e-3	1.6576	9.2843e-5	2.7435	7.2358e-6	3.7025
0.5	1/4	2.2343e-1	_	9.9015e-2	_	5.5287e-2	_
	1/8	8.8408e-2	1.3375	1.8250e-2	2.4397	5.2092e-3	3.4078
	1/16	3.2846e-2	1.4285	3.2643e - 3	2.4831	4.7478e-4	3.4557
	1/32	1.1950e-2	1.4588	5.7926e-4	2.4945	4.2526e-5	3.4809
0.75	1/4	$5.5005e{-1}$		2.6502e-1	_	$1.4931e{-1}$	_
	1/8	2.4642e-1	1.1585	5.8219e-2	2.1865	1.6704e-2	3.1600
	1/16	1.0622e-1	1.2140	1.2381e-2	2.2334	1.7908e-3	3.2216
	1/32	4.5201e-2	1.2327	2.6109e-3	2.2455	1.8939e-4	3.2411

Table 2.1: Absolute errors and convergence orders of schemes (2.3) for computing ${}_{0}^{C}D_{t}^{\gamma}[\sin(\pi t)]$.

$$\leq \frac{M_{m+1}}{\Gamma(1-\gamma)} \left[\frac{1}{(rm)^{\gamma}} + \frac{1}{m^{\gamma}} + \frac{1}{(1-\gamma)m^{\gamma-1}} \right] h^{m-\gamma+1},$$

where $M_{m+1} = \max_{t \in [t_0,T]} \|y^{(m+1)}(t)\|_{\infty}$. Therefore, this completes the proof.

We have noted that Li et al. [22] also used the *m*-th Lagrange interpolation to approximate the γ -order Caputo dervative. When m = 1, their scheme is the same as scheme (2.3), which is just L1 method (see e.g. [30,36]). However, when m > 1, they use $L_{j,j}(t)$ to approximate y(t) on $\in [t_{j-1}, t_j]$ ($0 \le j \le m$), which leads to that the approximation can not arrive at the accuracy of order $m - \gamma + 1$ on $[t_0, t_j]$ unless $y^{(j)}(t_0) = 0$ ($0 \le j \le m$). While, in our approach, y(t) is approximated by the *m*-th interpolation $L_{m,m}(t)$ on $[t_0, t_j]$ ($1 \le j \le m$), which, together with Theorem 2.1, implies a high-accuracy numerical approximation for y(t) can be achieved. This also can be testified by the following numerical example.

Example 2.1. In the following, we use the above two numerical schemes to compute the γ -order Caputo derivative of function $y(t) = \sin(\pi t)$ on [0, 5]. Let

$$\operatorname{err}_{m}(h) = \max_{1 \le n \le N} |R_{m}^{n}|, \quad \bar{p}_{m} = \log_{2} \left[\frac{\operatorname{err}_{m}(h)}{\operatorname{err}_{m}(h/2)} \right],$$

to characterize the absolute errors and convergence orders of the schemes, respectively. Taking $h = 1/2^i$ (i = 2, 3, 4, 5) and m = 1, 2, 3, respectively, and then applying scheme (2.3) and Li et al.'s scheme (7) in [22] to compute ${}_0^C D_t^{\gamma}[\sin(\pi t)]$ $(\gamma = 0.25, 0.5, 0.75)$ on [0, 5]. The derived numerical results are displayed in Tables 2.1-2.2, which show that, when m = 1, the both schemes have the same accuracy under the same stepsize; and when m > 1, scheme (2.3) can arrive at the convergence order $m - \gamma + 1$, but Li et al's scheme doesn't arrive at the desired order $m - \gamma + 1$.

3. The Extended BVMs for CFDEs

In the recent years, BVMs have been used successfully to solve various differential equations (see e.g. [1, 2, 4–6, 10–12, 15, 16, 18, 19, 38, 39, 41–43]). A detailed introduction on BVMs for

		m=1		m=2	2	m=3	
γ	h	$\operatorname{err}_1(h)$	\bar{p}_1	$\operatorname{err}_2(h)$	\bar{p}_2	$\operatorname{err}_3(h)$	\bar{p}_3
0.25	1/4	7.0241e-2	_	3.0561e-2	_	2.9155e-2	_
	1/8	2.4071e-2	1.5450	5.1022e-3	2.5825	4.5778e-3	2.6710
	1/16	7.7592e-3	1.6333	8.2352e-4	2.6312	7.3963e-4	2.6298
	1/32	2.4595e-3	1.6576	1.3124e-4	2.6496	1.1220e-4	2.7208
0.5	1/4	2.2343e-1	_	$1.0680e{-1}$	_	$1.0365e{-1}$	_
	1/8	8.8408e-2	1.3375	2.0414e-2	2.3873	1.9073e-2	2.4421
	1/16	3.2846e-2	1.4285	3.7627e - 3	2.4397	$3.5217\mathrm{e}{-3}$	2.4372
	1/32	1.1950e-2	1.4588	6.8363e-4	2.4605	6.3636e-4	2.4684
0.75	1/4	$5.5005e{-1}$	_	2.7312e-1	_	$2.6966e{-1}$	_
	1/8	2.4642e-1	1.1585	6.0728e-2	2.1691	5.9183e-2	2.1879
	1/16	1.0622e-1	1.2140	1.3031e-2	2.2204	$1.2575\mathrm{e}{-2}$	2.2346
	1/32	4.5201e-2	1.2327	$2.7655e{-}3$	2.2363	2.6915e-3	2.2241

Table 2.2: Absolute errors and convergence orders of Li et al's scheme for computing ${}_{0}^{C}D_{t}^{\gamma}[\sin(\pi t)]$.

ordinary differential equations (ODEs) refers to Brugnano and Trigiante's monograph [4], where BVMs are verified to have the better stability than the classical linear multistep methods. We note that, up to now, no result has been found on BVMs for CFDEs (1.1). Hence, in this section, we will extend BVMs to solve (1.1).

For convenience, we first give a brief review to the BVMs for the following d-dimensional problems of ODEs:

$$y'(t) = f(t, y(t)), \quad t \in [t_0, T]; \quad y(t_0) = y_0.$$
 (3.1)

For solving (3.1), the BVMs with k_1 initial values $\{y_i\}_{i=0}^{k_1-1}$ and $k_2(=k-k_1)$ final values $\{y_i\}_{i=N-k_2+1}^N$ ($N \in \mathbb{N}$) can be defined as follows:

$$\sum_{i=-k_1}^{k_2} \alpha_{i+k_1} y_{n+i} = h \sum_{i=-k_1}^{k_2} \beta_{i+k_1} f_{n+i}, \qquad n = k_1, k_1 + 1 \dots, N - k_2,$$
(3.2)

$$\sum_{i=0}^{k} \alpha_i^{(j)} y_i = h \sum_{i=0}^{k} \beta_i^{(j)} f_i, \qquad j = 1, 2, \dots, k_1 - 1, \qquad (3.3)$$

$$\sum_{i=0}^{k} \alpha_{k-i}^{(j)} y_{N-i} = h \sum_{i=0}^{k} \beta_{k-i}^{(j)} f_{N-i}, \qquad j = N - k_2 + 1, N - k_2 + 2, \dots, N, \qquad (3.4)$$

where $y_n \approx y(t_n)$, $f_n = f(t_n, y_n)$, and $\alpha_i, \beta_i, \alpha_i^{(j)}$ and $\beta_i^{(j)}$ are some real coefficients such that schemes (3.2)-(3.4) have the same consistency order. Let \otimes be the Kronecker product, I_d the

