Convergence of a Block-By-Block Method for Nonlinear Volterra Integro-Differential Equations

By Athena Makroglou*

Abstract. The theory of a block-by-block method for solving Volterra integral equations is extended to nonsingular Volterra integro-differential equations. Convergence is proved and a rate of convergence is found. The convergence results obtained are analogous to those obtained by Weiss [12] for Volterra integral equations. Several numerical examples are included.

1. Introduction. Consider the nonlinear Volterra integro-differential equation

\[ y'(x) = G\left(x, y(x), \int_0^x K(x, t, y(t)) dt\right) \quad (x \geq 0) \]

given \(y(0)\). Methods applied to the integro-differential equation (1.1), seen as a differential equation, have been discussed by Linz [7], Brunner and Lambert [3], Tavernini [11], and Neves [10]. It is relevant to observe that the initial value problem (1.1) can be written in the form

\[ y(x) = \int_0^x G(s, y(s), z(s)) ds + y(0) \quad (x \geq 0), \]

with

\[ z(x) = \int_0^x K(x, t, y(t)) dt \quad (x \geq 0), \]

in which the initial condition is incorporated. If we put \(F(x, t, \varphi) = (F_1(t), F_2(t))^T = (G(t, \varphi_1, \varphi_2), K(x, t, \varphi_1))^T\) with \(\varphi = (\varphi_1, \varphi_2)^T\), then (1.2), (1.3) can be written

\[ f(x) = \int_0^x F(x, s, f(s)) ds + c \quad (x \geq 0), \]

where \(f(x) = (y(x), z(x))^T\) and \(c = (y(0), 0)^T\). Methods applied to the integro-differential equation (1.1), seen as an integral (or a coupled pair of integral) equation(s), have been discussed by Day [5], and Mocarsky [9].

In this paper we too shall consider the equation (1.1) written in the form (1.2), (1.3) and we shall extend a method first applied by Weiss [12] to Volterra integral
equations of the form

\[(1.5) \quad f(x) = \int_0^x F(x, s, f(s)) \, ds + g(x) \quad (x > 0).\]

This method is an implicit block-by-block method. Such methods have the advantage over linear multistep methods and step-by-step methods that they can be of high order and still be self starting. In a block-by-block method we seek approximate values of the solution for \(0 \leq x_{m,j} \leq X\), where \(x_{m,j} = mh + uh, j = 0, 1, \ldots, p, 0 \leq u_0 < u_1 < \cdots < u_p, p\) integer, \(m = 0, 1, \ldots, N-1\), such that \(Nh = X\).

Weiss [12] has derived two schemes based on interpolatory quadrature rules such that

\[(1.6) \quad \int_0^u \varphi(x) \, dx \approx \sum_{k=0}^p w_k \varphi(u_k),\]

and

\[(1.7) \quad \int_0^1 \varphi(x) \, dx \approx \sum_{k=0}^p w_k \varphi(u_k),\]

where

\[(1.8) \quad w_k = \int_0^u L_k(x) \, dx,\]

\[(1.9) \quad w_k = w_k^p = \int_0^1 L_k(x) \, dx,\]

and

\[(1.10) \quad L_k(x) = \prod_{j=0, j \neq k}^p (x - u_j)/(u_k - u_j),\]

see Eqs. (2.11) and (2.12) in [12], that is

\[(1.11) \quad f_{m,j} = h \sum_{i=0}^{m-1} \sum_{k=0}^p w_k F(x_{m,j}, x_{i,k}, f_{i,k}) + h \sum_{k=0}^p w_k F(x_{m,j}, x_{m,k}, f_{m,k}),\]

and

\[(1.12) \quad f_{m,j} = h \sum_{i=0}^{m-1} \sum_{k=0}^p w_k F(x_{m,j}, x_{i,k}, f_{i,k}) + hu_j \sum_{k=0}^p w_k F(x_{m,j}, x_{m,j,k}, \sum_{r=0}^p L_r(u_ju_k)f_{m,r}).\]

Scheme (1.11) has the disadvantage over scheme (1.12) that it needs the evaluation of \(F(x, s, y(s))\) at points where \(s > x\), where \(F(x, s, y(s))\) might, for example, not be defined; see [12]. Weiss [12] indicated that the generalization of schemes (1.11), (1.12)
to a system of Volterra integral equations of the second kind follows immediately. Equation (1.1) can be formulated as a system of Volterra integral equations of the form (1.4). But, because of the special form of $F(t) = G(t, y(t), z(t))$, simplifications occur and thus we chose to treat each of the equations (1.2), (1.3) separately. We so produced three schemes A, B, and C, which we called block-by-block methods after Weiss, although they are new methods for integro-differential equations. We present here scheme C which in Makrogloou [8] was proved to be the best of all three, namely the simplest, the least computer-time consuming and on the whole equally as accurate as A and B. It also can be successfully used for Volterra integro-differential equations with weakly-singular kernels, as will be shown in a sequel to this; see also [8], scheme GC. The description of the scheme is given in Section 2 below. In Section 3 is given the convergence proof of scheme C. Some numerical results are included in Section 4. The results obtained by using schemes A, B, C in [8] were compared with those obtained by using linear multistep methods extended for the solution of (1.1) by Linz [7] and by Brunner and Lambert [3], hybrid methods applied to (1.1) by Makroglou [8] and some step-by-step methods discussed by Mocarsky [9]. We have not, however, undertaken a systematic assessment of the relative merits of the methods on a class of test problems. The results of the comparisons are stated in Section 4. For detailed results see [8].

