A New Error Analysis for a Cubic Spline Approximate Solution of a Class of Volterra Integro-Differential Equations

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Abstract. In this paper a third-order numerical method is considered which utilizes a twice continuously differentiable third degree spline to approximate the solution of

$$\dot{x}(t) = F\left(t, x(t), \int_a^t K(t, u, x(u)) du\right),$$

$$x(a) = x_0,$$

at discrete points in the interval [a, b]. The error analysis uses a technique usually associated with linear multistep methods.

I. Introduction. In this paper, consideration is directed to the Volterra integrodifferential equation

(1)
$$\dot{x}(t) = F\left(t, x(t), \int_a^t K(t, u, x(u)) du\right), \quad a \leq t \leq b,$$

with the initial condition $x(a) = x_0$. A third-order numerical method is considered which utilizes a twice continuously differentiable third degree spline to approximate the solution x at discrete points in the interval [a, b].

Other authors, e.g. Hung [5], have applied cubic splines to obtain an approximate solution of a scalar Volterra integro-differential equation. This paper considers the method as applied to vector equations. More important, however, is the error analysis presented herein. This analysis uses a lemma usually associated with linear multistep methods. The utilization of this lemma allows the cubic spline method to be applied to a larger class of equations than considered by Hung with, however, a corresponding reduction in the order of the errors. In particular, Hung requires the solution of (1) be of class $C^6[a, b]$ while the analysis presented here requires only $C^4[a, b]$. Accordingly, Hung achieves a discretization error $O(h^4)$ while this analysis achieves $O(h^3)$.

II. Notation and Assumptions. Let R(F) and R(K) be the regions defined by

$$R(F) = \{(t, x, y) : a \le t \le b; x, y \in E^n\}$$

and

$$R(K) = \{(t, u, v) : a \le u \le t \le b; v \in E^n\},$$

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where E^n is real Euclidean *n*-space. Moreover, let the *n*th order matrices $F^{(2)}$, $F^{(3)}$, and $K^{(3)}$ be such that the respective elements are given by

(2)
$$F_{i,j}^{(2)} = \frac{\partial F_i}{\partial x_i}, \qquad F_{i,j}^{(3)} = \frac{\partial F_i}{\partial y_j}, \quad \text{and} \quad K_{i,j}^{(3)} = \frac{\partial K_i}{\partial y_j}$$

Then, the following assumptions are made:

- (a) Equation (1) has a unique solution.
- (b) F and K are continuous mappings of R(F) and R(K) to E^n , respectively.
- (c) The matrix elements (2) are continuous and bounded.

Assumption (c) has two important implications. First, there exist constants $\bar{F}^{(2)}$, $\bar{F}^{(3)}$ and $\bar{K}^{(3)}$ such that $||F^{(2)}|| \leq \bar{F}^{(2)}$, $||F^{(3)}|| \leq \bar{F}^{(3)}$ and $||K^{(3)}|| \leq \bar{K}^{(3)}$, where $||\cdot||$ will be used interchangeably to denote compatible matrix and vector norms. Secondly, Buck [1, p. 268], for (t, x, y), $(t, \bar{x}, y) \in R(F)$ there exist $p_i \in E^n$, $i = 1, \dots, n$, such that

$$F(t, x, y) - F(t, \bar{x}, y) = F^{(2)}(x - \bar{x})$$

where $F^{(2)} = (F_{i,i}^{(2)}(t, p_i, y))$. Similarly, for (t, x, y), $(t, x, \bar{y}) \in R(F)$ and for (t, u, v), $(t, u, \bar{v}) \in R(K)$, there are $q_i, r_i \in E^n$, $i = 1, \dots, n$, such that

$$F(t, x, y) - F(t, x, \bar{y}) = F^{(3)}(y - \bar{y})$$

and

$$K(t, u, v) - K(t, u, \bar{v}) = K^{(3)}(v - \bar{v})$$

where $F^{(3)} = (F_{i,j}^{(3)}(t, x, q_i))$ and $K^{(3)} = (K_{i,j}^{(3)}(t, u, r_i))$.

