Simplest results presented here are the stability criteria of collocation methods for the second-order Volterra integrodifferential equation (VIDE) by polynomial spline functions. The polynomial spline collocation method is stable if all eigenvalues of a matrix are in the unit disk and all eigenvalues with $|\lambda| = 1$ belong to a $1 \times 1$ Jordan block. Also many other conditions are derived depending upon the choice of collocation parameters used in the solution procedure.

1. Introduction

In order to discuss the numerical stability, we consider the linear second-order Volterra integrodifferential equation of the form

$$y^{(2)}(t) = q(t) + \sum_{i=0}^{1} p_i(t)y^{(i)}(t) + \sum_{i=0}^{1} \int_{0}^{t} k_i(t,s)y^{(i)}(s)\,ds, \quad t \in I := [0,T], \tag{1.1}$$

with

$$y(0) = y_0, \quad y^{(1)}(0) = y_1, \tag{1.2}$$

where $q : I \to R$, $p_i : I \to R$, and $k_i : D \to R (i = 0,1)$ (with $D := \{(t,s) : 0 \leq s \leq t \leq T\}$) are given functions and are assumed to be (at least) continuous in the respective domains. For more details of these equations, many other interesting methods for the approximated solution and stability procedure are available in earlier literature [1, 2, 3, 4, 5, 6, 9, 10, 12, 13]. The above equation is usually known as basis test equation and is suggested by Brunner and Lambert in [4]. Since then, it has been widely used for analyzing the stability properties of various methods.

Second-order VIDEs of the above form (1.1) will be solved numerically using polynomial spline spaces. In order to describe these approximating polynomial spline spaces,
let $\prod_N : 0 = t_0 < t_1 < \cdots < t_N = T$ be the mesh for the interval $I$, and set

$$
\sigma_n := [t_n, t_{n+1}], \quad h_n := t_{n+1} - t_n, \quad n = 0, 1, \ldots, N - 1, \\
h = \max\{h_n : 0 \leq n \leq N - 1\}, \quad (\text{mesh diameter}), \\
Z_N := \{t_n : n = 1, 2, \ldots, N - 1\}, \quad \bar{Z}_N = Z_N \cup \{T\}. \tag{1.3}
$$

Let $\pi_{m+d}$ be the set of (real) polynomials of degree not exceeding $m + d$, where $m \geq 1$ and $d \geq -1$ are given integers. The solution to initial-value problem (1.1) will be approximated by an element $u$ in the polynomial spline space,

$$
S^{(d)}_{m+d}(Z_N) := \left\{ u := u(t) \mid t \in \sigma_n := u_n(t) \in \pi_{m+d}, n = 0, 1, \ldots, N - 1, \\
u^{(j)}_{n-1}(t_n) = u^{(j)}(t_n) \text{ for } j = 0, 1, \ldots, d, \quad t_n \in Z_N \right\}; \tag{1.4}
$$

that is, by a polynomial spline function of degree $m + d$ which possesses the knots $Z_N$ and is $d$ times continuously differentiable on $I$. If $d = -1$, then the elements of $S^{(-1)}_{m-1}(Z_N)$ may have jump discontinuities at the knots $Z_N$. There are many other papers which had treated such problem using $S^{(0)}_m(Z_N)$ and $S^{(1)}_m(Z_N)$ [3, 4, 6] polynomial spline spaces.

According to Miculá et al. [11], an element $u \in S^{(d)}_{m+d}(Z_N)$ has the following form (for all $n = 0, 1, \ldots, N - 1$ and $t \in \sigma_n$):

$$
u(t) = u_n(t) = \sum_{r=0}^{d} \frac{u^{(r)}_{n-1}(t_n)}{r!} (t - t_n)^r + \sum_{r=1}^{m} a_{n,r}(t - t_n)^{d+r}, \tag{1.5}\n$$

where

$$
u^{(r)}_{n-1}(0) := \left[ \frac{d^r}{dt^r} u(t) \right]_{t=0} = y^{(r)}(0), \quad r = 0, 1, \ldots, d. \tag{1.6}\n$$

From (1.5), we see that the element $u \in S^{(d)}_{m+d}(Z_N)$ is well defined provided that the coefficients $\{a_{n,r}\}_{r=1,\ldots,m}$ for all $n = 0, 1, \ldots, N - 1$ are known. In order to determine these coefficients, we consider a set of collocation parameters $\{c_j\}_{j=1,\ldots,m}$, where $0 < c_1 < \cdots < c_m \leq 1$, and define the set of collocation points as

$$
X(N) := \bigcup_{n=0}^{N-1} X_n, \quad \text{with } X_n := \{t_{n,j} := t_n + c_j h_n, \quad j = 1, 2, \ldots, m\}. \tag{1.7}\n$$

The approximate solution $u \in S^{(d)}_{m+d}(Z_N)$ will be determined by imposing the condition that $u$ satisfies the initial-value problem (1.1) on $X(N)$ and the initial conditions, that is,

$$
u^{(2)}(t) = q(t) + \sum_{i=0}^{1} p_i(t) u^{(i)}(t) + \sum_{i=0}^{1} \int_{0}^{t} k_i(t,s) u^{(i)}(s) \, ds, \quad \forall t \in X(N), \tag{1.8}\n$$

with

$$
u(0) = y_0, \quad \nu^{(1)}(0) = y_1 \tag{1.9}\n$$

with a uniform mesh sequence $\{\prod_N\}, h_n = h$, for all $n = 0, 1, \ldots, N - 1$.  


2. Numerical stability

In [7], Danciu studied the numerical stability of the collocation method for first-order integrodifferential equations. He studied the behavior of the method as applied to the initial-value problem integrodifferential test equation

\[ y'(t) = q(t) + \alpha_0 y(t) + \nu \int_0^t y(s) \, ds, \quad \nu \neq 0. \]  

Equation (2.1) has been suggested by Brunner and Lambert in 1974 (see [4]), since then it has been extensively used as a basis for investigating the stability properties of several other methods.