Throughout the paper, $y_{m,j}$ shall denote an approximation to $y(x_{m,j})$, $m = 0, 1, \ldots, N - 1; j = 0, 1, \ldots, p$, and $z_{m,j}$ an approximation to $z(x_{m,j})$. Also $x_{m,v_0,\ldots,v_k}$ shall denote the point $mh + u_0 \cdot \cdots \cdot u_k h$, $0 \leq v_i \leq p$, $i = 0, 1, \ldots, k$, $v_i$ integer.

2. Description of the Method.

2.1. Scheme C. Consider Eqs. (1.2), (1.3). For $x = x_{m,j}$ we have

$$y(x_{m,j}) = \int_0^{x_m} G(s, y(s), z(s)) ds$$

$$+ \int_{x_m}^{x_{m,j}} G(s, y(s), z(s)) + y(0),$$

(2.1)

and

$$z(x_{m,j}) = \int_0^{x_m} K(x_{m,j}, t, y(t)) dt$$

$$+ \int_{x_m}^{x_{m,j}} K(x_{m,j}, t, y(t)) dt$$

(2.2)

$$m = 0, 1, \ldots, N - 1; j = 0, 1, \ldots, p.$$

Then using the quadrature rules (1.6), (1.8), (1.10) and (1.7), (1.9), (1.10) in the equations (2.1) and (2.2), respectively, we have
\[ y_{m,i} = y(0) + h \sum_{i=0}^{m-1} \sum_{k=0}^{p} w_k G(x_{i,k}, y_{i,k}, z_{i,k}) \]

(2.3)

\[ + h \sum_{k=0}^{p} w_k^j G(x_{m,k}, y_{m,k}, z_{m,k}), \]

\[ z_{m,i} = h \sum_{i=0}^{m-1} \sum_{k=0}^{p} w_k K(x_{m,i}, x_{i,k}, y_{i,k}) \]

(2.4)

\[ + hu_j \sum_{k=0}^{p} w_k K(x_{m,j}, x_{m,j,k}, \sum_{r=0}^{p} L_r(u_j \mu_k) y_{m,r}), \]

\[ m = 0, 1, \ldots, N - 1; j = 0, 1, \ldots, p, \]

where, to obtain an approximation of \( y(x_{m,j,k}) \), we have used Lagrangian interpolation, that is,

(2.5)

\[ y(x_{m,j,k}) \approx \sum_{r=0}^{p} L_r(u_j \mu_k) y_{m,r}. \]

We note that in (1.2) \( G(s, y(s), z(s)) \) does not depend on \( x \) and so the evaluation of \( G \) at points where \( s > x \) will not matter.

In the case when \( u_0 = 0 \) Eqs. (2.3) simplify to

(2.6)

\[ y_{m,i} = y_{m,0} + h \sum_{k=0}^{p} w_k^j G(x_{m,k}, y_{m,k}, z_{m,k}). \]

(We cannot make a similar simplification to Eq. (2.4), because there the sum from 0 to \( m - 1 \) is dependent on \( j \).)

When \( u_0 \neq 0 \) we can still simplify (2.3), by differencing on \( m \) thus

(2.7)

\[ y_{0,i} = s(0,i) + y(0), \]

\[ y_{m,i} - hs(m,i) = y_{m-1,i} + hs(m-1,i) \]

\[ + h \sum_{k=0}^{p} w_k G(x_{m-1,k}, y_{m-1,k}, z_{m-1,k}), \]

\[ m \geq 1; j = 0, 1, \ldots, p, \]

where we have put

\[ s(m,i) = \sum_{k=0}^{p} w_k G(x_{m,k}, y_{m,k}, z_{m,k}), \quad m = 0, 1, \ldots, N - 1. \]

At the \( m \)-stage in Eq. (2.7) \( y_{m-1,i}, z_{m-1,i} \), and also \( s(m-1,i) \) are known from the previous stage for \( j = 0, 1, \ldots, p \), and we have to compute only one term of the sum from 0 to \( m-1 \) on the right-hand side of (2.3), namely the one for \( i = m-1 \).
We may note that the scheme (Eqs. (2.3), (2.4)) is self starting and that for
\( p = 1, u_0 = 0, u_1 = 1 \), it reduces to the scheme obtained, if we apply the trapezium
quadrature rule to the integrals occurring.

3. Convergence. In this section we are concerned with a convergence proof for
scheme C (Eqs. (2.3), (2.4)).

We shall establish a bound on the error in approximations (2.3), (2.4), see
Theorem 1. This bound is achieved in terms of certain discretization errors (\( M_1(h),
T_1(h), T_2(h), T_3(h), T_4(h) \) in the analysis to follow) by using a lemma [6, p. 313],
given as Lemma 2 below. We shall then deduce convergence, see Corollary 1, and the
rate of convergence of the method by using Lemma 1 ([12], [1]), see Corollary 2.

Before proceeding to the proofs, we need the following preliminaries; see Weiss
[12]. In relation to the quadrature rules (1.6)—(1.10), we define

\[
(3.1) \quad g(x) = (x - u_0)(x - u_1) \cdots (x - u_p) = \sum_{j=0}^{p+1} c_j x^{p+1-j}, \quad c_0 = 1.
\]

We denote the relation \( \int_0^1 g(x) \, dx \neq 0 \) by \( g(x) \in P_0 \) and the relations \( \int_0^1 x^i g(x) \, dx = 0, i = 0, 1, \ldots, v - 1, \int_0^1 x^v g(x) \, dx \neq 0 \), by \( g(x) \in P_v \). Let

\[
(3.2) \quad E_j(\varphi) = \int_0^u \varphi(x) \, dx - \sum_{k=0}^p w_k^j \varphi(u_k), \quad j = 0, 1, \ldots, p.
\]

Then the following result is valid.