 $d \times d$ identity matrix, $Y = \left(y_1^T, y_2^T, \dots, y_N^T\right)^T$, $F(Y) = \left(f_1^T, f_2^T, \dots, f_N^T\right)^T$,

$$A^{e} := \begin{bmatrix} a_{0} | A \end{bmatrix} = \begin{bmatrix} \alpha_{0}^{(1)} & \alpha_{1}^{(1)} & \cdots & \alpha_{k}^{(1)} & & & \\ \vdots & \vdots & \cdots & \vdots & & \\ \alpha_{0}^{(k_{1}-1)} & \alpha_{1}^{(k_{1}-1)} & \cdots & \alpha_{k}^{(k_{1}-1)} & & & \\ \alpha_{0} & \alpha_{1} & \cdots & \alpha_{k} & & & \\ & \ddots & \ddots & \ddots & \ddots & \ddots & & \\ & & \alpha_{0} & \alpha_{1} & \cdots & \alpha_{k} & & \\ & & \ddots & \ddots & \ddots & \ddots & \ddots & \\ & & & \alpha_{0}^{(N-k_{2}+1)} & \alpha_{1}^{(N-k_{2}+1)} & \cdots & \alpha_{k}^{(N-k_{2}+1)} \\ & & \vdots & \vdots & \cdots & \vdots \\ & & & & \alpha_{0}^{(N)} & \alpha_{1}^{(N)} & \cdots & \alpha_{k}^{(N)} \end{bmatrix} \in \mathbb{R}^{N \times (N+1)}$$

and $B^e := [b_0|B]$, which is defined similar to A^e by replacing α_i (resp. $\alpha_i^{(j)}$) with β_i (resp. $\beta_i^{(j)}$). Then BVMs (3.2)-(3.4) can be written in a compact form:

$$(A \otimes I_d)Y + a_0 \otimes y_0 = h(B \otimes I_d)F(Y) + hb_0 \otimes f(t_0, y_0).$$

$$(3.5)$$

A BVM (3.5) is called *consistent of order q* if its local truncation error

$$\check{\delta} := (A \otimes I_d)\bar{Y} + a_0 \otimes y_0 - h(B \otimes I_d)F(\bar{Y}) - hb_0 \otimes f(t_0, y_0) = \mathcal{O}(h^{q+1}), \tag{3.6}$$

where $\overline{Y} = (y(t_1)^T, \dots, y(t_N)^T)^T$. Write $Z = (z_1^T, \dots, z_N^T)^T$, where

$$z_n = \begin{cases} \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^m \theta_{i,n}^m y_i, & 1 \le n \le m, \\ \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^n \theta_{i,n}^m y_i, & m < n \le N \end{cases}$$
(3.7)

denotes the interpolation approximation for $z(t_n) := {}_{t_0}^C D_t^{\gamma} y(t_n)$, $\mathbf{0} = (0, \ldots, 0)^T \in \mathbb{R}^d$ and $F(Y, Z) = \left(f(t_1, y_1, z_1)^T, \ldots, f(t_N, y_N, z_N)^T\right)^T$. Adapting BVMs (3.5) to problem (1.1) yields the following extended BVMs:

$$(A \otimes I_d)Y + a_0 \otimes y_0 = h(B \otimes I_d)F(Y,Z) + hb_0 \otimes f(t_0, y_0, \mathbf{0}).$$

$$(3.8)$$

Let $\overline{Z} = (z(t_1)^T, \dots, z(t_N)^T)^T$. An extended BVM (3.8) is called *consistent of order* q if its local truncation error

$$\tilde{\delta} := (A \otimes I_d)\bar{Y} + a_0 \otimes y(t_0) - h(B \otimes I_d)F(\bar{Y},\bar{Z}) - hb_0 \otimes f(t_0,y(t_0),\mathbf{0}) = \mathcal{O}(h^{q+1}).$$
(3.9)

Remark 3.1. Since, for problem (3.1) it holds that

$$F(\bar{Y}) = \left([y'(t_1)]^T, \dots, [y'(t_N)]^T \right)^T, \quad f(t_0, y_0) = y'(t_0)$$

and for problem (1.1) it holds that

$$F(\bar{Y},\bar{Z}) = \left([y'(t_1)]^T, \dots, [y'(t_N)]^T \right)^T, \quad f(t_0, y(t_0), \mathbf{0}) = y'(t_0),$$

we conclude from (3.6) and (3.9) that an extended BVM (3.8) and its corresponding underlying BVM (3.5) have the same consistency order.

4. Local Stability and Unique Solvability of the Extended BVMs

Let $\langle \cdot, \cdot \rangle$ be the Euclidean inner product on \mathbb{R}^d and $\|\cdot\|_2$ the induced norm by this inner product. For any given vectors $U = (u_1^T, \ldots, u_N^T)^T$ and $V = (v_1^T, \ldots, v_N^T)^T$ in \mathbb{R}^{Nd} , we further define the inner product $\langle \cdot, \cdot \rangle_h$ and the corresponding norm $\|\cdot\|_h$ as follows:

$$\langle U, V \rangle_h = h \sum_{i=1}^N \langle u_i, v_i \rangle, \quad \|U\|_h = \sqrt{h \sum_{i=1}^N \|u_i\|_2^2}.$$

Based on vector norm $\|\cdot\|_h$, we also introduce the following matrix norm $\|\cdot\|_h$:

$$\|\mathcal{M}\|_{h} = \max_{\|U\|_{h}=1} \|\mathcal{M}U\|_{h}, \quad \forall \mathcal{M} \in \mathbb{R}^{(Nd) \times (Nd)}.$$

It is easy to check that $\|\mathcal{M}\|_h = \|\mathcal{M}\|_2$ for all $\mathcal{M} \in \mathbb{R}^{(Nd) \times (Nd)}$. In the following, we always assume that there exist constants $L_1, L_2 > 0$ such that function f in (1.1) satisfies that

$$\langle f(t, y, z) - f(t, \hat{y}, z), y - \hat{y} \rangle \le L_1 \| y - \hat{y} \|_2^2, \quad \forall t \in [t_0, T], \ y, \ \hat{y}, \ z \in \mathbb{R}^d,$$
(4.1)

$$\|f(t, y, z) - f(t, y, \hat{z})\|_{2} \le L_{2} \|z - \hat{z}\|_{2}, \quad \forall t \in [t_{0}, T], \ y, \ z, \ \hat{z} \in \mathbb{R}^{d}.$$

$$(4.2)$$

On the basis of the above settings, we will deal with local stability and unique solvability of the extended BVMs (3.8). Let $\hat{Y} = (\hat{y}_1^T, \dots, \hat{y}_N^T)^T$ be the solution of the following perturbed equation with local perturbation $\delta \in \mathbb{R}^{Nd}$:

$$(A \otimes I_d)\hat{Y} + a_0 \otimes y_0 = h(B \otimes I_d)F(\hat{Y}, \hat{Z}) + hb_0 \otimes f(t_0, y_0, \mathbf{0}) + \delta,$$

$$(4.3)$$

where

$$\hat{z}_{n} = \begin{cases} \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^{m} \theta_{i,n}^{m} \hat{y}_{i}, & 1 \le n \le m, \\ \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^{n} \theta_{i,n}^{m} \hat{y}_{i}, & m < n \le N, \end{cases} \qquad \hat{y}_{0} = y_{0}, \quad \hat{Z} = \left(\hat{z}_{1}^{T}, \dots, \hat{z}_{N}^{T}\right)^{T}.$$

Subtracting (3.8) from (4.3) yields that

$$(A \otimes I_d)\mathbf{v} = h(B \otimes I_d)\mathbf{w} + \delta, \tag{4.4}$$

where $\mathbf{v} = \hat{Y} - Y$, $\mathbf{w} = F(\hat{Y}, \hat{Z}) - F(Y, Z)$. An extended BVM (3.8) is called *locally stable* if there exists a constant c > 0 such that $\|\mathbf{v}\|_h \leq c \|\delta\|_h$. In order to derive the local stability criterion, we first present the following lemma.

