A Mechanical Quadrature Method for Solving Delay Volterra Integral Equation with Weakly Singular Kernels

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In this work, a mechanical quadrature method based on modified trapezoid formula is used for solving weakly singular Volterra integral equation with proportional delays. An improved Gronwall inequality is testified and adopted to prove the existence and uniqueness of the solution of the original equation. Then, we study the convergence and the error estimation of the mechanical quadrature method. Moreover, Richardson extrapolation based on the asymptotic expansion of error not only possesses a high accuracy but also has the posterior error estimate which can be used to design self-adaptive algorithm. Numerical experiments demonstrate the efficiency and applicability of the proposed method.

1. Introduction

In recent years, Volterra integral equation with delay has received a considerable amount of attention. This paper considers the following weakly singular Volterra integral equations with proportional delays:

\[ u(t) = g(t) + (Iu)(t) + (I_{\theta}u)(t), \quad t \in I = [0, T], \]  

(1)

with

\[ (Iu)(t) = \int_{0}^{t} K_1(t, s) u(s) \, ds, \]  

(2a)

\[ (I_{\theta}u)(t) = \int_{0}^{\theta(t)} K_2(t, s) u(s) \, ds, \]  

(2b)

and

\[ \theta(t) = q\bar{t}, \quad 0 < q < 1, \]

\[ K_1(t, s) = s^{\alpha}k_1(t, s), \]  

(3a)

\[ K_2(t, s) = s^{\gamma}k_2(t, s), \]  

(3b)

\[ -1 < \alpha < 0, \quad -1 < \gamma \leq 0, \]

where \( k_1(t, s) \) and \( k_2(t, s) \) are known continuous functions defined on the domains \( D = \{(t, s) : 0 \leq s \leq t \leq T\} \) and \( D_0 = \{(t, s) : 0 \leq s \leq \theta(t), t \in I = [0, T]\} \), respectively. \( g(t) \) is a known function and \( u(t) \) is an unknown function. In practice, the delay arguments are consistent with the real phenomena which make the models more realistic for simulation. Delay integral equation and partial differential equation have been widely used in many population growth and relevant phenomena in mathematical biology [1–3].

Numerical algorithms for implementing delay models should be designed specially according to the nature of the equations. There are many numerical techniques for the delay differential equations [4–8] and integral-differential equations [9–11]. Simultaneously, vast researchers also focused their interests on the numerical techniques of delay integral equations with continuous kernels, such as least squares approximation method [12], spectral method [13], Bernoulli wavelet method [14], and collocation method [15]. Xie et al. in [16] handled the Volterra integral equation with the delay function vanishing at the initial point in the given interval; they found that the iterated collocation solution possessed local superconvergence at the mesh points. In [17], the authors adopted multistep method based on Hermite collocation method and it turned out that the numerical method had uniform order \( 2m + 2r \) with \( m \) collocation points and \( r \) previous time steps. In [18], an hp-spectral collocation method was used for nonlinear Volterra integral equations with vanishing variable delays.
There are a few researches on the delay integral equation with weakly singular kernels, such as [19]. In this paper, we concentrate on (1) whose upper limit of integral is a delay function and the integral kernels are weakly singular functions, which increase the computational complexity and the theoretical difficulty. To the best of our knowledge, there are no studies in (1) by the mechanical quadrature method in recent years. A further advantage of this method for (1) is that the error has an asymptotic expansion. Thus, we can improve the accuracy order of approximation. Simultaneously, the theoretical analysis is complete and the calculation is simplified.

In this paper, we firstly deduce the improved Gronwall inequality; the existence and uniqueness of the solution for (1) are testified via the improved Gronwall inequality. Then the existence and uniqueness of the solution for (1) is simplified. We use the iterative method for solving the approximate equation. The existence and uniqueness of the solution for the approximate equation are testified via the discrete Gronwall inequality. Finally, we prove that the convergent order is \( O(h^2) \), in order to achieve a higher accuracy order \( O(h^2) \), the Richardson-\( h^{2+\min\{\alpha,\gamma\}} \) extrapolation based on the asymptotic expansion of error is adopted; moreover, a posterior error estimate is realized conveniently.

The layouts of this paper are as follows. In Section 2, we prove the existence and uniqueness of the solution for (1). In Section 3, we introduce the quadrature method, iterative method, and interpolation technique. In Section 4, the existence and uniqueness of the solution for the approximate equation are discussed. In Section 5, the convergence and the error estimation are obtained to ensure the reliability of the method. In Section 6, the asymptotic expansion of the error is achieved, a higher accuracy order is realized by extrapolation, and a posterior error estimate is derived. In Section 7, some numerical examples are demonstrated to illustrate the theoretical results. Some concluding remarks are provided in Section 8.

2. The Existence and Uniqueness of the Solution for the Original Equation

In this section, we will verify the existence and uniqueness of the solution for (1). We first prove the improved Gronwall inequality.

**Lemma 1.** Suppose that \( g(t), h(t), \) and \( u(t) \) are nonnegative integrable functions, \( t \in [0, T] \), \( 0 < q < 1 \), and \( A \geq 0 \); based on the inequality

\[
u(t) \leq A + \int_0^t g(s) u(s) ds + \int_0^q h(s) u(s) ds, \tag{4}\]

we have

\[
u(t) \leq Ae^{\int_0^q (g(s)+h(s))ds}. \tag{5}\]

**Proof.** Due to the fact that \( 0 < q < 1, qt < t \) and

\[
u(t) \leq A + \int_0^t g(s) u(s) ds + \int_0^q h(s) u(s) ds \leq A + \int_0^t (g(s) + h(s)) u(s) ds. \tag{6}\]

Let \( H(s) = g(s) + h(s) \); we can deduce

\[
\frac{H(t)u(t)}{A + \int_0^t H(s) u(s) ds} \leq H(t). \tag{7}
\]

Integrate on both sides

\[
\ln (A + \int_0^t H(s) u(s) ds) - \ln A \leq \int_0^t H(s) ds,
\]

and then

\[
A + \int_0^t H(s) u(s) ds \leq Ae^{\int_0^q H(s)ds}; \tag{9}\]

namely,

\[
u(t) \leq Ae^{\int_0^q H(s)ds}. \tag{10}\]

The proof of the Lemma 1 is completed.