**Lemma 1** ([12], [1], [8]). If \( g(x) \in P_v \), then

\[
(3.3) \quad E_j(s^{p+1+r}) = \int_0^u x^r g(x) \, dx - \sum_{i=1}^r c_i E_j(s^{p+1+r-i}), \quad r = 0, 1, \ldots, p,
\]

and

\[
(3.4) \quad E_p(s^{p+1+r}) = 0 \quad \text{for } r < v - 1.
\]

From (3.4) we may conclude that the degree of precision of the formula (1.7)
is \( p + 1 + v - 1 = p + v \). The following lemma gives an estimation of the growth of
the solution of nonhomogeneous difference equations.

**Lemma 2** ([6, p. 313]). If \( |q_i| \leq A \sum_{i=0}^{n-1} |q_i| + B \) for \( n = s, s + 1, \ldots \)
with \( A > 0, B > 0 \) and \( \sum_{i=0}^{s-1} |q_i| < P \), then \( |q_i| \leq (B + AP)(1 + A)^{n-2}, n = s, s + 1, \ldots \).
Furthermore, if \( A = hk \) and \( nh = x \), then \( |q_i| \leq (B + hkP)\exp(kx) \).

We shall also make use of the following notation:

\[
(3.5) \quad \Gamma(x) \equiv G(x, y(x), z(x)), \quad K_1(x, s) \equiv K(x, s, y(s)), \quad 0 \leq s < x < X,
\]

\[
(3.6) \quad \Gamma^{(m)}(s) \equiv \frac{\partial^m}{\partial \eta^m} \Gamma(\eta)|_{\eta=s}, \quad K_1^{(m)}(x, s) \equiv \frac{\partial^m}{\partial \eta^m} K_1(x, \eta)|_{\eta=s}.
\]
We now define

\[ L = \sup_{0 < r, j, k < p} |L_r(u_j u_k)|, \quad \Lambda_p = \left\| \sum_{j=0}^{p} L_j(x) \right\|_\infty, \]

\[ W = \sup_{0 < k < p} |w_k|, \quad \overline{W}' = \sup_{0 < j, k < p} |w'_j|, \]

\[ \Delta_1 = \{(x, y, z) : 0 < x < X, |y| < \infty, |z| < \infty\}, \]

\[ \Delta_2 = \{(x, s, y) : 0 < s < x < X, |y| < \infty\}, \]

\[ M_1(h) = \max_{0 < m < N-1} \left| \int_{x_m}^{x_{m+1}} K_1(x_m, s) ds - h \sum_{i=0}^{m-1} \sum_{k=0}^{p} w_k K_1(x_m, x_i, x_{i+k}) \right|, \]

\[ M_2(h) = \max_{0 < m < N-1} \left| \int_{x_m}^{x_{m+1}} \Gamma(x) dx - h \sum_{i=0}^{m-1} \sum_{k=0}^{p} w_k \Gamma(x_i, x_{i+k}) \right|, \]

\[ (3.7) \quad T_1(h) = \max_{0 < m < N-1} \left| \int_{x_m}^{x_{m+1}} K_1(x_m, y_m, s) ds \right|, \]

\[ T_2(h) = \max_{0 < m < N-1} \left| \int_{x_m}^{x_{m+1}} \Gamma(x) dx - h \sum_{k=0}^{p} w_k \Gamma(x_m, x_{m+k}) \right|, \]

\[ T_3(h) = \max_{0 < m < N-1} \left| \int_{x_m}^{x_{m+1}} \Gamma(x) dx - h \sum_{r=0}^{p} L_r(u_j u_k) \Gamma(x_m, x_{r+k}) \right|, \]

\[ e_m, j = y(x_m, j) - y_m, \quad m = 0, 1, \ldots, N-1; j = 0, 1, \ldots, p, \]

\[ e_m = \max_{0 < j < p} |e_m, j|, \quad m = 0, 1, \ldots, N-1, \]

and proceed to the proof of the convergence theorems.

**Theorem 1.** Let \( L_1, L_2 \) be the Lipschitz constants of \( G(x, y, z) \) with respect to the second and third variable, respectively, and \( L_3 \) the Lipschitz constant of \( K(x, s, y) \) with respect to the third variable on the sets \( \Delta_1, \Delta_2 \). Then for the error \( e_m \) in the approximations (2.3), (2.4) we have, for \( h \) sufficiently small, that

\[ e_m \leq (C' + hE'E') \exp((m - 1)hE') \]

and

\[ e_0 \leq (T_2(h) + hW'L_2(p + 1)[T_1(h) + hWL_3(p + 1)T_3(h)])/(1 - D(p + 1)), \]

where

\[ C' = C/(1 - D(p + 1)), \quad E' = E(p + 1)/(1 - D(p + 1)), \]

License or copyright restrictions may apply to redistribution; see https://www.ams.org/journal-terms-of-use
\[ C = M_2(h) + W L_2(p + 1)(M_1(h) + T_1(h)) + h W^2 L_2 L_3 X(p + 1)^2 T_3(h) \]
\[ + T_2(h) + h W' L_2(p + 1)[M_1(h) + T_1(h) + h W L_3(p + 1) T_3(h)], \]
\[ D = h W' L_1 + h^2 W W' L_2 L_3 L(p + 1)^2, \]
\[ E = W L_1 + W^2 L_2 L_3 X(p + 1) + h W^2 L_2 L_3 L(p + 1)^2 + h W W' L_2 L_3(p + 1), \]

and \( \xi \) is such that \( e_0 < \xi \).

