Travelling-Wave Solution of Volterra Lattice by the Optimal Homotopy Analysis Method

Qi Wang
Department of Applied Mathematics, Shanghai University of Finance and Economics, Shanghai 200433, PR China

Reprint requests to Q. W.; E-mail: wangqee@gmail.com

Received August 17, 2011

The travelling-wave solution of the Volterra lattice has been constructed by the optimal homotopy analysis method. The optimal method used here contains three auxiliary convergence-control parameters to adjust and control the convergence region of the solution. By minimizing the averaged residual error, the optimal convergence-control parameters can be obtained, which give much better approximations than those given by the usual homotopy analysis method. The obtained results show that the optimal homotopy analysis method is also very efficient for differential-difference equations.

Key words: Volterra Lattice; Optimal Homotopy Analysis Method; Travelling-Wave Solution.

PACS numbers: 02.30.Xx; 02.30.Mv; 02.60.Lj

1. Introduction

For better understanding the meaning of nonlinear partial differential equations (PDEs), it is crucial to search for their exact analytic solutions. The exact solutions, if available, of those nonlinear PDEs can facilitate the verification of numerical solvers and aid in the stability analysis of solutions. Many powerful methods have been generalized to construct solutions of PDEs such as the inverse scattering method [1, 2], the Bäcklund transformation [3, 4], the Darboux transformation [5], the Lie group method [6], the Hirota method [7], etc. Among them, the homotopy analysis method (HAM) which is based on the idea of homotopy in topology, is a general analytic method for nonlinear problems [8]. Unlike the traditional methods (for example, perturbation techniques and so on), HAM contains many auxiliary parameters which provide us with a simple way to adjust and control the convergence region and rate of convergence of the series solution, and it has been successfully employed to solve explicit analytic solutions for many types of nonlinear problems [9 – 15].

However, unlike the widely applications in continuous cases of HAM, there is very few work on constructing solutions for differential-difference equations (DDEs) [16, 17] which are semi-discretized with some (or all) of their spacial variables while time is usually kept continuous. They play an important role in numerical simulations of nonlinear PDEs, queuing problems, and discretization in solid state and quantum physics [18 – 22]. Moreover as illustrated in [23], the usual HAM has only one convergence-control parameter $c_0$ but unfortunately the curves for the convergence-control parameter (i.e. $c_0$-curves) can not tell us which value of $c_0 \in \mathbb{R}$ gives the fastest convergent series. Recently, to overcome this shortcoming, Liao [23] proposed an optimal HAM with more than one convergence-control parameter. In this optimal method, Liao also introduced the so called averaged residual error to get the optimal convergence-control parameters efficiently, which compared with the exact square residual error can greatly decrease the computing time (CPU time) and also give a good enough approximation. In general, the optimal HAM can greatly modify the convergence of homotopy series solution for PDEs [23, 24].

The aim of this paper is to extend the optimal HAM to consider the travelling-wave solutions of the Volterra lattice. The method proposed here contains three convergence-control parameters to accelerate the convergence of homotopy series solution. The optimal convergence-control parameters can be determined by minimizing the averaged residual error. The results ob-
tained in this paper show that the optimal HAM is also very efficient for DDEs. The solutions obtained by the optimal HAM give much better approximations and convergence much faster than those obtained by the usual HAM.

2. Optimal HAM for Volterra Lattice

The Volterra lattice system [25]
\[
\frac{\partial a_n}{\partial t} = a_n (b_n - b_{n-1}), \\
\frac{\partial b_n}{\partial t} = b_n (a_{n+1} - a_n),
\]
(1)
in which \(a_n(t)\) and \(b_n(t)\) are functions of continuous variable \(t\) and discrete variable \(n \in \mathbb{Z}_+\). The Volterra type equations are discretizations of the Korteweg–de Vries (KdV) and modified KdV equations [26, 27]. The exact solutions of (1) have been obtained by the tanh method [28]. Set
\[
a_n = u_{2n-1}, \quad b_n = u_{2n},
\]
(2)
the Volterra lattice system (1) become the more convenient form of a single polynomial equation
\[
\frac{\partial u_n}{\partial t} = u_n (u_{n+1} - u_{n-1}).
\]
(3)