In order to discuss the numerical stability for second-order integrodifferential equations, we study the numerical stability of the collocation spline method when applied to the initial-value problem integrodifferential test equation of the following form:

\[ y''(t) = q(t) + \alpha_0 y(t) + \alpha_1 y'(t) + \nu \int_0^t y(s) \, ds, \quad \nu \neq 0, \quad \text{(2.2)} \]

\[ y(0) = y_0, \quad y'(0) = y_1, \quad \text{(2.3)} \]

where \(\alpha_1, \alpha_2,\) and \(\nu\) are constants, and the given function \(g : I \to \mathbb{R}\) is sufficiently smooth.

For simplicity, we use a polynomial spline collocation method in the space \(S_{m+d}(Z_N)\), as an \((m,d)\)-method (see [8]).

Definition 2.1. An \((m,d)\)-method is said to be stable if all solutions \(\{u(t_n)\}\) remain bounded, as \(n \to \infty, h \to 0\) while \(h_n\) remains fixed.

From (1.5), we observe that the first \(d+1\) coefficients of the \(u \in S_{m+d}(Z_N)\) are determine by the smooth conditions, and then the collocation conditions can be used to determine the last \(m\) coefficients. Thus, for convenience, we introduce the following notations:

\[ \eta_n := (\eta_{n,r})_{r=0,\ldots,d}, \quad \text{with} \quad \eta_{n,r} := \frac{u^{(r)}_{n-1}(t_n)}{r!} h^r; \]

\[ \beta_n := (\beta_{n,r})_{r=1,\ldots,m}, \quad \text{with} \quad \beta_{n,r} := a_{n,r} h^{d+r}, \quad n = 0, 1, \ldots, N. \]  

Using (2.4) in (1.5), for all \(t := t_n + \tau h \in \sigma_n\), we have the following equation:

\[ u(t) = u_n(t_n + \tau h) = \sum_{r=0}^d \eta_{n,r} \tau^r + \sum_{r=1}^m \beta_{n,r} \tau^{d+r}, \quad \forall \tau \in (0,1], \quad n = 0, 1, \ldots, N. \]  

By applying the collocation method to test (2.2) for the case \(d \geq 2\) and using (2.5) we have the following collocation equation:

\[ V \beta_n = W \eta_n + h^2 R_n, \quad \forall n = 0, 1, \ldots, N - 1, \]  

where $V$ is the $m \times m$ matrix, $W$ is the $m \times (d+1)$ matrix, and $R_n$ is the $m$-vector, whose elements are given by

$$V_{j,r} := \begin{cases} (d+r)(d+r-1) - \alpha_0 h^2 c_j^2 - \alpha_1 h(d+r) c_j - \frac{\gamma h^3}{(d+r+1)} c_j^3 & \text{if } r = 0, \\ \gamma h^3 c_j & \text{if } r = 1, \\ \alpha_0 h^2 c_j + \frac{\gamma h^3}{2} c_j^2 & \text{if } r = 2, \\ \alpha_0 h^2 c_j^2 + 2\alpha_1 h c_j + \frac{\gamma h^3}{3} c_j^3 & \text{if } 3 \leq r \leq d, \\ \left[\alpha_0 h^2 c_j^2 + \alpha_1 h r c_j + \frac{\gamma h^3}{r+1} c_j^3 - r(r-1)\right] c_j^{r-2} & \text{if } 3 \leq r \leq d, \end{cases}$$

(2.7)

$$W_{j,r} := \begin{cases} q(t_{0,j}) - q(t_0) & \text{if } n = 0, \\ q(t_{n,j}) - q(t_{n-1,m}) + u''_{n-1}(t_{n-1,m}) - u''_{n-1}(t_n) & \text{if } n > 0. \end{cases}$$

$$R_{n,j} := \begin{cases} +\alpha_0 [u_{n-1}(t_n) - u_{n-1}(t_{n-1,m})] + \alpha_1 [u'_{n-1}(t_n) - u'_{n-1}(t_{n-1,m})] \\ +\gamma h \int_{t_m}^{t_{n-1}} u_{n-1}(t_{n-1} + \tau h) d\tau & \text{if } n > 0. \end{cases}$$

By direct differentiation of (2.5) and using the smooth conditions of the approximation $u \in S_{m+d}(Z_N)$, we get a relationship between vector $\eta_{n+1}$ and vectors $\eta_n$ and $\beta_n$, as follows:

$$\eta_{n+1} = A \eta_n + B \beta_n, \quad \forall \ n = 0,1,\ldots,N-1,$$

(2.8)

where $A$ is the $(d+1) \times (d+1)$ upper triangular matrix and $B$ is the $(d+1) \times m$ matrix, whose elements are given by

$$a_{j,r} := \begin{cases} 0 & \text{if } r < j, \\ (r \choose j) & \text{if } r \geq j, \end{cases} \quad b_{j,r} := (d+r \choose j).$$

(2.9)

For $h$ small enough, the matrix $V$ is invertible since the determinant of $V$ is a Vandermonde-type determinant for $h = 0$. Hence from (2.6) and (2.8), we have

$$\eta_{n+1} = A \eta_n + BV^{-1} [W \eta_n + h^2 R_n] = (A + BV^{-1} W) \eta_n + h^2 BV^{-1} R_n.$$

(2.10)

Thus we have the following recurrence relation:

$$\eta_{n+1} = M \eta_n + h^2 BV^{-1} R_n,$$

(2.11)

where

$$M := A + BV^{-1} W.$$
Therefore, we have the following theorem which represents a stability criterion for the present method. The proof of this theorem is quite similar to the proof given by Danciu [7] for first-order VIDEs.

**Theorem 2.2.** An \((m,d)\)-method is stable if and only if all eigenvalues of matrix \(M = A + BV^{-1}W\) are in the unit disk and all eigenvalues with \(|\lambda| = 1\) belong to a \(1 \times 1\) Jordan block.

**Remark 2.3.** Note that the dimension of the matrix \(M\) is \(d + 1\). Moreover, if we denote by \(M_0\) the matrix \(M\) with \(h = 0\), and by \(\lambda^{(0)}\) and \(\lambda\) the eigenvalues of \(M_0\) and \(M\), respectively, then it follows that the matrix \(M_0\) has \(\lambda_1^{(0)} = \lambda_2^{(0)} = \lambda_3^{(0)} = 1\), for \(m \geq 1\) and \(d \geq 2\).