Lemma 4.1. Let $D = diag(d_1, d_2, ..., d_N)$ be a given $N \times N$ positive diagonal matrix and conditions (4.1) and (4.2) hold. Then the vectors \mathbf{v} and \mathbf{w} in (4.4) satisfy that

$$\langle \mathbf{v}, (D \otimes I_d) \mathbf{w} \rangle_h \le \mu \langle \mathbf{v}, (D \otimes I_d) \mathbf{v} \rangle_h,$$
(4.5)

where

$$\mu = L_1 + \frac{L_2 h^{-\gamma} \omega_1 \omega_2}{\Gamma(1-\gamma)}, \qquad \omega_2 = \sqrt{\sum_{i=1}^N \sum_{j=1}^N \frac{d_i}{d_j}},$$
$$\omega_1 = \max\left\{\max_{0 \le i \le m} \max_{1 \le n \le m} |\theta_{i,n}^m|, \max_{0 \le i \le n} \max_{m < n \le N} |\theta_{i,n}^m|\right\}.$$

Proof. It follows from (4.1), (4.2) and the discrete Cauchy-Schwartz inequality that

$$\langle \mathbf{v}, (D \otimes I_d) \mathbf{w} \rangle_h = h \sum_{i=1}^N d_i \langle \hat{y}_i - y_i, f(t_i, \hat{y}_i, \hat{z}_i) - f(t_i, y_i, z_i) \rangle$$

$$= h \sum_{i=1}^N d_i \langle \hat{y}_i - y_i, f(t_i, \hat{y}_i, \hat{z}_i) - f(t_i, y_i, \hat{z}_i) \rangle + h \sum_{i=1}^N d_i \langle \hat{y}_i - y_i, f(t_i, y_i, \hat{z}_i) - f(t_i, y_i, z_i) \rangle$$

$$\leq L_1 h \sum_{i=1}^N d_i \| \hat{y}_i - y_i \|_2^2 + L_2 h \sum_{i=1}^N d_i \| \hat{y}_i - y_i \|_2 \| \hat{z}_i - z_i \|_2$$

$$\leq L_1 \langle \mathbf{v}, (D \otimes I_d) \mathbf{v} \rangle_h + \frac{L_2 h^{1-\gamma} \omega_1}{\Gamma(1-\gamma)} \sum_{i=1}^N d_i \| \hat{y}_i - y_i \|_2 \sum_{j=1}^N \| \hat{y}_j - y_j \|_2$$

$$\leq L_1 \langle \mathbf{v}, (D \otimes I_d) \mathbf{v} \rangle_h + \frac{L_2 h^{1-\gamma} \omega_1}{\Gamma(1-\gamma)} \sqrt{\sum_{i=1}^N d_i} \sqrt{\sum_{i=1}^N d_i \| \hat{y}_i - y_i \|_2^2} \sqrt{\sum_{i=1}^N \frac{1}{d_i}} \sqrt{\sum_{i=1}^N d_i \| \hat{y}_i - y_i \|_2^2}$$

Hence the lemma is proven.

Refer to references [10, 15, 37], we introduce the following hypothesis:

 \mathcal{H} : There exist $h_0 > 0$, $N \times N$ positive diagonal matrices \tilde{D} and \hat{D} , and a positive bounded function S(h) on $(0, h_0]$ such that

$$\lambda_{\min}\left(\frac{\mathcal{A}\mathcal{B}^{T} + \mathcal{B}\mathcal{A}^{T}}{2} - h\mu\mathcal{B}\mathcal{B}^{T}\right) \ge S(h), \quad h \in (0, h_{0}],$$
(4.6)

where $\mathcal{A} = (\tilde{D}A\hat{D}) \otimes I_d$, $\mathcal{B} = (\tilde{D}B\hat{D}) \otimes I_d$ and $\lambda_{\min}(\cdot)$ denotes the minimum eigenvalue of a matrix.

With the above arguments, a local stability criterion can be stated as follows.

Theorem 4.1. Assume that conditions (4.1), (4.2) and \mathcal{H} are satisfied. Then the extended BVM (3.8) is locally stable with

$$\|\mathbf{v}\|_{h} \leq \left[\frac{\|\hat{D}\|_{h} \|\mathcal{B}^{T}\|_{h} \|\tilde{D}\|_{h}}{S(h)}\right] \|\delta\|_{h}, \quad h \in (0, h_{0}].$$
(4.7)

Proof. Since S(h) is a positive bounded function on $(0, h_0]$, it suffices to prove that (4.7) is true. Let $\tilde{\mathbf{v}}$ be a vector with $\mathcal{B}^T \tilde{\mathbf{v}} = (\hat{D}^{-1} \otimes I_d) \mathbf{v}$. Then equation (4.4) can be written as

$$\mathcal{A}\mathcal{B}^T \tilde{\mathbf{v}} = h\mathcal{B}(\hat{D}^{-1} \otimes I_d)\mathbf{w} + (\tilde{D} \otimes I_d)\delta.$$
(4.8)

Taking the inner product with $\tilde{\mathbf{v}}$ on both sides of (4.8) yields that

$$\left\langle \tilde{\mathbf{v}}, \mathcal{A}\mathcal{B}^T \tilde{\mathbf{v}} \right\rangle_h = h \left\langle \tilde{\mathbf{v}}, \mathcal{B}(\hat{D}^{-1} \otimes I_d) \mathbf{w} \right\rangle_h + \left\langle \tilde{\mathbf{v}}, (\tilde{D} \otimes I_d) \delta \right\rangle_h.$$
 (4.9)

In terms of the usual properties of inner product, (4.9) is equivalent to

$$\left\langle \tilde{\mathbf{v}}, \mathcal{AB}^T \tilde{\mathbf{v}} \right\rangle_h = h \left\langle \mathbf{v}, (\hat{D}^{-2} \otimes I_d) \mathbf{w} \right\rangle_h + \left\langle \tilde{\mathbf{v}}, (\tilde{D} \otimes I_d) \delta \right\rangle_h.$$
 (4.10)

Also, it follows from Lemma 4.1 that

$$\left\langle \mathbf{v}, (\hat{D}^{-2} \otimes I_d) \mathbf{w} \right\rangle_h \le \mu \left\langle \mathbf{v}, (\hat{D}^{-2} \otimes I_d) \mathbf{v} \right\rangle_h = \mu \left\langle \tilde{\mathbf{v}}, \mathcal{B} \mathcal{B}^T \tilde{\mathbf{v}} \right\rangle_h.$$
 (4.11)