**Proof.** Subtracting (2.3) from (2.1) we find

\[ e_{m,i} = A + B, \]

where \( A = A_m, B = B_{m,i} \) with

\[ A = \int_0^x \Gamma(x) \, dx - h \sum_{i=0}^{m-1} \sum_{k=0}^p w_k G(x_{i,k}, y_{i,k}, z_{i,k}), \]

(3.13)
\[ B = \int_0^x \Gamma(x) \, dx - h \sum_{k=0}^p w_k G(x_{m,k}, y_{m,k}, z_{m,k}), \]

and \( z_{i,k} \) is the approximation to the integral \( z(x_{i,k}) = \int_0^{x_{i,k}} K_1(x_{i,k}, s) \, ds, i = 0, 1, \ldots, m; k = 0, 1, \ldots, p, \) given from Eq. (2.4) by

\[ z_{i,k} = h \sum_{\lambda=0}^{i-1} \sum_{\mu=0}^p w_\lambda K(x_{i,k}, x_{\lambda,\mu}, y_{\lambda,\mu}) \]
\[ + h u_k \sum_{\mu=0}^p w_\mu K\left(x_{i,k}, x_{i,k,\mu}, \sum_{r=0}^p l_r u_r y_{i,r}\right). \]

(3.14)

We shall proceed using the "add and subtract" procedure; see, for example, Mocarsky [9]. So if we denote by \( z'(x_{i,k}) \) the right-hand side of (3.14) with \( y_{\lambda,\mu}, y_{i,r} \) replaced by \( y(x_{\lambda,\mu}), y(x_{i,r}) \), we have

\[ |e_{m,i}| \leq \sum_{l=0}^4 |A_l| + \sum_{l=0}^4 |B_l|, \]

where

\[ A_1 = \int_0^x \Gamma(x) \, dx - h \sum_{i=0}^{m-1} \sum_{k=0}^p w_k \Gamma(x_{i,k}), \]

\[ A_2 = h \sum_{i=0}^{m-1} \sum_{k=0}^p w_k [\Gamma(x_{i,k}) - G(x_{i,k}, y(x_{i,k}), z'(x_{i,k}))], \]

(3.16)
\[ A_3 = h \sum_{i=0}^{m-1} \sum_{k=0}^p w_k [G(x_{i,k}, y(x_{i,k}), z'(x_{i,k})) - G(x_{i,k}, y_{i,k}, z'(x_{i,k}))], \]

\[ A_4 = h \sum_{i=0}^{m-1} \sum_{k=0}^p w_k [G(x_{i,k}, y_{i,k}, z'(x_{i,k})) - G(x_{i,k}, y_{i,k}, z_{i,k})], \]

License or copyright restrictions may apply to redistribution; see https://www.ams.org/journal-terms-of-use
and
\[ B_1 = \int_{x_m}^{x_m+1} \Gamma(x) \, dx - h \sum_{k=0}^{p} w_k \Gamma(x_m, k), \]
\[ B_2 = h \sum_{k=0}^{p} w_k [\Gamma(x_m, k) - G(x_m, k, y(x_m, k), z'(x_m, k))], \]
\[ B_3 = h \sum_{k=0}^{p} w_k [G(x_m, k, y(x_m, k), z'(x_m, k)) - G(x_m, y_m, k, z'(x_m, k))], \]
\[ B_4 = h \sum_{k=0}^{p} w_k [G(x_m, k, y_m, k, z'(x_m, k)) - G(x_m, y_m, z_m, k)]. \]

Using the Lipschitz conditions on \( G \) and \( K \), the definitions (3.7) and the triangle inequality, we obtain the following inequalities (see [8, pp. 178, 179, 197]),
\[ |A_1| \leq M_2(h), \]
\[ |A_2| \leq W_2(h+1) (M_1(h) + T_1(h)) + hW_2L_3X(p + 1)^2 T_3(h), \]
\[ |A_3| \leq hWL_1 \sum_{i=0}^{m-1} \sum_{k=0}^{p} |e_{i,k}|, \]
\[ |A_4| \leq hW_2L_3X(p + 1)^2 \sum_{\lambda=0}^{m-1} \sum_{\mu=0}^{p} |e_{\lambda,\mu}| + h^2W_2L_3L(p + 1)^2 \sum_{i=0}^{m-1} \sum_{r=0}^{p} |e_{i,r}|, \]

and
\[ |B_1| \leq T_2(h), \]
\[ |B_2| \leq hW_2^2(p + 1) [M_1(h) + T_1(h) + hWL_3(p + 1) T_3(h)], \]
\[ |B_3| \leq hWL_1 \sum_{k=0}^{p} |e_{m,k}|, \]
\[ |B_4| \leq hW_2^2(p + 1) \left[ hWL_3 \sum_{i=0}^{m-1} \sum_{r=0}^{p} |e_{i,r}| + hWL_3L(p + 1) \sum_{r=0}^{p} |e_{m,r}| \right]. \]

From (3.15), (3.18), (3.19) we then find
\[ |e_{m,j}| \leq C + D \sum_{r=0}^{p} |e_{m,r}| + hE \sum_{i=0}^{m-1} \sum_{r=0}^{p} |e_{i,r}|, \]
or
\[ e_m \leq C' + hE' \sum_{i=0}^{m-1} e_i. \]
Applying Lemma 2 now to (3.21) for $h$ sufficiently small and $e_0 \leq \xi$ we obtain the result (3.8) of the theorem. The result (3.9) can be proved proceeding similarly and taking into account the fact that for $m = 0$ we have $M_1(h) = 0$ and $M_2(h) = 0$.