3. Applications

In this section, we will investigate the following special cases.

(I) For the case \(d = 2\), the approximation space is \(S_{m+2}^{(2)}(Z_N)\). From Theorem 2.2 and Remark 2.3, we have the following theorem.

**Theorem 3.1.** For every choice of the collocation parameters \(\{c_j\}_{j=1,m}\), an \((m,2)\)-method is stable for all \(m \geq 1\).

(II) For the case \(m = 1\), this choice of \(m\) corresponds to a classical spline function, that is, the approximate solution \(u \in S_{1+d}^{(d)}(Z_N)\). Using notations from Remark 2.3 (i.e., \(M_0\) is the matrix \(M\) with \(h = 0\), and \(\lambda^{(0)}\) and \(\lambda\) are the respective eigenvalues of \(M_0\) and \(M\)), we have

\[
\lambda = \lambda^{(0)} + O(h). \tag{3.1}
\]

If \(c_1 \in (0,1]\) is the collocation parameter, then for all \(d \geq 1\), using the binomial expansion, we find that for matrix \(M_0\) the trace is,

\[
\text{Tr}(M_0) = d + 2 + \frac{1}{c_1^{d-1}} - \left(1 + \frac{1}{c_1}\right)^{d-1}. \tag{3.2}
\]

As regard the stability of the spline collocation method, we have the following result.

**Theorem 3.2.** A \((1,d)\)-method is stable if and only if one of the following conditions is true:

(i) \(d = 2\) and \(c_1 \in (0,1]\),

(ii) \(d = 3\) and \(c_1 = 1\).

**Proof.** For the case \(d = 2\), this theorem follows from Theorem 3.1. If \(d = 3\), then the fourth eigenvalue of \(M_0\) is \(\lambda_4^{(0)} = 1 - (2/c_1) \leq -1\) for \(c_1 \in (0,1]\), and its absolute value is 1, if and only if \(c_1 = 1\). If \(d \geq 4\), then setting \(p = d - 1\) in (3.2), we have

\[
\text{Tr}(M_0) = p + 3 + \frac{1}{c_1^p} - \left(1 + \frac{1}{c_1}\right)^p, \tag{3.3}
\]

so if \(d > 4\) and \(c_1 \in (0,1]\), then

\[
\infty < \text{Tr}(M_0) < -p = -(d - 1). \tag{3.4}
\]
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Since \( \text{Tr}(M_0) = \lambda_1^{(0)} + \lambda_2^{(0)} + \cdots + \lambda_{d+1}^{(0)} < -d + 1 \), and \( \lambda_1^{(0)} = 1 \), it results that there exists an eigenvalue \( \lambda^{(0)} \) whose value is smaller than \(-1\). If \( d = 4 \) then from (3.2), \( \lambda_4^{(0)} + \lambda_5^{(0)} \leq -4 \), and therefore \( \lambda_4^{(0)} < -1 \) or \( \lambda_5^{(0)} < -1 \). Thus from Theorem 2.2 we have that, for \( d \geq 4 \), a \((1,d)\)-method is unstable for any choice of the collocation parameter \( c_1 \in (0,1) \). □

(III) For \( m = 2 \), we can prove the following result but the proof is the same as in [7].

Theorem 3.3. Let \( 0 < c_1 < c_2 \leq 1 \) be the collocation parameters, then

(i) a \((2,2)\)-method is stable for every choice of the collocation parameters,
(ii) a \((2,3)\)-method is stable if and only if \( c_1 + c_2 \geq 3/2 \),
(iii) if \( c_2 = 1 \), then a \((2,d)\)-method is unstable for all \( d \geq 4 \).

(IV) For the case \( d = 3 \), the approximation \( u \in S_{m+3}^{(3)}(Z_N) \) and the dimension of the matrix \( M_0 \) is 4 and its first \( p + 1 \) eigenvalues are \( \lambda_1^{(0)} = \lambda_2^{(0)} = \lambda_3^{(0)} = 1 \). To compute the fourth eigenvalue, we need the following results. But, first we introduce the following notations:

\[
S_k := S_k(c_1, \ldots, c_m) = \sum_{1 \leq i_1 < \cdots < i_k \leq m} c_{i_1} c_{i_2} \cdots c_{i_k}, \quad \text{for } 1 \leq k \leq m,
\]

\[
S_0 := S_0(c_1, \ldots, c_m) = 1,
\]

\[
S_{k,j} := S_k(c_1, \ldots, c_{j-1}, c_{j+1}, \ldots, c_m), \quad \text{for } 1 \leq k \leq m-1, 1 \leq j \leq m.
\]

Lemma 3.4. Let \( 0 < c_1 < c_2 < \cdots < c_m \leq 1 \) be the collocation parameters, then

\[
\begin{vmatrix}
1 & c_1 & c_2 & \cdots & c_1^{i-1} & c_1^{i+1} & \cdots & c_i^m \\
1 & c_2 & c_3 & \cdots & c_2^{i-1} & c_2^{i+1} & \cdots & c_i^m \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_m & c_{m+1} & \cdots & c_m^{i-1} & c_m^{i+1} & \cdots & c_i^m
\end{vmatrix} = S_{m-i} \prod_{1 \leq k < j \leq m} (c_j - c_k).
\] (3.6)

Proof. We will prove the lemma by induction on the dimension of the matrix, starting with \( 2 \times 2 \) matrices. For the \( 2 \times 2 \) matrices, the result is clearly true. For \( m \times m \) matrices \((m > 2)\), we define

\[
f(x) := \begin{vmatrix}
1 & c_1 & c_2 & \cdots & c_1^{i-1} & c_1^{i+1} & \cdots & c_i^m \\
1 & c_2 & c_3 & \cdots & c_2^{i-1} & c_2^{i+1} & \cdots & c_i^m \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_{m-1} & c_m & \cdots & c_{m-1}^{i-1} & c_{m-1}^{i+1} & \cdots & c_i^m \\
1 & x & x^2 & \cdots & x^{i-1} & x^{i+1} & \cdots & x^m
\end{vmatrix}.
\] (3.7)
Note that
\[
\begin{vmatrix}
1 & c_1 & c_1^2 & \cdots & c_1^{i-1} & c_1^{i+1} & \cdots & c_1^m \\
1 & c_2 & c_2^2 & \cdots & c_2^{i-1} & c_2^{i+1} & \cdots & c_2^m \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_m & c_m^2 & \cdots & c_m^{i-1} & c_m^{i+1} & \cdots & c_m^m
\end{vmatrix}
= f(c_m).
\tag{3.8}
\]

