

A numerical approximation for the one-dimensional Burger–Fisher equation

Abstract

In this paper, an implicit finite difference method based on the Crank–Nicolson method is proposed for the numerical solution of the one-dimensional Burger–Fisher equation. The Crank–Nicolson scheme provides a system of nonlinear difference equations, which is solved by an integration of the Jacobian-Free-Newton-Krylov (JFNK) and GMRES methods. Various numerical examples are given to demonstrate the efficiency of the proposed scheme. Comparison of the computed solutions with the analytical ones demonstrates the accuracy of this proposed method.

Keywords: Crank–Nicolson scheme, Burger–Fisher equation, Jacobian-Free-Newton-Krylov method, GMRES method.

1. Introduction

Being a combination of convection, diffusion, and reaction mechanisms, the Burger–Fisher equation is highly nonlinear. This equation has many applications in many scientific fields such as gas dynamics, number theory, elasticity, heat conduction, etc. [1]. Recently, several methods have been proposed to solve it. Here, we briefly discuss the methods of some researchers.

Ismail et al. [2] and Golbabai and Javidi [3] studied Adomian decomposition methods. Ismail and Rabboh [4] developed a restrictive Padé approximation. Kaya and Sayed [5] found explicit solutions of the generalized Burger–Fisher equation. The tanh method is proposed by Wazwaz [6]. Javidi et al. [7] introduced the spectral collocation method. Khattak [8] presented a computational meshless method. Rashidi et al. [9] proposed an analytical solution using the homotopy perturbation method. Zhu and Kang [1] studied the Burger–Fisher equation by applying the B-spline quasi-interpolation method. Zhang and Yan [10] investigated a lattice Boltzmann model. Sari et al. [11] introduced a compact finite difference method. Kocacoban et al. [12] solved the Burger–Fisher equation using the reduced differential transformation method. Zhang et al. [13] studied the local discontinuous Galerkin. Mickens and Gumel [14] developed a nonstandard finite difference scheme. The exact finite difference scheme and nonstandard finite difference schemes were used by Zhang et al. [15]. In [16], two implicit finite difference schemes were designed to solve the one-dimensional nonlinear Burger–Fisher equation.

A generalized form of the one-dimensional Burger–Fisher equation is as follows

$$\frac{\partial U}{\partial t} - \frac{\partial^2 U}{\partial x^2} + \alpha(t) U^\delta \frac{\partial U}{\partial x} = \beta(t) U (1 - U^\delta), \quad x \in \Omega, 0 \leq t \leq T, \quad (1)$$

where $\Omega = \{x | a \leq x \leq b \text{ and } a, b \in \mathbb{R}\}$, $\alpha(t)$ and $\beta(t)$ are continuous functions, and $U(x, t)$ is an unknown continuous function. The initial and boundary conditions of this problem are as follows.

$$U(x, 0) = \Phi(x), \quad (2)$$

$$U(a, t) = \Psi_1(x, t), U(b, t) = \Psi_2(x, t), \quad , t > 0. \quad (3)$$

The remainder of this paper is structured as follows. In Section 2, the numerical method is explained. Some numerical experiments are reported in Section 3. Section 4 is dedicated to the conclusion.

2. The method of solution

In this section, the discretization of equation (1) is explained, and the Jacobian-Free-Newton-Krylov (JFNK) method is explained to solve the obtained system of nonlinear equations.

2.1. The discretization method

Consider Δx as a grid size in the space such that $\{x_i | x_i = i \Delta x, i = 0, 1, \dots, I\}$ covers Ω . Consider a positive integer N . The grid size in time (Δt) for the finite difference scheme is $\frac{T}{N}$. Consider U_i^n as the value of $U(x_i, t_n)$. The Crank–Nicolson approximation of (1) can be written as follows.

$$\begin{aligned} & \frac{U_i^n - U_i^{n-1}}{\Delta t} - \frac{1}{2} \left[\frac{U_{i+1}^n - 2U_i^n + U_{i-1}^n}{(\Delta x)^2} + \frac{U_{i+1}^{n-1} - 2U_i^{n-1} + U_{i-1}^{n-1}}{(\Delta x)^2} \right] \\ & + \frac{1}{2} \left[\alpha(t^n)(U^\delta)_i^n \frac{U_{i+1}^n - U_i^n}{(\Delta x)} + \alpha(t^{n-1})(U^\delta)_i^{n-1} \frac{U_{i+1}^{n-1} - U_i^{n-1}}{(\Delta x)} \right] \\ & = \frac{1}{2} \left[\beta(t^n) U_i^n (1 - (U^\delta)_i^n) + \beta(t^{n-1}) U_i^{n-1} (1 - (U^\delta)_i^{n-1}) \right], \end{aligned} \quad (4)$$

$i = 1, 2, \dots, I, \quad n = 1, 2, \dots, N$

To solve the system (4) of nonlinear equations, the JFNK method is used together with the GMRES method. The main advantage of the JFNK method [17] is that using it, no cost is incurred for the creation and storage of the Jacobian matrix.