Substituting (4.11) into (4.10) and then applying the Cauchy-Schwartz inequality and equality: $\|\tilde{D} \otimes I_d\|_h = \|\tilde{D}\|_h$ derive that

$$\left\langle \tilde{\mathbf{v}}, \mathcal{A}\mathcal{B}^{T}\tilde{\mathbf{v}} \right\rangle_{h} \leq h\mu \left\langle \tilde{\mathbf{v}}, \mathcal{B}\mathcal{B}^{T}\tilde{\mathbf{v}} \right\rangle_{h} + \left\langle \tilde{\mathbf{v}}, (\tilde{D} \otimes I_{d})\delta \right\rangle_{h} \leq h\mu \left\langle \tilde{\mathbf{v}}, \mathcal{B}\mathcal{B}^{T}\tilde{\mathbf{v}} \right\rangle_{h} + \|\tilde{\mathbf{v}}\|_{h}\|\tilde{D}\|_{h}\|\delta\|_{h}.$$
(4.12)

Moreover, applying identity: $\langle \tilde{\mathbf{v}}, \mathcal{AB}^T \tilde{\mathbf{v}} \rangle_h = \langle \tilde{\mathbf{v}}, (\frac{\mathcal{AB}^T + \mathcal{BA}^T}{2}) \tilde{\mathbf{v}} \rangle_h$ to (4.12) gives that

$$\left\langle \tilde{\mathbf{v}}, \left(\frac{\mathcal{A}\mathcal{B}^T + \mathcal{B}\mathcal{A}^T}{2} - h\mu\mathcal{B}\mathcal{B}^T \right) \tilde{\mathbf{v}} \right\rangle_h \le \|\tilde{\mathbf{v}}\|_h \|\tilde{D}\|_h \|\delta\|_h.$$
(4.13)

Whereas, by the property of inner product, the following inequality holds:

$$\left\langle \tilde{\mathbf{v}}, \left(\frac{\mathcal{A}\mathcal{B}^{T} + \mathcal{B}\mathcal{A}^{T}}{2} - h\mu\mathcal{B}\mathcal{B}^{T}\right) \tilde{\mathbf{v}} \right\rangle_{h} \geq \lambda_{\min} \left(\frac{\mathcal{A}\mathcal{B}^{T} + \mathcal{B}\mathcal{A}^{T}}{2} - h\mu\mathcal{B}\mathcal{B}^{T}\right) \|\tilde{\mathbf{v}}\|_{h}^{2}.$$
 (4.14)

A combination of (4.6), (4.13) and (4.14) generates that

$$\|\tilde{\mathbf{v}}\|_{h} \le \frac{\|\tilde{D}\|_{h} \|\delta\|_{h}}{S(h)}, \quad h \in (0, h_{0}].$$
(4.15)

Since $\mathbf{v} = (\hat{D} \otimes I_d) \mathcal{B}^T \tilde{\mathbf{v}}$, inequality (4.7) can be followed immediately by (4.15).

With Theorem 4.1, we can obtain a unique solvability criterion of the extended BVMs (3.8).

Theorem 4.2. Assume that conditions (4.1), (4.2) and \mathcal{H} are satisfied. Then the extended *BVM* (3.8) is uniquely solvable.

Proof. Write

$$Q^{e} := \begin{bmatrix} q_{0} | Q \end{bmatrix} = \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \begin{bmatrix} \theta_{0,1}^{m} & \theta_{1,1}^{m} & \cdots & \theta_{m,1}^{m} \\ \vdots & \vdots & \cdots & \vdots \\ \theta_{0,m}^{m} & \theta_{1,m}^{m} & \cdots & \theta_{m,m}^{m} \\ \theta_{0,m+1}^{m} & \theta_{1,m+1}^{m} & \cdots & \theta_{m,m+1}^{m} & \theta_{m+1,m+1}^{m} \\ \vdots & \vdots & \cdots & \vdots & \vdots & \ddots \\ \theta_{0,N}^{m} & \theta_{1,N}^{m} & \cdots & \theta_{m,N}^{m} & \theta_{m+1,N}^{m} & \cdots & \theta_{N,N}^{m} \end{bmatrix} \in \mathbb{R}^{N \times (N+1)}.$$

Then, it follows from (3.7) that $Z = q_0 \otimes y_0 + (Q \otimes I_d)Y$. Let \check{Y} be a vector such that $\mathcal{B}^T\check{Y} = (\hat{D}^{-1} \otimes I_d)Y$. With this, the unique solvability of the extended BVM (3.8) is equivalent to the unique solvability of the following equation:

$$\mathcal{AB}^{T}\breve{Y} + (\tilde{D}a_{0}) \otimes y_{0} = h\mathcal{B}(\hat{D}^{-1} \otimes I_{d})F\left((\hat{D} \otimes I_{d})\mathcal{B}^{T}\breve{Y}, q_{0} \otimes y_{0} + ((Q\hat{D}) \otimes I_{d})\mathcal{B}^{T}\breve{Y}\right) + h(\tilde{D}b_{0}) \otimes f(t_{0}, y_{0}, \mathbf{0}).$$

$$(4.16)$$

Firstly, we show that the existence of equation (4.16)'s solution. For this, we introduce function $G : \mathbb{R}^{Nd} \to \mathbb{R}^{Nd}$ as follows:

$$G(U) = \mathcal{A}\mathcal{B}^{T}U + (\tilde{D}a_{0}) \otimes y_{0} - h\mathcal{B}(\hat{D}^{-1} \otimes I_{d})F\left((\hat{D} \otimes I_{d})\mathcal{B}^{T}U, q_{0} \otimes y_{0} + ((Q\hat{D}) \otimes I_{d})\mathcal{B}^{T}U\right)$$

$$-h(\tilde{D}b_0)\otimes f(t_0,y_0,\mathbf{0}), \quad U\in\mathbb{R}^{Nd}.$$

In terms of Theorem 6.4.4 in [31], if we assume that equation (4.16) has its solution, it suffices to prove that G(U) is uniformly monotone. Namely, we need only to show that there exists a constant $\eta > 0$ such that

$$\left\langle G(U) - G(\tilde{U}), U - \tilde{U} \right\rangle_h \ge \eta \|U - \tilde{U}\|_h^2, \quad \forall \ U, \tilde{U} \in \mathbb{R}^{Nd}.$$

$$(4.17)$$

In fact, it follows from condition \mathcal{H} , Lemma 4.1 and the properties of inner product that