Now, since \( f(c_1) = f(c_2) = \cdots = f(c_{m-1}) = 0 \), we have
\[
f(x) = a(x - b) \prod_{i=1}^{m-1} (x - c_i),
\tag{3.9}
\]
where \( a, b \) are constants to be determined. By the induction hypothesis, we obtain
\[
a = S_{m-1-i}(c_1, \ldots, c_{m-1}) \prod_{k<j}^{m-1} (c_j - c_k). \tag{3.10}
\]
Moreover, from (3.9),
\[
f(0) = a(-1)^m c_1 c_2 \cdots c_{m-1} b.
\tag{3.11}
\]
On the other hand, from the definition of \( f \) and by the induction hypothesis, we have
\[
f(0) = (-1)^{m+1} \begin{vmatrix}
c_1 & c_1^2 & \cdots & c_1^{i-1} & c_1^{i+1} & \cdots & c_1^m \\
c_2 & c_2^2 & \cdots & c_2^{i-1} & c_2^{i+1} & \cdots & c_2^m \\
\vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
c_{m-1} & c_{m-1}^2 & \cdots & c_{m-1}^{i-1} & c_{m-1}^{i+1} & \cdots & c_{m-1}^m
\end{vmatrix}
= (-1)^{m+1} c_1 c_2 \cdots c_{m-1} S_{m-1}(c_1, \ldots, c_{m-1}) \prod_{k<j}^{m-1} (c_j - c_k).
\tag{3.12}
\]
Thus, from (3.11) and (3.12), we have
\[
-ab = S_{m-1}(c_1, \ldots, c_{m-1}) \prod_{k<j}^{m-1} (c_j - c_k),
\tag{3.13}
\]
and so
\[
f(c_m) = a(c_m - b) \prod_{i=1}^{m-1} (c_m - c_i)
= \left[ c_m S_{m-1-i}(c_1, \ldots, c_{m-1}) \prod_{k<j}^{m-1} (c_j - c_k) \right] \prod_{i=1}^{m-1} (c_m - c_i).
\tag{3.14}
\]
However, since

\[
\begin{align*}
  cmS_{m-1-i}(c_1, \ldots, c_{m-1}) + S_{m-i}(c_1, \ldots, c_{m-1}) &= S_{m-i}(c_1, \ldots, c_m) = S_{m-i}, \\
  \prod_{k<j} (c_j - c_k) \prod_{i=1}^{m-1} (c_m - c_i) &= \prod_{k<j} (c_j - c_k),
\end{align*}
\]  

we have

\[
  f(c_m) = S_{m-i} \prod_{k<j} (c_j - c_k),
\]

which proves the lemma. \(\square\)

Remark 3.5. Note that in Lemma 3.4, if \(i = m\), then we have the Vandermonde determinant.

Corollary 3.6. Let \(V_0\) be the matrix \(V\) with \(h = 0, d = 3\), that is, \(V_0\) is the \(m \times m\) matrix whose elements are

\[
(V_0)_{j,r} := ((r+3)(r+2))C_j^{r+1}.
\]

Then, \(V_0^{-1}\) is the matrix whose elements are

\[
(V_0^{-1})_{r,j} = \frac{1}{\det(V_0)} (-1)^r \prod_{l<k, l \neq j} (c_k - c_l) \prod_{k=1, k \neq r}^{m} (k+2)(k+3),
\]

where

\[
\det(V_0) = \left[ \prod_{k=1}^{m} (k+2)(k+3) \prod_{l<k} (c_k - c_l) \right] S_m^2.
\]

Proof. From Lemma 3.4, we have

\[
\det(V_0) = \left[ \prod_{k=1}^{m} (k+2)(k+3) \prod_{l<k} (c_k - c_l) \right] S_m^2.
\]

Now

\[
V_0^{-1} = \frac{\text{Adj}(V_0)}{\det(V_0)},
\]
where Adj($V_0$) is the adjoint matrix of $V_0$, however,

$$\text{Adj}(V_0)_{r,j} = (-1)^{r+j} S_{m-1,j}^2 \prod_{k=1, (k \neq r)}^{m} (k+2)(k+3)$$

\[
\begin{vmatrix}
1 & c_1 & c_1^2 & \cdots & c_1^{r-2} & c_1^r & \cdots & c_1^{m-1} \\
1 & c_2 & c_2^2 & \cdots & c_2^{r-2} & c_2^r & \cdots & c_2^{m-1} \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_{j-1} & c_{j-1}^2 & \cdots & c_{j-1}^{r-2} & c_{j-1}^r & \cdots & c_{j-1}^{m-1} \\
1 & c_{j+1} & c_{j+1}^2 & \cdots & c_{j+1}^{r-2} & c_{j+1}^r & \cdots & c_{j+1}^{m-1} \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_m & c_m^2 & \cdots & c_m^{r-2} & c_m^r & \cdots & c_m^{m-1} \\
\end{vmatrix} \times
\begin{vmatrix}
1 & c_1 & c_1^2 & \cdots & c_1^{r-2} & c_1^r & \cdots & c_1^{m-1} \\
1 & c_2 & c_2^2 & \cdots & c_2^{r-2} & c_2^r & \cdots & c_2^{m-1} \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_{j-1} & c_{j-1}^2 & \cdots & c_{j-1}^{r-2} & c_{j-1}^r & \cdots & c_{j-1}^{m-1} \\
1 & c_{j+1} & c_{j+1}^2 & \cdots & c_{j+1}^{r-2} & c_{j+1}^r & \cdots & c_{j+1}^{m-1} \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_m & c_m^2 & \cdots & c_m^{r-2} & c_m^r & \cdots & c_m^{m-1} \\
\end{vmatrix} = S_{m-r,j} \prod_{l<k, (l \neq j)}^{m} (c_k - c_l).$$

Again, by Lemma 3.4 and using relations

$$S_{m-1-(r-1)}(c_1, \ldots, c_{j-1}, c_{j+1}, \ldots, c_m) = S_{m-r}(c_1, \ldots, c_{j-1}, c_{j+1}, \ldots, c_m) = S_{m-r,j},$$

we have

$$\text{det} \left( V_0 \right)_{r,j} = \frac{1}{S_m} (-1)^{r+j} S_{m-1,j}^2 S_{m-r,j} \prod_{l<k, (l \neq j)}^{m} (c_k - c_l) \prod_{k=1, (k \neq r)}^{m} (k+2)(k+3),$$

which completes the proof of the corollary. □

Now, we can derive a formula for computing the $p+2$-eigenvalue of the matrix $M_0$.

