

Deviation of the Error Estimation for Second Order Fredholm-Volterra Integro Differential Equations

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Abstract. In this paper we study the deviation of the error estimation for the second order Fredholm-Volterra integro-differential equations. We prove that for m degree piecewise polynomial collocation method, our method provides $\mathcal{O}(h^{m+1})$ as the order of the deviation of the error. Also numerical results in the final section are included to confirm the theoretical results.

Keywords: deviation of the error, finite difference, exact finite difference, integro-differential.

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1 Introduction

In this paper we consider the second order Fredholm-Volterra integro-differential (SFVID) equations as follows

$$y''(t) = F(t, y(t), y'(t), z_{\mathbf{f}}[y](t), z_{\mathbf{v}}[y](t)), \quad t \in I := [a, b], \quad (1.1)$$

$$y(a) = r_1, \quad y(b) = r_2, \quad (1.2)$$

where

$$z_{\mathbf{f}}[y](t) = \int_a^b K_{\mathbf{f}}(t, s, y(s), y'(s), y''(s)) ds,$$

$$z_{\mathbf{v}}[y](t) = \int_a^t K_{\mathbf{v}}(t, s, y(s), y'(s), y''(s)) ds$$

and $a, b, r_1, r_2 \in R = (-\infty, \infty)$. W and S are defined as follows

$$W := \{(t, y, y', z_f, z_v); t \in I \text{ and } y, y', z_f, z_v \in \mathbb{R}\},$$

$$S := \{(t, s, u, u', u''); t, s \in I \text{ and } u, u', u'' \in \mathbb{R}\}.$$

In this paper we shall assume that F is uniformly continuous in W . Also we assume that $z_f[y](t), z_v[y](t)$ are uniformly continuous in S . We say that $z_f[y](t)$ and $z_v[y](t)$ are linear if we can write $z_f[y](t)$ and $z_v[y](t)$ as

$$z_f[y](t) = \sum_{l=0}^2 \int_a^b A_{l,f}(t, s) y^{(l)}(s) ds, \quad z_v[y](t) = \sum_{l=0}^2 \int_a^t A_{l,v}(t, s) y^{(l)}(s) ds,$$

where $A_{l,j}(t, s)$ ($l = 0, 1, 2$ & $j = f, v$) are sufficiently smooth in $J := \{(t, s); t, s \in I\}$. Also we say that F is linear if we can write it as

$$F(t, y(t), y'(t), z_f[y](t), z_v[y](t)) = \sum_{l=1}^2 a_l(t) y^{(2-l)}(t) + \sum_{l=f,v} z_l[y](t) + a_3(t).$$

In the nonlinear case we assume that $F(t, y, y', z_f, z_v), F_l(t, y, y', z_f, z_v)$ for any $l = t, y, y', z_f, z_v$ are Lipschitz-continuous. When $z_f[y](t)$ and $z_v[y](t)$ are nonlinear we assume that $K_j(t, s, u, u', u'')$ and $(K_j)_l(t, s, u, u', u'')$ ($j = f, v$ & $l = u, u', u''$) are Lipschitz-continuous. We say SFVID equation with boundary condition (1.2) is linear if we can write (1.1) as follows

$$y''(t) = a_1(t)y'(t) + a_2(t)y(t) + a_3(t) + z_f[y](t) + z_v[y](t), \quad t \in [a, b] \quad (1.3)$$

with linear $z_f[y](t)$ and $z_v[y](t)$. Also, in the linear case we assume that $a_i(t), i = 1, 2, 3$ are sufficiently smooth in I . In this paper we use the defect correction principle, more details about this can be found in [4, 9]. The piecewise polynomial collocation method for integro-differential equations can be found in [5, 6, 7, 8]. Also other methods for the integro-differential equations are studied in [11, 12]. The deviation of the error estimation for linear and nonlinear first and second order boundary value problem is studied in [1, 2, 3]. The error estimation based on locally weighted defect that we will use in this manuscript, has been introduced in [1, 3].

The rest of this paper is organized as follows. In Section 2, the method is described and we introduce some details about the deviation of the error for SFVID. In Section 3, the analysis of the deviation of the error is given. Also the main results of the paper are formulate in Theorems 4-5. In Section 4, we study the special case of SFVID equation. And we show that in this case for m degree piecewise polynomial collocation method, our method provides $\mathcal{O}(h^{m+2})$ as the order of the deviation of the error. In Section 5, we present the numerical experiments that demonstrate our theoretical results. A summary is given at the end of the paper in Conclusion section.

2 Description of the method

In this section, we introduce some details about the deviation of the error estimation, collocation method, finite differences and exact difference schemes.

2.1 Collocation method

In the first step we consider τ_i, ρ_i as follows

$$a = \tau_0 < \tau_1 < \dots < \tau_n = b, \quad (n \geq 1), \quad 0 = \rho_0 < \rho_1 < \dots < \rho_m < \rho_{m+1} = 1.$$

DEFINITION 1. In this paper we define

$$X_i := \{t_{i,j} := \tau_i + \rho_j h_i; j = 1, \dots, m\}, \quad Z_n := \{t_{i,0} := \tau_i; i = 0, \dots, n\},$$

$$S_{m+1}^{(1)}(Z_n) := \{p \in C^1(I); p \upharpoonright [\tau_i, \tau_{i+1}] \in \Pi_{m+1}([\tau_i, \tau_{i+1}]) (i = 0, \dots, n - 1)\},$$

where $h_i := \tau_{i+1} - \tau_i$ and $\Pi_{m+1}([\tau_i, \tau_{i+1}])$ is space of real polynomial functions on $[\tau_i, \tau_{i+1}]$ of degree $\leq m + 1$. Also we define h (the diameter of gird Z_n) and h' as

$$h := \max\{h_i; i = 0, \dots, n - 1\}, \quad h' := \min\{h_i; i = 0, \dots, n - 1\}.$$

The set $X(n) := \bigcup_{i=0}^{n-1} X_i$ is called the set of collocation points.

In the piecewise polynomial collocation method we are looking to find a $p \in S_{m+1}^{(1)}(Z_n)$ so that (1.1)–(1.2) holds for all $t_{i,j} \in X(n)$. In the collocation method, since always we can not determine exact value for $z_l[p](t)$ ($l = \mathbf{f}, \mathbf{v}$), therefore we use the following quadrature method to determine $z_l[p](t_{i,j})$ ($l = \mathbf{f}, \mathbf{v}$).

$$z_{\mathbf{f}}[p](t_{i,j}) \approx \sum_{k=0}^{n-1} \sum_{z=0}^{m+1} \alpha_{k,z} K_{\mathbf{f}}(t_{i,j}, t_{k,z}, p(t_{k,z}), p'(t_{k,z}), p''(t_{k,z})) =: \tilde{z}_{\mathbf{f}}[p](t_{i,j}), \tag{2.1}$$

$$z_{\mathbf{v}}[p](t_{i,j}) \approx \sum_{k=0}^{i-1} \sum_{z=0}^{m+1} \alpha_{k,z} K_{\mathbf{v}}(t_{i,j}, t_{k,z}, p(t_{k,z}), p'(t_{k,z}), p''(t_{k,z})) + (t_{i,j} - \tau_i)$$

$$\times \sum_{z=0}^{m+1} \beta_z K_{\mathbf{v}}(t_{i,j}, \bar{t}_{i,j,z}, p(\bar{t}_{i,j,z}), p'(\bar{t}_{i,j,z}), p''(\bar{t}_{i,j,z})) =: \tilde{z}_{\mathbf{v}}[p](t_{i,j}), \tag{2.2}$$

where $\bar{t}_{i,j,z} := \tau_i + \rho_z(t_{i,j} - \tau_i)$ and

$$\alpha_{k,z} := \int_{\tau_k}^{\tau_{k+1}} L_z^{[\tau_k, \tau_{k+1}]}(s) ds, \quad \beta_z := \int_0^1 L_z(s) ds$$

with

$$L_j(\rho) := \prod_{\substack{i=0 \\ i \neq j}}^{m+1} \frac{\rho - \rho_i}{\rho_j - \rho_i}, \quad L_j^{[a', b']}(a) := L_j\left(\frac{\rho - a'}{b' - a'}\right), \quad a \leq a' < b' \leq b.$$

According to [10] we have the following theorem.

Theorem 1. (Interpolation Error Theorem) *If the function f has an $(n + 1)$ st derivative and $P_n(x)$ be a polynomial of degree at most n that interpolates f at $n + 1$ distinct points $x_i (i = 0, \dots, n)$, then for every argument \bar{x} there exists a number ζ in the smallest interval $I[x_0, \dots, x_n, \bar{x}]$ which contains \bar{x} and all support abscissas x_i , satisfying*

$$f(\bar{x}) - P_n(\bar{x}) = \frac{w(\bar{x})f^{(n+1)}(\zeta)}{(n + 1)!},$$

where $w(x) := (x - x_0) \dots (x - x_n)$.

For the above method we have the following lemma.

Lemma 1. *For sufficiently smooth f , the following estimate holds*

$$|z_l[f](t_{i,j}) - \tilde{z}_l[f](t_{i,j})| = \mathcal{O}(h^{m+2}), \quad l = \mathbf{f}, \mathbf{v}, \tag{2.3}$$

where $\tilde{z}_{\mathbf{f}}[\cdot](t_{i,j})$ and $\tilde{z}_{\mathbf{v}}[\cdot](t_{i,j})$ are defined in (2.1)–(2.2).