**Theorem 2.** Assume that \( k_i(t,s) \) \( (i = 1, 2) \) are known continuous functions defined on the domains \( D = \{(t,s) : 0 \leq s \leq t \leq T\} \) and \( D_0 = \{(t,s) : 0 \leq s \leq \theta(t), t \in I = [0, T]\} \), respectively; then the solution of (1) is existent uniquely.

**Proof.** We construct the sequence \( u_k(t) \) \( (k \in \mathbb{N}) \), satisfying

\[
u_0(t) = u(0) = g(t), \]

\[
u_k(t) = g(t) + \int_0^t K_1(t,s) u_{k-1}(s) ds + \int_0^q K_2(t,s) u_{k-1}(s) ds, \tag{11}\]

where \( K_1(t,s) \) and \( K_2(t,s) \) are defined in (3), with \( k \in \mathbb{N} \). \( k_i(t,s) \) \( (i = 1, 2) \) are continuous functions; then there exists a constant \( C \) such that \( |k_i(t,s)| \leq C \). We have

\[
|u_2(t) - u_1(t)| \leq C \int_0^t s^\alpha |u_1(s) - u_0(s)| ds + C \int_0^q s^\gamma |u_1(s) - u_0(s)| ds \tag{12}\]

\[
|u_2(t) - u_1(t)| \leq C \int_0^t (s^{\alpha + \gamma} + s^\gamma) |u_1(s) - u_0(s)| ds \leq Ca \int_0^t s^\alpha ds + Ca \int_0^q s^\gamma ds = 2Ca \frac{\mu^\mu+1}{\mu + 1}.
\]
Complexity

with \( a = \max_{0 \leq t \leq T} |u_1(t) - u_0(t)|; \mu = \min(\alpha, \gamma) \). Now, we can deduce

\[
|u_k(t) - u_{k-1}(t)| \leq a \frac{(2C)^{k-1}}{(k-1)! (\mu + 1)^{k-1}} t^{(k-1)(\mu + 1)}.
\]

By means of the mathematical induction, when \( n = k + 1 \), we obtain

\[
|u_{k+1}(t) - u_k(t)| \leq C \int_0^t s^{\alpha} |u_k(s) - u_{k-1}(s)| ds + C \int_0^t s^{\gamma} |u_k(s) - u_{k-1}(s)| ds
\]

\[
\leq C \int_0^t s^{\alpha} |u_k(s) - u_{k-1}(s)| ds + 2C \int_0^t s^{\gamma} |u_{k-1}(s)| ds
\]

\[
\leq a \frac{(2C)^k}{k! (\mu + 1)^k} t^{k(\mu + 1)}.
\]

Next, we prove that \( u_k(t) \) is the basic sequence in \( C[0, T] \); in fact,

\[
|u_k(t) - u_{k+m}(t)|
\]

\[
\leq |u_{k+1}(t) - u_k(t)| + |u_{k+2}(t) - u_{k+1}(t)| + \ldots
\]

\[
+ |u_{k+m}(t) - u_{k+m-1}(t)|
\]

\[
\leq a \frac{(2C)^k}{k! (\mu + 1)^k} t^{k(\mu + 1)} + \ldots
\]

\[
+ a \frac{(2C)^{k+m-1}}{(k+m-1)! (\mu + 1)^{k+m-1}} t^{(k+m-1)(\mu + 1)}.
\]

For sufficiently small \( \epsilon > 0 \), there exists a positive integer \( N \) such that when \( n > N \) and any \( m > 0 \), we have

\[
|u_n(t) - u_{n+m}(t)| \leq \epsilon.
\]

According to Cauchy's test for convergence, the sequence \( u_n(t) \) \( (n \in \mathbb{N}) \) is convergent uniformly to \( u(t) \) which is the solution of (1).

Suppose that both \( u \) and \( v \) are the solutions of (1); let \( |w(t)| = |u - v| \); then we get

\[
|w(t)|
\]

\[
= \left| \int_0^t s^\alpha k_1(t, s) w(s) ds + \int_0^t s^\gamma k_2(t, s) w(s) ds \right|
\]

\[
\leq \int_0^t s^\alpha |k_1(t, s)| |w(s)| ds
\]

\[
+ \int_0^t s^\gamma |k_2(t, s)| |w(s)| ds
\]

\[
\leq \int_0^t s^\alpha |k_1(t, s)| |w(s)| ds + \int_0^t s^\gamma |k_2(t, s)| |w(s)| ds.
\]

\[
\leq \int_0^t (s^\alpha |k_1(t, s)| + s^\gamma |k_2(t, s)|) |w(s)| ds.
\]

\[
\leq \int_0^t (s^\alpha |k_1(t, s)| + s^\gamma |k_2(t, s)|) |w(s)| ds.
\]

Next, we verify that \( s^\alpha |k_1(t, s)| + s^\gamma |k_2(t, s)| \) is integrable;

\[
\int_0^t s^\alpha |k_1(t, s)| + s^\gamma |k_2(t, s)| ds \leq C \int_0^t (s^\alpha + s^\gamma) ds
\]

\[
\leq C \left( \frac{t^{1+\alpha}}{1+\alpha} + \frac{t^{1+\gamma}}{1+\gamma} \right).
\]

According to Lemma 1, we can derive that \( |w(t)| = 0 \); the solution is unique.

\[
\square
\]

3. The Quadrature Method and the Iterative Algorithm

Let the delay function \( \theta(t) = t - \tau(t) \geq 0 \), \( t \in [0, T] \), satisfy the following conditions [21]:

(1) \( \theta(0) = 0 \), and \( \theta(t) > 0 \) with \( t \in (0, T) \) (vanishing delay);

(2) \( \theta(t) \leq q_1 t \) on \( I = [0, T] \) for some \( q_1 \in (0, 1) \);

\( \theta(t) \geq q_0 > 0 \); \( q_1, i = 0, 1 \) are constants;

(3) \( \theta(t) \in C^1(I) \).

For \( 0 < q < 1 \), the special case is \( \tau(t) = (1 - q)t \); we get \( \theta(t) = t - \tau(t) = qt \); (1) is a weakly integral equation with vanishing delay. In order to solve (1), the modified trapezoid quadrature formula is considered to deal with the integrals \( (Iu)(t) \) and \( (I\alpha)(u)(t) \).