III. The Method. Let [a, b] be divided into N equal subintervals of length h = (b - a)/N with endpoints t_0, t_1, \dots, t_N , called nodes. Let x_k, \dot{x}_k and \ddot{x}_k denote approximations for $x(t_k)$, $\dot{x}(t_k)$ and $\ddot{x}(t_k)$, respectively. The n-dimensional cubic spline S on [a, b] is defined as follows: For $t \in [t_k, t_{k+1}]$, S is denoted by S_k and is defined by

(3)
$$S_k(t) = x_k + (t - t_k)\dot{x}_k + \frac{(t - t_k)^2}{2}\ddot{x}_k + \frac{(t - t_k)^3}{3h^2}(\dot{x}_{k+1} - \dot{x}_k - h\ddot{x}_k).$$

Note that $S_k(t_k) = x_k$, $\dot{S}_k(t_k) = \dot{x}_k$, $\ddot{S}_k(t_k) = \ddot{x}_k$ and $\dot{S}_k(t_{k+1}) = \dot{x}_{k+1}$.

The approximate solution to (1) is obtained by replacing the integral by a numerical quadrature formula and requiring that the resulting equation be satisfied at the nodes. Thus, if the cubic spline S replaces x in this equation, (1) is replaced by

(4)
$$\dot{S}_k(t_{k+1}) = F\left(t_{k+1}, S_k(t_{k+1}), h \sum_{i=0}^{k+1} w_i K(t_{k+1}, t_i, S_{i-1}(t_i))\right)$$

where the weights w_i are bounded and depend on the numerical quadrature formula used and where $S_{-1}(t_0) \equiv x_0$. Then, using $x_k = S_{k-1}(t_k)$, $\dot{x}_k = \dot{S}_{k-1}(t_k)$, $\ddot{x}_k = \ddot{S}_{k-1}(t_k)$ and $\dot{x}_{k+1} = \dot{S}_k(t_{k+1})$, (4) becomes

$$\dot{x}_{k+1} = H(\dot{x}_{k+1})$$

where

$$H(\dot{x}_{k+1}) = F\left(t_{k+1}, x_k + h\dot{x}_k + \frac{h^2}{2}\ddot{x}_k + \frac{h}{3}(\dot{x}_{k+1} - \dot{x}_k - h\ddot{x}_k), q_{k+1}\right)$$

with

$$q_{k+1} = h \sum_{i=0}^{k} w_i K(t_{k+1}, t_i, x_i)$$

$$+ h w_{k+1} K \left(t_{k+1}, t_{k+1}, x_k + h \dot{x}_k + \frac{h^2}{2} \ddot{x}_k + \frac{h}{3} (\dot{x}_{k+1} - \dot{x}_k - h \ddot{x}_k) \right)$$

All quantities in (5) are known except \dot{x}_{k+1} . Since \dot{x}_{k+1} determines S_k , the values $x_{k+1} = S_k(t_{k+1})$ and $\ddot{x}_{k+1} = \ddot{S}_k(t_{k+1})$ follow. (Although (5) is used to determine x_k , it is convenient to use (4) in the error analysis to follow.)

It follows, in the usual straightforward manner, from assumption (c) that, for x, $\bar{x} \in E^n$,

$$||H(x) - H(\bar{x})|| \le \frac{h\bar{F}^{(2)} + h^2 |w_{k+1}| \bar{F}^{(3)} \bar{K}^{(3)}}{3} ||x - \bar{x}||.$$

Thus, for h sufficiently small the mapping given by (5) is a contraction. This proves the following theorem.

THEOREM 1. For H as defined by (5) and with assumption (c) satisfied, it follows that, for sufficiently small h, H is a contraction mapping.

Thus, (5) can be used iteratively to determine x_i , i = r, \cdots , N, where r depends on the starting method used.