Proof. For nonlinear $\tilde{z}_{\mathbf{f}}[\cdot]$, we can write

$$\begin{aligned} z_{\mathbf{f}}[f](t_{i,j}) - \tilde{z}_{\mathbf{f}}[f](t_{i,j}) &= \int_a^b K_{\mathbf{f}}(t_{i,j}, s, f(s), f'(s), f''(s)) ds - \sum_{k=0}^{n-1} \sum_{z=0}^{m+1} \alpha_{k,z} \\ &\times K_{\mathbf{f}}(t_{i,j}, t_{k,z}, f(t_{k,z}), f'(t_{k,z}), f''(t_{k,z})) = \sum_{k=0}^{n-1} \int_{\tau_k}^{\tau_{k+1}} I_1(s) ds. \end{aligned} \tag{2.4}$$

By using Interpolation error theorem, we can get that $I_1 = \mathcal{O}(h^{m+2})$. Then we can rewrite (2.4) as follows

$$z_{\mathbf{f}}[f](t_{i,j}) - \tilde{z}_{\mathbf{f}}[f](t_{i,j}) \leq nh\mathcal{O}(h^{m+2}) \leq \frac{h}{h'}(b - a)\mathcal{O}(h^{m+2}) = \mathcal{O}(h^{m+2}).$$

Also for $l = \mathbf{v}$, we get

$$z_{\mathbf{v}}[f](t_{i,j}) - \tilde{z}_{\mathbf{v}}[f](t_{i,j}) = \sum_{k=0}^{i-1} \int_{\tau_k}^{\tau_{k+1}} I_2(s) ds + (t_{i,j} - \tau_i) \int_0^1 I_3(s) ds,$$

where

$$\begin{aligned} I_2(s) &:= K_{\mathbf{v}}(t_{i,j}, s, f(s), f'(s), f''(s)) \\ &\quad - \sum_{z=0}^{m+1} L_z^{[\tau_k, \tau_{k+1}]}(s) K_{\mathbf{v}}(t_{i,j}, t_{k,z}, f(t_{k,z}), f'(t_{k,z}), f''(t_{k,z})), \\ I_3(s) &:= K_{\mathbf{v}}(t_{i,j}, \check{t}_{i,j,s}, f(\check{t}_{i,j,s}), f'(\check{t}_{i,j,s}), f''(\check{t}_{i,j,s})) \\ &\quad - \sum_{z=0}^{m+1} L_z(s) K_{\mathbf{v}}(t_{i,j}, \bar{t}_{i,j,z}, f(\bar{t}_{i,j,z}), f'(\bar{t}_{i,j,z}), f''(\bar{t}_{i,j,z})), \end{aligned}$$

where $\check{t}_{i,j,s} := \tau_i + s(t_{i,j} - \tau_i)$. By using Interpolation error theorem, we can say $I_2 = \mathcal{O}(h^{m+2})$ and $I_3 = \mathcal{O}(h^{m+2})$. Then we get

$$z_{\mathbf{v}}[f](t_{i,j}) - \tilde{z}_{\mathbf{v}}[f](t_{i,j}) \leq \frac{h}{h'}(\tau_i - \tau_0)\mathcal{O}(h^{m+2}) + \mathcal{O}(h^{m+2}) = \mathcal{O}(h^{m+2}),$$

which completes the proof. Similarly, we can find (2.3) for linear case. \square

For above collocation method we have the following theorem [5].

Theorem 2. *Assume that the SFVID problem (1.1)–(1.2) has a unique and sufficiently smooth solution $y(t)$. Also assume that $p(t)$ is a piecewise polynomial collocation solution of degree $\leq m + 1$. Then for sufficiently small h , the collocation solution $p(t)$ is well-defined and the following uniform estimates at least hold:*

$$\begin{aligned} \|y^{(j)}(t) - p^{(j)}(t)\|_{\infty} &= \mathcal{O}(h^m), \quad j = 0, 1, 2, \\ \|y^{(j)}(t) - p^{(j)}(t)\|_{\infty} &= \mathcal{O}(h^{m+2-j}), \quad j = 3, \dots, m + 1. \end{aligned}$$

Also in the piecewise polynomial collocation method when m is odd and the nodes ρ_i are symmetrically distributed we have

$$\|y^{(j)}(t) - p^{(j)}(t)\|_{\infty} = \mathcal{O}(h^{m+1}), \quad j = 0, 1.$$

Lemma 2. *For linear and nonlinear $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$) we have*

$$|\tilde{z}_l[p](t_{i,j}) - \tilde{z}_l[y](t_{i,j})| = \mathcal{O}(h^m), \quad l = \mathbf{f}, \mathbf{v}.$$

Proof. For linear case by using Lemma 1, Theorem 2 and the Integral mean value theorem, we get

$$\begin{aligned} \tilde{z}_{\mathbf{f}}[p](t_{i,j}) - \tilde{z}_{\mathbf{f}}[y](t_{i,j}) &= z_{\mathbf{f}}[e](t_{i,j}) + \mathcal{O}(h^{m+2}) \\ &= \sum_{l=0}^2 \int_a^b \Lambda_{l,\mathbf{f}}(t_{i,j}, s) e^{(l)}(s) ds + \mathcal{O}(h^{m+2}) \\ &= \sum_{l=0}^2 (b-a) \Lambda_{l,\mathbf{f}}(t_{i,j}, \zeta_{i,j}^l) \underbrace{e^{(l)}(\zeta_{i,j}^l)}_{\mathcal{O}(h^m)} + \mathcal{O}(h^{m+2}) = \mathcal{O}(h^m), \end{aligned}$$

where $\zeta_{i,j}^l \in [a, b]$. Also we can obtain

$$\begin{aligned} \tilde{z}_{\mathbf{v}}[p](t_{i,j}) - \tilde{z}_{\mathbf{v}}[y](t_{i,j}) &= z_{\mathbf{v}}[e](t_{i,j}) + \mathcal{O}(h^{m+2}) \\ &= \sum_{l=0}^2 \int_a^{t_{i,j}} \Lambda_{l,\mathbf{v}}(t_{i,j}, s) e^{(l)}(s) ds + \mathcal{O}(h^{m+2}) \\ &= \sum_{l=0}^2 (t_{i,j} - a) \Lambda_{l,\mathbf{v}}(t_{i,j}, \bar{\zeta}_{i,j}^l) \underbrace{e^{(l)}(\bar{\zeta}_{i,j}^l)}_{\mathcal{O}(h^m)} + \mathcal{O}(h^{m+2}) = \mathcal{O}(h^m), \end{aligned}$$

where $\zeta_{i,j}^l \in [a, b]$. Also for nonlinear case by using Lemma 1 and the Lipschitz condition for K_l ($l = \mathbf{f}, \mathbf{v}$) we can find

$$\begin{aligned}
 & |\tilde{z}_{\mathbf{f}}[y](t_{i,j}) - \tilde{z}_{\mathbf{f}}[p](t_{i,j})| \\
 & \leq |\tilde{z}_{\mathbf{f}}[y](t_{i,j}) - z_{\mathbf{f}}[y](t_{i,j}) - \tilde{z}_{\mathbf{f}}[p](t_{i,j}) + z_{\mathbf{f}}[p](t_{i,j}) + z_{\mathbf{f}}[y](t_{i,j}) - z_{\mathbf{f}}[p](t_{i,j})| \\
 & \leq \left| \int_a^b \left(K_{\mathbf{f}}(t_{i,j}, s, y(s), y'(s), y''(s)) - K_{\mathbf{f}}(t_{i,j}, s, p(s), p'(s), p''(s)) \right) ds \right| \\
 & \quad + \mathcal{O}(h^{m+2}) \leq C \sum_{l=0}^2 \int_a^b |y^{(l)}(s) - p^{(l)}(s)| ds + \mathcal{O}(h^{m+2}) = \mathcal{O}(h^m), \\
 & |\tilde{z}_{\mathbf{v}}[y](t_{i,j}) - \tilde{z}_{\mathbf{v}}[p](t_{i,j})| \\
 & \leq |\tilde{z}_{\mathbf{v}}[y](t_{i,j}) - z_{\mathbf{v}}[y](t_{i,j}) - \tilde{z}_{\mathbf{v}}[p](t_{i,j}) + z_{\mathbf{v}}[p](t_{i,j}) + z_{\mathbf{v}}[y](t_{i,j}) - z_{\mathbf{v}}[p](t_{i,j})| \\
 & \leq \left| \int_a^{t_{i,j}} \left(K_{\mathbf{v}}(t_{i,j}, s, y(s), y'(s), y''(s)) - K_{\mathbf{v}}(t_{i,j}, s, p(s), p'(s), p''(s)) \right) ds \right| \\
 & \quad + \mathcal{O}(h^{m+2}) \leq C \sum_{l=0}^2 \int_a^{t_{i,j}} |y^{(l)}(s) - p^{(l)}(s)| ds + \mathcal{O}(h^{m+2}) = \mathcal{O}(h^m),
 \end{aligned}$$

which completes the proof. \square

2.2 Finite difference scheme

DEFINITION 2. In this subsection we define

$$\begin{aligned}
 \Delta_{i,j} & := \{(l, k); l = 0, \dots, i - 1 \ \& \ k = 0, \dots, m\} \cup \{(i, k); k = 0, \dots, j - 1\}, \\
 \mathcal{A} & := \{(i, j); t_{i,j} \in X(n) \cup Z_n\}, \quad \mathcal{B} := \mathcal{A} - \{(0, 0), (n, 0)\}, \quad \mathcal{T} := \mathcal{A} - \{(n, 0)\}.
 \end{aligned}$$

Also we define

$$\delta_{i,j} := t_{i,j+1} - t_{i,j}, \quad \widehat{\delta}_{i,j} := \frac{\delta_{i,j-1} + \delta_{i,j}}{2}, \quad \widehat{\alpha}_{i,j} := \frac{\delta_{i,j-1}}{\widehat{\delta}_{i,j}}, \quad \widehat{\beta}_{i,j} := \frac{\delta_{i,j}}{\widehat{\delta}_{i,j}}.$$

A general one-step finite difference scheme can be written as follows.

$$\begin{aligned}
 (L_{\mathcal{A}}^{(2)}\eta)_{i,j} & = F(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j}, \chi^{\mathbf{f}}[\eta]_{i,j}, \chi^{\mathbf{v}}[\eta]_{i,j}), \quad (i, j) \in \mathcal{B}, \quad (2.5) \\
 \eta_{0,0} & = r_1, \quad \eta_{n,0} = r_2,
 \end{aligned}$$

where

$$\begin{aligned}
 (L_{\mathcal{A}}^{(2)}\eta)_{i,j} & := \frac{\widehat{\alpha}_{i,j}\eta_{i,j+1} - 2\eta_{i,j} + \widehat{\beta}_{i,j}\eta_{i,j-1}}{\widehat{\alpha}_{i,j}\widehat{\beta}_{i,j}\widehat{\delta}_{i,j}^2}, \quad (L_{\mathcal{A}}^{(1)}\eta)_{i,j} := \frac{\eta_{i,j+1} - \eta_{i,j}}{\delta_{i,j}}, \quad (2.6) \\
 \chi^{\mathbf{f}}[\eta]_{i,j} & := \sum_{(l,v) \in \mathcal{T}} \delta_{l,v} K_{\mathbf{f}}(t_{i,j}, t_{l,v}, \eta_{l,v}, (L_{\mathcal{A}}^{(1)}\eta)_{l,v}, (L_{\mathcal{A}}^{(2)}\eta)_{l,v}), \\
 \chi^{\mathbf{v}}[\eta]_{i,j} & := \sum_{(l,v) \in \Delta_{i,j}} \delta_{l,v} K_{\mathbf{v}}(t_{i,j}, t_{l,v}, \eta_{l,v}, (L_{\mathcal{A}}^{(1)}\eta)_{l,v}, (L_{\mathcal{A}}^{(2)}\eta)_{l,v}).
 \end{aligned}$$