$$\begin{split} \left\langle G(U) - G(\tilde{U}), U - \tilde{U} \right\rangle_{h} \\ &= \left\langle \mathcal{A}\mathcal{B}^{T}(U - \tilde{U}), U - \tilde{U} \right\rangle_{h} - h \left\langle \mathcal{B}(\hat{D}^{-1} \otimes I_{d})F\left((\hat{D} \otimes I_{d})\mathcal{B}^{T}U, q_{0} \otimes y_{0} + ((Q\hat{D}) \otimes I_{d})\mathcal{B}^{T}U\right) \right. \\ &- \mathcal{B}(\hat{D}^{-1} \otimes I_{d})F\left((\hat{D} \otimes I_{d})\mathcal{B}^{T}\tilde{U}, q_{0} \otimes y_{0} + ((Q\hat{D}) \otimes I_{d})\mathcal{B}^{T}\tilde{U}\right), U - \tilde{U} \right\rangle_{h} \\ &= \left\langle \mathcal{A}\mathcal{B}^{T}(U - \tilde{U}), U - \tilde{U} \right\rangle_{h} - h \left\langle (\hat{D}^{-2} \otimes I_{d})F\left((\hat{D} \otimes I_{d})\mathcal{B}^{T}U, q_{0} \otimes y_{0} + ((Q\hat{D}) \otimes I_{d})\mathcal{B}^{T}U\right) \right. \\ &- (\hat{D}^{-2} \otimes I_{d})F\left((\hat{D} \otimes I_{d})\mathcal{B}^{T}\tilde{U}, q_{0} \otimes y_{0} + ((Q\hat{D}) \otimes I_{d})\mathcal{B}^{T}\tilde{U}\right), (\hat{D} \otimes I_{d})\mathcal{B}^{T}(U - \tilde{U}) \right\rangle_{h} \\ &\geq \left\langle \mathcal{A}\mathcal{B}^{T}(U - \tilde{U}), U - \tilde{U} \right\rangle_{h} - h\mu \left\langle (\hat{D}^{-1} \otimes I_{d})\mathcal{B}^{T}(U - \tilde{U}), (\hat{D} \otimes I_{d})\mathcal{B}^{T}(U - \tilde{U}) \right\rangle_{h} \\ &= \left\langle \mathcal{A}\mathcal{B}^{T}(U - \tilde{U}), U - \tilde{U} \right\rangle_{h} - h\mu \left\langle \mathcal{B}\mathcal{B}^{T}(U - \tilde{U}), U - \tilde{U} \right\rangle_{h} \\ &= \left\langle \left(\frac{\mathcal{A}\mathcal{B}^{T} + \mathcal{B}\mathcal{A}^{T}}{2} - h\mu \mathcal{B}\mathcal{B}^{T} \right) (U - \tilde{U}), U - \tilde{U} \right\rangle_{h} \end{split}$$

Since S(h) is a positive bounded function on $(0, h_0]$, (4.17) holds and hence the existence of the solution of the extended BVM (3.8) is proven.

The uniqueness of the solution of (3.8) can be shown with a direct application of Theorem 4.1 when setting $\delta = 0$. Hence the proof is completed.

5. Convergence of the Extended BVMs

This section will deal with convergence of the extended BVMs. An extended BVM (3.8) is called *convergent of order* p if it has global error $\|\bar{Y} - Y\|_h = \mathcal{O}(h^p)$. A convergence theorem of methods (3.8) can be stated as follows.

Theorem 5.1. Assume that conditions (4.1), (4.2) and \mathcal{H} hold and the extended BVM (3.8) has consistent order q. Then, when method (3.8) is applied to problem (1.1) with $y(t) \in C^{(m+1)}([t_0,T])$, the derived numerical solution Y is convergent of order min $\{q, m - \gamma + 1\}$.

Proof. Let
$$\hat{\delta} = h(B \otimes I_d)[F(\bar{Y}, \bar{Z}) - F(\bar{Y}, \tilde{Z})], \quad \tilde{Z} = (\tilde{z}_1^T, \dots, \tilde{z}_N^T)^T$$
 and

$$\tilde{z}_n = \begin{cases} \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^m \theta_{i,n}^m y(t_i), & 1 \le n \le m, \\ \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{i=0}^n \theta_{i,n}^m y(t_i), & m < n \le N, \end{cases}$$

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Taking use of the above symbols and (3.9), we have

$$(A \otimes I_d)\bar{Y} + a_0 \otimes y(t_0) = h(B \otimes I_d)F(\bar{Y},\tilde{Z}) + hb_0 \otimes f(t_0,y(t_0),\mathbf{0}) + \tilde{\delta} + \hat{\delta}.$$
 (5.1)

It follows from Theorem 2.1 that there exists a constant $c_0 > 0$ such that

$$||z(t_n) - \tilde{z}_n||_{\infty} \le c_0 h^{m-\gamma+1}, \quad 1 \le n \le N.$$
 (5.2)

This, together with condition (4.2) and $Nh = T - t_0$, implies that

$$\|\delta\|_{h} \leq h\|B\|_{h}\|F(Y,Z) - F(Y,Z)\|_{h}$$

$$= h\|B\|_{h} \sqrt{h \sum_{i=1}^{N} \|f(t_{i}, y(t_{i}), z(t_{i})) - f(t_{i}, y(t_{i}), \tilde{z}_{i})\|_{2}^{2}}$$

$$\leq h\|B\|_{h} L_{2} \sqrt{h \sum_{i=1}^{N} \|z(t_{i}) - \tilde{z}_{i}\|_{2}^{2}} \leq c_{0} L_{2}\|B\|_{h} \sqrt{Ndh^{m-\gamma+\frac{5}{2}}}$$

$$= c_{0} L_{2}\|B\|_{h} \sqrt{d(T-t_{0})} h^{m-\gamma+2}.$$
(5.3)

Subtracting (3.8) from (5.1) yields

$$(A \otimes I_d) \left(\bar{Y} - Y \right) = h(B \otimes I_d) \left[F(\bar{Y}, \tilde{Z}) - F(Y, Z) \right] + \tilde{\delta} + \hat{\delta}.$$
(5.4)

Applying Theorem 4.1 to (5.4) yields for all $h \in (0, h_0]$ that

$$\|\bar{Y} - Y\|_{h} \le \left[\frac{\|\hat{D}\|_{h} \|\mathcal{B}^{T}\|_{h} \|D\|_{h}}{S(h)}\right] \|\tilde{\delta} + \hat{\delta}\|_{h} \le \left[\frac{\|\hat{D}\|_{h} \|\mathcal{B}^{T}\|_{h} \|D\|_{h}}{S(h)}\right] (\|\tilde{\delta}\|_{h} + \|\hat{\delta}\|_{h}).$$
(5.5)

Also, since by hypothesis \mathcal{H} that S(h) is a positive bounded function S(h) on $(0, h_0]$, there exist constants $c_1 > 0$ and $h_1 \in (0, h_0]$ such that

$$S(h) \ge c_1 h, \quad \forall h \in (0, h_1].$$

$$(5.6)$$

Moreover, the q-order consistency of the method implies that there exists a constant $c_2 > 0$ such that

$$\|\tilde{\delta}\|_h \le c_2 h^{q+1}.\tag{5.7}$$

Therefore, a combination of (5.3), (5.5)–(5.7) concludes that the extended BVM (3.8) is convergent of order min $\{q, m - \gamma + 1\}$. This completes the proof.