To find the travelling-wave solutions of (3), it is convenient to introduce a new dependent variable \(w_n(\xi_n)\) defined by
\[
u_n(t) = a \ w_n(\xi_n),
\]
(4)
where \(\xi_n = qt + gn\), \(a\) is the amplitude, \(k\) is the wave speed, and \(q\) is an arbitrary non-zero constant. Substitution of \(u_n\) given by (4) into (3) gives
\[
k w_n' = a w_n (w_{n+1} - w_{n-1}),
\]
(5)
where the prime denotes the differentiation with respect to \(\xi_n\). Assume that the dimensionless wave solution \(w_n(\xi_n)\) arrives its maximum at the origin. Obviously, \(w_n(\xi_n)\) and its derivatives tend to zero when \(\xi_n \to \infty\). Thus, the boundary conditions of the travelling-wave solutions are
\[
w_n(0) = 1, \quad w_n(\infty) = 0, \quad w_n'(\infty) = 0.
\]
(6)

According to (5) and the boundary conditions (6), the travelling-wave solution can be expressed by
\[
w_n(\xi_n) = \sum_{m=1}^{+\infty} d_m e^{-m\xi_n},
\]
(7)
where \(d_m (m = 1, 2, \ldots)\) are coefficients to be determined. Moreover, according to the rule of solution expression denoted by (7) and the boundary conditions (6), it is natural to choose \(w^*(\xi_n) = e^{-\xi_n}\) as the initial approximation of \(w_n(\xi_n)\).

Let \(p \in [0, 1]\) denote the embedding parameter, \(c_0 \neq 0\) denote an auxiliary parameter, called the convergence-control parameter, and \(\phi_n(\xi_n; p)\) denote a kind of continuous mapping of \(w_n(\xi_n)\), respectively. We can construct such a generalized homotopy
\[
(1 - C(p)) \mathcal{L}[\phi_n(\xi_n; p) - w^*(\xi_n)] = c_0 B(p) \mathcal{N}[\phi_n(\xi_n; p)],
\]
(8)
where
\[
\mathcal{L}[\phi_n(\xi_n; p)] = \left( \frac{\partial^2}{\partial \xi_n^2} + \frac{\partial}{\partial \xi_n} \right) \phi_n(\xi_n; p)
\]
is an auxiliary linear operator, with the property
\[
\mathcal{L}[C_1 e^{-\xi_n/2} + C_2] = 0,
\]
(10)
where \(C_1\) and \(C_2\) are constants. From (5), we define the nonlinear operator
\[
\mathcal{N}[\phi_n(\xi_n; p)] = k \frac{\partial \phi_n(\xi_n; p)}{\partial \xi_n} - a \phi_n(\xi_n; p) (\phi_{n+1}(\xi_n+1; p) - \phi_{n-1}(\xi_n-1; p)).
\]
(11)
In (11), \(B(p)\) and \(C(p)\) are the so-called deformation functions satisfying
\[
B(0) = C(0) = 0, \quad B(1) = C(1) = 1,
\]
(12)
whose Taylor series
\[
B(p) = \sum_{m=1}^{+\infty} v_m p^m, \quad C(p) = \sum_{m=1}^{+\infty} \sigma_m p^m
\]
exist and are convergent for \(|p| < 1\).

Then when \(p = 0\), according to the definition of \(\mathcal{L}\) and \(w^*(\xi_n)\), it is obvious that \(\phi_n(\xi_n; 0) = w^*(\xi_n)\). When \(p = 1\), according to the definition (11), (8) is equivalent to the original (3), provided \(\phi_n(\xi_n; 1) = w_n(\xi_n)\). Thus, as \(p\) increases from 0 to 1, the solution \(\phi_n(\xi_n; p)\)
varies (or deforms) continuously from the initial guess $w^*(\xi_n)$ to the solution $w_n(\xi_n)$ of (3).