It is challenging because the integral upper limit is a delay function \( \theta(t) \) and the integral kernels are weakly singular at the endpoint.

Lemma 3 (see [22]). Let \( I(G) = \int_a^b G(x) dx = \int_a^b (x - a)^\alpha g(x) dx, -1 < \alpha < 0 \), and \( g(x) \in C^{2m}[a, b] \); then the modified trapezoidal rule is

\[
Q_N(G) = -h^{1+\alpha} g(a) \zeta(-\alpha) + h \sum_{i=1}^{N-1} G(x_i) + \frac{h^2}{2} G(b),
\]

where \( \zeta \) is the zeta function. Further, \( I(G) \) has the following asymptotic expansion:

\[
E_N(G) = Q_N(G) - I(G)
\]

\[
= \sum_{j=1}^{m-1} \frac{B_{2j}}{(2j)!} G^{(2j-1)}(b) h^{2j} + \sum_{j=1}^{2m-1} \frac{(1-j) \zeta(-\alpha-j) g^{(j)}(a) h^{j+1}}{j!} + O \left( h^{2m} \right).
\]

where \( B_{(2j)}, j = 1, \ldots, m - 1 \), are the Bernoulli numbers.
Lemma 4. Let \( u \in C^3[0, 1] \), and \( z = \beta x + (1 - \beta)y \) with \( \beta \in [0, 1], x, y \in [0, T] \); then there is
\[
u(z) = \beta u(x) + (1 - \beta) u(y)
- \frac{\beta(1 - \beta)}{2}(x - y)^2 u''(z) + O((x - y)^3).
\]
(21)

Proof. The Taylor expansion of function \( u(x) \) at the point \( z \) is
\[
u(x) = u(\beta x + (1 - \beta)y)\]
\[
u(x) = u(z + (1 - \beta)(x - y))\]
\[
u(x) = u(z) + (1 - \beta)(x - y)u'(z)
+ \frac{(1 - \beta)^2}{2}(x - y)^2 u''(z) + O((x - y)^3).
\]
Similarly, the Taylor expansion of function \( u(y) \) at the point \( z \) is
\[
u(y) = u(z - \beta(x - y))\]
\[
u(y) = u(z - \beta(x - y))u'(z) + \frac{\beta^2}{2}(x - y)^2 u''(z) + O((x - y)^3).
\]
Combining (22) with (23), we derive (21).
\[

Now, we structure the quadrature algorithm; by Lemma 3 and the trapezoidal quadrature rule, we obtain
\[
I_1 = \int_0^t s^2 k_1(t, s) u(s) ds = \sum_{k=1}^{i-1} \int_{[q_i]}^t s^2 k_1(t, s) u(t_k)\]
\[
+ \frac{h}{2} q^v t_{q_t} k(t, t) u(t) - \zeta (-\alpha) h^{1+\alpha} k(t, t_0) u(t_0)
+ E_{1i}(s^2 k_1(t, s)),
\]
(26)
\[
I_2 = \int_{[q_i]}^t s^2 k_2(t, s) u(s) ds = \sum_{k=1}^{i-1} \int_{[q_i]}^t s^2 k_2(t, s) u(t_k)\]
\[
+ \frac{h}{2} q^v t_{q_t} k(t, t_0) u(t) + E_{2i}(s^2 k_2(t, s)),
\]
(27)
\[
I_3 = \int_{[q_i]}^t s^2 k_2(t, s) u(s) ds
\]
\[
= \frac{q^v t_{q_t} - t_{[q_i]}}{2} \left[ t_{[q_i]} k_2(t, t_{[q_i]}) u(t_{[q_i]}) + (q t)^{\gamma}
\cdot u(t_k) + \frac{h^2}{2} q^v t_{[q_i]} k_2(t, t_{[q_i]}) u(t_{[q_i]}) - \zeta (-\gamma)
\cdot h^{1+\gamma} k_2(t, t_0) u(t_0) + E_{3i}(s^2 k_2(t, s))\right]
\]
(28)
where \( E_{1i}, E_{2i}, \) and \( E_{3i} \) are error functions, which are, respectively, equal to
\[
E_{1i} = \left[ k_1(t, s) u(s) \right]_{s=q_i}^t \zeta (-\alpha) h^{3+\alpha}
\]
\[
+ \frac{1}{2!} \left[ k_1(t, s) u(s) \right]_{s=q_i}^{t_{[q_i]}} \zeta (-\alpha - 2) + o (h^{3+\alpha})
= T_1(t_j) h^{3+\alpha} + O (h^{3+\alpha}),
\]
\[
E_{2j} = \left[ k_2(t, s) u(s) \right]_{s=q_i}^t \zeta (-\gamma) h^{3+\gamma}
\]
\[
+ \frac{1}{2!} \left[ k_2(t, s) u(s) \right]_{s=q_i}^{t_{[q_i]}} \zeta (-\gamma - 2) h^{3+\gamma} + o (h^{3+\gamma})
= T_2(t_j) h^{3+\gamma} + O (h^{3+\gamma}),
\]
\[
E_{3j} = \left[ \beta u(t_{[q_i]}) + (1 - \beta) u(t_{[q_i]+1}) \right] \frac{(q t)^{\gamma}}{2}
\cdot \left[ k_2(t, s) u(s) \right]_{s=q_i}^{t_{[q_i]}} + o (h^{3+\gamma})
\]
\[
\cdot \left[ k_2(t, s) u(s) \right]_{s=q_i}^{t_{[q_i]}} + o (h^{3+\gamma})
\]
\[
\cdot \int_{[q_i]}^t \frac{\partial^2}{\partial s^2} [k_2(t, s) u(s)] ds + O (h^2) = T_3(t_j)
\]
\[ h^2 + \frac{(q_i - t_{[|q|]})^2 - h^2}{12} \]
\[ \int_{t_{[|q|]}}^{\tau_{[|q|]} \{ k_2 (t, s) u (s) s' \} ds + O (h^3) = T_3 (t_i) + h^2 + O (h^3) , \]
\[ (29) \]

where
\[ T_1 (t_i) = \left[ k_1 (t_i, s) u (s) \right]_{s=0}^{\infty} \zeta (-\alpha - 1) , \]
\[ T_2 (t_i) = \left[ k_2 (t_i, s) u (s) \right]_{s=0}^{\infty} \zeta (-\gamma - 1) , \]
\[ T_3 (t_i) = -\beta \frac{(1 - \beta)}{2} u'' (q_i) (q_i) k_2 (t_i, q_i) \]
\[ \cdot \left( \beta u (t_{[|q|]}) + (1 - \beta) u (t_{[|q|] + 1}) \right) + \frac{1}{12} \]
\[ \cdot \int_{t_{[|q|]}}^{\tau_{[|q|]} \{ k_2 (t, s) u (s) s' \} ds . \]