**Theorem 3.7.** For the case $d = 3$ and $m \geq 1$, the $p+2$-eigenvalue of $M_0$ can be computed by using the following relation:

$$\lambda_{4}^{(0)} = \frac{S_m - 2S_{m-1} + 3S_{m-2} + \cdots + (-1)^{m-1}mS_1 + (-1)^m(m+1)}{S_m}.$$
Proof. Let $V_0$ and $W_0$ be the matrices $V$ and $W$, respectively, with $h = 0$, then for $d = 3$, $W_0$ is an $m \times 4$ matrix whose elements are given by

$$
(W_0)_{j,r} := \begin{cases} 
0 & \text{if } r = 0, 1, 2, \\
-6c_j & \text{if } r = 3.
\end{cases}
$$

(3.27)

Now, the fourth eigenvalue of $M_0 = A + BV_0^{-1}W_0$ is

$$
\lambda_4^{(0)} = 1 + \sum_{r=1}^{m} (B)_{4,r} (V_0^{-1}W_0)_{r,4}.
$$

(3.28)

From (2.8), the entries of the last row of matrix $B$ are

$$
(B)_{4,r} = \begin{pmatrix} 3 + r \\ 3 \end{pmatrix}.
$$

(3.29)

Moreover, from (3.27) and Corollary 3.6, we have

$$
(V_0^{-1}W_0)_{r,4} = -\frac{6}{\det(V_0)} \sum_{j=1}^{m} \left[ (-1)^{(r+j)} S_{m-1,j}^2 S_{m-r,j} c_j \right. \\
\times \left. \prod_{l<k, (l,k \neq j)} (c_k - c_l) \prod_{k=1, (k \neq r)}^{m} (k+2)(k+3) \right].
$$

(3.30)

Therefore,

$$
\lambda_4^{(0)} = 1 + \frac{6}{\det(V_0)} \sum_{r=1}^{m} \sum_{j=1}^{m} \left[ \frac{3 + r}{3} \right] (-1)^{(r+j+1)} S_{m-1,j}^2 S_{m-r,j} c_j \\
\times \prod_{l<k, (l,k \neq j)} (c_k - c_l) \prod_{k=1, (k \neq r)}^{m} (k+2)(k+3) \right].
$$

(3.31)

By using relations

$$
c_j S_{m-1,j}^2 = S_m S_{m-1,j},
$$

$$
6 \left( \frac{3 + r}{3} \right) \prod_{k=1, (k \neq r)}^{m} (k+2)(k+3) = (r+1) \prod_{k=1}^{m} (k+2)(k+3),
$$

(3.32)

and $\det(V_0)$, the above expression can be simplified as follows:

$$
\lambda_4^{(0)} = 1 + \frac{\sum_{r=1}^{m} (-1)^r (r+1) \sum_{j=1}^{m} (-1)^{(j+1)} S_{m-1,j} S_{m-r,j} \prod_{l<k, (l,k \neq j)} (c_k - c_l)}{S_m \prod_{l<k} (c_k - c_l)}.
$$

(3.33)
However, from Lemma 3.4, we have

\[
\sum_{j=1}^{m} (-1)^{(j+1)} S_{m-1,j} S_{m-r,j} \prod_{l<k,(l+k)} (c_k - c_l) = \left| \begin{array}{ccccccc}
1 & c_1 & c_1^2 & \cdots & c_1^{r-1} & c_1^{r+1} & \cdots & c_1^{m} \\
1 & c_2 & c_2^2 & \cdots & c_2^{r-1} & c_2^{r+1} & \cdots & c_2^{m} \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
1 & c_m & c_m^2 & \cdots & c_m^{r-1} & c_m^{r+1} & \cdots & c_m^{m} \\
\end{array} \right| \]

\[
= S_{m-r} \prod_{l<k} (c_k - c_l).
\]

(3.34)

Hence,

\[
\lambda_p^{(0)} = 1 + \sum_{r=1}^{m} (-1)^{r}(r+1) S_{m-r} \\
= \sum_{r=0}^{m} (-1)^{r}(r+1) S_{m-r} \\
= \frac{S_m - 2S_{m-1} + 3S_{m-2} + \cdots + (-1)^{m-1}mS_1 + (-1)^{m}(m+1)}{S_m},
\]

which concludes the proof of Theorem 3.7.

Remark 3.8. Theorem 3.7 proves the conjecture asserted by Danciu [7] for first-order integrodifferential equations \((p = 1, d = 2)\).

As an application to Theorem 3.7, we can prove the following results. The proofs are quite similar to [7] for the first-order Volterra integrodifferential equation.

**Corollary 3.9.** An \((m,3)\)-method is stable if and only if

\[
\left| \frac{[d/dt (t \cdot R_m(t))]_{t=1}}{R_m(0)} \right| \leq 1.
\]

(3.36)

where \(R_m(t)\) is the polynomial of degree \(m\) whose zeroes are the collocation parameters \(\{c_j\}_{j=1,...,m}\).

Regarding the stability of local superconvergent solution \(u \in S_{m+4}^{(3)}(Z_n)\), we have the following corollary.

**Corollary 3.10.** (i) If the collocation parameters \(\{c_j\}_{j=1,...,m}\) are uniformly distributed in \((0,1)\) (i.e., \(c_j = j/m\), for all \(j = 1,2,\ldots,m\)), then \((m,3)\)-method is stable for \(m \geq 1\).