IV. Error Analysis. Let E(t) = x(t) - S(t). Then, from (1) and (4), there follows

$$\dot{E}(t_k) = F\left(t_k, x(t_k), \int_{t_0}^{t_k} K(t_k, u, x(u)) du\right) - F\left(t_k, S_{k-1}(t_k), \int_{t_0}^{t_k} K(t_k, u, x(u)) du\right) \\
+ F\left(t_k, S_{k-1}(t_k), \int_{t_0}^{t_k} K(t_k, u, x(u)) du\right) \\
- F\left(t_k, S_{k-1}(t_k), h \sum_{i=0}^{k} w_i K(t_k, t_i, x(t_i))\right) \\
+ F\left(t_k, S_{k-1}(t_k), h \sum_{i=0}^{k} w_i K(t_k, t_i, x(t_i))\right) \\
- F\left(t_k, S_{k-1}(t_k), h \sum_{i=0}^{k} w_i K(t_k, t_i, S_{i-1}(t_i))\right).$$

Thus, in view of assumption (c),

(6)
$$\dot{E}(t_k) = F_{(k)}^{(2)} E(t_k) + \tilde{O}(h^p) + h F_{(k)}^{(3)} \sum_{i=0}^k w_i K_{(k,i)}^{(3)} E(t_i)$$

where $F_{(k)}^{(2)}$ and $F_{(k)}^{(3)}$ indicate the matrices $F^{(2)}$ and $F^{(3)}$ depend on the index k and $K_{(k,i)}^{(3)}$ indicates the matrix depends on the indices k and i. Furthermore, the numerical quadrature formula is assumed to be such that

$$\int_{t_0}^{t_k} K(t_k, u, x(u)) du - h \sum_{i=0}^k w_i K(t_k, t_i, x(t_i)) = \tilde{O}(h^p)$$

where $\tilde{O}(h^{\nu})$ is a vector with components all $O(h^{\nu})$.

The error analysis development is to obtain an equation involving E and \dot{E} at the nodes. Then, (6) is used to provide an equation in E only. The error information at the nodes is then used to obtain error bounds at nonnodal points.

To proceed with the error analysis at the nodes, it is assumed the solution $x \in C^{(4)}[a, b]$. Then, for $t \in [t_k, t_{k+1}]$,

(7)
$$E(t) = E(t_k) + (t - t_k)\dot{E}(t_k) + \frac{(t - t_k)^2}{2} \ddot{E}(t_k) + \frac{(t - t_k)^3}{6} \ddot{E}(t_k) + \tilde{O}(h^4),$$

(8)
$$\dot{E}(t) = \dot{E}(t_k) + (t - t_k) \ddot{E}(t_k) + \frac{(t - t_k)^2}{2} \ddot{E}(t_k) + \tilde{O}(h^3),$$

and

(Since $S^{(4)} = 0$, the error terms involve only the solution x and not the spline S.) Evaluation of (7) and (8) at t_{k+1} and elimination of $E(t_k)$ provides

(10)
$$E(t_{k+1}) = E(t_k) + \frac{2h}{3} \dot{E}(t_k) + \frac{h}{3} \dot{E}(t_{k+1}) + \frac{h^2}{6} \ddot{E}(t_k) + \tilde{O}(h^4)$$

while elimination of $\ddot{E}(t_k)$ provides

(11)
$$\frac{h^3}{12} \overset{\cdots}{E}(t_k) = E(t_k) - E(t_{k+1}) + \frac{h}{2} \dot{E}(t_k) + \frac{h}{2} \dot{E}(t_{k+1}) + \tilde{O}(h^4).$$

Evaluation of (9) at t_{k+1} and substitution of $E(t_k)$ from (11) yields

(12)
$$\ddot{E}(t_{k+1}) = \ddot{E}(t_k) + \frac{12}{h^2} E(t_k) - \frac{12}{h^2} E(t_{k-1}) + \frac{6}{h} \dot{E}(t_k) + \frac{6}{h} \dot{E}(t_{k+1}) + \tilde{O}(h^2).$$

Reduction of subscripts by one in (12) and substitution of the resulting expression for $E(t_{k-1})$ into the equation which results from the reduction of the subscripts by one in (10) yields

$$\frac{h^2}{6} : E(t_k) = -E(t_k) + E(t_{k-1}) + \frac{2h}{3} \dot{E}(t_k) + \frac{h}{3} \dot{E}(t_{k-1}) + \tilde{O}(h^4),$$

which when substituted in (10) provides

(13)
$$E(t_{k+1}) - E(t_{k-1}) + \frac{h}{3} \left[\dot{E}(t_{k-1}) + 4\dot{E}(t_k) + \dot{E}(t_{k+1}) \right] + \tilde{O}(h^4).$$