Theorem 3. Let f be a sufficiently smooth function on the interval $[a, b]$. Then we have

$$|\chi^l[f]_{i,j} - z_l[f](t_{i,j})| = \mathcal{O}(h), \quad l = \mathbf{f}, \mathbf{v}. \tag{2.7}$$

Proof. We can find

$$\begin{aligned} & |K_{\mathbf{f}}(t_{i,j}, t_{l,v}, f(t_{l,v}), f'(t_{l,v}), f''(t_{l,v})) \\ & - K_{\mathbf{f}}(t_{i,j}, t_{l,v}, f(t_{l,v}), (L_{\mathcal{A}}^{(1)} f)_{l,v}, (L_{\mathcal{A}}^{(2)} f)_{l,v})| = \mathcal{O}(h). \end{aligned} \tag{2.8}$$

Also we get

$$\begin{aligned} & K_{\mathbf{f}}(t_{i,j}, s, f(s), f'(s), f''(s)) - K_{\mathbf{f}}(t_{i,j}, t_{l,v}, f(t_{l,v}), f'(t_{l,v}), f''(t_{l,v})) \\ & = (s - t_{l,v}) \frac{\partial K_{\mathbf{f}}}{\partial s}(t_{i,j}, \zeta_s^{l,v}, f(\zeta_s^{l,v}), f'(\zeta_s^{l,v}), f''(\zeta_s^{l,v})), \end{aligned} \tag{2.9}$$

where $\zeta_s^{l,v} \in [t_{l,v}, t_{l,v+1}]$. Therefore by using (2.8) and (2.9) we get

$$\begin{aligned} & \int_{t_{l,v}}^{t_{l,v+1}} K_{\mathbf{f}}(t_{i,j}, s, f(s), f'(s), f''(s)) ds \\ & - \delta_{l,v} K_{\mathbf{f}}(t_{i,j}, t_{l,v}, f(t_{l,v}), (L_{\mathcal{A}}^{(1)} f)_{l,v}, (L_{\mathcal{A}}^{(2)} f)_{l,v}) \\ & = \int_{t_{l,v}}^{t_{l,v+1}} K_{\mathbf{f}}(t_{i,j}, s, f(s), f'(s), f''(s)) ds \\ & - \delta_{l,v} K_{\mathbf{f}}(t_{i,j}, t_{l,v}, f(t_{l,v}), f'(t_{l,v}), f''(t_{l,v})) + \mathcal{O}(h^2) \\ & = \frac{\delta_{l,v}^2}{2} \frac{\partial K_{\mathbf{f}}}{\partial s}(t_{i,j}, \zeta_s^{l,v}, f(\zeta_s^{l,v}), f'(\zeta_s^{l,v}), f''(\zeta_s^{l,v})) + \mathcal{O}(h^2) \\ & \leq \frac{h^2}{2} \frac{\partial K_{\mathbf{f}}}{\partial s}(t_{i,j}, \zeta_s^{l,v}, f(\zeta_s^{l,v}), f'(\zeta_s^{l,v}), f''(\zeta_s^{l,v})) + \mathcal{O}(h^2), \end{aligned}$$

then we can obtain

$$\begin{aligned} & \sum_{(l,v) \in \mathcal{T}} \left(\int_{t_{l,v}}^{t_{l,v+1}} K_{\mathbf{f}}(t_{i,j}, s, f(s), f'(s), f''(s)) ds \right. \\ & \quad \left. - \delta_{l,v} K_{\mathbf{f}}(t_{i,j}, t_{l,v}, f(t_{l,v}), (L_{\mathcal{A}}^{(1)} f)_{l,v}, (L_{\mathcal{A}}^{(2)} f)_{l,v}) \right) \\ & \leq \sum_{(l,v) \in \mathcal{T}} \frac{h^2}{2} \frac{\partial K_{\mathbf{f}}}{\partial s}(t_{i,j}, \zeta_s^{l,v}, f(\zeta_s^{l,v}), f'(\zeta_s^{l,v}), f''(\zeta_s^{l,v})) + \mathcal{O}(h) \\ & \leq n(m+1) \frac{h^2}{2} \max_{s \in [a,b]} \frac{\partial K_{\mathbf{f}}}{\partial s}(t_{i,j}, s, f(s), f'(s), f''(s)) + \mathcal{O}(h) \tag{2.10} \\ & \leq \frac{(b-a)(m+1)h}{2h'} h \max_{s \in [a,b]} \frac{\partial K_{\mathbf{f}}}{\partial s}(t_{i,j}, s, f(s), f'(s), f''(s)) + \mathcal{O}(h) = \mathcal{O}(h), \end{aligned}$$

then by using (2.10), we can say that

$$\chi^{\mathbf{f}}[f]_{i,j} - z_{\mathbf{f}}[f](t_{i,j}) = \mathcal{O}(h).$$

Similarly, we can find (2.7) for linear case and $l = \mathbf{v}$. \square

DEFINITION 3. For any function u , we define

$$\mathcal{R}(u) := \{u(t_{i,j}); \quad (i, j) \in \mathcal{A}\},$$

also we define

$$\eta := \{\eta_{i,j}; \quad (i, j) \in \mathcal{A}\}, \quad L_{\mathcal{A}}^{(l)}\eta := \{(L_{\mathcal{A}}^{(l)}\eta)_{i,j}; \quad (i, j) \in \mathcal{A}\}, \quad l = 1, 2.$$

For the above finite difference scheme we have the following estimate

$$\|\eta - \mathcal{R}(y)\|_{\infty} = \mathcal{O}(h), \quad \|L_{\mathcal{A}}^{(l)}\eta - \mathcal{R}(y^{(l)})\|_{\infty} = \mathcal{O}(h^l), \quad l = 1, 2,$$

where η and $L_{\mathcal{A}}^{(l)}\eta$ is defined in the Definition 3.

2.3 Deviation of the error estimation

We study the deviation of the error estimation for (1.1)–(1.2). We consider the Dirichlet problem

$$y''(t) = f(t), \quad a \leq t \leq b, \quad y(a) = y(b) = 0, \tag{2.11}$$

where $f(t)$ is permitted to have jump discontinuities in the points belonging to Z_n . For the discretization form of (2.11), i.e.,

$$(L_{\mathcal{A}}^{(2)}\eta)_{i,j} = f(t_{i,j}), \quad (i, j) \in \mathcal{B}, \tag{2.12}$$

$$\eta_{0,0} = 0, \quad \eta_{n,0} = 0, \tag{2.13}$$

according to [1, 3], we have the following lemmas.

Lemma 3. *The unique solution η of (2.12)–(2.13) is given by*

$$\eta_{i,j} = \sum_{(l,v) \in \mathcal{B}} \widehat{\delta}_{l,v} G(t_{i,j}, t_{l,v}) f(t_{l,v}),$$

where $G(t, \tau)$ is Green's function

$$G(t, \tau) = \begin{cases} (b-t)(a-\tau)/b-a, & a \leq \tau \leq t \leq b, \\ (b-\tau)(a-t)/b-a, & a \leq t \leq \tau \leq b. \end{cases}$$

Lemma 4. *For $v \in \widehat{\mathcal{C}}_2[t_{i,j-1}, t_{i,j}, t_{i,j+1}]$, where*

$$\widehat{\mathcal{C}}_2[t_{i,j-1}, t_{i,j}, t_{i,j+1}] := \{v \in \mathcal{C}^1[t_{i,j-1}, t_{i,j+1}] : v'' \text{ continuous on } [t_{i,j-1}, t_{i,j}] \cup (t_{i,j}, t_{i,j+1}], \lim_{t \uparrow t_{i,j}} v'' \in \mathbb{R}, \lim_{t \downarrow t_{i,j}} v'' \in \mathbb{R} \text{ exist}\},$$

we have

$$(L_{\mathcal{A}}^{(2)}v)_{i,j} = \int_{-\widehat{\alpha}_{i,j}}^{\widehat{\beta}_{i,j}} R_{i,j}(\xi) v''(t_{i,j} + \widehat{\delta}_{i,j}\xi) d\xi,$$

with kernel

$$R_{i,j}(\xi) = \begin{cases} 1 + \xi/\widehat{\alpha}_{i,j}, & \xi \in [-\widehat{\alpha}_{i,j}, 0], \\ 1 - \xi/\widehat{\beta}_{i,j}, & \xi \in [0, \widehat{\beta}_{i,j}]. \end{cases}$$

As [1,3] we can find “the exact finite difference” for (1.1) as follows

$$(L_{\mathcal{A}}^{(2)}p)_{i,j} = \mathcal{I}_{\mathcal{A}}\left(F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j}))\right),$$

where

$$\mathcal{I}_{\mathcal{A}}(w(t_{i,j})) := \int_{-\hat{\alpha}_{i,j}}^{\hat{\beta}_{i,j}} R_{i,j}(\xi)w(t_{i,j} + \hat{\delta}_{i,j}\xi)d\xi.$$

We can say that a solution of problem (1.1)–(1.2) satisfies in the exact finite difference scheme. Since according to the collocation method, we can say that

$$p''(t_{i,j}) - F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j})) \equiv 0, (i, j) \in X(n),$$

therefore we define the defect at $t_{i,j}$ as follows

$$D_{i,j} := (L_{\mathcal{A}}^{(2)}p)_{i,j} - \mathcal{I}_{\mathcal{A}}\left(F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j}))\right), (i, j) \in \mathcal{B}.$$