Using the result of Theorem 1, we shall now establish convergence; that is, we shall prove

**Corollary 1.** Let

(i) the assumptions of Theorem 1 be valid,

(ii) $y(x), z(x), G(x, y(x), z(x)), K(x, y(s))$ be continuous for $x \in [0, X]$, $s \in [0, x]$, and

(iii) $g(x) = (x - u_0)(x - u_1) \cdots (x - u_p) \in P_v$.

Then $e_m \to 0$, $m < N$ as $h \to 0$, $Nh = X$.

**Proof.** We shall prove that

\[
\lim_{h \to 0} M_1(h) = \lim_{h \to 0} M_2(h) = \lim_{h \to 0} T_1(h) = \lim_{h \to 0} T_2(h) = \lim_{h \to 0} T_3(h) = 0,
\]

and then from (3.8), (3.9) the required result follows.

From Lemma 1 we have that the degree of precision of the quadrature rule (1.7) is $p + v > p$. Thus,

\[M_2(h) \leq 6X(1 + \Lambda_p)\omega(\Gamma; h/2p)\quad ([8, result I-2.20]),\]

where $\Lambda_p$ is defined in (3.7) and

\[\omega(\varphi; \delta) = \sup_{x_1, x_2 \in [a, b]} \frac{|\varphi(x_1) - \varphi(x_2)|}{|x_1 - x_2| < \delta}\]

is the modulus of continuity of a function $\varphi(x)$. So $\lim_{h \to 0; Nh = X} M_2(h) = 0$. Also,

\[M_1(h) \leq 6X(1 + \Lambda_p)\max_{0 < m < N - 1} \omega(K_1(x_{m,j}, s); h/2p),\]

where $K_1(x, s) \equiv K(x, y(s)), 0 \leq s \leq x \leq X$, is uniformly continuous in $0 \leq s \leq x \leq X$. So

\[\max_{0 < m < N - 1} \omega(K_1(x_{m,j}, s); h/2p) \to 0 \quad \text{as} \quad h \to 0,
\]

which implies that $\lim_{h \to 0; Nh = X} M_1(h) = 0$. For $T_1(h)$ we have

\[T_1(h) \leq 6h(1 + \Lambda_p)\max_{0 < m < N - 1} \omega(K_1(x_{m,j}, mh + uhs); 1/2p),\]

and therefore $\lim_{h \to 0; Nh = X} T_1(h) = 0$. For $T_2(h)$, since the degree of precision of rule (1.6) is $\geq 0$ and the continuity assumption of $\Gamma(x)$ on $0 \leq x \leq X$ implies the
bounded and Riemann integrable of $\Gamma(x)$, we have $\lim_{h \to 0 ; N h = X} T_2(h) = 0$; see [5, p. 25]. Finally, for the error $T_3(h)$ in Lagrangian interpolation, using result (i-2.7) and (i) of (Jackson) Theorem 1-2 in [8] we have $T_3(h) \leq 6(1 + \Lambda_p)\omega(y; h/2p)$.

$\gamma(x) \in C([0, X])$ implies that $\gamma(x)$ is uniformly continuous in $[0, X]$, ($X < \infty$). Thus, $\lim_{h \to 0 ; N h = X} \omega(y; h/2p) = 0$ and so $\lim_{h \to 0 ; N h = X} T_3(h) = 0$.

Having examined convergence, we shall examine the rate of convergence; that is, we shall prove

**Corollary 2.** Let

(i) the assumptions of Theorem 1 be valid,

(ii) $g(x) \in P_v$,

(iii) $K(x, s, y)$ is $p + v + 2$ times continuously differentiable with respect to $x$, $s$ and $y$, respectively, on $0 \leq s < x$, $0 \leq x < X$, $|y| \leq \bar{y}$, where $\bar{y} = \max_{0 < x < X} |\gamma(x)|$,

(iv) $G(x, y, z)$ is $p + v + 2$ times continuously differentiable with respect to $x$, $y$ and $z$, respectively on $0 \leq x < X$, $|y| \leq \bar{y}$, $|z| \leq \bar{z}$, where $\bar{y}$ is as in (iii) and $\bar{z} = \max_{0 < x < X} |\gamma(x)|$.

(v) $\gamma(x)$ is $p + 2$ times continuously differentiable on $0 \leq x < X$.