(ii) If the collocation parameters \(\{c_j\}_{j=1,...,m}\) are the Radau II points in the interval \((0,1)\), then \((m,3)\)-method is unstable for \(m \geq 2\).

(iii) If the collocation parameters \(\{c_j\}_{j=1,...,m}\) are the Gauss points in the interval \((0,1)\), then \((m,3)\)-method is unstable for \(m \geq 2\).

(iv) If the first \(m - 1\) collocation parameters \(\{c_j\}_{j=1,...,m}\) are the Gauss points in the interval \((0,1)\) and the last collocation parameter is \(c_m = 1\), then \((m,3)\)-method is stable for \(m \geq 2\).
4. Stability of $S_m^{(0)}(Z_n)$

In this section, we will investigate the stability when $d = 0$.

From (2.5), the restriction of $u \in S_m^{(0)}(Z_n)$ to the subinterval $\sigma_n$ is given by

$$u(t) = u_n(t_n + \tau h) = u_{n-1}(t_n) + \sum_{r=1}^{m} \beta_{n,r} \tau^r, \quad \text{for } \tau \in (0, 1], \ n = 0, 1, \ldots, N - 1. \quad (4.1)$$

If we denote by $u_{n+1}$ and by $u''_{n+1}$ the vectors with $m$ elements

$$u_{n+1} := (u_n(t_n + c_j h))^T_{j=1,\ldots,m}, \quad u''_{n+1} := (u''_n(t_n + c_j h))^T_{j=1,\ldots,m}, \quad (4.2)$$

then from (4.1), we obtain

$$u_{n+1} = (1, 1, \ldots, 1)^T u_{n-1}(t_n) + E \beta_n, \quad \text{for } n = 0, 1, \ldots, N - 1, \quad (4.3)$$

$$u''_{n+1} = h^{-2} E'' \beta_n, \quad \text{for } n = 0, 1, \ldots, N - 1. \quad (4.4)$$

Here the matrices $E$ and $E''$ are $m \times m$ matrices defined by

$$E := (c_j^r)_{j,r=1,\ldots,m} \quad \text{and} \quad E' := (r(r-1)c_j^{r-2})_{j,r=1,\ldots,m}.$$ 

In this case, the collocation equation becomes

$$V \beta_n = h^2 W'(u_{n-1}(t_n), u'_{n-1}(t_n), u''_{n-1}(t_n))^T + h^2 R_n, \quad (4.5)$$

for $n = 0, 1, \ldots, N - 1$, where $W'$ is the $m \times 3$ matrix whose elements are

$$(W')_{j,r} := \begin{cases} 
\nu h c_j & \text{if } r = 1, \\
-\alpha_1 & \text{if } r = 2, \\
1 & \text{if } r = 3,
\end{cases} \quad (4.6)$$

and the matrix $V$ and the vector $R_n$ are defined in (2.6) when $d = 0$.

Since $V = E'' + O(h)$, the elimination of $\beta_n$ between (4.4) and (4.5) yields

$$u''_n(t_n, j) = (1 + O(h)) u''_{n-1}(t_n) + (1 + O(h)) u'_n(t_n)$$

$$+ O(h) u_{n-1}(t_n) + (1 + O(h)) R_{n,j}, \quad \text{for } j = 1, 2, \ldots, m \quad (n = 0, 1, \ldots, N - 1). \quad (4.7)$$

For $\tau \in [0, 1]$, the second derivatives of the approximation $u \in S_m^{(0)}(Z_n)$ may be written in the form

$$u''_n(t + \tau h) = \sum_{j=1}^{m} L_j(\tau) u''_n(t_n, j), \quad \text{for } n = 0, 1, \ldots, N - 1, \quad (4.8)$$

where

$$L_j(\tau) := \prod_{j=1, i \neq j}^{m} \frac{(\tau - c_i)}{(c_j - c_i)}, \quad \text{for } j = 0, 1, \ldots, m, \quad (4.9)$$
are the Lagrange fundamental polynomials associated with the collocation parameters \( \{c_j\}_{j=1,m} \). Now, replacing \( u''(t_{n,j}) \) in (4.8) with its values given by (4.7), for \( n = 0,1,\ldots, N-1 \), we obtain

\[
\begin{align*}
    u''_n(t_{n+1}) &= (1 + O(h)) \left( u''_{n-1}(t_n) + u'_n(t_n) + \sum_{j=1}^{m} L_j(1)R_{n,j} \right) \\
    &
\end{align*}
\]

(4.10)

By integrating relation (4.8), for \( \tau \in [0,1] \), and using relation (4.7), we obtain

\[
\begin{align*}
    u'_n(t_{n+1}) &= h(1 + O(h)) u''_{n-1}(t_n) + (1 + h(1 + O(h))) u'_n(t_n) \\
    &+ hO(h) u_{n-1}(t_n) + h(1 + O(h)) \int_{0}^{1} \left( \sum_{j=1}^{m} L_j(\tau)R_{n,j} \right) d\tau,
\end{align*}
\]

(4.11)

for \( j = 1,2,\ldots,m \) \( (n = 0,1,\ldots,N-1) \).

Integrating (4.8) one more time and using relation (4.7) yields

\[
\begin{align*}
    u_n(t_{n+1}) &= h^2 (1 + O(h)) (u''_{n-1}(t_n) + u'_n(t_n)) + (1 + h^2 O(h)) u_{n-1}(t_n) \\
    &+ h^2 (1 + O(h)) \int_{0}^{1} \int_{0}^{s} \left( \sum_{j=1}^{m} L_j(\tau)R_{n,j} \right) d\tau ds,
\end{align*}
\]

(4.12)

for \( j = 1,2,\ldots,m \) \( (n = 0,1,\ldots,N-1) \).