In order to compute the integral in this expression, we use a quadrature formula. When $t_{i,j} \in X(n)$ we have [1,3]

$$\begin{aligned} \mathcal{I}_{\mathcal{A}}(F(\cdot, p, p', z_{\mathbf{f}}[p], z_{\mathbf{v}}[p]), t_{i,j}) &\approx Q_{\mathcal{A}}\left(F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), \tilde{z}_{\mathbf{f}}[p](t_{i,j}), \tilde{z}_{\mathbf{v}}[p](t_{i,j}))\right) \\ &:= \sum_{k=0}^{m+1} \gamma_{i,j}^k F(t_{i,k}, p(t_{i,k}), p'(t_{i,k}), \tilde{z}_{\mathbf{f}}[p](t_{i,k}), \tilde{z}_{\mathbf{v}}[p](t_{i,k})), \end{aligned}$$

where $\gamma_{i,j}^k = \int_{-\hat{\alpha}_{i,j}}^{\hat{\beta}_{i,j}} R_{i,j}(\xi)L_k(\rho_j + \xi \frac{\hat{\delta}_{i,j}}{h_i})d\xi$. Also for $t_{i,0} = \tau_i$ we have [1,3]

$$\begin{aligned} \mathcal{I}_{\mathcal{A}}(F(t_{i,0}, p(t_{i,0}), p'(t_{i,0}), z_{\mathbf{f}}[p](t_{i,0}), z_{\mathbf{v}}[p](t_{i,0})) &\approx Q_{\mathcal{A}}\left(F(t_{i,0}, p(t_{i,0}), p'(t_{i,0}), \tilde{z}_{\mathbf{f}}[p](t_{i,0}), \tilde{z}_{\mathbf{v}}[p](t_{i,0}))\right) \\ &:= \sum_{k=0}^{m+1} \gamma_{i,0}^{k+} F(t_{i,k}, p(t_{i,k}), p'(t_{i,k}), \tilde{z}_{\mathbf{f}}[p](t_{i,k}), \tilde{z}_{\mathbf{v}}[p](t_{i,k})) \\ &\quad + \sum_{k=0}^{m+1} \gamma_{i,0}^{k-} F(t_{i-1,k}, p(t_{i-1,k}), p'(t_{i-1,k}), \tilde{z}_{\mathbf{f}}[p](t_{i-1,k}), \tilde{z}_{\mathbf{v}}[p](t_{i-1,k})), \end{aligned}$$

where

$$\gamma_{i,0}^{k+} = \int_0^{\hat{\beta}_{i,0}} R_{i,0}(\xi)L_k(\xi \frac{\hat{\delta}_{i,0}}{h_i})d\xi, \quad \gamma_{i,0}^{k-} = \int_{-\hat{\alpha}_{i,0}}^0 R_{i,0}(\xi)L_k(1 + \xi \frac{\hat{\delta}_{i,0}}{h_{i-1}})d\xi.$$

For the above quadrature formula we can find the following lemma.

Lemma 5. For sufficiently smooth f the following error holds

$$\mathcal{I}_{\mathcal{A}}(f, t_{i,j}) - Q_{\mathcal{A}}(f, t_{i,j}) = \mathcal{O}(h^{m+2}).$$

Also when m is odd and the nodes ρ_i are distributed symmetrically, we have the following relation

$$\mathcal{I}_{\mathcal{A}}(f, t_{i,j}) - Q_{\mathcal{A}}(f, t_{i,j}) = \mathcal{O}(h^{m+3}).$$

Then we consider defect at $t_{i,j}$ as follows

$$D_{i,j} \approx (L_{\mathcal{A}}^{(2)} p)_{i,j} - Q_{\mathcal{A}}(F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j}))), \quad (i, j) \in \mathcal{B}. \tag{2.14}$$

At this step we define $\pi = \{\pi_{i,j}; (i, j) \in \mathcal{A}\}$ as the solution of the following finite difference

$$(L_{\mathcal{A}}^{(2)} \pi)_{i,j} = F(t_{i,j}, \pi_{i,j}, (L_{\mathcal{A}}^{(1)} \pi)_{i,j}, \chi^{\mathbf{f}}[\pi]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j}) + D_{i,j}, \quad (i, j) \in \mathcal{B}, \tag{2.15}$$

$$\pi_{0,0} = r_1, \quad \pi_{n,0} = r_2.$$

We define $\mathbf{D} := \{D_{i,j}; (i, j) \in \mathcal{B}\}$. For small value \mathbf{D} , we have

$$\pi - \mathcal{R}(p) \approx \eta - \mathcal{R}(y).$$

We define ε and e as $\varepsilon := \pi - \eta \approx \mathcal{R}(p) - \mathcal{R}(y) =: e$. An estimate for the error e can be found in Theorem 2. The deviation of the error can be written in the following form $\theta := e - \varepsilon$. By using (2.14) and Lemma 5 for linear and nonlinear case we can easily prove the following lemmas.

Lemma 6. *The defined defect in (2.14) has order $\mathcal{O}(h^m)$.*

Lemma 7. *The $\pi - \eta$ has order $\mathcal{O}(h^m)$.*

3 Analysis of the deviation of the error

Lemma 8. *For the linear $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$) we have*

$$|\chi^l[e]_{i,j} - \tilde{z}_l[e](t_{i,j})| = \mathcal{O}(h^m), \quad l = \mathbf{f}, \mathbf{v}. \tag{3.1}$$

Proof. When $l = \mathbf{f}$ by using Lemma 1 we can write

$$|\chi^{\mathbf{f}}[e]_{i,j} - \tilde{z}_{\mathbf{f}}[e](t_{i,j})| \leq |\chi^{\mathbf{f}}[e]_{i,j} - z_{\mathbf{f}}[e](t_{i,j})| + \underbrace{|z_{\mathbf{f}}[e](t_{i,j}) - \tilde{z}_{\mathbf{f}}[e](t_{i,j})|}_{\mathcal{O}(h^{m+2})},$$

also by using Theorem 2 and Theorem 3 we can write

$$\begin{aligned} |\chi^{\mathbf{f}}[e]_{i,j} - z_{\mathbf{f}}[e](t_{i,j})| &= \left| \sum_{(l,v) \in \mathcal{T}} \sum_{m=0}^2 \left(\int_{t_{l,v}}^{t_{l,v+1}} A_{m,\mathbf{f}}(t_{i,j}, s) e^{(m)}(s) ds \right. \right. \\ &\quad \left. \left. - \delta_{l,v} A_{m,\mathbf{f}}(t_{i,j}, t_{l,v}) e^{(m)}(t_{l,v}) \right) \right| + \mathcal{O}(h^m) \leq \frac{(b-a)(m+1)h}{2h^l} h \\ &\times \underbrace{\left| \max_{s \in [a,b]} \frac{\partial (\sum_{m=0}^2 A_{m,\mathbf{f}}(t_{i,j}, s) e^{(m)}(s))}{\partial s} \right|}_{\mathcal{O}(h^{m-1})} + \mathcal{O}(h^m) = \mathcal{O}(h^m). \end{aligned}$$

In a similar way to the $l = \mathbf{f}$, we can prove (3.1) for $l = \mathbf{v}$. \square

3.1 Linear case

Theorem 4. Consider the SFVID equation (1.3) with boundary conditions (1.2). Assume that the SFVID problem has a unique and sufficiently smooth solution. Then the following estimate holds

$$\|\theta\|_\infty = \|e - \varepsilon\|_\infty = \mathcal{O}(h^{m+1}),$$

where e is error, ε is the error estimate and θ is the deviation of the error estimate.

Proof. Since F and $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$) are linear then by using (2.5), (2.14) and (2.15) we get

$$\begin{aligned} (L_A^{(2)}\theta)_{i,j} &= a_1(t_{i,j})(L_A^{(1)}\theta)_{i,j} + a_2(t_{i,j})\theta_{i,j} + \sum_{k=\mathbf{f},\mathbf{v}} \chi^k[\theta]_{i,j} \\ &+ \underbrace{\mathcal{I}_A(a_1e' + a_2e, t_{i,j}) - (a_1(t_{i,j})(L_A^{(1)}e)_{i,j} + a_2(t_{i,j})e(t_{i,j}))}_{I_4} \\ &+ \underbrace{(Q_A - \mathcal{I}_A)(a_1p' + a_2p + a_3, t_{i,j})}_{I_5} + \underbrace{\sum_{k=\mathbf{f},\mathbf{v}} (\mathcal{I}_A(z_k[e], t_{i,j}) - \chi^k[e]_{i,j})}_{I_6} \\ &+ \underbrace{\sum_{k=\mathbf{f},\mathbf{v}} (Q_A(\tilde{z}_k[p](t_{i,j})) - \mathcal{I}_A(z_k[p], t_{i,j}))}_{I_7}, \end{aligned} \tag{3.2}$$

by using Lemma 6.1 in [1] and [3] we can say

$$I_4 = (g_{i,j} - \dot{g}_{i,j} - \ddot{g}_{i,j}) + \frac{1}{2\hat{\delta}_{i,j}}(\delta_{i,j}\phi_{i,j+\frac{1}{2}} - \delta_{i,j-1}\phi_{i,j-\frac{1}{2}}) + \mathcal{O}(h^{m+2}),$$

where

$$\begin{aligned} g_{i,j} &:= \int_{-\hat{\alpha}_{i,j}}^0 \left(\frac{\xi}{\hat{\alpha}_{i,j}} + 1\right)a_1(t_{i,j} + \xi\hat{\delta}_{i,j})e'(t_{i,j} + \xi\hat{\delta}_{i,j})d\xi, \\ \dot{g}_{i,j} &:= \int_0^{\hat{\beta}_{i,j}} \left(\frac{\xi}{\hat{\beta}_{i,j}} - 1\right)a_1(t_{i,j} + \xi\hat{\delta}_{i,j})e'(t_{i,j} + \xi\hat{\delta}_{i,j})d\xi, \\ \ddot{g}_{i,j} &:= \frac{a_1(t_{i,j})}{\hat{\beta}_{i,j}} \int_0^{\hat{\beta}_{i,j}} e'(t_{i,j} + \xi\hat{\delta}_{i,j})d\xi, \\ \phi_{i,j+\frac{1}{2}} &:= \delta_{i,j} \int_{-\frac{1}{2}}^{\frac{1}{2}} \left(u^2 + \frac{1}{4}\right)Q(t_{i,j+\frac{1}{2}} + \delta_{i,j}u)du, \\ \phi_{i,j-\frac{1}{2}} &:= \delta_{i,j-1} \int_{-\frac{1}{2}}^{\frac{1}{2}} \left(u^2 + \frac{1}{4}\right)Q(t_{i,j-\frac{1}{2}} + \delta_{i,j-1}u)du \end{aligned}$$

with $Q(t) = (a_2(t)e(t))'$. We can easily prove that

$$g_{i,j} = \mathcal{O}(h^{m+1}), \quad \dot{g}_{i,j} = \mathcal{O}(h^{m+1}), \quad \ddot{g}_{i,j} = \mathcal{O}(h^m), \quad \phi_{i,j\pm\frac{1}{2}} = \mathcal{O}(h^{m+1}).$$

Also according to Theorem 6.1 in [1] and [3], we can obtain $I_5 = \mathcal{O}(h^{m+2})$. Also since $p \in \Pi_{m+1}$ and $\Lambda_{l,k}(t, s)$ ($l = 0, 1, 2$ & $k = \mathbf{f}, \mathbf{v}$) are sufficiently smooth then by using lemma 5 we can say that $I_7 = \mathcal{O}(h^{m+2})$.