Then, there are constants $C_1$, $C_2$, $C_3$ such that,

$$e_0 \leq C_1 h^{p+2},$$

$$e_m \leq C_2 h^{p+1}, \quad m = 1, 2, \ldots, N - 1, \text{ if } v = 0,$$

$$e_m \leq C_3 h^{p+2}, \quad m = 1, 2, \ldots, N - 1, \text{ if } v > 0.$$

**Proof.** Consider the definitions in (3.5), (3.6). Then for the errors $M_2(h)$, $M_1(h)$, $T_1(h)$, $T_2(h)$ in (3.7), extending the analysis of Weiss [12], we find

$$M_2(h) = \max_{0 < m < N - 1} \frac{h^{p+v+2}}{(p + v + 1)!} \left| E_p(s^{p+v+1}) \sum_{i=0}^{m-1} \Gamma(p+v+1)(x_i) \right| + O(h^{p+v+3}),$$

or

$$M_2(h) = \max_{0 < m < N - 1} \frac{h^{p+v+1}}{(p + v + 1)!} \left| E_p(s^{p+v+1}) \int_0^{x_m} \Gamma(p+v+1)(s) ds \right|$$

$$+ O(h^{p+v+2}),$$

(3.24)

where we have used Taylor series expansion for $\Gamma(x_i + hs)$, $\Gamma(x_i + u_k h)$ and that $E_p(s^{p+1+r}) = 0$ if $r < v - 1$ (Lemma 1). Similarly we find

$$M_1(h) = \max_{0 < m < N - 1} \frac{h^{p+v+1}}{(p + v + 1)!} \left| E_p(s^{p+v+1}) \int_0^{x_m} K_1(p+v+1)(x_m, j, s) ds \right|$$

$$+ O(h^{p+v+2}),$$

(3.25)

For $T_2(h)$, using Taylor series expansion for $\Gamma(x_m + h(s - u_j))$ and
\[ T_2(h) = \max_{0 < m < N-1} \max_{0 < u_j \leq p} h \sum_{r=p+1}^{p+v} \frac{h^r}{r!} \Gamma^{(r)}(x_{m,i}) E_j((s-u_j)') + O(h^{p+u+2}), \]

\[ (E_j(s') = 0 \text{ for } r = 0, 1, \ldots, p). \]

For \( T_1(h) \), using Taylor series expansion for \( K_1(x_{m,i}, x_{m,j} + u_jh(s-1)) \) and \( K_1(x_{m,i}, x_{m,j} + u_jh(u_k - 1)) \), we find

\[ T_1(h) = \max_{0 < m < N-1} \max_{0 < u_j \leq p} u_j h \sum_{r=p+1}^{p+v} \frac{u_j^r h^r}{r!} K_1^{(r)}(x_{m,i}, x_{m,j}) E_j((s-1)'), \]

\[ + O(h^{p+u+2}), \]

\[ (E_j(s') = 0 \text{ for } r = 0, 1, \ldots, p). \]

For the Lagrangian interpolation, using the well-known error formula, we have

\[ T_3(h) = \max_{0 < m < N-1} \max_{0 < j, k \leq p} h^{p+1} |u_j^{(p+1)}(\xi) \Gamma^p(u_k)| / (p+1)! + O(h^{p+2}), \]

where \( \xi \) lies between \( x_{m,0}, x_{m,p} \text{ and } x_{m,i,k} \).

Combining now the results above with (3.8), (3.9), we obtain the required result (3.23).

The same rate of convergence was found in [8] for the other two schemes mentioned in the introduction, but for scheme A under the additional assumption that \( z(x) \) is \( p+2 \) times continuously differentiable on \( 0 < x < X \).

4. Numerical Results. We now display some numerical results obtained by testing scheme C on three examples ((a), (b), (c) below) for the cases \( p = 2, p = 3, u_0 = 0, u_i = i/p, i = 1, \ldots, p, \) with \( v = 1 \) and \( v = 0, \) respectively. The results verify order of convergence \( O(h^{p+2}) = O(h^4) \) and \( O(h^{p+1}) = O(h^4) \), respectively. Example (c) is a “stiff” (constructed) example. In general, Eqs. (2.3), (2.4) form a nonlinear system for \( y_{m,0}, y_{m,1}, \ldots, y_{m,p}; z_{m,0}, z_{m,1}, z_{m,2}, \ldots, z_{m,p} \), which we solved by applying a Newton iteration. It might seem preferable to use the method with \( z_{m,0}, \ldots, z_{m,p} \) eliminated (see [8], methods “with elimination of the \( z \)-variable”) in order to reduce computing time; a sample of actual computing time (see [8, p. 146]) though, shows this to be false.

(a) \( y(x) = 1 + y(x) - x \exp(-x^2) - 2 \int_0^x x \exp(-y^2(s)) ds, 0 \leq x \leq 1, y(0) = 0, y(x) = x \) (Mocarsky [9, p. 239], Linz [7, p. 301], Makroglou [8, Example 2, p. 93]).

(b) \( y'(x) = 1 + 2x - y(x) + \int_0^x x(1 + 2x) \exp(s(x-s))y(s) ds, 0 \leq x \leq 1, y(0) = 1, y(x) = \exp(x^2) \) (Linz [7, p. 300], Makroglou [8, Example 6, p. 94]).

(c) \( y'(x) = y'(0) + \int_0^x -20(y(s)) ds + 21y(0) - 21y(s), y(0) = 1, y'(0) = 1, y(x) = (21/19)\exp(-x) - (2/19)\exp(-20x) \) (Makroglou [8, Example 7b, pp. 94, 95]).
Example (a)  Scheme C
\[ u_0 = 0, u_i = i/p \quad (i = 1, 2, \ldots, p) \]