Finally, from (6) and (13), there follows

$$E(t_{k+1}) - E(t_{k-1}) = \frac{h}{3} \left[F_{(k-1)}^{(2)} E(t_{k-1}) + 4 F_{(k)}^{(2)} E(t_k) + F_{(k+1)}^{(2)} E(t_{k+1}) \right]$$

$$+ \frac{h^2}{3} \left[F_{(k-1)}^{(3)} \sum_{i=0}^{k-1} w_i K_{(k-1,i)}^{(3)} E(t_i) + 4 F_{(k)}^{(3)} \sum_{i=0}^{k} w_i K_{(k,i)}^{(3)} E(t_i) \right]$$

$$+ F_{(k+1)}^{(3)} \sum_{i=0}^{k+1} w_i K_{(k+1,i)}^{(3)} E(t_i)$$

$$+ \tilde{O}(h^{\min(p+1,4)}).$$

In order to bound the discretization errors at the nodes, the following lemma is used, the proof of which is similar to that for Lemma 5.6 (Henrici, [3, p. 243]) and Linz's [7] lemma.

LEMMA. Let $z_m \in E^n$, $m \ge r$, be the solution of

$$\rho_k z_{m+k} + \cdots + \rho_0 z_m = h(\beta_{k,m} z_{m+k} + \beta_{k-1,m+k-1} z_{m+k-1} + \cdots + \beta_{0,m} z_m)$$

$$+ h^2 \left(\sum_{i=0}^{m+k} \mu_{m+k,i} z_i + \sum_{i=0}^{m+k-1} \mu_{m+k-1,i} z_i + \cdots + \sum_{i=0}^{m} \mu_{m,i} z_i \right) + \lambda_m$$

where all $\beta_{i,j}$ and $\mu_{i,j}$ are nth order matrices and the ρ_i are scalars. Assume the polynomial $\rho_k \xi^k + \rho_{k-1} \xi^{k-1} + \cdots + \rho_0$ satisfies the Dahlquist stability condition (Henrici [3, p. 218]). Thus, if

$$1/(\rho_k + \rho_{k-1}\xi + \cdots + \rho_0\xi^k) \equiv \gamma_0 + \gamma_1\xi + \cdots,$$

where $\rho_k \neq 0$, then $\Gamma \equiv \sup_i |\gamma_i| < \infty$ (Henrici [3, p. 242]). Furthermore, assume $||z_i|| \leq Z$, $i = 0, 1, \dots, k + r - 1$, and for all $i, j, ||\beta_{i,j}|| \leq \beta$, $||\mu_{i,j}|| \leq \mu$, $||\lambda_i|| \leq \lambda$. Then, for sufficiently small h,

$$||z_n|| \leq K^* e^{nhL^*}, \quad n = 0, 1, \dots, N,$$

where

$$K^* = [k\Gamma AZ + hb\mu^*\Gamma rZ + N\lambda\Gamma]/[1 - h\Gamma(\beta + b\mu^*)],$$
 $L^* = [\beta^*\Gamma + b\mu^*\Gamma]/[1 - h\Gamma(\beta + b\mu^*)],$
 $A = |\rho_0| + \cdots + |\rho_n|,$
 $\beta^* = \beta(k+1)$ and $\mu^* = \mu(k+1).$