In a similar way to the I_4 , we can find $I_6 = \Upsilon_{\mathbf{f}}(t_{i,j}) + \Upsilon_{\mathbf{v}}(t_{i,j})$,

$$\begin{aligned} \Upsilon_{\mathbf{f}}(t_{i,j}) &:= \sum_{(w,v) \in \mathcal{T}} \left(\dot{\Omega}_{\mathbf{f}}^{w,v}(t_{i,j}) - \ddot{\Omega}_{\mathbf{f}}^{w,v}(t_{i,j}) \right), \\ \Upsilon_{\mathbf{v}}(t_{i,j}) &:= \sum_{(w,v) \in \Delta_{i,j}} \left(\dot{\Omega}_{\mathbf{v}}^{w,v}(t_{i,j}) - \ddot{\Omega}_{\mathbf{v}}^{w,v}(t_{i,j}) \right) \\ &+ \sum_{l=0}^2 \int_{-\widehat{\alpha}_{i,j}}^{\widehat{\beta}_{i,j}} \int_{t_{i,j}}^{t_{i,j} + \xi \widehat{\delta}_{i,j}} \Lambda_{l,v}(t_{i,j} + \xi \widehat{\delta}_{i,j}, s) e^{(l)}(s) ds d\xi, \end{aligned}$$

where

$$\begin{aligned} \dot{\Omega}_k^{w,v}(t_{i,j}) &:= \Psi_k^{w,v}(t_{i,j}) - \check{\Psi}_k^{w,v}(t_{i,j}), \quad k = \mathbf{f}, \mathbf{v}, \\ \ddot{\Omega}_k^{w,v}(t_{i,j}) &:= \bar{\Psi}_k^{w,v}(t_{i,j}) - \check{\check{\Psi}}_k^{w,v}(t_{i,j}), \quad k = \mathbf{f}, \mathbf{v} \end{aligned}$$

with

$$\begin{aligned} \Psi_k^{w,v}(t_{i,j}) &:= \sum_{l=0}^2 \int_{t_{w,v}}^{t_{w,v+1}} \int_{-\widehat{\alpha}_{i,j}}^{\widehat{\beta}_{i,j}} \left(\frac{\xi}{\widehat{\alpha}_{i,j}} + 1 \right) \Lambda_{l,k}(t_{i,j} + \xi \widehat{\delta}_{i,j}, s) e^{(l)}(s) d\xi ds, \\ \bar{\Psi}_k^{w,v}(t_{i,j}) &:= \sum_{l=0}^2 \int_{t_{w,v}}^{t_{w,v+1}} \int_0^{\widehat{\beta}_{i,j}} \left(\frac{\xi}{\widehat{\beta}_{i,j}} - 1 \right) \Lambda_{l,k}(t_{i,j} + \xi \widehat{\delta}_{i,j}, s) e^{(l)}(s) d\xi ds, \\ \check{\Psi}_k^{w,v}(t_{i,j}) &:= \int_{-\widehat{\alpha}_{i,j}}^0 \frac{1}{2} \delta_{w,v} \left(\Lambda_{0,k}(t_{i,j}, t_{w,v}) e(t_{w,v}) \right. \\ &\quad \left. + \sum_{l=1}^2 \Lambda_{l,k}(t_{i,j}, t_{w,v}) (L_{\mathcal{A}}^{(l)} e)_{w,v} \right) d\xi, \\ \check{\check{\Psi}}_k^{w,v}(t_{i,j}) &:= - \int_0^{\widehat{\beta}_{i,j}} \frac{1}{2} \delta_{w,v} \left(\Lambda_{0,k}(t_{i,j}, t_{w,v}) e(t_{w,v}) \right. \\ &\quad \left. + \sum_{l=1}^2 \Lambda_{l,k}(t_{i,j}, t_{w,v}) (L_{\mathcal{A}}^{(l)} e)_{w,v} \right) d\xi. \end{aligned}$$

By using Theorem 2, we obtain

$$\Upsilon_k(t_{i,j}) = \mathcal{O}(h^m), \quad k = \mathbf{f}, \mathbf{v}.$$

Therefore we can rewrite (3.2) as follows

$$\begin{aligned} (L_{\mathcal{A}}^{(2)} \theta)_{i,j} &= a_1(t_{i,j}) (L_{\mathcal{A}}^{(1)} \theta)_{i,j} + a_2(t_{i,j}) \theta_{i,j} + \sum_{k=\mathbf{f}, \mathbf{v}} \chi^k [\theta]_{i,j} \\ &+ (g_{i,j} - \dot{g}_{i,j} - \ddot{g}_{i,j}) + \frac{1}{2\widehat{\delta}_{i,j}} (\delta_{i,j} \phi_{i,j+\frac{1}{2}} - \delta_{i,j-1} \phi_{i,j-\frac{1}{2}}) \\ &+ \Upsilon_{\mathbf{f}}(t_{i,j}) + \Upsilon_{\mathbf{v}}(t_{i,j}) + \mathcal{O}(h^{m+2}). \end{aligned}$$

In this step we define

$$H := \{g_{i,j} - \dot{g}_{i,j} - \ddot{g}_{i,j}; (i, j) \in \mathcal{B}\}, \Theta = \{\mathcal{Y}_{\mathbf{f}}(t_{i,j}) + \mathcal{Y}_{\mathbf{v}}(t_{i,j}); (i, j) \in \mathcal{B}\},$$

$$\Phi := \{\delta_{i,j} \phi_{i,j+\frac{1}{2}} - \delta_{i,j-1} \phi_{i,j-\frac{1}{2}}; (i, j) \in \mathcal{B}\}.$$

Then by using Lemma 3 we find

$$\theta_{i,j} = ((L_{\mathcal{A}}^2)^{-1}H)_{i,j} + \frac{1}{2\widehat{\delta}_{i,j}}((L_{\mathcal{A}}^2)^{-1}\Phi)_{i,j} + ((L_{\mathcal{A}}^2)^{-1}\Theta)_{i,j}.$$

Let

$$\widehat{g} = \max_{i,j} |H| = \mathcal{O}(h^m), \quad \widehat{\phi} = \max_{i,j} |\Phi| = \mathcal{O}(h^{m+1}), \quad \widehat{\Upsilon} = \max_{i,j} |\Theta| = \mathcal{O}(h^m),$$

therefore we get

$$\|\theta\|_{\infty} \leq h \left(\max_{w,x} \sum_{(i,j) \in \mathcal{B}} G(t_{w,x}, t_{i,j}) \widehat{g} \right) + \frac{1}{2} \left(\max_{w,x} \sum_{(i,j) \in \mathcal{B}} G(t_{w,x}, t_{i,j}) \widehat{\phi} \right)$$

$$+ h \left(\max_{w,x} \sum_{(i,j) \in \mathcal{B}} G(t_{w,x}, t_{i,j}) \widehat{\Upsilon} \right) = \mathcal{O}(h^{m+1}).$$

□

3.2 Nonlinear case

DEFINITION 4. For nonlinear and linear $z_l[\cdot]$ ($l = \mathbf{f}, \mathbf{v}$) we define

$$\bar{\chi}^{\mathbf{f}}[\varepsilon]_{i,j} := \sum_{(l,v) \in \mathcal{T}} \delta_{l,v} \left(\Gamma_{0,\mathbf{f}}^{\varepsilon}(t_{i,j}, t_{l,v}) \varepsilon_{l,v} + \sum_{m=1}^2 \Gamma_{m,\mathbf{f}}^{\varepsilon}(t_{i,j}, t_{l,v}) (L_{\mathcal{A}}^{(m)} \varepsilon)_{l,v} \right),$$

$$\bar{\chi}^{\mathbf{v}}[\varepsilon]_{i,j} := \sum_{(l,v) \in \Delta_{i,j}} \delta_{l,v} \left(\Gamma_{0,\mathbf{v}}^{\varepsilon}(t_{i,j}, t_{l,v}) \varepsilon_{l,v} + \sum_{m=1}^2 \Gamma_{m,\mathbf{v}}^{\varepsilon}(t_{i,j}, t_{l,v}) (L_{\mathcal{A}}^{(m)} \varepsilon)_{l,v} \right),$$

where, for linear case

$$\Gamma_{m,k}^{\varepsilon}(t_{i,j}, t_{l,v}) := \Lambda_{m,k}(t_{i,j}, t_{l,v}), \quad m = 0, 1, 2, \quad k = \mathbf{f}, \mathbf{v}$$

and for nonlinear case and $k = \mathbf{f}, \mathbf{v}$ we define

$$\Gamma_{m,k}^{\varepsilon}(t_{i,j}, t_{l,v}) := \begin{cases} \int_0^1 (K_k)_u(t_{i,j}, t_{l,v}, \eta_{l,v} + \tau \varepsilon_{l,v}, (L_{\mathcal{A}}^{(1)} \pi)_{l,v}, (L_{\mathcal{A}}^{(2)} \pi)_{l,v}) d\tau, & m = 0, \\ \int_0^1 (K_k)_{u'}(t_{i,j}, t_{l,v}, \eta_{l,v}, (L_{\mathcal{A}}^{(1)} \eta)_{l,v} + \tau (L_{\mathcal{A}}^{(1)} \varepsilon)_{l,v}, (L_{\mathcal{A}}^{(2)} \pi)_{l,v}) d\tau, & m = 1, \\ \int_0^1 (K_k)_{u''}(t_{i,j}, t_{l,v}, \eta_{l,v}, (L_{\mathcal{A}}^{(1)} \eta)_{l,v}, (L_{\mathcal{A}}^{(2)} \eta)_{l,v} + \tau (L_{\mathcal{A}}^{(2)} \varepsilon)_{l,v}) d\tau, & m = 2. \end{cases}$$

DEFINITION 5. For nonlinear and linear $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$) we define

$$\bar{\varkappa}_{\mathbf{f}}[e](t_{i,j}) := \sum_{m=0}^2 \int_a^b \Gamma_{m,\mathbf{f}}^e(t_{i,j}, s) e^{(m)}(s) ds,$$

$$\bar{\varkappa}_{\mathbf{v}}[e](t_{i,j}) := \sum_{m=0}^2 \int_a^{t_{i,j}} \Gamma_{m,\mathbf{v}}^e(t_{i,j}, s) e^{(m)}(s) ds,$$

where, for linear case we define

$$\Gamma_{m,k}^e(t_{i,j}, s) := \Lambda_{m,k}(t_{i,j}, s), \quad m = 0, 1, 2, \quad k = \mathbf{f}, \mathbf{v}$$

and for nonlinear case and $k = \mathbf{f}, \mathbf{v}$ we define

$$\Gamma_{m,k}^e(t_{i,j}, s) := \begin{cases} \int_0^1 (K_k)_u(t_{i,j}, s, y(s) + \tau e(s), p'(s), p''(s)) d\tau, & m = 0, \\ \int_0^1 (K_k)_{u'}(t_{i,j}, s, y(s), y'(s) + \tau e'(s), p''(s)) d\tau, & m = 1, \\ \int_0^1 (K_k)_{u''}(t_{i,j}, s, y(s), y'(s), y''(s) + \tau e''(s)) d\tau, & m = 2. \end{cases}$$

Now we can easily find the following lemma.