Remark 5.1. In Theorem 5.1, in order to assure that the convergence order of numerical solution Y can arrive at min $\{q, m - \gamma + 1\}$, we ask that $y(t) \in C^{(m+1)}([t_0, T]) \ (m \ge 1)$. The same assumption can be seen in [8, 22, 27] and the references therein. However, this type of strong smooth assumption could not be satisfied at initial point t_0 for some realistic problems of the form (1.1). Hence, Theorem 5.1 is only applicable to the problems with smooth initial data. As to the high-order convergence condition for the case of nonsmooth initial data, it keeps unknown at present because of the lack of analytical techniques. A similar open issue was also proposed in Ford & Yan [13]. Although it is difficult to give a high-order convergence criterion for the problems with nonsmooth initial data, luckily, some numerical treatment methods have

been suggested (see e.g. [23]). Inspired by the idea in Li, Liang & Yan [23], in the following, we will adopt a technique to improve the computational accuracy of methods (3.8) when the problem has the nonsmooth initial data. Firstly, we divide the subinterval $[t_0, t_m]$ $(m \ge 1)$ by the equispaced nodes with stepsize \tilde{h} :

$$t_0 = t_m^{(0)} < t_m^{(1)} < \dots < t_m^{(n_1)} = t_m$$
, where $n_1 = \left\lfloor mh^{1 - \frac{\min\{q, m - \gamma + 1\}}{2 - \gamma}} \right\rfloor$

which leads to $\mathcal{O}\left(\tilde{h}^{2-\gamma}\right) \approx \mathcal{O}\left(h^{\min\{q,m-\gamma+1\}}\right)$. Secondly, we apply the second-order extended trapezoidal rule (cf. [4]) and the piecewise linear interpolation to problem (1.1) on $[t_0, t_m]$. It is well-known that the piecewise linear interpolation for Caputo derivative (2.3) is just *L*1 method (cf. [30]) and its convergence order is $\mathcal{O}\left(\tilde{h}^{2-\gamma}\right)$ (cf. [22]). Hence we have that

$$\|\tilde{y}_m - y(t_m)\|_h = \mathcal{O}\left(\tilde{h}^{2-\gamma}\right), \quad \text{where } \tilde{y}_m \approx y(t_m).$$

Then, we choose \tilde{y}_m as the computational initial value and apply the extended BVMs (3.8) to problem (1.1) on $[t_m, T]$. In this way, the numerical solution Y can arrive at the theoretical accuracy $\mathcal{O}(h^{\min\{q,m-\gamma+1\}})$.

6. Numerical Experiments

In this section, to illustrate the computational efficiency, accuracy and comparability of the extended BVMs, we present several numerical examples. Combining the underlying BVMs: second-order ETR, third-order GBDF and fourth-order ETR₂ in Brugnano & Trigiante [4] with m-th (m = 1, 2, 3) interpolation (3.7) for the Caputo derivatives, respectively, we can obtain a series of extended BVMs for problems (1.1). For convenience, correspondingly, we write these extended BVMs as ETR(2, 1), GBDF(3, 2) and ETR₂(4, 3). Moreover, in order to show the computational advantages of the extended BVMs, we will consider an adapted version of the product integration methods (PIMs) in [28,29] for (1.1). The adapted k-step ($k + 1 - \gamma$)-order PIMs (APIM(k)) can be expressed as follows:

$$\sum_{l=1}^{k+1} \frac{\nabla^l y_n}{l} = hf(t_n, y_n, z_n), \quad z_n = \frac{h^{-\gamma}}{\Gamma(1-\gamma)} \sum_{l=1}^k \sum_{j=k+1}^n a_l(n-j) \nabla^l y_n, \quad k+1 \le n \le N,$$
(6.1)

where ∇ is the backward difference operator and

$$a_l(n-j) = \int_{-1}^0 \frac{1}{(n-j-\xi)^{\gamma} l!} \frac{d}{d\xi} [\xi(\xi+1)\cdots(\xi+l-1)] d\xi.$$

For the implementation of APIM(k), we will use some (k + 1)-order one-step methods to get its starting values. Besides the comparison with APIM(k), we will also consider another class of extended BVMs, which are constructed by combining the underlying BVMs and Li et al's interpolation scheme (7) in [22]. For example, we can combine the second-order ETR, thirdorder GBDF and fourth-order ETR₂ with Li et al's interpolation scheme with m = 1, 2, 3and write the derived methods as AETR(2, 1), AGBDF(3, 2) and AETR₂(4, 3), respectively. In the following, global error and convergence order of the above methods will be computed respectively by the formulas:

$$\operatorname{err}(h) := \|\bar{Y} - Y\|_h, \quad \hat{p} := \log_2 \left[\frac{\operatorname{err}(h)}{\operatorname{err}(h/2)} \right].$$

Table 6.1: Global errors and convergence orders of ETR(2, 1), GBDF(3, 2) and $\text{ETR}_2(4, 3)$ for problem (6.2).

	ETR(2, 1)		GBDF((3, 2)	$\mathrm{ETR}_{2}(4)$	4, 3)
h	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}
1/2	8.3408e-2	_	3.1622e-2	_	1.6331e-2	_
1/4	3.2184e-2	1.3738	$6.8565e{}3$	2.2054	1.7017e-3	3.2625
1/8	1.2028e-2	1.4199	1.1524e-3	2.5728	1.4692e-4	3.5339
1/16	4.4184e-3	1.4448	2.2832e-4	2.3355	1.1208e-5	3.7124

Table 6.2: Global errors and convergence orders of APIM(k) (k = 1, 2, 3) for problem (6.2).

	$\operatorname{APIM}(1)$		APIM	APIM(2)		I(3)
h	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}
1/2	$1.7484e{-1}$	_	3.2093e-2	_	3.1642e-2	_
1/4	6.5514e-2	1.4162	7.6183e - 3	2.0747	7.2058e-3	2.1346
1/8	2.5109e-2	1.3836	$2.6825e{}3$	1.5059	1.9667e-3	1.8734
1/16	9.5292e-3	1.3978	1.1017e-3	1.2838	6.7490e-4	1.5430

Table 6.3: Global errors and convergence orders of AETR(2, 1), AGBDF(3, 2) and $AETR_2(4, 3)$ for problem (6.2).

	AETR(2, 1)		AGBDF	AGBDF(3, 2)		(4, 3)
h	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}
1/2	8.3408e-2	_	3.8400e-2	_	1.7947e-2	_
1/4	3.2184e-2	1.3738	8.0270e-3	2.2582	1.9063e-3	3.2349
1/8	1.2028e-2	1.4199	1.5110e-3	2.4094	2.0790e-4	3.1968
1/16	4.4184e - 3	1.4448	3.4198e-4	2.1435	3.2784e-5	2.6648

Example 6.1. Consider the nonlinear problem of CFDEs:

$$y'(t) = -\frac{y^2(t)}{1+y^2(t)} + \frac{1}{5}\cos\left({}_0^C D_t^{0.5} y(t)\right) + g(t), \quad t \in [0,3]; \quad y(0) = 0, \tag{6.2}$$

where g(t) is a given function such that problem (6.2) has the exact solution $y(t) = t\sqrt{t}$.

Taking stepsizes $h = 1/2^i$ (i = 1, 2, 3, 4) and then applying ETR(2, 1), GBDF(3, 2) and ETR₂(4, 3) to problem (6.2), respectively, we can obtain a series of numerical solutions of (6.2), where, for attaining the theoretical accuracy presented in Theorem 5.1, we also use the algorithm in Remark 5.1 to cope with the nonsmooth initial condition. As an example, the numerical solution y_n solved by ETR₂(4, 3) with h = 1/16 is plotted in Fig. 6.1(a), and the error $\varepsilon_n = |y(t_n) - y_n|$ is shown in Fig. 6.1(b). A detailer description on the global errors and convergence orders of the above extended BVMs for (6.2) is displayed in Table 6.1. These numerical results confirm the computational effectiveness of the extended BVMs and the theoretical accuracy shown in Theorem 5.1.