According to [23], there are an infinite number of deformation functions satisfying the properties (12) and (13). And in theory, the more convergence-control parameters are used, the better approximation one should obtain by this generalized HAM. But for the sake of computation efficiency, we just use the following one-parameter deformation functions:

$$B(c_1; p) = \sum_{m=1}^{+\infty} v_m(c_1)p^m,$$

$$C(c_2; p) = \sum_{m=1}^{+\infty} \sigma_m(c_2)p^m,$$

where $|c_1| < 1$ and $|c_2| < 1$ are constants, which are convergence-control parameters too, and

$$v_1(c_1) = 1 - c_1, \quad v_m(c_1) = (1 - c_1)c_1^{m-1}, \quad m > 1,$$

$$\sigma_1(c_2) = 1 - c_2, \quad \sigma_m(c_2) = (1 - c_2)c_2^{m-1}, \quad m > 1.$$

The different values of $c_1$ give different paths of $B(c_1; p)$, as shown in Figure 1. Note that $B(c_1; p)$ and $C(c_2; p)$ contain the convergence-control parameters $c_1$ and $c_2$, respectively. So, we have at most three unknown convergence-control parameters $c_0, c_1,$ and $c_2$, which can be used to ensure the convergence of solutions series, as shown later.

Then the so-called zeroth-order deformation equation becomes

$$(1 - C(c_2; p))\mathcal{L}(\phi_n(\xi_n) - w^*(\xi_n)) = c_0B(c_1; p)N[\phi_n(\xi_n); p],$$

and according to (6), it should subject to following boundary conditions:

$$\phi_n(0; p) = 1, \quad \phi_n(\infty; p) = 0,$$

$$\frac{\partial \phi_n(\xi_n; p)}{\partial \xi}\big|_{\xi=\infty} = 0.$$

Obviously, $\phi_n(\xi_n; p)$ is determined by the auxiliary linear operator $\mathcal{L}$, the initial guess $w^*(\xi_n)$ and the convergence-control parameters $c_0, c_1,$ and $c_2$. Note that we have great freedom to choose all of them. Assuming that all of them are so properly chosen that the Taylor series

$$\phi_n(\xi_n; p) = w^*(\xi_n) + \sum_{m=1}^{+\infty} w_{m,n}(\xi_n)p^m,$$

exist and converge at $p = 1$, we have following homotopy series solution

$$w_n(\xi_n) = w^*(\xi_n) + \sum_{m=1}^{+\infty} w_{m,n}(\xi_n),$$

where

$$w_{m,n}(\xi_n) = \frac{1}{m!} \frac{\partial^m \phi_n(\xi_n; p)}{\partial p^m} \big|_{p=0}.$$
Let \( R_1(\xi_n) = \sum_{i=0}^{l} w_{i,n} - a \sum_{i=0}^{j} \xi_i (w_{i,n+1} - w_{i,n-1}) \) (25) and

\[
\chi_m = \begin{cases} 
0 & \text{if } m = 1, \\
1 & \text{if } m > 1.
\end{cases}
\] (26)

Let \( w_m(\xi_n) \) denote a special solution of (23) and \( \mathcal{L}^{-1} \) the inverse operator of \( \mathcal{L} \), respectively. Then we have

\[
w_{m,n}(\xi_n) = \chi_m \sum_{i=1}^{m-1} \sigma_{m-i} (c_2) w_{i,n}(\xi_n) + c_0 \sum_{i=0}^{m-1} v_{m-i} (c_1) \mathcal{L}^{-1}(R_1(\xi_n)).
\] (27)

So the common solution of (23) reads

\[
w_{m,n}(\xi_n) = w_m(\xi_n) + C_1 e^{-\xi_n/2} + C_2.
\] (28)

Under the rule of solution expression (7), \( C_1 = C_2 = 0 \). Then we just need to identify a special solution of (23).