2.2. The JFNK method

The JFNK method integrates the Newton method to solve a system of nonlinear equations and a Krylov subspace method to solve the system of linear equations resulted from the Newton method. In this method, the creation and storage of the Jacobian matrix are not required. The integration of the JFNK with GMRES methods is described in [18]. Assume a system of nonlinear equations as follows

$$F(u) = 0,$$

where u is the unknown vector. In each iteration r of the Newton method, the system of linear equations

$$J^r \delta u^r = -F(u^r), \quad (5)$$

where J^r is the Jacobian matrix in iteration r , is solved by the GMRES method. Then, the approximate solution is updated as

$$u^{r+1} = u^r + \delta u^r, \quad r = 0, 1, 2, \dots,$$

where u^0 is an initial approximate solution for the Newton method. The Newton method is stopped by the following criterion

$$\|F(u^{r+1})\| \leq \varepsilon_{\text{Newton}},$$

where $\varepsilon_{\text{Newton}}$ is a tolerance for stopping the Newton method. In the JFNK method, the product $J^r \cdot x_j$ is approximated as

$$J^r \cdot x_j \approx \frac{F(u^r + \varepsilon x_j) - F(u^r)}{\varepsilon}.$$

The value of ε can be determined by various approaches. See [19], [20], and [21]. In this work, as in [18], the following value is used.

$$\varepsilon = 10^{-8} \frac{\|u^r\|_2}{\|x_j\|_2}.$$

Algorithm 1 gives one iteration of this integrated method [22].

Algorithm 1[22]: $u^{r+1} := \text{JFNK}(F, u^r)$

/* This Algorithm solves $A\delta u^r = b$ by the GMRES method. where $A = J^r$, $b = -F(u^r)$, u^r is the approximate solution for $F(u) = 0$. Then it updates u^r */

/* e_1 is the unit vector: $e_1 = [1 \ 0 \ \dots \ 0]^T$ */

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/* H is Hessenberg matrix:  $H = \{h_{ij}\}$  */
/* V is unitary matrix:  $V = [v_1, v_2, \dots]$  */
1. Choose  $\varepsilon_G$  /*A tolerance to stop the GMRES method */
2. Choose  $\delta u^r$  /* An initial approximate solution */
   (In this experience  $\delta u^r = 0$ )
3.  $\varepsilon := 10^{-8}$ 
4.  $r_0 := -F(u^r) - \frac{F(u^r + \varepsilon \delta u^r) - F(u^r)}{\varepsilon}$ 
5.  $\gamma := \|r_0\|_2$ 
6.  $v_1 = \frac{r_0}{\gamma}$ 
7. For  $j = 1, 2, \dots$ , until convergence, Do
    $\varepsilon := (10^{-8}) \frac{\|u^r\|_2}{\|v_j\|_2}$ 
    $w_j := \frac{F(u^r + \varepsilon v_j) - F(u^r)}{\varepsilon}$ 
   For  $i=1, \dots, j$ , Do
      $h_{ij} := (w_j, v_i)$ 
      $w_j := w_j - h_{ij} v_i$ 
   End For
    $h_{j+1,j} := \|w_j\|_2$ 
   Compute  $y_j$ , as the minimizer of  $\|\gamma e_1 - Hy\|_2$ 
    $v_{j+1} := \frac{w_j}{h_{j+1,j}}$ 
   If  $\|\gamma e_1 - Hy_j\|_2 < \varepsilon_G$ 
      $\delta u^r := \delta u^r + V y_j$ , Go to the step 8, End If
   End For
8.  $u^{r+1} := u^r + \delta u^r$ , End.

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3. Numerical experiments

In this section, numerical solutions of problem (1) with initial and boundary conditions (2)-(3) are computed using the approximation scheme (4). Then, these are compared with the analytical solutions.