<table>
<thead>
<tr>
<th>( x )</th>
<th>( p = 2 )</th>
<th>( p = 3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>( h = 0.1 )</td>
<td>( h = 0.05 )</td>
</tr>
<tr>
<td>0.1</td>
<td>( 3.56 \times 10^{-11} )</td>
<td>( 2.26 \times 10^{-12} )</td>
</tr>
<tr>
<td>0.2</td>
<td>( 5.77 \times 10^{-10} )</td>
<td>( 4.29 \times 10^{-11} )</td>
</tr>
<tr>
<td>0.3</td>
<td>( 3.22 \times 10^{-9} )</td>
<td>( 2.29 \times 10^{-10} )</td>
</tr>
<tr>
<td>0.4</td>
<td>( 1.06 \times 10^{-8} )</td>
<td>( 7.29 \times 10^{-10} )</td>
</tr>
<tr>
<td>0.5</td>
<td>( 2.59 \times 10^{-8} )</td>
<td>( 1.74 \times 10^{-9} )</td>
</tr>
<tr>
<td>0.6</td>
<td>( 5.26 \times 10^{-8} )</td>
<td>( 3.48 \times 10^{-9} )</td>
</tr>
<tr>
<td>0.7</td>
<td>( 9.37 \times 10^{-8} )</td>
<td>( 6.13 \times 10^{-9} )</td>
</tr>
<tr>
<td>0.8</td>
<td>( 1.52 \times 10^{-7} )</td>
<td>( 9.85 \times 10^{-9} )</td>
</tr>
<tr>
<td>0.9</td>
<td>( 2.29 \times 10^{-7} )</td>
<td>( 1.48 \times 10^{-8} )</td>
</tr>
<tr>
<td>1.0</td>
<td>( 3.28 \times 10^{-7} )</td>
<td>( 2.10 \times 10^{-8} )</td>
</tr>
<tr>
<td>1.5</td>
<td>( 1.24 \times 10^{-6} )</td>
<td>( 7.85 \times 10^{-8} )</td>
</tr>
<tr>
<td>2.0</td>
<td>( 3.43 \times 10^{-6} )</td>
<td>( 2.16 \times 10^{-7} )</td>
</tr>
</tbody>
</table>

Example (b)  Scheme C
\[ u_0 = 0, u_i = i/p \quad (i = 1, 2, \ldots, p) \]

<table>
<thead>
<tr>
<th>( x )</th>
<th>( p = 2 )</th>
<th>( p = 3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>( h = 0.1 )</td>
<td>( h = 0.05 )</td>
</tr>
<tr>
<td>0.1</td>
<td>( 2.20 \times 10^{-7} )</td>
<td>( 1.38 \times 10^{-8} )</td>
</tr>
<tr>
<td>0.2</td>
<td>( 4.71 \times 10^{-7} )</td>
<td>( 2.95 \times 10^{-8} )</td>
</tr>
<tr>
<td>0.3</td>
<td>( 7.68 \times 10^{-7} )</td>
<td>( 4.81 \times 10^{-8} )</td>
</tr>
<tr>
<td>0.4</td>
<td>( 1.12 \times 10^{-6} )</td>
<td>( 7.04 \times 10^{-8} )</td>
</tr>
<tr>
<td>0.5</td>
<td>( 1.56 \times 10^{-6} )</td>
<td>( 9.75 \times 10^{-8} )</td>
</tr>
<tr>
<td>0.6</td>
<td>( 2.08 \times 10^{-6} )</td>
<td>( 1.31 \times 10^{-7} )</td>
</tr>
<tr>
<td>0.7</td>
<td>( 2.72 \times 10^{-6} )</td>
<td>( 1.71 \times 10^{-7} )</td>
</tr>
<tr>
<td>0.8</td>
<td>( 3.46 \times 10^{-6} )</td>
<td>( 2.17 \times 10^{-7} )</td>
</tr>
<tr>
<td>0.9</td>
<td>( 4.26 \times 10^{-6} )</td>
<td>( 2.68 \times 10^{-7} )</td>
</tr>
<tr>
<td>1.0</td>
<td>( 4.98 \times 10^{-6} )</td>
<td>( 3.14 \times 10^{-7} )</td>
</tr>
<tr>
<td>1.5</td>
<td>( 3.14 \times 10^{-5} )</td>
<td>( 1.90 \times 10^{-6} )</td>
</tr>
<tr>
<td>2.0</td>
<td>( 1.71 \times 10^{-3} )</td>
<td>( 1.05 \times 10^{-4} )</td>
</tr>
</tbody>
</table>
### Example (c) Scheme C

\( u_0 = 0, u_i = i/p \) \( (i = 1, 2, \ldots, p) \)