Lemma 9. For linear and nonlinear $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$), we get

$$\chi^k[\pi]_{i,j} - \chi^k[\eta]_{i,j} = \bar{\chi}^k[\varepsilon]_{i,j}, \quad k = \mathbf{f}, \mathbf{v}, \tag{3.3}$$

$$z_k[p](t_{i,j}) - z_k[y](t_{i,j}) = \bar{z}_k[e](t_{i,j}), \quad k = \mathbf{f}, \mathbf{v}. \tag{3.4}$$

Lemma 10. For linear and nonlinear $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$), we have

$$|\bar{\chi}^k[\varepsilon]_{i,j} - \bar{z}_k[e](t_{i,j})| = \mathcal{O}(h^m), \quad k = \mathbf{f}, \mathbf{v}. \tag{3.5}$$

Proof. In the linear case by using Lemma 7, Theorem 2 and the Integral mean value theorem we get

$$\begin{aligned} \bar{\chi}^{\mathbf{f}}[\varepsilon]_{i,j} - \bar{z}_{\mathbf{f}}[e](t_{i,j}) &= \chi^{\mathbf{f}}[\varepsilon]_{i,j} - z_{\mathbf{f}}[e](t_{i,j}) \\ &= \sum_{(l,v) \in \mathcal{T}} \delta_{l,v} (\Lambda_{0,\mathbf{f}}(t_{i,j}, t_{l,v}) \varepsilon_{l,v} + \sum_{m=1}^2 \Lambda_{m,\mathbf{f}}(t_{i,j}, t_{l,v}) (L_{\mathcal{A}}^{(m)} \varepsilon)_{l,v}) \\ &\quad - \sum_{l=0}^2 \int_a^b \Lambda_{l,\mathbf{f}}(t_{i,j}, s) e^{(l)}(s) ds \\ &\leq \frac{(b-a)(m+1)h}{h'} \mathcal{O}(h^m) \sum_{l=0}^2 \max_{(l,v) \in \mathcal{A}} \Lambda_{l,\mathbf{f}}(t_{i,j}, t_{l,v}) \\ &\quad + \mathcal{O}(h^m)(b-a) \sum_{l=0}^2 \Lambda_{l,\mathbf{f}}(t_{i,j}, \zeta_{i,j}^l) = \mathcal{O}(h^m), \end{aligned}$$

where $\zeta_{i,j}^l \in (a, b)$. In this step we study nonlinear case. According to (3.3) and (3.4) we obtain

$$\begin{aligned} |\bar{\chi}^{\mathbf{f}}[\varepsilon]_{i,j}| &= |\chi^{\mathbf{f}}[\pi]_{i,j} - \chi^{\mathbf{f}}[\eta]_{i,j}| \\ &= \sum_{(l,v) \in \mathcal{T}} \delta_{l,v} \left(K_{\mathbf{f}}(t_{i,j}, t_{l,v}, \pi_{l,v}, (L_{\mathcal{A}}^{(1)} \pi)_{l,v}, (L_{\mathcal{A}}^{(2)} \pi)_{l,v}) \right. \\ &\quad \left. - K_{\mathbf{f}}(t_{i,j}, t_{l,v}, \eta_{l,v}, (L_{\mathcal{A}}^{(1)} \eta)_{l,v}, (L_{\mathcal{A}}^{(2)} \eta)_{l,v}) \right) \leq \sum_{(l,v) \in \mathcal{T}} \delta_{l,v} \left(C |\pi_{l,v} - \eta_{l,v}| \right. \\ &\quad \left. + C |(L_{\mathcal{A}}^{(1)} \pi)_{l,v} - (L_{\mathcal{A}}^{(1)} \eta)_{l,v}| + C |(L_{\mathcal{A}}^{(2)} \pi)_{l,v} - (L_{\mathcal{A}}^{(2)} \eta)_{l,v}| \right) \\ &\leq C \mathcal{O}(h^m) h n (m+1) \leq C \frac{(b-a)(m+1)h}{h'} \mathcal{O}(h^m) = \mathcal{O}(h^m), \end{aligned}$$

also

$$\begin{aligned} |\bar{z}_{\mathbf{f}}[e](t_{i,j})| &= |z_{\mathbf{f}}[p](t_{i,j}) - z_{\mathbf{f}}[y](t_{i,j})| \\ &\leq \int_a^b |K_{\mathbf{f}}(t_{i,j}, s, p(s), p'(s), p''(s)) - K_{\mathbf{f}}(t_{i,j}, s, y(s), y'(s), y''(s))| ds \\ &\leq C(b-a) \sum_{l=0}^2 |p^{(l)}(s) - y^{(l)}(s)| = \mathcal{O}(h^m), \end{aligned}$$

therefore by using the triangle inequality we have

$$|\bar{z}_{\mathbf{f}}[e](t_{i,j}) - \bar{\chi}_{\mathbf{f}}[\varepsilon]_{i,j}| \leq |\bar{z}_{\mathbf{f}}[e](t_{i,j})| + |\bar{\chi}_{\mathbf{f}}[\varepsilon]_{i,j}| = \mathcal{O}(h^m).$$

In a similar way we can find (3.5) for $k = \mathbf{v}$. \square

DEFINITION 6. We define $b_l(t_{i,j})$ and $c_l(t_{i,j})$ ($l = 1, 2, \mathbf{f}, \mathbf{v}$) as follows

$$\begin{aligned} b_1(t_{i,j}) &:= \int_0^1 F_y(t_{i,j}, \eta_{i,j} + \tau\varepsilon_{i,j}, (L_{\mathcal{A}}^{(1)}\pi)_{i,j}, \chi^{\mathbf{f}}[\pi]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j}) d\tau, \\ b_2(t_{i,j}) &:= \int_0^1 F_{y'}(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j} + \tau(L_{\mathcal{A}}^{(1)}\varepsilon)_{i,j}, \chi^{\mathbf{f}}[\pi]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j}) d\tau, \\ b_{\mathbf{f}}(t_{i,j}) &:= \int_0^1 F_{z_{\mathbf{f}}}(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j}, \chi^{\mathbf{f}}[\eta]_{i,j} + \tau\bar{\chi}^{\mathbf{f}}[\varepsilon]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j}) d\tau, \\ b_{\mathbf{v}}(t_{i,j}) &:= \int_0^1 F_{z_{\mathbf{v}}}(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j}, \chi^{\mathbf{f}}[\eta]_{i,j}, \chi^{\mathbf{v}}[\eta]_{i,j} + \tau\bar{\chi}^{\mathbf{v}}[\varepsilon]_{i,j}) d\tau, \\ c_1(t_{i,j}) &:= \int_0^1 F_y(t_{i,j}, y(t_{i,j}) + \tau e(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j})) d\tau, \\ c_2(t_{i,j}) &:= \int_0^1 F_{y'}(t_{i,j}, y(t_{i,j}), y'(t_{i,j}) + \tau e'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j})) d\tau, \\ c_{\mathbf{f}}(t_{i,j}) &:= \int_0^1 F_{z_{\mathbf{f}}}(t_{i,j}, y(t_{i,j}), y'(t_{i,j}), z_{\mathbf{f}}[y](t_{i,j}) + \tau\bar{z}_{\mathbf{f}}[e](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j})) d\tau, \\ c_{\mathbf{v}}(t_{i,j}) &:= \int_0^1 F_{z_{\mathbf{v}}}(t_{i,j}, y(t_{i,j}), y'(t_{i,j}), z_{\mathbf{f}}[y](t_{i,j}), z_{\mathbf{v}}[y](t_{i,j}) + \tau\bar{z}_{\mathbf{v}}[e](t_{i,j})) d\tau. \end{aligned}$$

Lemma 11. We have

$$|b_l(t_{i,j}) - c_l(t_{i,j})| = \mathcal{O}(h), \quad l = 1, 2, \mathbf{f}, \mathbf{v}. \tag{3.6}$$