Equations (4.7), (4.11), and (4.12) together form a system which may be written in the form

\[
\begin{pmatrix}
    u_n(t_{n+1}) \\
    u'_n(t_{n+1}) \\
    u''_n(t_{n+1})
\end{pmatrix}
= M' \begin{pmatrix}
    u_{n-1}(t_n) \\
    u'_{n-1}(t_n) \\
    u''_{n-1}(t_n)
\end{pmatrix}
+ (1 + O(h)) R'_n \quad \text{for } n = 0,1,\ldots,N-1,
\]

(4.13)

where

\[
M' := \begin{pmatrix}
    (1 + h^2 O(h)) & h^2 (1 + O(h)) & h^2 (1 + O(h)) \\
    hO(h) & (1 + h(1 + O(h))) & h(1 + O(h)) \\
    O(h) & (1 + O(h)) & (1 + O(h))
\end{pmatrix},
\]

\[
R'_n := \begin{pmatrix}
    h^2 \int_{0}^{1} \int_{0}^{s} \left( \sum_{j=1}^{m} L_j(\tau)R_{n,j} \right) d\tau ds \\
    h \int_{0}^{1} \left( \sum_{j=1}^{m} L_j(\tau)R_{n,j} \right) d\tau \\
    \sum_{j=1}^{m} L_j(1)R_{n,j}
\end{pmatrix}.
\]

Equation (4.13) has the same form as (2.11). Since \( h = 0 \) implies that the matrix \( M' \) has eigenvalues \( \lambda'_1 = \lambda'_2 = \lambda'_3 = 1 \), we can prove the following theorem.
Theorem 4.1. For every choice of the collocation parameters \( \{c_j\}_{j=1}^{m} \), an \((m,0)\)-method is stable for all \( m \geq 1 \).

5. Stability of \( S_{m+1}^{(1)}(Z_n) \)

In this section, we will investigate the stability when \( d = 1 \).

From (2.5), the restriction of \( u \in S_{m+1}^{(1)}(Z_n) \) to the subinterval \( \sigma_n \) is given by

\[
u(t) = u_n(t + \tau h) = u_{n-1}(t_n) + u_n'(t_n)\tau + \sum_{r=1}^{m} \beta_{n,r} \tau^{r+1},
\]

for \( \tau \in (0,1] \), \( n = 0,1,\ldots,N-1 \).

In this case, the collocation equation becomes

\[
V\beta_n = h^2 W''(u_{n-1}(t_n),u_n'(t_n),u_n''(t_n))^T + h^2 R_n,
\]

for \( n = 0,1,\ldots,N-1 \), where \( W'' \) is the \( m \times 3 \) matrix whose elements are

\[
(W'')_{j,r} := \begin{cases} 
\nu h c_j & \text{if } r = 1, \\
c_j h \left( \alpha_0 + \frac{\nu h c_j}{2} \right) & \text{if } r = 2, \\
1 & \text{if } r = 3,
\end{cases}
\]

and the matrix \( V \) and the vector \( R_n \) are defined in (2.6) when \( d = 1 \).

Using the same procedure as in Section 4, we can derive the system

\[
\begin{pmatrix}
u_n(t_{n+1}) \\
u_n'(t_{n+1}) \\
u_n''(t_{n+1})
\end{pmatrix} = M''
\begin{pmatrix}
u_{n-1}(t_n) \\
u_{n-1}'(t_n) \\
u_{n-1}''(t_n)
\end{pmatrix} + (1 + O(h)) R''_n,
\]

for \( n = 0,1,\ldots,N-1 \), (5.4)

where

\[
M'' := \begin{pmatrix}
1 + h^2 O(h) & h^2 O(h) & h^2 (1 + O(h)) \\
h O(h) & 1 + h O(h) & h (1 + O(h)) \\
o h & O(h) & 1 + O(h)
\end{pmatrix},
\]

\[
R''_n := \begin{pmatrix}
h \int_0^1 \left( \sum_{j=1}^{m} L_j(\tau) R_{n,j} \right) d\tau ds \\
\int_0^1 \left( \sum_{j=1}^{m} L_j(\tau) R_{n,j} \right) d\tau \\
\sum_{j=1}^{m} L_j(1) R_{n,j}
\end{pmatrix}.
\]

Equation (5.4) has the same form as (2.11). Since \( h = 0 \) implies that the matrix \( M'' \) has eigenvalues \( \lambda_1' = \lambda_2' = \lambda_3' = 1 \), we can prove the following theorem.
Theorem 5.1. For every choice of the collocation parameters \( \{ c_j \}_{j=1,...,m} \), an \((m,1)\)-method is stable for all \( m \geq 1 \).

6. Numerical examples

The \((3,d)\)-method is tested using the following three examples in the interval \([0,1]\) with step size \( h = 0.05 \). The following notations will be used in the presentation.

\[ e_1 := |y(t_1) - u(t_1)|, \quad e_{N/2} := |y(0.5) - u(0.5)|, \quad e_N := |y(1) - u(1)|, \quad (6.1) \]

where \( u \in S_{3+d}(Z_n) \) is the approximate solution.

Example 6.1. Consider the following integrodifferential equation of second order:

\[ y''(t) = 1 + \frac{1}{2} y(t) + \frac{1}{2} \int_0^t y(s) \, ds, \quad y(0) = 2, \quad y'(0) = 2, \quad (6.2) \]

where \( y(t) = 2e^t \) is the exact solution.

Example 6.2. Consider the following integrodifferential equation of second order:

\[ y''(t) = q(t) - \frac{t^2}{16} y'(t) + \int_0^t t^2 y'(s) \, ds, \quad y(0) = 1, \quad y'(0) = 4, \quad (6.3) \]

where \( q(t) \) is chosen so that \( y(t) = \sin 4t \) is the exact solution.

Example 6.3. Consider the following integrodifferential equation of second order:

\[ y''(t) = q(t) + p_1(t) y(t) + p_2(t) y'(t) + \int_0^t y(s) \, ds \\
+ \int_0^t t s^2 y'(s) \, ds, \quad y(0) = 2, \quad y'(0) = 0, \quad (6.4) \]

with

\[ p_1(t) = -t^2 + 2t - 1, \quad p_2(t) = 1 - 2t^2, \quad (6.5) \]

where \( q(t) \) is chosen so that \( y(t) = 1 + \cos t \) is the exact solution.

(a) If the collocation parameters are uniformly distributed, that is, \( c_1 = 1/3, c_2 = 2/3, \) and \( c_3 = 1 \), then we have Tables 6.1, 6.2, and 6.3 corresponding to Examples 6.1, 6.2, and 6.3, respectively.