In order to exhibit the computational advantage of the extended BVMs, we further apply APIM(k) (k = 1, 2, 3), AETR(2, 1), AGBDF(3, 2) and AETR₂(4, 3) with $h = 1/2^i$ (i = 1, 2, 3, 4) to problem (6.2), where the nonsmooth initial condition is treated by an algorithm similar to Remark 5.1. The errors and convergence orders of the above methods are listed in Tables 6.2-6.3. By comparing the numerical results in Tables 6.1-6.3, we can find that,

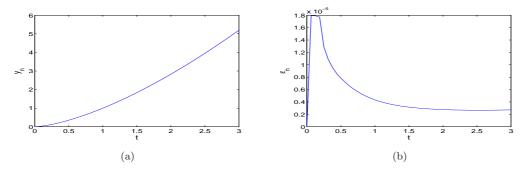


Fig. 6.1. (a) Numerical solution of problem (6.2) by $\text{ETR}_2(4, 3)$ with h = 1/16; (b) Error of $\text{ETR}_2(4, 3)$ with h = 1/16 for problem (6.2).

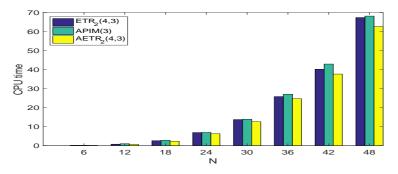


Fig. 6.2. The CPU times (in seconds) of $\text{ETR}_2(4, 3)$, APIM(3) and $\text{AETR}_2(4, 3)$ with N = (i+1)6 (i = 0, 1, ..., 7) for problem (6.2).

when the stepsize is small enough, the extended BVMs (3.8) have higher accuracy than the other two classes methods with the same order, and APIM(k) have the phenomenon of order reduction. Moreover, we also plot the CPU times of $\text{ETR}_2(4, 3)$, APIM(3) and AETR₂(4, 3) with N = (i + 1)6 (i = 0, 1, ..., 7) for problem (6.2) in Fig. 6.2, which shows that, in computational efficiency, the extended BVMs (3.8) is not better than these with Li et al's interpolation scheme but better than APIM(k).

Example 6.2. Consider the following fractal mobile/immobile transport models (cf. [26, 33]):

$$\begin{cases} \frac{\partial u(x,t)}{\partial t} + {}_{0}^{C} D_{t}^{\gamma} u(x,t) = \frac{\partial^{2} u(x,t)}{\partial x^{2}} + f(u) + g(x,t), \quad (x,t) \in [0,1] \times [0,3], \\ u(x,0) = 0, \quad x \in [0,1], \\ u(0,t) = u(1,t) = 0, \quad t \in [0,3], \end{cases}$$

$$(6.3)$$

where $\gamma \in (0,1)$, $f(u) = u(1-u^2)$ and g(x,t) is a given function such that problem (6.3) has the exact solution $u(x,t) = [x(1-x)]^4 \sin(\pi t)$. In the following, we will adopt the method of lines to solve this problem.

Firstly, similar to the discretization idea for delay-reaction-diffusion equations in paper [20], we apply a compact difference scheme to discretize the space variable x. Let M be a positive integer, $x_i = i\Delta x$ (i = 0, 1, ..., M) spatial grid-points with stepsize $\Delta x = 1/M$, and $\mathcal{W} := \{v_i: 0 \le i \le M\}$ grid function space on grid set $\Omega_{\Delta x} := \{x_i: 0 \le i \le M\}$. Define the

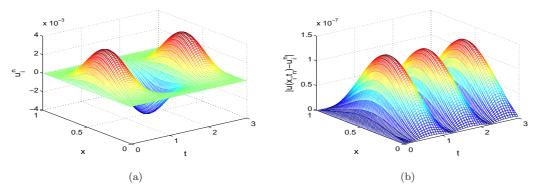


Fig. 6.3. (a) Numerical solution of problem (6.3) with $\gamma = 0.5$ by ETR₂(4, 3) with h = 1/16 and $\Delta x = 1/80$; (b) Error of ETR₂(4, 3) with h = 1/16 and $\Delta x = 1/80$ for problem (6.3) with $\gamma = 0.5$.

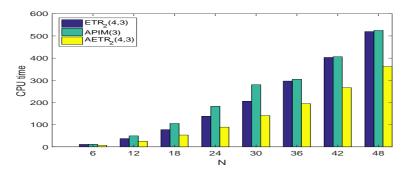


Fig. 6.4. The CPU times (in seconds) of $\text{ETR}_2(4, 3)$, APIM(3) and $\text{AETR}_2(4, 3)$ with $\Delta x = 1/80$ and N = (i+1)6 (i = 0, 1, ..., 7) for problem (6.3) with $\gamma = 0.5$.

following difference operators:

$$\delta_x^2 v_i = \frac{1}{\Delta x^2} (v_{i+1} - 2v_i + v_{i-1}), \quad \mathcal{D}v_i = \frac{1}{12} (v_{i+1} + 10v_i + v_{i-1}).$$

Then a compact difference scheme for (6.3) can be obtained as follows:

$$\mathcal{D}\frac{\partial u_i(t)}{\partial t} + \mathcal{D}_0^C D_t^\gamma u_i(t) = \delta_x^2 u_i(t) + \mathcal{D}f(u_i(t)) + \mathcal{D}g(x_i, t), \quad u_i(0) = 0, \quad 1 \le i \le M, \quad (6.4)$$

where $u_i(t)$ is an approximation to $u(x_i, t)$. When the following notations are introduced:

$$Q(t) = \begin{pmatrix} f(u_{1}(t)) + g(x_{1}, t) \\ f(u_{2}(t)) + g(x_{2}, t) \\ \vdots \\ f(u_{M-2}(t)) + g(x_{M-2}, t) \\ f(u_{M-1}(t)) + g(x_{M-1}, t) \end{pmatrix}, \quad q(t) = \frac{1}{12} \begin{pmatrix} f(u_{0}(t)) + g(x_{0}, t) \\ 0 \\ \vdots \\ 0 \\ f(u_{M}(t)) + g(x_{0}, t) \end{pmatrix},$$
$$P = \begin{bmatrix} \frac{5}{6} & \frac{1}{12} & 0 & \cdots & 0 \\ \frac{1}{12} & \frac{5}{6} & \frac{1}{12} & \cdots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & \frac{1}{12} & \frac{5}{6} & \frac{1}{12} \\ 0 & \cdots & 0 & \frac{1}{12} & \frac{5}{6} \end{bmatrix}, \quad J = \begin{bmatrix} -2 & 1 & 0 & \cdots & 0 \\ 1 & -2 & 1 & \cdots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 1 & -2 & 1 \\ 0 & \cdots & 0 & 1 & -2 \end{bmatrix} \in \mathbb{R}^{(M-1) \times (M-1)},$$