In this way, we can derive \( w_{m,n}(\xi_n) \) for \( m = 0, 1, 2, 3, \ldots \) successively. Then from (4) and (20), we can obtain the travelling-wave solution of the Volterra lattice. At the \( M \)th-order approximation, we have the analytic solution of (3), namely

\[
u_n(x,t) = a w_{m,n}(\xi_n) \approx a W_{m,n}(\xi_n)
\] (29)

In usual HAM [11], there is only one unknown convergence-control parameter \( c_0 \). By the so-called \( c_0 \)-curve, we can determine the possible valid region of \( c_0 \), but unfortunately it can not tell us the optimal value of \( c_0 \) which gives the fastest convergent series. To find the possible optimal values of convergence-control parameters, it usually needs to minimize the exact square residual error [23]. But it is a pity that the calculation needs too much CPU time even if the order of approximation is not very high, and thus is often useless in practice. Moreover, in the expression of the obtained solution, there are three unknown convergence-control parameters \( c_0, c_1, \) and \( c_2 \) to make sure the convergence of the solutions. So in this paper, as in [23], we just determine the possible optimal values of convergence-control parameters by minimizing the averaged residual error

\[
E_M = \frac{1}{L} \sum_{j=0}^{L} \| W_j(j\Delta x), K_M \|^2,
\] (30)

where we usually choose \( \Delta x = 1/10 \), \( L = 20 \), and \( M = 10 \) in this paper.

3. Comparisons of Different Approaches

In this section, we will give optimal homotopy analysis approaches with different numbers of unknown convergence-control parameters, and compare them in details. For ease of comparison, we suppose \( a = t = 1 \), \( k = 0.5 \), \( q = 0.9 \), and take three different cases of unknown convergence-control parameters as in [23].

3.1. Optimal \( c_0 \) in Case of \( c_1 = c_2 = 0 \)

In this case, the method proposed above degenerates into the usual HAM and there is only one unknown convergence-control parameter \( c_0 \). In usual HAM, we can investigate the influence of \( c_0 \) on the series of \( w_{m,n}(\xi_n) \) by means of the so-called \( c_0 \)-curves. As pointed by Liao [11], the valid region of \( c_0 \) is a horizontal line segment. Thus, the valid region of \( c_0 \) in this example as shown in Figure 2 is \(-0.4 < c_0 < 2 \). So we can just determine the possible valid region of \( c_0 \). However, usually the \( c_0 \)-curves can not tell us the optimal value of \( c_0 \) which gives the fastest convergent series, and it is a pity that the exact square residual error needs too much CPU time to calculate even if the order of approximation is not very high, and thus is often useless in practice.

To overcome this shortcoming, in [23], Liao advised to determine the possible optimal value of \( c_0 \) by the minimum of averaged residual error \( E_{10} \), corresponding to the nonlinear algebraic equation \( E_{10} = 0 \). And as shown in [23, 24], the averaged residual error can greatly decrease the CPU time and also give good enough approximation of the optimal convergence-control parameter. Hence, using the symbolic computation software Maple, by minimizing the averaged residual error (30), we can directly get the optimal convergence-control parameter \( c_0 = 0.8658 \). According to Table 1, by means of \( c_0 = 0.8658 \), the value of residual error converges much faster to 0 than the
Table 1. Comparison of averaged residual error given by different \( c_0 \) in case of \( c_1 = c_2 = 0 \).

<table>
<thead>
<tr>
<th>( m ), order of approximation</th>
<th>Optimal value of ( E_m ) when ( c_2 = 1 )</th>
<th>Minimum value of ( E_m ) when ( c_1 = c_2 = 0 )</th>
<th>Value of ( E_m ) when ( c_1 = c_2 = 0 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>0.6283 ( \cdot 10^{-6} )</td>
<td>0.8693 ( \cdot 10^{-12} )</td>
<td>0.2126 ( \cdot 10^{-9} )</td>
</tr>
<tr>
<td>10</td>
<td>0.6283 ( \cdot 10^{-6} )</td>
<td>0.8693 ( \cdot 10^{-12} )</td>
<td>0.2126 ( \cdot 10^{-9} )</td>
</tr>
</tbody>
</table>

Table 2. Comparison of averaged residual error given by different \( c_1 = c_2 \) in case of \( c_0 = -1 \).