(b) If the collocation parameters are the Radau II points, that is, \( c_1 = (4 - \sqrt{6})/10, \ c_2 = (4 + \sqrt{6})/10, \) and \( c_3 = 1 \), then we have Tables 6.4, 6.5, and 6.6 corresponding to Examples 6.1, 6.2, and 6.3, respectively.

(c) If the collocation parameters are the Gauss points, that is, \( c_1 = (5 - \sqrt{15})/10, c_2 = 1/2, \) and \( c_3 = (5 + \sqrt{15})/10 \), then we have Tables 6.7, 6.8, and 6.9 corresponding to Examples 6.1, 6.2, and 6.3, respectively.

(d) If the first two collocation parameters are the Gauss points, that is, \( c_1 = (3 - \sqrt{3})/6, \ c_2 = (3 + \sqrt{3})/6, \) and \( c_3 = 1 \), then we have Tables 6.10, 6.11, and 6.12 corresponding to Examples 6.1, 6.2, and 6.3, respectively.
From these numerical examples, we observe that a \((3, d)\)-method is stable for \(d = 2\) and it is unstable for \(d = 4\). In the case \(d = 3\), this method is stable if the collocation parameters are uniformly distributed (i.e., \(c_j := j/3\), for \(j = 1, 2, 3\)) as in case a, or \(c_1 = (3 - \sqrt{3})/6\),
Table 6.7. Approximate error for Example 6.1.

<table>
<thead>
<tr>
<th>$d$</th>
<th>$e_1$</th>
<th>$e_{N/2}$</th>
<th>$e_N$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>$1.55 \times 10^{-6}$</td>
<td>$2.55 \times 10^{-3}$</td>
<td>$2.46 \times 10^{-2}$</td>
</tr>
<tr>
<td>3</td>
<td>$9.30 \times 10^{-6}$</td>
<td>$7.84 \times 10^{-6}$</td>
<td>$1.80 \times 10^{-5}$</td>
</tr>
<tr>
<td>4</td>
<td>$1.01 \times 10^{-4}$</td>
<td>$5.42 \times 10^{-2}$</td>
<td>$5.03 \times 10^{-11}$</td>
</tr>
</tbody>
</table>

Table 6.8. Approximate error for Example 6.2.

<table>
<thead>
<tr>
<th>$d$</th>
<th>$e_1$</th>
<th>$e_{N/2}$</th>
<th>$e_N$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>$1.6 \times 10^{-9}$</td>
<td>$3.59 \times 10^{-4}$</td>
<td>$1.00 \times 10^{-4}$</td>
</tr>
<tr>
<td>3</td>
<td>$1.10 \times 10^{-9}$</td>
<td>$4.19 \times 10^{-2}$</td>
<td>$3.78 \times 10^{-5}$</td>
</tr>
<tr>
<td>4</td>
<td>$9.40 \times 10^{-9}$</td>
<td>$4.80 \times 10^{-8}$</td>
<td>$4.47 \times 10^{-12}$</td>
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</table>

Table 6.9. Approximate error for Example 6.3.

<table>
<thead>
<tr>
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</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>0</td>
<td>$4.83 \times 10^{-3}$</td>
<td>$3.65 \times 10^{-2}$</td>
</tr>
<tr>
<td>3</td>
<td>0</td>
<td>$3.40 \times 10^{-3}$</td>
<td>$1.44 \times 10^{-5}$</td>
</tr>
<tr>
<td>4</td>
<td>$2.00 \times 10^{-4}$</td>
<td>$9.96 \times 10^{-2}$</td>
<td>$9.40 \times 10^{-12}$</td>
</tr>
</tbody>
</table>

Table 6.10. Approximate error for Example 6.1.

<table>
<thead>
<tr>
<th>$d$</th>
<th>$e_1$</th>
<th>$e_{N/2}$</th>
<th>$e_N$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>$1.00 \times 10^{-9}$</td>
<td>$4.00 \times 10^{-9}$</td>
<td>$1.20 \times 10^{-8}$</td>
</tr>
<tr>
<td>3</td>
<td>0</td>
<td>0</td>
<td>$7.00 \times 10^{-9}$</td>
</tr>
<tr>
<td>4</td>
<td>0</td>
<td>$4.45 \times 10^{13}$</td>
<td>$5.12 \times 10^{43}$</td>
</tr>
</tbody>
</table>

Table 6.11. Approximate error for Example 6.2.

<table>
<thead>
<tr>
<th>$d$</th>
<th>$e_1$</th>
<th>$e_{N/2}$</th>
<th>$e_N$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>$1.10 \times 10^{-9}$</td>
<td>$3.32 \times 10^{-4}$</td>
<td>$2.11 \times 10^{-3}$</td>
</tr>
<tr>
<td>3</td>
<td>$6.00 \times 10^{-10}$</td>
<td>$3.32 \times 10^{-4}$</td>
<td>$2.11 \times 10^{-3}$</td>
</tr>
<tr>
<td>4</td>
<td>$3.00 \times 10^{-10}$</td>
<td>$5.43 \times 10^{16}$</td>
<td>$6.26 \times 10^{16}$</td>
</tr>
</tbody>
</table>

Table 6.12. Approximate error for Example 6.3.

<table>
<thead>
<tr>
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<th>$e_1$</th>
<th>$e_{N/2}$</th>
<th>$e_N$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>0</td>
<td>$8.36 \times 10^{-6}$</td>
<td>$1.04 \times 10^{-3}$</td>
</tr>
<tr>
<td>3</td>
<td>$1.00 \times 10^{-10}$</td>
<td>$8.36 \times 10^{-6}$</td>
<td>$1.04 \times 10^{-3}$</td>
</tr>
<tr>
<td>4</td>
<td>$1.00 \times 10^{-10}$</td>
<td>$3.06 \times 10^{15}$</td>
<td>$3.54 \times 10^{45}$</td>
</tr>
</tbody>
</table>

$c_2 = (3 + \sqrt{3})/6$, and $c_3 = 1$ as in case (d). These examples illustrate the conclusions of Theorem 5.1.
1066 Stability of the collocation methods

References


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