Application of the lemma to (14) yields

$$||E(t_k)|| \leq K^* e^{L^* t_k}$$

with

$$K^* = [4Z + 4hbw \bar{K}^{(3)} \bar{F}^{(3)} r Z + bO(h^{\min{(3,p)}})]/[1 - h(\frac{4}{3}\bar{F}^{(2)} + 4bw \bar{F}^{(3)} \bar{K}^{(3)})]$$
 and

$$L^* = [4\bar{F}^{(2)} + 4bw\bar{F}^{(3)}\bar{K}^{(3)}]/[1 - h(\frac{4}{3}\bar{F}^{(2)} + 4bw\bar{F}^{(3)}\bar{K}^{(3)})]$$

where $|w_i| \le w$, $i = 0, \dots, N$. Z and r depend on the starting method while p depends on the numerical integration method used. From K^* and L^* , it is readily seen that

minimally the starting method should be $\tilde{O}(h^3)$ and p = 3. For this case $||E(t_k)|| = O(h^3)$. There follows from (6) and (10), respectively, that $||\dot{E}(t_k)|| = O(h^3)$ and $||\dot{E}(t_k)|| = O(h)$. This proves the following theorem.

THEOREM 2. If assumptions (a), (b) and (c) are satisfied, $x \in C^{(4)}[a, b]$ and the starting method and numerical integration method are both $\tilde{O}(h^3)$, then $||E(t_k)|| = O(h^3)$, $||\dot{E}(t_k)|| = O(h^3)$ and $||\dot{E}(t_k)|| = O(h)$.

The error analysis at nonnodal points proceeds by setting $t = t_{k+1}$ in (7) and (8), solving the resulting equations for $\ddot{E}(t_k)$ and $\ddot{E}(t_k)$ and substituting these equations back into (7) and (8) to obtain

$$E(t) = \left[1 - \frac{3(t - t_k)^2}{h^2} + \frac{2(t - t_k)^3}{h^3}\right] E(t_k) + \left[\frac{3(t - t_k)^2}{h^2} - \frac{2(t - t_k)^3}{h^3}\right] E(t_{k+1})$$

$$+ \left[(t - t_k) - \frac{2(t - t_k)^2}{h} + \frac{(t - t_k)^3}{h^2}\right] \dot{E}(t_k)$$

$$+ \left[\frac{(t - t_k)^3}{h^2} - \frac{(t - t_k)^2}{h}\right] \dot{E}(t_{k+1}) + \tilde{O}(h^4)$$

and

$$\dot{E}(t) = \left[\frac{6(t-t_k)^2}{h^3} - \frac{6(t-t_k)}{h^2} \right] E(t_k) + \left[\frac{6(t-t_k)}{h^2} - \frac{6(t-t_k)^2}{h^3} \right] E(t_{k+1}) \\
+ \left[1 - \frac{4(t-t_k)}{h} + \frac{9(t-t_k)^2}{h^2} \right] \dot{E}(t_k) + \left[\frac{3(t-t_k)^2}{h^2} - \frac{2(t-t_k)}{h} \right] \dot{E}(t_{k+1}) \\
+ \tilde{O}(h^3).$$

Hence, for $t_k \leq t \leq t_{k+1}$,

$$||E(t)|| \leq ||E(t_k)|| + ||E(t_{k+1})|| + h||\dot{E}(t_k)|| + h||\dot{E}(t_{k+1})|| + O(h^4)$$

and

$$||\dot{E}(t)|| \leq 6 \left(\frac{1}{h} ||E(t_k)|| + \frac{1}{h} ||E(t_{k+1})|| + ||\dot{E}(t_k)|| + ||\dot{E}(t_{k+1})|| \right) + O(h^3).$$

Thus, $||E(t)|| = O(h^3)$ and $||\dot{E}(t)|| = O(h^2)$. From (9) and (11) there follows $||\dot{E}(t)|| = O(h)$. This proves the following theorem.

THEOREM 3. If the conditions of Theorem 2 are satisfied, then, for $t \in [a, b]$, t a nonnodal point, $||E(t)|| = O(h^3)$, $||\dot{E}(t)|| = O(h^2)$ and $||\ddot{E}(t)|| = O(h)$.