The discrete forms of (1) are obtained:
\[ u (t_0) = g (t_0) ; \]
\[ u (t_i) = g (t_i) + h \sum_{k=1}^{i-1} t_{[|q|]} k_1 (t_i, t_k) u (t_k) + h^2 k_1 (t_i, t_i) \]
\[ \cdot u (t_i - \zeta (-\alpha) h^{1+\alpha} k_1 (t_i, t_0) u (t_0) + \]
\[ + h \sum_{k=1}^{i-1} t_{[|q|]} k_2 (t_i, t_k) u (t_k) + h^2 k_2 (t_i, t_{[|q|]}) \]
\[ \cdot u (t_{[|q|]}) - \zeta (-\gamma) h^{1+\gamma} k_2 (t_i, t_0) u (t_0) + \]
\[ + \frac{q_i - t_{[|q|]}}{2} \left[ r_{[|q|]} k_3 (t_i, t_{[|q|]}) u (t_{[|q|]}) + \right] \]
\[ \cdot \left( \beta u (t_{[|q|]}) + (1 - \beta) u (t_{[|q|] + 1}) \right) \]
\[ + E_{1,i} + E_{2,i} + E_{3,i} , \]
\[ i = 1, \ldots, N . \]

Let \( u_i \) be the approximate solution of \( u(t_i) \) and ignore the error function; (32) becomes
\[ u_0 = g (t_0) ; \]
\[ u_i = g (t_i) + h \sum_{k=1}^{i-1} t_{[|q|]} k_1 (t_i, t_k) u_k + \]
\[ h \sum_{k=1}^{i-1} t_{[|q|]} k_2 (t_i, t_k) u_k \]
\[ - \zeta (-\alpha) h^{1+\alpha} k_1 (t_i, t_0) u_0 + \]
\[ + h \sum_{k=1}^{i-1} t_{[|q|]} k_2 (t_i, t_k) u_k \]
\[ + \frac{q_i - t_{[|q|]}}{2} \left[ r_{[|q|]} k_3 (t_i, t_{[|q|]}) u (t_{[|q|]}) + \right] \]
\[ \cdot \left( \beta u (t_{[|q|]}) + (1 - \beta) u (t_{[|q|] + 1}) \right) \]
\[ + \frac{q_i - t_{[|q|]}}{2} \left[ r_{[|q|]} k_3 (t_i, t_{[|q|]}) u (t_{[|q|]}) + \right] \]
\[ \cdot \left( \beta u (t_{[|q|]}) + (1 - \beta) u (t_{[|q|] + 1}) \right) . \]

The iterative algorithm is built to solve (33).

**Iterative Algorithm**

**Step 1.** Take sufficiently small \( \epsilon > 0 \) and set \( u_0 = g (t_0) ; i = 1 ; \)

**Step 2.** Let \( u_i^0 = u_{i-1}, m = 0 \); then compute \( u_i^{m+1} (i \leq N) \) as follows:
\[ u_i^{m+1} = g (t_i) + h \sum_{k=1}^{i-1} t_{[|q|]} k_1 (t_i, t_k) u_k^m + \]
\[ \frac{h}{2} t_{[|q|]} k_2 (t_i, t_i) u_i^m \]
\[ - \zeta (-\alpha) h^{1+\alpha} k_1 (t_i, t_0) u_0 + \]
\[ + \sum_{k=1}^{i-1} t_{[|q|]} k_2 (t_i, t_k) u_k^m \]
\[ + \frac{h}{2} t_{[|q|]} k_2 (t_i, t_{[|q|]}) u_i^m \]
\[ + \frac{q_i - t_{[|q|]}}{2} \left[ r_{[|q|]} k_3 (t_i, t_{[|q|]}) u (t_{[|q|]}) + \right] \]
\[ \cdot \left( \beta u (t_{[|q|]}) + (1 - \beta) u (t_{[|q|] + 1}) \right) ; \]

**Step 3.** If \( |u_i^{m+1} - u_i^m| \leq \epsilon \), set \( u_i = u_i^{m+1} \) and \( i = i + 1 \) and return to Step 2; else, let \( m = m + 1 \) and return to Step 2.

**4. The Existence and Uniqueness of the Solution for the Approximate Equation**

Now, we prove the existence and uniqueness of the solution for the approximate equation. We first introduce the following lemma.

**Lemma 5** (see [23]). Suppose that the sequence \( \{ w_n \}, n = 0, \ldots, N, \) satisfies
\[ |w_n| \leq h \sum_{k=1}^{n} B_k |w_k| + A , \quad 0 \leq n \leq N , \]
where \( A \) and \( B_k (k = 1, \ldots, N) \) are nonnegative constants. Let \( h \max_{1 \leq k \leq N} B_k \leq 1/2 \) with \( h = 1/N \); then we can derive
\[ \max_{0 \leq k \leq N} |w_n| \leq A \exp \left( 2h \sum_{k=1}^{N} B_k \right) . \]

**Theorem 6.** Assume that \( h \) is sufficiently small; then the solution of (34) is existent uniquely, and the algorithm converges at a geometrical rate.

**Proof.** From the nature of the delay function \( \theta t \), we discuss the existence and uniqueness of the solution for the approximate equation under two situations.
First, we prove that the solution of (34) is existent under two situations.