<table>
<thead>
<tr>
<th>( x )</th>
<th>( p = 2 )</th>
<th>( p = 3 )</th>
<th>( x )</th>
<th>( p = 2 )</th>
<th>( p = 3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.4</td>
<td>2.37 \times 10^{-2}</td>
<td>7.95 \times 10^{-3}</td>
<td>1.0</td>
<td>5.72 \times 10^{-2}</td>
<td>3.53 \times 10^{-2}</td>
</tr>
<tr>
<td>0.8</td>
<td>5.35 \times 10^{-3}</td>
<td>5.96 \times 10^{-4}</td>
<td>2.0</td>
<td>3.13 \times 10^{-2}</td>
<td>1.20 \times 10^{-2}</td>
</tr>
<tr>
<td>1.2</td>
<td>1.20 \times 10^{-3}</td>
<td>4.31 \times 10^{-5}</td>
<td>3.0</td>
<td>1.72 \times 10^{-2}</td>
<td>3.98 \times 10^{-3}</td>
</tr>
<tr>
<td></td>
<td>4.0</td>
<td>9.43 \times 10^{-3}</td>
<td></td>
<td>1.36 \times 10^{-3}</td>
<td></td>
</tr>
<tr>
<td>3.2</td>
<td>4.46 \times 10^{-6}</td>
<td>5.80 \times 10^{-7}</td>
<td>5.0</td>
<td>5.19 \times 10^{-3}</td>
<td>4.48 \times 10^{-4}</td>
</tr>
<tr>
<td>3.6</td>
<td>3.74 \times 10^{-6}</td>
<td>4.37 \times 10^{-7}</td>
<td>6.0</td>
<td>2.85 \times 10^{-3}</td>
<td>1.56 \times 10^{-4}</td>
</tr>
<tr>
<td>4.0</td>
<td>2.87 \times 10^{-6}</td>
<td>3.26 \times 10^{-7}</td>
<td>7.0</td>
<td>1.57 \times 10^{-3}</td>
<td>5.02 \times 10^{-5}</td>
</tr>
<tr>
<td>4.4</td>
<td>2.14 \times 10^{-6}</td>
<td>2.40 \times 10^{-7}</td>
<td>8.0</td>
<td>8.63 \times 10^{-4}</td>
<td>1.78 \times 10^{-5}</td>
</tr>
<tr>
<td></td>
<td>9.0</td>
<td>4.74 \times 10^{-4}</td>
<td></td>
<td>5.61 \times 10^{-6}</td>
<td></td>
</tr>
<tr>
<td>6.0</td>
<td>5.90 \times 10^{-7}</td>
<td>6.61 \times 10^{-8}</td>
<td>10.0</td>
<td>2.60 \times 10^{-4}</td>
<td>2.04 \times 10^{-6}</td>
</tr>
<tr>
<td>6.4</td>
<td>4.22 \times 10^{-7}</td>
<td>4.73 \times 10^{-8}</td>
<td>11.0</td>
<td>1.43 \times 10^{-4}</td>
<td>6.24 \times 10^{-7}</td>
</tr>
<tr>
<td></td>
<td>12.0</td>
<td>7.86 \times 10^{-5}</td>
<td></td>
<td>2.36 \times 10^{-7}</td>
<td></td>
</tr>
<tr>
<td>8.0</td>
<td>1.06 \times 10^{-7}</td>
<td>1.19 \times 10^{-8}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>8.4</td>
<td>7.49 \times 10^{-8}</td>
<td>8.40 \times 10^{-9}</td>
<td>15.0</td>
<td>1.30 \times 10^{-5}</td>
<td>7.57 \times 10^{-9}</td>
</tr>
<tr>
<td>11.6</td>
<td>4.22 \times 10^{-9}</td>
<td>4.73 \times 10^{-10}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>12.0</td>
<td>2.93 \times 10^{-9}</td>
<td>3.28 \times 10^{-10}</td>
<td>20.0</td>
<td>6.48 \times 10^{-7}</td>
<td>4.42 \times 10^{-11}</td>
</tr>
</tbody>
</table>

All the above results were obtained on a CDC 7600 and represent absolute errors \( y(x_n) - y_n, n = 0, 1, \ldots \).

In Makroglou [8], scheme C was tested on one more example for \( p = 1, 2, 3, 4, 5, u_0 = 0, u_i = i/p, i = 1, 2, \ldots, p \). In the same reference are also included results with \( u_i \) being the Lobatto points \( (p = 2, u_0 = 0.5(1 - 0.2(1 + \sqrt{6})), u_1 = 0.5(1 + 0.2(\sqrt{6} - 1)), u_2 = 1), (p = 3, u_0 = 0, u_1 = 0.5(1 - \sqrt{0.2}), u_2 = 0.5(1 + \sqrt{0.2}), u_3 = 1), (p = 4, u_0 = 0, u_1 = 0.5(1 - \sqrt{3/7}), u_2 = 0.5, u_3 = 0.5(1 + \sqrt{3/7}), u_4 = 1) \). Using these points, order of convergence, higher than the expected from the convergence proof given here, was observed. Weiss [12] has proved an \( O(h^{p+q+1}) \) order of convergence for \( e_{m,p} \) in solving equations of the form (1.5) by (1.11) and (1.12). The adaptation of his work to equations of the form (1.1) solved by scheme C will be tried next.

In [8] some examples were also tested for \( p = 5 \) and \( u_i \) equidistant. This case though, did not seem to have any advantage over the one with \( p = 4 \); for some cases it was even worse in accuracy. This is perhaps due to increased round-off errors because of the increased size of the algebraic equations to be solved.
Compared with the methods mentioned in the introduction of the same order, tested on the same examples, scheme C was found the most accurate. Especially on the "stiff" example (example (c)) we obtained very accurate results with errors decreasing as $x$ increases, even with a stepsize as big as $h = 1$. For some of the other methods the error grows catastrophically as $x$ increases; see [8, Tables 35, 36].

For stability results obtained for the linear test equation $y'(x) = \xi y(x) + \eta \int_0^x y(s) \, ds$, $y(0)$ given, $\xi, \eta$ real constants, see [2], [8].

Acknowledgement. This work forms part of the author's Ph. D. thesis written at the University of Manchester under the supervision of Dr. C. T. H. Baker and Dr. Ian Gladwell (for six months), both of whom the author wishes to thank from here too, for their valuable advice, encouragement and criticism.

The author also wishes to acknowledge that the part about convergence with low differentiability requirements (Corollary 1) was added by Dr. Baker.

The M. Sc. and Ph. D. work of the author were supported mainly by the Greek State Scholarships Foundation.

Statistical Unit
Athens University
Panepistemiopolis
Athens-621, Greece