V. A Numerical Example. The scalar equation considered here is

$$\dot{x}(t) = -\frac{3}{16}(t-1)x(t) + \int_0^t x(u) du + \frac{13}{3}(t-1)^{10/3} + \frac{3}{16},$$

$$x(0) = -1,$$

which has the solution $x(t) = (t-1)^{13/3}$. Gregory's third-order formula,

$$\int_{t_0}^{t_k} f \doteq h \left(\frac{5}{12} f_0 + \frac{13}{12} f_1 + f_2 + \cdots + f_{k-2} + \frac{13}{12} f_{k-1} + \frac{5}{12} f_k \right),$$

is used for numerical quadrature. In order to apply Gregory's formula, the values x_0 , x_1 , x_2 and x_3 are needed. x_0 is known. To obtain the other starting values, let G(t, x(t)) represent the right side of the above equation, i.e., $\dot{x}(t) = G(t, x(t))$, and let $G_k = G(t_k, x(t_k))$. Then

$$x(t_{k+1}) = x(t_k) + \int_{t_k}^{t_{k+1}} G(t, x(t)) dt$$

and

$$x(t_{k+2}) = x(t_k) + \int_{t_k}^{t_{k+2}} G(t, x(t)) dt.$$

Thus, using Simpson's rule,

(15)
$$x_{k+1} = x_k + \frac{h}{6} (G_k + 4G_{k+1/2} + G_{k+1})$$

and

(16)
$$x_{k+2} = x_k + \frac{h}{3} (G_k + 4G_{k+1} + G_{k+2})$$

where $t_{k+1/2} = t_k + h/2$. The quadratic equation through the points (t_k, x_k) , (t_{k+1}, x_{k+1}) and (t_{k+2}, x_{k+2}) evaluated at $t_{k+1/2}$ provides

$$(17) x_{k+1/2} = \frac{3}{8}x_k + \frac{3}{4}x_{k+1} - \frac{1}{8}x_{k+2},$$

and, similarly,

(18)
$$G_{k+1/2} = \frac{3}{8}G_k + \frac{3}{4}G_{k+1} - \frac{1}{8}G_{k+2}.$$

TABLE OF ERRORS

Step Size 2 ^{-P} P	<i>E</i> (1.5)	$ \dot{E}(1.5) $	$\max E(t_i) , i = 1, 2, 3$ (for starting method)
2	$.865964 \times 10^{-3}$	$.170270 \times 10^{-2}$	$.636479 \times 10^{-2}$
3	$.915525 \times 10^{-3}$ (.108248 × 10^{-3})	$.101191 \times 10^{-2}$ (.212838 × 10^{-3})	$.423563 \times 10^{-3}$
4	$.227987 \times 10^{-3}$ (.114481 × 10^{-3})	$.218648 \times 10^{-3}$ (.126889 × 10^{-3})	$.274553 \times 10^{-4}$
5	$.380606 \times 10^{-4}$ ($.284984 \times 10^{-4}$)	$.352258 \times 10^{-4}$ (.273310 × 10^{-4})	174974×10^{-5}
6	$.545480 \times 10^{-5}$ (.475757 $\times 10^{-5}$)	$.498293 \times 10^{-5}$ (.440322 × 10^{-5})	$.110465 \times 10^{-6}$
7	$.728986 \times 10^{-6}$ ($.681850 \times 10^{-6}$)	$.662161 \times 10^{-6}$ ($.622616 \times 10^{-6}$)	$.693943 \times 10^{-8}$
8	$.941891 \times 10^{-7}$ (.911232 × 10^{-7})	$.853280 \times 10^{-7}$ ($.827701 \times 10^{-7}$)	.434832 × 10 ⁻⁹

Equations (15), (16), (17) and (18) are used iteratively to find x_1 and x_2 , then x_3 and x_4 . The value x_4 is not a needed starting value but it is obtained simultaneously with x_3 . The method is used until the difference between two consecutive iterates does not exceed 10^{-12} . This starting method is $O(h^4)$.

Once the starting values are known, the method of Section III is used with Gregory's formula for the numerical integration. Note the method is applied to \dot{x}_i , not x_i . Once \dot{x}_i is known, then

$$x_i = S_{i-1}(t_i)$$
 and $\ddot{x}_i = \ddot{S}_{i-1}(t_i)$.

The above table summarizes the errors corresponding to various step sizes. According to the theory, if the step size is halved the errors in E and \dot{E} should be reduced by approximately one-eighth. The numbers in parenthesis are one-eighth the error for the previous step size, i.e., the error predicted by the theory when the step size is halved.

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