(1) One situation is $[qi] + 1 = i$; that is, when $i < 1/(1 - q)$, we can easily obtain

$$
\left| u_i^{m+1} - u_i^m \right| \leq \frac{h_i}{2} \left| k_1(t_i, t_i) \right| \left| u_i^m - u_i^{m-1} \right|
+ \frac{q t_i - t_{[qi]}}{2} \left( k_2(t_i, t_{[qi]}) \right) \left| z_{[qi]} \right| \\
\cdot \left| t_{[qi]} \right| C \left[ z_{[qi]} \right] + C \frac{q t_i - t_{[qi]}}{2} \left[ t_{[qi]} \right] \left| z_{[qi]} \right| \\
+ \left( q t_i \right)^\gamma \left( \beta_i \right) \left| z_{[qi]} \right| + \left( 1 - \beta_i \right) \left| z_{[qi] + 1} \right|,
$$

(37)

with $t_i = ih$, $i = 1, \ldots, N$. Let $\mu = \min(\alpha, \gamma)$.

(1) The first situation is $[qi] + 1 = i$, that is, when $i < 1/(1 - q)$. Let $h B_j \leq 1/2$ for a sufficiently small $h$; then (40) can be written as follows:

$$
\left| z_i \right| \leq C \sum_{j=1}^{[qi]-1} h t_i^\mu \left| z_j \right| + C h t_i^\mu \left| z_{[qi]} \right| + C h t_i^\mu \left| z_i \right|
+ C \sum_{j=1}^{[qi]-1} h t_i^\mu \left| z_j \right| + C h t_i^\mu \left| z_{[qi]} \right| + C h t_i^\mu \left| z_i \right|
+ r_{[qi]} \left( \beta_i \right) \left| z_{[qi]} \right| + \left( 1 - \beta_i \right) \left| z_{[qi] + 1} \right|,
$$

(40)

with $t_i = ih$, $i = 1, \ldots, N$. Let $\mu = \min(\alpha, \gamma)$.

(2) The other situation is $[qi] + 1 < i$; that is, when $i > 1/(1 - q)$, then

$$
\left| u_i^{m+1} - u_i^m \right| \leq \frac{h_i}{2} \left| k_1(t_i, t_i) \right| \left| u_i^m - u_i^{m-1} \right|
\leq C h t_i^\mu \left| z_i \right|
$$

(38)

Let $C(h/2)^\mu t_i^\mu \leq 1/2$ for a sufficiently small $h$; then $[ui]^{m+1} - u_i^m \leq (1/2)[u_i^m - u_i^{m-1}]$ holds.

With the discussion of the above two situations, one can conclude that the iterative algorithm is convergent geometrically, and the limit is the solution of (34); therefore, the solution of (34) is existent.

Next, we prove that the solution of (34) is unique. If $u_i$ and $v_i$ are solutions of (34), the difference can be represented as $z_i = |u_i - v_i|, 1 \leq i \leq N$, and

$$
|z_0| = 0;
$$

$$
|z_i| = \sum_{j=1}^{[qi]-1} h t_j^\mu \left| k_1(t_j, t_j) \right| \left| z_j \right| + \frac{h_i}{2} \left| k_1(t_i, t_i) \right| \left| z_i \right|
+ \left\{ \begin{array}{ll}
\zeta (-\alpha) h^{1+\alpha} \left| k_1(t_i, t_0) \right| \left| z_0 \right| \\
+ \sum_{j=1}^{[qi]} h t_j^\mu \left| k_2(t_j, t_j) \right| \left| z_j \right| + \frac{h_i}{2} \left| k_2(t_i, t_{[qi]}) \right| \left| z_{[qi]} \right|
\end{array} \right.
$$

(39)

and $|k_1(t, s)| \leq C (i = 1, 2)$, because $k_1(t, s)$ are continuous on bounded domains, with $qt_i - t_{[qi]} \leq h$; then

$$
|z_0| = 0;
$$

$$
|z_i| \leq C \sum_{j=1}^{[qi]-1} h t_j^\mu \left| z_j \right| + \frac{h_i}{2} \left| z_i \right| + \sum_{j=1}^{[qi]-1} h t_j^\mu C \left| z_j \right| + \frac{h_i}{2} \left| t_{[qi]} \right| C \left| z_{[qi]} \right|
+ \left( q t_i \right)^\gamma \left( \beta_i \right) \left| z_{[qi]} \right| + \left( 1 - \beta_i \right) \left| z_{[qi] + 1} \right|,
$$

(40)
(2) The second situation is $[qi] + 1 < i$, that is, when $i > 1/(1 - q)$. Let $hB_j \leq 1/2$ for a sufficiently small $h$; then (40) can be written as

$$|z_i| \leq C \sum_{j=1}^{[qi]} h_{t_j}^\mu |z_j| + C h_{t_{[qi]}}^\mu |z_{[qi]}|$$

$$+ C h_{t_{[qi]+1}}^\mu |z_{[qi]+1}| + \sum_{j=[qi]+1}^{i-1} h_{t_j}^\mu |z_j| + C h_{t_j}^\mu |z_j|$$

$$+ C \sum_{j=[qi]+1}^{i-1} h_{t_j}^\mu |z_j| + C h_{t_j}^\mu |z_j| + C$$

$$\cdot \frac{h}{2} \left[ t_{[qi]}^\mu |z_{[qi]}| + t_{[qi]}^\mu (\beta_i |z_j| + (1 - \beta_i) |z_{[qi]+1}|) \right]$$

$$= 2Ch \sum_{j=1}^{[qi]} t_{t_j}^\mu |z_j| + \left( 2Ch_{t_{[qi]}}^\mu + C h_{t_{[qi]}}^\mu \beta_i \right) |z_{[qi]}|$$

$$+ \left( Ch_{t_{[qi]}}^\mu + C h_{t_{[qi]}}^\mu (1 - \beta_i) \right) |z_{[qi]+1}|$$

$$+ \sum_{j=[qi]+1}^{i-1} h_{t_j}^\mu |z_j| + C h_{t_j}^\mu |z_j| ;$$

namely,

$$|z_i| \leq \sum_{j=1}^{i} B_j |z_j|$$

$$= 2Ch \sum_{j=1}^{[qi]-1} t_{t_j}^\mu |z_j| + \left( 2Ch_{t_{[qi]}}^\mu + C h_{t_{[qi]}}^\mu \beta_i \right) |z_{[qi]}|$$