<table>
<thead>
<tr>
<th>( m ), order of approximation</th>
<th>Optimal value of ( E_m ) when ( c_1 = c_2 = 0 )</th>
<th>Minimum value of ( E_m ) when ( c_1 = c_2 = 0 )</th>
<th>Value of ( E_m ) when ( c_1 = c_2 = 0 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>0.6283 ( \cdot 10^{-6} )</td>
<td>0.8693 ( \cdot 10^{-12} )</td>
<td>0.2126 ( \cdot 10^{-9} )</td>
</tr>
<tr>
<td>10</td>
<td>0.6283 ( \cdot 10^{-6} )</td>
<td>0.8693 ( \cdot 10^{-12} )</td>
<td>0.2126 ( \cdot 10^{-9} )</td>
</tr>
</tbody>
</table>

Table 3. Comparison of averaged residual error given by different \( c_1 \neq c_2 \) in case of \( c_0 = -1 \).

<table>
<thead>
<tr>
<th>( m ), order of approximation</th>
<th>Optimal value of ( c_1 = c_2 )</th>
<th>Minimum value of ( E_m ) when ( c_1 = c_2 = 0 )</th>
<th>Value of ( E_m ) when ( c_1 = c_2 = 0 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>0.6283 ( \cdot 10^{-6} )</td>
<td>0.8693 ( \cdot 10^{-12} )</td>
<td>0.2126 ( \cdot 10^{-9} )</td>
</tr>
<tr>
<td>10</td>
<td>0.6283 ( \cdot 10^{-6} )</td>
<td>0.8693 ( \cdot 10^{-12} )</td>
<td>0.2126 ( \cdot 10^{-9} )</td>
</tr>
</tbody>
</table>

may not be the best value for the usual HAM. So, even the one-parameter optimal HAM can give much better approximations.

3.2. Optimal \( c_1 = c_2 \) in Case of \( c_0 = -1 \)

Here, we investigate another one-parameter optimal approach in case \( c_0 = -1 \) with the unknown \( c_1 = c_2 \). Using the symbolic computation software Maple too, we can directly get the optimal convergence-control parameter \( c_1 = c_2 = 0.501 \). It is found that the homotopy approximations given by \( c_0 = -1 \) and \( c_1 = c_2 = 0.501 \) converge much faster than those given by the usual HAM in case of \( c_0 = -1 \) and \( c_1 = c_2 = 0 \), as shown in Table 2. This further illustrates that the second one-parameter optimal HAM is as good as the first one mentioned above.

3.3. Optimal \( c_1 \neq c_2 \) in Case of \( c_0 = -1 \)

Here, we investigate the two-parameter optimal approach in the case \( c_0 = -1 \) with the unknown \( c_1 \neq c_2 \). According to above section, we can directly get the optimal convergence-control parameter \( c_1 = 0.63904 \) and \( c_2 = -0.66415 \). As shown in Table 3, it is found that the homotopy approximations given by \( c_0 = -1 \), \( c_1 = 0.63904 \), and \( c_2 = -0.66415 \) converge much faster than those given by the usual HAM in case of \( c_0 = -1 \) and \( c_1 = c_2 = 0 \), too. This further proves that the two-parameter optimal homotopy analysis approach is efficient, too.

4. Conclusions

In this paper, the optimal HAM is extended to construct the travelling-wave solution of the Volterra lat-
The obtained results show that the optimal HAM is also effective for DDEs. Unlike the usual HAM, the optimal HAM used three convergence-control parameters to guarantee the convergence of the homotopy series solution. As shown in this paper, by minimizing the averaged residual error, we can get the possible optimal value of the convergence-control parameters which may give the fastest convergent series. Note that the linear operator $L$ and the nonlinear operator $N$ in (17) are rather general so that the above mentioned optimal HAM can be employed to find the travelling-wave solutions with more fast convergence for different types of equations with strong nonlinearity, such as fractional differential equations, supersymmetric equations, stochastic differential equations, which we will consider in following works.

Acknowledgements

This work was supported by Leading Academic Discipline Program, 211 Project for Shanghai University of Finance and Economics (the 3rd phase). The author would like to thank the City University of Hong Kong for warm hospitality.