Proof. By using the Lipschitz condition for $F_y, F_{y'}, F_{z_{\mathbf{f}}}$ and $F_{z_{\mathbf{v}}}$ we can find

$$\begin{aligned} &\left| F_y(t_{i,j}, \eta_{i,j} + \tau\varepsilon_{i,j}, (L_{\mathcal{A}}^{(1)}\pi)_{i,j}, \chi^{\mathbf{f}}[\pi]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j}) \right. \\ &\quad \left. - F_y(t_{i,j}, y(t_{i,j}) + \tau e(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j})) \right| \end{aligned}$$

$$\begin{aligned}
&\leq C\left(|\eta_{i,j} - y(t_{i,j})| + \tau|\varepsilon_{i,j} - e(t_{i,j})|\right) + C\left|(L_{\mathcal{A}}^{(1)}\pi)_{i,j} - p'(t_{i,j})\right| \\
&\quad + C\left|\chi^{\mathbf{f}}[\pi]_{i,j} - z_{\mathbf{f}}[p](t_{i,j})\right| + C\left|\chi^{\mathbf{v}}[\pi]_{i,j} - z_{\mathbf{v}}[p](t_{i,j})\right| = \mathcal{O}(h), \quad (3.7) \\
&\left|F_{y'}(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j} + \tau(L_{\mathcal{A}}^{(1)}\varepsilon)_{i,j}, \chi^{\mathbf{f}}[\pi]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j})\right. \\
&\quad \left.- F_{y'}(t_{i,j}, y(t_{i,j}), y'(t_{i,j}) + \tau e'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j}))\right| \\
&\leq C|\eta_{i,j} - y(t_{i,j})| + C\left(|(L_{\mathcal{A}}^{(1)}\eta)_{i,j} - y'(t_{i,j})| + \tau|(L_{\mathcal{A}}^{(1)}\varepsilon)_{i,j} - e'(t_{i,j})|\right) \\
&\quad + C\left|\chi^{\mathbf{f}}[\pi]_{i,j} - z_{\mathbf{f}}[p](t_{i,j})\right| + C\left|\chi^{\mathbf{v}}[\pi]_{i,j} - z_{\mathbf{v}}[p](t_{i,j})\right| = \mathcal{O}(h), \\
&\left|F_{z_{\mathbf{f}}}(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j}, \chi^{\mathbf{f}}[\eta]_{i,j} + \tau\bar{\chi}^{\mathbf{f}}[\varepsilon]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j})\right. \\
&\quad \left.- F_{z_{\mathbf{f}}}(t_{i,j}, y(t_{i,j}), y'(t_{i,j}), z_{\mathbf{f}}[y](t_{i,j}) + \tau\bar{z}_{\mathbf{f}}[e](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j}))\right| \\
&\leq C|\eta_{i,j} - y(t_{i,j})| + C|(L_{\mathcal{A}}^{(1)}\eta)_{i,j} - y'(t_{i,j})| + C(|\chi^{\mathbf{f}}[\eta]_{i,j} - z_{\mathbf{f}}[y](t_{i,j})| \\
&\quad + \tau|\bar{\chi}^{\mathbf{f}}[\varepsilon]_{i,j} - \bar{z}_{\mathbf{f}}[e](t_{i,j})|) + C|\chi^{\mathbf{v}}[\pi]_{i,j} - z_{\mathbf{v}}[p](t_{i,j})| = \mathcal{O}(h), \\
&\left|F_{z_{\mathbf{v}}}(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j}, \chi^{\mathbf{f}}[\eta]_{i,j}, \chi^{\mathbf{v}}[\eta]_{i,j} + \tau\bar{\chi}^{\mathbf{v}}[\varepsilon]_{i,j})\right. \\
&\quad \left.- F_{z_{\mathbf{v}}}(t_{i,j}, y(t_{i,j}), y'(t_{i,j}), z_{\mathbf{f}}[y](t_{i,j}), z_{\mathbf{v}}[y](t_{i,j}) + \tau\bar{z}_{\mathbf{v}}[e](t_{i,j}))\right| \\
&\leq C|\eta_{i,j} - y(t_{i,j})| + C|(L_{\mathcal{A}}^{(1)}\eta)_{i,j} - y'(t_{i,j})| + C|\chi^{\mathbf{f}}[\eta]_{i,j} - z_{\mathbf{v}}[y](t_{i,j})| \\
&\quad + C(|\chi^{\mathbf{v}}[\eta]_{i,j} - z_{\mathbf{v}}[y](t_{i,j})| + \tau|\bar{\chi}^{\mathbf{v}}[\varepsilon]_{i,j} - \bar{z}_{\mathbf{v}}[e](t_{i,j})|) = \mathcal{O}(h).
\end{aligned}$$

Now we study (3.6). For $l = 1$, by using (3.7) we can get

$$\begin{aligned}
|b_1(t_{i,j}) - c_1(t_{i,j})| &\leq \int_0^1 \left|F_y(t_{i,j}, \eta_{i,j} + \tau\varepsilon_{i,j}, (L_{\mathcal{A}}^{(1)}\pi)_{i,j}, \chi^{\mathbf{f}}[\pi]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j})\right. \\
&\quad \left.- F_y(t_{i,j}, y(t_{i,j}) + \tau e(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j}))\right| d\tau = \mathcal{O}(h).
\end{aligned}$$

In a similar way we can find (3.6) for $l = 2, \mathbf{f}$ and \mathbf{v} . \square

When F is nonlinear we have the following theorem.

Theorem 5. Consider the SFVID equation (1.1) with boundary conditions (1.2), where $F(t, y, y', z_{\mathbf{f}}, z_{\mathbf{v}})$, $F_l(t, y, y', z_{\mathbf{f}}, z_{\mathbf{v}})$ ($l = y, y', z_{\mathbf{f}}, z_{\mathbf{v}}$) are Lipschitz-continuous. Also for nonlinear $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$) we let $K_l(t, s, u, u', u'')$ and $(K_l)_j(t, s, u, u', u'')$ ($l = \mathbf{f}, \mathbf{v}$ & $j = u, u', u''$) are Lipschitz-continuous. Assume that the SFVID problem has a unique and sufficiently smooth solution. Then the following estimate holds

$$\|\theta\|_{\infty} = \|e - \varepsilon\|_{\infty} = \mathcal{O}(h^{m+1}),$$

where e is error, ε is the error estimate and θ is the deviation of the error estimate.

Proof. We have

$$\begin{aligned}
 (L_{\mathcal{A}}^{(2)}\theta)_{i,j} &= (L_{\mathcal{A}}^{(2)}e)_{i,j} - (L_{\mathcal{A}}^{(2)}\varepsilon)_{i,j} \\
 &= \mathcal{I}_{\mathcal{A}} \left(\underbrace{F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j}))}_{R_1} \right. \\
 &\quad \left. - \underbrace{F(t_{i,j}, y(t_{i,j}), y'(t_{i,j}), z_{\mathbf{f}}[y](t_{i,j}), z_{\mathbf{v}}[y](t_{i,j}))}_{R_2} \right) \\
 &\quad - \underbrace{\left(F(t_{i,j}, \pi_{i,j}, (L_{\mathcal{A}}^{(1)}\pi)_{i,j}, \chi^{\mathbf{f}}[\pi]_{i,j}, \chi^{\mathbf{v}}[\pi]_{i,j}) \right)}_{R_3} \\
 &\quad - \underbrace{F(t_{i,j}, \eta_{i,j}, (L_{\mathcal{A}}^{(1)}\eta)_{i,j}, \chi^{\mathbf{f}}[\eta]_{i,j}, \chi^{\mathbf{v}}[\eta]_{i,j})}_{R_4} \\
 &\quad + Q_{\mathcal{A}} \left(F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), \tilde{z}_{\mathbf{f}}[p](t_{i,j}), \tilde{z}_{\mathbf{v}}[p](t_{i,j})) \right) \\
 &\quad - \mathcal{I}_{\mathcal{A}} \left(F(t_{i,j}, p(t_{i,j}), p'(t_{i,j}), z_{\mathbf{f}}[p](t_{i,j}), z_{\mathbf{v}}[p](t_{i,j})) \right). \tag{3.8}
 \end{aligned}$$

We can get

$$\begin{aligned}
 R_1 - R_2 &= \sum_{l=0}^1 c_{l+1}(t_{i,j})e^{(l)}(t_{i,j}) + \sum_{l=\mathbf{f},\mathbf{v}} c_l(t_{i,j})\bar{z}_l[e](t_{i,j}), \\
 R_3 - R_4 &= b_1(t_{i,j})\varepsilon_{i,j} + b_2(t_{i,j})(L_{\mathcal{A}}^{(1)}\varepsilon)_{i,j} + \sum_{l=\mathbf{f},\mathbf{v}} b_l(t_{i,j})\bar{\chi}^l[\varepsilon]_{i,j}.
 \end{aligned}$$

Therefore we can rewrite (3.8) as follows

$$\begin{aligned}
 (L_{\mathcal{A}}^{(2)}\theta)_{i,j} &= b_1(t_{i,j})\theta_{i,j} + b_2(t_{i,j})(L_{\mathcal{A}}^{(1)}\theta)_{i,j} + \sum_{l=\mathbf{f},\mathbf{v}} b_l(t_{i,j})\bar{\chi}^l[\theta]_{i,j} \\
 &\quad + \mathcal{I}_{\mathcal{A}} \left(c_1(t_{i,j})e(t_{i,j}) + c_2(t_{i,j})e'(t_{i,j}) + \sum_{l=\mathbf{f},\mathbf{v}} c_l(t_{i,j})\bar{z}_l[e](t_{i,j}) \right) \\
 &\quad - \left(c_1(t_{i,j})e(t_{i,j}) + c_2(t_{i,j})(L_{\mathcal{A}}^{(1)}e)_{i,j} + \sum_{l=\mathbf{f},\mathbf{v}} c_l(t_{i,j})\bar{\chi}^l[e]_{i,j} \right) \\
 &\quad + \underbrace{(c_1(t_{i,j}) - b_1(t_{i,j}))e(t_{i,j})}_{\mathcal{O}(h^m)} + \underbrace{(c_2(t_{i,j}) - b_2(t_{i,j}))(L_{\mathcal{A}}^{(1)}e)_{i,j}}_{\mathcal{O}(h^m)} \\
 &\quad + \sum_{l=\mathbf{f},\mathbf{v}} \underbrace{(c_l(t_{i,j}) - b_l(t_{i,j}))\bar{\chi}^l[e]_{i,j}}_{\mathcal{O}(h^m)} + Q_{\mathcal{A}}(F(\cdot, p, p', \tilde{z}_{\mathbf{f}}[p], \tilde{z}_{\mathbf{v}}[p]), t_{i,j}) \\
 &\quad - \mathcal{I}_{\mathcal{A}}(F(\cdot, p, p', z_{\mathbf{f}}[p], z_{\mathbf{v}}[p]), t_{i,j})). \tag{3.9}
 \end{aligned}$$

Then we rewrite (3.9) as follows

$$\begin{aligned}
 (L_{\mathcal{A}}^{(2)}\theta)_{i,j} &= b_1(t_{i,j})\theta_{i,j} + b_2(t_{i,j})(L_{\mathcal{A}}^{(1)}\theta)_{i,j} + \sum_{l=\mathbf{f},\mathbf{v}} b_l(t_{i,j})\bar{\chi}^l[\theta]_{i,j} \\
 &\quad + \mathcal{I}_{\mathcal{A}}(c_1(t_{i,j})e(t_{i,j}) + c_2(t_{i,j})e'(t_{i,j}) + \sum_{l=\mathbf{f},\mathbf{v}} c_l(t_{i,j})\bar{z}_l[e](t_{i,j}))
 \end{aligned}$$

$$\begin{aligned}
 & - (c_1(t_{i,j})e(t_{i,j}) + c_2(t_{i,j})(L_{\mathcal{A}}^{(1)}e)_{i,j} + \sum_{l=\mathbf{f},\mathbf{v}} c_l(t_{i,j})\bar{\chi}^l[e]_{i,j}) \\
 & + Q_{\mathcal{A}}(F(\cdot, p, p', \tilde{z}_{\mathbf{f}}[p], \tilde{z}_{\mathbf{v}}[p]), t_{i,j}) \\
 & - \mathcal{I}_{\mathcal{A}}(F(\cdot, p, p', z_{\mathbf{f}}[p], z_{\mathbf{v}}[p]), t_{i,j}) + \mathcal{O}(h^{m+1}).
 \end{aligned}$$