$$+ \left( Ch_{t_{[qi]}}^\mu + C h_{t_{[qi]}}^\mu (1 - \beta_i) \right) |z_{[qi]+1}|$$

$$+ \sum_{j=[qi]+1}^{i-1} h_{t_j}^\mu |z_j| + C h_{t_j}^\mu |z_j| .$$

And

$$B_j = \begin{cases} 2C \sum_{j=1}^{[qi]-1} t_{t_j}^\mu, & j = 1, \ldots, [qi] - 1, \\ 2Ch_{t_{[qi]}}^\mu + C h_{t_{[qi]}}^\mu \beta_i, & j = [qi], \\ Ch_{t_{[qi]}}^\mu + C h_{t_{[qi]}}^\mu (1 - \beta_i), & j = [qi] + 1, \\ \sum_{j=[qi]+1}^{i-1} t_{t_j}^\mu, & j = [qi] + 1, \ldots, i - 1, \\ C h_{t_{t_j}}^\mu, & j = i. \end{cases}$$

Based on Lemma 5, $|z_i| = 0$ with $A = 0$ and the solution of the discrete equation (34) is unique. The proof of Theorem 6 is completed.

### 5. The Error Estimation

In this section, we give the error estimate between the approximation solution and the exact solution of (1).

**Theorem 7.** Let $u(t)$ be the exact solution of (1); the kernel functions $k_i(t, s) = s^\gamma k_i(t, s), k_i(t, s) = s^\gamma k_i(t, s), -1 < \alpha, \gamma \leq 0$, and the functions $k_i(t, s)$ and $k_i(t, s)$ are continuous in the domains $D = \{(t, s) : 0 \leq s \leq t \leq T\}$ and $D = \{(t, s) : 0 \leq s \leq \theta(t), t \in I\}$, respectively. Then there is a positive constant $b$ independent of $h$ such that $|k_i(t, s)| \leq C$ and $|e_i| = |u(t_i) - u_i|$ ($i = 0, \ldots, N$) have the following estimation:

$$\max_{1 \leq i \leq N} |e_i| \leq bh^{2+\mu}.$$

**Proof.** From (26) and (27), we have

$$E_{1j} = \frac{|k_1(t_j, s) u(s)|^\alpha}{2!} (-\alpha - 1) h^{2+\alpha}$$

$$+ \frac{|k_1(t_j, s) u(s)|^\alpha}{2!} (-\alpha - 2) h^{3+\alpha}$$

$$+ O(h^{4+\alpha}) = T_1(t_i) h^{2+\alpha} + O(h^{3+\alpha}),$$

$$E_{2j} = \frac{|k_2(t_j, s) u(s)|^\alpha}{2!} (-\alpha - 1) h^{2+\gamma}$$

$$+ \frac{|k_2(t_j, s) u(s)|^\alpha}{2!} (-\alpha - 2) h^{3+\gamma}$$

$$+ O(h^{4+\gamma}) = T_2(t_j) h^{2+\gamma} + O(h^{3+\gamma}).$$

By the trapezoidal formula and Lemma 4, we have

$$E_{3j} = \frac{\beta (1 - \beta)}{2} h^2 u^\tau(q_t) \left( q_t^\tau q_t \right)^\gamma k_2(t, q_t)$$

$$\cdot \left( \beta u_{t_{t_j}} + (1 - \beta_j) u_{t_{t_j+1}} + \frac{(q_{t_j} - t_{t_j})^2}{12} \right)$$

$$+ \sum_{t_{t_j} \in [q_{t_j}, q_{t_j+1}]} Q^\alpha \frac{\partial^2}{\partial s^2} k_2(t, s) u(s) s^\gamma ds + O(h^3) = T_3(t_j) h^2$$

$$+ \frac{(q_{t_j} - t_{t_j})^2 - h^2}{12} \int_{t_{t_j}}^{q_{t_j}} \frac{\partial^2}{\partial s^2} k_2(t, s) u(s) s^\gamma ds$$

$$+ O(h^3) = T_3(t) h^2 + O(h^3),$$

where

$$T_3(t_j) = \frac{\beta (1 - \beta)}{2} h^2 u^\tau(q_t) \left( q_t^\tau q_t \right)^\gamma k_2(t, q_t)$$

$$\cdot \left( \beta u_{t_{t_j}} + (1 - \beta_j) u_{t_{t_j+1}} + \frac{1}{12} \right)$$

$$+ \int_{t_{t_j}}^{q_{t_j}} \frac{\partial^2}{\partial s^2} k_2(t, s) u(s) s^\gamma ds.$$
Then, we have

\[
\begin{align*}
    u(t) &= g(t) + Q_N + Q_N^* + T_1(t) h^{2\alpha} \\
    &+ T_2(t) h^{2\gamma} + T_3(t) h^2 + O(h^2),
\end{align*}
\]

and the analysis is the same as (41) and (44); we have

\[
|e_i| \leq A + \sum_{j=1}^{i} B_j |e_j|, \quad 1 \leq i \leq N.
\]

From Lemma 5, there is a positive constant \( b \) independent of \( h \) such that

\[
\max_{1 \leq i \leq N} |e_i| \leq bh^{2\mu}.
\]

The proof of Theorem 7 is completed.

### 6. Error Asymptotic Expansion and Extrapolation Algorithm

In this section, we present the main theoretical result of the error asymptotic expansions and the relevant extrapolation algorithm.