		ETR(2	, 1)	GBDF(3, 2)		$\operatorname{ETR}_2(4, 3)$	
γ	h	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}
0.25	1/2	1.6977e-3	_	2.3258e-3	_	6.4404e-4	_
	1/4	3.2183e-4	2.3992	2.9792e-4	2.9647	4.6988e-5	3.7768
	1/8	8.1896e-5	1.9744	3.4452e-5	3.1123	3.2047e-6	3.8740
	1/16	$2.2643\mathrm{e}{-5}$	1.8547	4.3265e-6	2.9933	2.0667e-7	3.9548
0.5	1/2	1.5969e-3	_	2.4827e-3	_	1.1515e-3	_
	1/4	4.7657e-4	1.7445	3.7071e-4	2.7436	1.1528e-4	3.3202
	1/8	1.7872e-4	1.4150	4.9302e-5	2.9106	1.0843e-5	3.4103
	1/16	6.6121e-5	1.4345	7.3897e-6	2.7381	9.5079e-7	3.5116
0.75	1/2	2.2629e-3	_	2.9855e-3	_	2.0329e-3	_
	1/4	1.0753e-3	1.0734	6.0230e-4	2.3094	2.8450e-4	2.8371
	1/8	4.8749e-4	1.1414	1.1175e-4	2.4302	$3.3280e{-5}$	3.0957
	1/16	2.1178e-4	1.2028	$2.2659\mathrm{e}{-5}$	2.3021	3.5662e-6	3.2222

Table 6.4: Global errors and convergence orders of ETR(2, 1), GBDF(3, 2) and ETR₂(4, 3) with $\Delta x = 1/80$ for problems (6.3) with $\gamma = 0.25, 0.5, 0.75$, respectively.

Table 6.5: Global errors and convergence orders of APIM(k) (k = 1, 2, 3) with $\Delta x = 1/80$ for problems (6.3) with $\gamma = 0.25, 0.5, 0.75$, respectively.

		$\operatorname{APIM}(1)$		APIM(2)		APIM(3)	
γ	h	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}
0.25	1/2	3.3106e-3	_	3.0310e-3	_	3.5194e-3	_
	1/4	1.0885e-3	1.6048	6.9830e-4	2.1179	3.8015e-4	3.2107
	1/8	2.9990e-4	1.8597	9.6552e-5	2.8545	$2.5729\mathrm{e}{-5}$	3.8851
	1/16	7.7447e-5	1.9532	1.2263e-5	2.9770	1.7029e-6	3.9173
0.5	1/2	3.5919e-3	_	3.1167e-3	_	3.6611e-3	_
	1/4	1.1938e-3	1.5891	7.3156e-4	2.0910	4.0402e-4	3.1798
	1/8	3.4644e-4	1.7849	1.0313e-4	2.8265	2.7769e-5	3.8629
	1/16	9.9953e-5	1.7933	1.3625e-5	2.9202	1.9315e-6	3.8457
0.75	1/2	4.3676e-3	_	3.4820e-3	_	4.0693e-3	_
	1/4	1.6370e-3	1.4158	8.9347e-4	1.9624	5.0667e-4	3.0057
	1/8	5.9337e-4	1.4640	1.4770e-4	2.5968	4.2661e-5	3.5700
	1/16	2.2884e-4	1.3746	2.5723e-5	2.5215	3.9640e-6	3.4279

and $y(t) = (u_1(t), \ldots, u_{M-1}(t))^T$, then (6.4) can be written in an equivalent form:

$$\begin{cases} y'(t) + {}_{0}^{C}D_{t}^{\gamma}y(t) = \frac{1}{\Delta x^{2}}P^{-1}Jy(t) + Q(t) + P^{-1}q(t), \quad t \in [0,3], \\ y(0) = \mathbf{0} \in \mathbb{R}^{M-1}, \end{cases}$$
(6.5)

Secondly, we take stepsizes $\Delta x = 1/80$, $h = 1/2^i$ (i = 1, 2, 3, 4) and then apply ETR(2, 1), GBDF(3, 2) and ETR₂(4, 3) to solve (6.5) with $\gamma = 0.25, 0.5, 0.75$, respectively. In this way, a series of effective numerical solutions for (6.3) can be obtained. For simplicity, in Fig. 6.3(a), we only plot the numerical solution of (6.3) with $\gamma = 0.5$ solved by ETR₂(4, 3) with $\Delta x = 1/80$ and h = 1/16, and the global errors $|u(x_i, t_n) - u_i^n|$ ($0 \le i \le 80, 0 \le n \le 16$) are plotted in Fig. 6.3(b). A whole description to the global errors and convergence orders of the above

		AETR(2, 1)		AGBDF(3, 2)		$\operatorname{AETR}_2(4, 3)$	
γ	h	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}	$\operatorname{err}(h)$	\hat{p}
0.25	1/2	1.6977e-3	_	2.2265e-3	_	6.7201e-4	_
	1/4	3.2183e-4	2.3992	2.9227e-4	2.9294	5.2910e-5	3.6669
	1/8	8.1896e-5	1.9744	3.6030e-5	3.0200	4.6995e-6	3.4930
	1/16	2.2643e-5	1.8547	4.6904e-6	2.9414	$3.9299e{-7}$	3.5799
0.5	1/2	1.5969e-3	_	2.2763e-3	_	1.2429e-3	_
	1/4	4.7657e-4	1.7445	3.5974e-4	2.6617	1.4566e-4	3.0930
	1/8	1.7872e-4	1.4150	5.4326e-5	2.7272	1.6634e-5	3.1304
	1/16	6.6121e-5	1.4345	8.6396e-6	2.6526	1.6936e-6	3.2960
0.75	1/2	2.2629e-3	_	2.7745e-3	_	2.2374e-3	_
	1/4	1.0753e-3	1.0734	6.0675e-4	2.1930	3.6124e-4	2.6308
	1/8	4.8749e-4	1.1414	1.2270e-4	2.3060	4.8412e-5	2.8995
	1/16	2.1178e-4	1.2028	$2.5055e{5}$	2.2920	5.7590e-6	3.0715

Table 6.6: Global errors and convergence orders of AETR(2, 1), AGBDF(3, 2) and AETR₂(4, 3) with $\Delta x = 1/80$ for problems (6.3) with $\gamma = 0.25, 0.5, 0.75$, respectively.

methods can be seen in Table 6.4. Again, the numerical results verify the computational effectiveness of the extended BVMs (3.8) and their theoretical accuracy shown in Theorem 5.1.

For giving a comparison, in Tables 6.5-6.6, we also present the global errors and convergence orders of APIM(k) (k = 1, 2, 3), AETR(2, 1), AGBDF(3, 2) and AETR₂(4, 3) with stepsizes $\Delta x = 1/80$ and $h = 1/2^i$ (i = 1, 2, 3, 4) for problems (6.3) with $\gamma = 0.25, 0.5, 0.75$, respectively. From Tables 6.4-6.6, we can find that the extend BVMs (3.8) have the best accuracy under the same order and stepsize. Moreover, in Fig. 6.4, we also plot CPU times (in seconds) of ETR₂(4, 3), APIM(3) and AETR₂(4, 3) with $\Delta x = 1/80$ and N = (i + 1)6 (i = 0, 1, ..., 7) for problem (6.3) with $\gamma = 0.5$. This, again, shows that the computational efficiency of the extended BVMs (3.8) is not better than that of the methods with Li et al's interpolation scheme but better than that of APIM(k).

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