In a similar way to the Theorem 4, we can complete the proof. \square

4 Improvement of the deviation of the error estimate

In particular case of (1.1)–(1.2) with

$$z_{\mathbf{f}}[y](t) = \int_a^b K_{\mathbf{f}}(t, s, y(s))ds, \quad z_{\mathbf{v}}[y](t) = \int_a^t K_{\mathbf{v}}(t, s, y(s))ds \quad (4.1)$$

we can find second order finite difference scheme. In this case by using the trapezoidal rule we can find the following finite difference scheme

$$\begin{aligned}
 (L_{\mathcal{A}}^{(2)}\eta)_{i,j} &= F(t_{i,j}, \eta_{i,j}, (\bar{L}_{\mathcal{A}}^{(1)}\eta)_{i,j}, \varrho^{\mathbf{f}}[\eta]_{i,j}, \varrho^{\mathbf{v}}[\eta]_{i,j}), \quad (i, j) \in \mathcal{B}, \\
 \eta_{0,0} &= r_1, \quad \eta_{n,0} = r_2,
 \end{aligned}$$

where $(L_{\mathcal{A}}^{(2)}\eta)_{i,j}$ is defined in (2.6) and

$$\begin{aligned}
 (\bar{L}_{\mathcal{A}}^{(1)}\eta)_{i,j} &:= \frac{\eta_{i,j+1} - \eta_{i,j-1}}{2\hat{\delta}_{i,j}}, \\
 \varrho^{\mathbf{f}}[\eta]_{i,j} &:= \sum_{(l,v) \in \mathcal{T}} \frac{\delta_{l,v}}{2} \left(K_{\mathbf{f}}(t_{i,j}, t_{l,v}, \eta_{l,v}) + K_{\mathbf{f}}(t_{i,j}, t_{l,v+1}, \eta_{l,v+1}) \right), \\
 \varrho^{\mathbf{v}}[\eta]_{i,j} &:= \sum_{(l,v) \in \Delta_{i,j}} \frac{\delta_{l,v}}{2} \left(K_{\mathbf{v}}(t_{i,j}, t_{l,v}, \eta_{l,v}) + K_{\mathbf{v}}(t_{i,j}, t_{l,v+1}, \eta_{l,v+1}) \right).
 \end{aligned}$$

For above case, in the same way that, as discussed in the Section 3, we can easily prove the following two theorems.

Theorem 6. Consider the SFVID equation (1.3) with (4.1) and boundary conditions (1.2). Assume that the SFVID problem has a unique and sufficiently smooth solution. Then the following estimate holds

$$\|\theta\|_{\infty} = \|e - \varepsilon\|_{\infty} = \mathcal{O}(h^{m+2}),$$

where e is error, ε is the error estimate and θ is the deviation of the error estimate.

Theorem 7. Consider the SFVID equation (1.1) with (4.1) and boundary conditions (1.2), where $F(t, y, y', z_{\mathbf{f}}, z_{\mathbf{v}})$ and $F_l(t, y, y', z_{\mathbf{f}}, z_{\mathbf{v}})$ ($l = y, y', z_{\mathbf{f}}, z_{\mathbf{v}}$) are Lipschitz-continuous. Also for nonlinear $z_l[\cdot](t)$ ($l = \mathbf{f}, \mathbf{v}$) we let $K_l(t, s, u)$ and $(K_l)_j(t, s, u)$ ($l = \mathbf{f}, \mathbf{v} \& j = u$) are Lipschitz-continuous. Assume that the SFVID problem has a unique and sufficiently smooth solution. Then the following estimate holds

$$\|\theta\|_{\infty} = \|e - \varepsilon\|_{\infty} = \mathcal{O}(h^{m+2}),$$

where e is error, ε is the error estimate and θ is the deviation of the error estimate.

5 Numerical Examples

In this Section we apply the numerical results. The examples 1-2 considered below are used as test for Theorem 4 and Theorem 5. Also example 3 and example 4 serve to illustrate Theorem 6 and Theorem 7. The results obtained by using Mathematica-9 programming. Also in the examples the boundary conditions are taken from the exact solution.

Example 1. We consider the linear case as

$$y''(t) = ty'(t) + y(t) + a_3(t) + \int_0^1 ts \sum_{l=0}^2 y^{(l)}(s)ds + \int_0^t ts^2 \sum_{l=0}^2 y^{(l)}(s)ds.$$

In this case $a_3(t)$ chosen so that exact solution is $y(t) = \exp(2t)$. In the Table 1 we choose $m = 4$ and n collocation subintervals of length $1/n$. Also in the Table 2 we choose $m = 2$ and assume that $\{\rho_0 = 0, \rho_1 = 0.283333, \rho_2 = 0.616667, \rho_3 = 1\}$.

Table 1. Numerical results for example 1.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
4	3.25842e-2	–	3.91123e-6	–
8	2.11365e-6	3.94637	1.41308e-7	4.79071
16	1.32181e-7	3.99914	4.75659e-9	4.89277
32	8.26373e-9	3.99958	1.54231e-10	4.94676
64	5.16669e-10	3.99948	4.92406e-12	4.96910

Table 2. Numerical results for example 1.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
4	1.71121e-2	–	2.90399e-3	–
8	4.54653e-3	1.91218	4.43366e-4	2.71146
16	1.15124e-3	1.98158	6.15327e-5	2.84907
32	2.89467e-4	1.99172	8.23059e-6	2.90229
64	7.25721e-5	1.99591	1.09647e-6	2.90814

Example 2. We consider the problem

$$y''(t) = ty'(t) + y^2(t) + a_3(t) + \int_0^1 ts \left(y(s) + y'(s) + (y''(s))^2 \right) ds + \int_0^t ts^2 (y(s)y'(s) + y''(s)) ds,$$

$a_3(t)$ chosen so that exact solution is $y(t) = \sin 2t$. For this example we choose n collocation subintervals of length $1/n$. In the Table 3 we choose $m = 4$ and assume that $\rho_i (i = 0, \dots, 5)$ are equidistant points. In the Table 4 we choose $m = 3$ and $\{\rho_0 = 0, \rho_1 = 0.21, \rho_2 = 0.46, \rho_3 = 0.71, \rho_4 = 1\}$.

Table 3. Numerical results for example 2.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
2	5.65455e-7	–	9.43288e-8	–
4	3.21333e-8	4.13727	3.58181e-9	4.71894
8	2.02451e-9	3.98843	1.21604e-10	4.88043
16	1.25615e-10	4.01049	3.94984e-12	4.94425

Table 4. Numerical results for example 2.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
2	1.20478e-5	–	4.38700e-6	–
4	1.58034e-6	2.93046	2.48796e-7	4.14020
8	2.02665e-7	2.96307	1.45576e-8	4.09512
16	2.57390e-8	2.97707	8.12608e-10	4.16307

Example 3. In this example we consider here the following linear problem

$$y''(t) = \cos(t)y'(t) + y(t) + a_3(t) + \int_0^1 y(s)ds + \int_0^t t \cos(s)y(s)ds,$$

$a_3(t)$ chosen so that exact solution is $y(t) = t \sin(t)$. For this example we choose n collocation subintervals of length $1/n$. In the Table 5 we choose $m = 2$ and assume that $\rho_i (i = 0, \dots, 3)$ are equidistant points. Also in the Table 6, we consider $m = 3$ and $\{\rho_0 = 0, \rho_1 = 0.2, \rho_2 = 0.55, \rho_3 = 0.72, \rho_4 = 1\}$.

Table 5. Numerical results for example 3.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
2	2.26647e-3	–	8.45242e-6	–
4	5.63358e-4	2.00832	4.92246e-7	4.10191
8	1.40539e-4	2.00308	2.73140e-8	4.17167
16	3.51146e-5	2.00083	1.61489e-9	4.08014

Table 6. Numerical results for example 3.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
2	1.25407e-5	–	4.98589e-7	–
4	2.63666e-6	2.24984	1.47117e-8	5.08282
8	4.24484e-7	2.63493	4.31918e-10	5.09006
16	6.00350e-8	2.82183	7.28798e-12	5.88910

Example 4. As a last study we consider the nonlinear case as

$$y''(t) = y'(t) + y^2(t) + a_3(t) + \left(\int_0^1 \sin(t)y(s)ds \right) \left(\int_0^t \cos(s)y(s)ds \right),$$

$a_3(t)$ chosen so that exact solution is $y(t) = \sin(t)$. In this example for Table 7 we choose $m = 2$ and assume that τ_i are equidistant point. Also In this table we choose equidistant points for ρ_i . In the Table 8 we choose $m = 3$ and $\{\rho_0 = 0, \rho_1 = 0.23, \rho_2 = 0.55, \rho_3 = 0.78, \rho_4 = 1\}$.

Table 7. Numerical results for example 4.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
2	3.68252e-4	–	1.38395e-6	–
4	8.78560e-5	2.06748	5.77515e-8	4.58279
8	2.28186e-5	1.94493	2.29570e-9	4.65285
16	5.68809e-6	2.00419	1.00123e-10	4.51909

Table 8. Numerical results for example 4.

n	$\ e\ _\infty$	Order	$\ \theta\ _\infty$	Order
2	1.80270e-6	–	9.53181e-8	–
4	2.48418e-7	2.85931	1.52154e-9	5.96915
8	3.46815e-8	2.84053	2.39238e-11	5.99094
16	4.55469e-9	2.92874	7.47957e-13	4.99934

Conclusions

In this paper, we have constructed efficient asymptotically correct a posteriori error estimates for the numerical approximation of second order Fredholm - Volterra integro differential equations. Also it is shown that when we use m degree piecewise polynomial collocation method, the order of the deviation of the error estimation is $\mathcal{O}(h^{m+1})$. In the previous section, numerical examples confirming the theoretical results are given.

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