**Theorem 8.** Based on the conditions of Theorem 7, there exist continue functions \( \hat{T}_i(t) \), \( i = 1, 2, 3 \) satisfying the asymptotic expansion

\[
\begin{align*}
    u_i &= u(t) + \hat{T}_1(t) h^{2\alpha} + \hat{T}_2(t) h^{2\gamma} + \hat{T}_3(t) h^2 \\
    &+ O(h^{3\mu}), \quad -1 < \alpha < 0, \quad -1 < \gamma \leq 0.
\end{align*}
\]

**Proof.** Suppose that \( \hat{T}_k(t), \ k = 1, 2, 3 \) satisfies the auxiliary delay equations:

\[
\begin{align*}
    \hat{T}_k(t) &= T_k(t) + \int_0^t s^\alpha k_1(s) u(s) \hat{T}_k(s) \, ds \\
    &+ \int_0^t s^\alpha k_2(s, t) u(s) \hat{T}_k(s) \, ds,
\end{align*}
\]

and \( \hat{T}_k(t) \), \( i = 1, \ldots, N \), satisfy the approximation equations:

\[
\begin{align*}
    \hat{T}_k(t_i) &= T_k(t_i) + \sum_{j=1}^{i-1} h \sum_{j=1}^{i-1} t^{\alpha} k_1(t_i, t_j) u_j \hat{T}_k(t_j) + \frac{h}{2} \\
    &+ \sum_{j=1}^{i-1} h \sum_{j=1}^{i-1} t^{\alpha} k_1(t_i, t_j) u_j \hat{T}_k(t_j) - \zeta (-\alpha) h^{1+\alpha} u_k(t_i, t_0) \\
    + T_1(t_i) h^{2\alpha} + T_2(t_i) h^{2\gamma} + T_3(t_i) h^2 + O(h^{3\mu})
\end{align*}
\]

Let

\[
    A = \left| T_1(t) h^{2\alpha} + T_2(t) h^{2\gamma} + T_3(t) h^2 + O(h^{3\mu}) \right|
\]

\[
= O(h^{2\mu}),
\]

\[
\begin{align*}
    \max_{1 \leq i \leq N} \left| \hat{T}_k(t_i) - T(t_i) \right| &\leq Ch^{2\mu}.
\end{align*}
\]
Let
\[ E_i = e_i - (T_1(t_i) h^{2+\alpha} + T_2(t_i) h^{2+\gamma} + T_3(t_i)) h^2. \] (60)

We have
\[ E_i = \sum_{j=1}^{i-1} h^{2+\gamma} k_1(t_j, t_j) u_j E_j + \frac{h^{2+\alpha}}{2} k_1(t_i, t_i) u_i E_i \]
\[ - \zeta (\alpha) h^{2+\alpha} u_0 k(t_i, t_0) E_0 + \sum_{j=1}^{i-1} h^{2+\gamma} k_2(t_i, t_j) u_j E_j \]
\[ + \frac{h^{2+\gamma}}{2} k_2(t_i, t_{[i]}(u_{[i]} \tilde{E}_{[i]} - \zeta (\gamma)) \]
\[ + (h^{2+\gamma}) k_2(t_i, t_{[i]}) (\beta u_{[i]} + (1 - \beta) u_{[i+1]}) \]
\[ + (1 - \beta) E_{[i+1]} \]
(61)

From Lemma 1, there exists a constant \( d \) such that
\[ \max_{1 \leq i \leq N} |E_i| \leq d h^{3+\mu}. \] (62)

The asymptotic expansion is (56).

Based on Theorem 8, we adopt the Richardson extrapolation to improve the accuracy.

**Extrapolation Algorithm**

**Step 1.** Assume that \( \mu = \min(\alpha, \gamma) = \alpha, \) and halve the step length to obtain
\[ u_i^{h/2} = u(t_i) + \tilde{T}_1(t_i) \left( \frac{h}{2} \right)^{2+\alpha} + \tilde{T}_2(t_i) \left( \frac{h}{2} \right)^{2+\gamma} \]
\[ + \tilde{T}_3(t_i) \left( \frac{h}{2} \right)^2 + O \left( \left( \frac{h}{2} \right)^{3+\alpha} \right). \] (63)

By combining (56) with (63), we get
\[ u_i^{h} = \frac{2^{2+\alpha} u_i^{h/2} - u_i^{h}}{2^{2+\alpha} - 1} \]
\[ = u(t_i) + \tilde{T}_2(t_i) h^{2+\gamma} + \tilde{T}_3(t_i) h^2 + O \left( h^{3+\alpha} \right). \] (64)

**Step 2.** We implement Richardson \( h^{2+\gamma} \) extrapolation:
\[ u_i^{1/2} = u(t_i) + \tilde{T}_2(t_i) \left( \frac{h}{2} \right)^{2+\gamma} + \tilde{T}_3(t_i) \left( \frac{h}{2} \right)^2 + O \left( \left( \frac{h}{2} \right)^{3+\alpha} \right). \] (65)

Combining (64) with (65), we get
\[ u_i^{2h} = \frac{2^{2+\gamma} u_i^{h/2} - u_i^{h}}{2^{2+\gamma} - 1} = u(t_i) + \tilde{T}_3(t_i) h^2 + O \left( h^{3+\alpha} \right). \] (66)

Moreover, a posterior asymptotic error estimate
\[ |u_i^{h/2} - u_i| = \frac{2^{2+\alpha} u_i^{h/2} - u_i^{h}}{2^{2+\alpha} - 1} = u(i) + \frac{u_i^{h} - u_i^{h/2}}{2^{2+\alpha} - 1} \]
\[ \leq \frac{2^{2+\alpha} u_i^{h/2} - u_i^{h}}{2^{2+\alpha} - 1} = u(i) + \frac{u_i^{h} - u_i^{h/2}}{2^{2+\alpha} - 1} \]
\[ + O \left( h^2 \right). \] (67)

The error \( u_i^{h/2} - u(i) \) is bounded by \( (u_i^{h} - u_i^{h/2})/(2^{2+\alpha} - 1) \), which is essential to construct adaptable algorithms.

**7. Numerical Experiments**

In this section, three examples will be presented to show the efficiency of the quadrature method. We design a set of grids on the interval \( I \); the absolute error is denoted by
\[ e_h(t_i) = |u(t_i) - u_i| \] (68)
with \( i = 1, \ldots, N \); \( u(t_i) \) and \( u_i \) are the exact solution and the approximate solution at \( t = t_i \), respectively. Set \( h = 1/N \), and the convergence order is defined by
\[ \text{Rate} = \frac{\log \left( e_h/e_{h/2} \right)}{\log 2}. \] (69)

**Example 1.** Consider the following equation:
\[ u(t) = g(t) - \int_0^t (t - s)^\alpha u(s) \, ds, \quad t \in [0, T], \] (70)
where \( T = 1, -1 < \alpha < 0 \), and the analytical solution is \( u(t) = \cos(t) \).

The numerical results at the point \( t = 1 \) with the partitions \( N = 2^1, 2^2, 2^3, 2^4, 2^5, 2^6 \) are addressed in Table 1. By the mechanical quadrature method and iterative method, the obtained absolute errors \( e_h \) with more refined partitions show a more accurate approximate solution, and the convergent rate is adjacent to \((3/2)-\)order. Based on the Richardson \( h^{2+\alpha} \) extrapolation, the errors are closer to the exact solution, and the convergence order is improved to 2-order. The posteriori error is also achieved. The two kinds of convergence orders are consistent with the theoretical analysis.
Table 1: Numerical results at point $t = 1$ of Example 1.

<table>
<thead>
<tr>
<th>$N$</th>
<th>$\varepsilon_h$</th>
<th>Rate</th>
<th>$R(-h^{2.5})$ extrapolation</th>
<th>Rate</th>
<th>Posterior errors</th>
</tr>
</thead>
<tbody>
<tr>
<td>$2^4$</td>
<td>9.27e-04</td>
<td>—</td>
<td>—</td>
<td>—</td>
<td>—</td>
</tr>
<tr>
<td>$2^5$</td>
<td>3.22e-04</td>
<td>$2.53$</td>
<td>9.65e-06</td>
<td>—</td>
<td>3.31e-04</td>
</tr>
<tr>
<td>$2^6$</td>
<td>1.12e-04</td>
<td>$2.52$</td>
<td>2.24e-06</td>
<td>$2.10$</td>
<td>1.14e-04</td>
</tr>
<tr>
<td>$2^7$</td>
<td>3.93e-05</td>
<td>$2.51$</td>
<td>5.32e-07</td>
<td>$2.08$</td>
<td>3.99e-05</td>
</tr>
<tr>
<td>$2^8$</td>
<td>1.38e-05</td>
<td>$2.51$</td>
<td>1.28e-04</td>
<td>$2.06$</td>
<td>1.39e-05</td>
</tr>
</tbody>
</table>

Table 2: Numerical results of Example 2.

<table>
<thead>
<tr>
<th>$t$</th>
<th>$\varepsilon_h$</th>
<th>$\varepsilon_{h/2}$</th>
<th>$\varepsilon_{h/4}$</th>
<th>Posterior error</th>
</tr>
</thead>
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<tr>
<td>0.2</td>
<td>3.89e-06</td>
<td>7.84e-07</td>
<td>1.77e-07</td>
<td>1.69e-06</td>
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<tr>
<td>0.4</td>
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<td>1.01e-06</td>
<td>2.28e-07</td>
<td>2.20e-06</td>
</tr>
<tr>
<td>0.6</td>
<td>5.67e-06</td>
<td>1.14e-06</td>
<td>2.55e-07</td>
<td>2.48e-06</td>
</tr>
<tr>
<td>0.8</td>
<td>5.98e-06</td>
<td>1.19e-06</td>
<td>2.68e-07</td>
<td>2.61e-06</td>
</tr>
<tr>
<td>1.0</td>
<td>6.07e-06</td>
<td>1.21e-06</td>
<td>2.71e-07</td>
<td>2.66e-06</td>
</tr>
</tbody>
</table>

Table 3: Numerical results of Example 3.

<table>
<thead>
<tr>
<th>$t$</th>
<th>$\varepsilon_h$</th>
<th>$\varepsilon_{h/2}$</th>
<th>$\varepsilon_{h/4}$</th>
<th>Posterior errors</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.2</td>
<td>1.43e-06</td>
<td>9.29e-07</td>
<td>7.18e-07</td>
<td>2.75e-07</td>
</tr>
<tr>
<td>0.4</td>
<td>1.23e-05</td>
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<td>2.90e-05</td>
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<td>3.25e-04</td>
<td>7.63e-05</td>
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</table>

Example 2. Consider the following delay Volterra integral equation with weakly singular kernel as

$$u(t) = g(t) - \int_0^t s^{-1/2} u(s) \, ds + \int_0^t (t^2 + s) u(s) \, ds,$$  \hspace{1cm} (71)

with $q = 0.9$, $T = 1$, and the initial value $u(0) = g(0) = 0$, and $g(t)$ is determined by the analytical solution $u(t) = t + 1$.

We denote the approximate solution by $u_h$, with $h = 1/32$. The error results at some interior points in the interval $t \in [0, 1]$ with different partition are listed in Table 2. The comparison of the exact solution and the approximate solution with partition $N = 2^3$ is shown in Figure 1. It is obvious that this paper provides a high accuracy algorithm for the weakly singular Volterra integral equation with proportional delays.

Example 3. Consider the following delay Volterra integral equation with weakly singular kernel as

$$u(t) = g(t) - \int_0^t s^{-1/4} u(s) \, ds + \int_0^t (t^2 + s) u(s) \, ds,$$  \hspace{1cm} (72)

with $q = 0.9$, $T = 1$, and the initial value $u(0) = g(0) = 0$, and $g(t)$ is determined by the analytical solution $u(t) = t + 1$.

The absolute error $|\epsilon_h(t)| = |u(t) - u_h(t)|$, $t \in [0, 1]$, and the posteriori errors at some interior points are listed in Table 3 when $N = 2^3$, $2^4$, $2^5$, which can observe that the error results decay quickly with the increasing of $N$. The absolute errors between the exact solution and the approximate solution with partition $N = 2^2$, $2^3$ are shown in Figure 2, which indicate that our algorithm is effective.

8. Conclusion

In this paper, we use the mechanical quadrature method and Romberg extrapolation for weakly Volterra integral equation with proportional delays. Most papers analyze the delay Volterra integral equation with continuous kernels; the study for the delay Volterra integral equation with weakly singular kernels still faces a real challenge for the relevant researchers in both numerical computation and theoretical analysis. The improved Gronwall inequality is adopted to prove the existence and uniqueness of the solution of the original equation. At the same time, the discrete Gronwall equality and iterative method are adopted to prove the existence and uniqueness of the solution of the approximate equation. Moreover, according to the error asymptotic expansion of the mechanical quadrature method and the extrapolation, a high order of accuracy can be achieved and a posterior error estimation can be obtained. Both the theoretical analysis and the numerical examples show that the presented method is efficient.

Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.
Complexity

Conflicts of Interest
The authors declare that there are no conflicts of interest regarding the publication of this article.

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