The accuracy of this proposed, numerical method is measured in terms of the relative error

$$L_{r\infty}(t) = \max_{1 \leq i \leq I-1} \left| \frac{\tilde{U}(x_i, t) - U(x_i, t)}{U(x_i, t)} \right|,$$

where \tilde{U} and U are the numerical approximation and the exact solution, respectively.

Example 1. We know from [13] that, if $\alpha, \beta, \delta \in \mathbb{R}$, then the exact solution of equation (1) is

$$U(x, t) = \left(\frac{1}{2} + \frac{1}{2} \tanh(\theta_1(x - \theta_2 t)) \right)^{\frac{1}{\delta}}, \quad (6)$$

where $\theta_1 = \frac{-\alpha\delta}{2(1+\delta)}$ and $\theta_2 = \frac{\alpha}{1+\delta} + \frac{\beta(1+\delta)}{\alpha}$. Consider equation (1) for $x \in [0,1]$, with initial and boundary conditions as in (6).

In Table 1, the maximum relative errors at $t = 0.5$ and $t = 1$ with $\Delta x = \Delta t = .01$ for $\delta = 1$ and various values of α and β are presented. In Figure 1, the function U introduced in relation (6) is plotted with $\beta = 8$, $\alpha = 10^{-4}$, and $\delta = 1$.

Table 1. The maximum relative errors at $t = 0.5$ and $t = 1$ with $\delta = 1$,

and various values of α and β in Example 1.

		$\alpha = 10^{-2}$	$\alpha = 10^{-4}$
$t=0.5$	$\beta = 1$	$L_{r_{\infty}} = 1.6618e-05$	$L_{r_{\infty}} = 1.6440e-05$
	$\beta = 8$	$L_{r_{\infty}} = 1.1716e-06$	$L_{r_{\infty}} = 1.2030e-06$
$t=1$	$\beta = 1$	$L_{r_{\infty}} = 1.6308e-05$	$L_{r_{\infty}} = 1.6319e - 05$
	$\beta = 8$	$L_{r_{\infty}} = 4.9791e - 06$	$L_{r_{\infty}} = 5.2182e - 06$

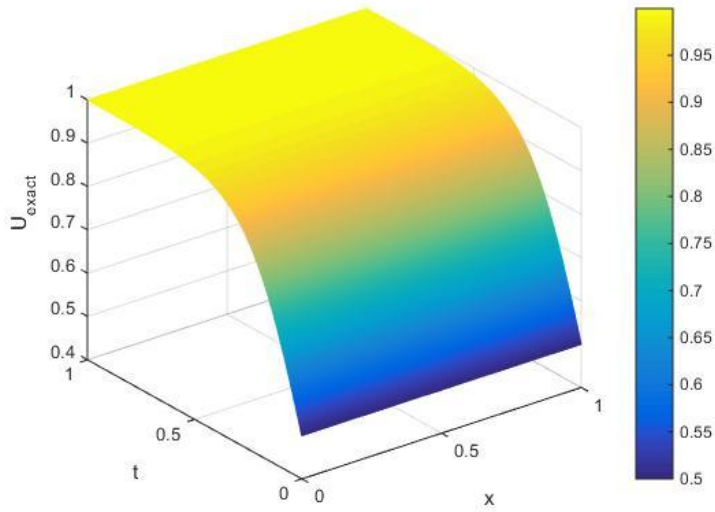


Figure 1. The function U with $\beta = 8$, $\alpha = 10^{-4}$, and $\delta = 1$, in Example 1.

Example 2. We know from [23] that the exact solution of equation (1) is

$$U(x, t) = \frac{\exp(kx - k^2t) + 1.5 - \exp(-kx + k^2t)}{\exp(kx - k^2t) + 2}, \quad (7)$$

where $\delta = 1$, k and β are arbitrary, and $\alpha = \frac{\beta}{k}$. Consider equation (1) for $x \in [0, .5]$, with initial and boundary conditions as in (7). Table 2 shows the maximum relative errors at $t = 0.5$ and $t = 1$ with $\beta = 100 + \cos(t)$ and various values of $k, \Delta t$, and Δx . In Figures 2 and 3, the function U introduced in relation (7) is plotted with $k = 3, 4$, respectively.

Table 2. The maximum relative errors at $t = 0.5$ and $t = 1$ with $\beta = 100 + \cos(t)$, various values of $k, \Delta t$, and Δx , and $\alpha = \frac{\beta}{k}$ in Example 2.

		$\Delta t = \Delta x = 0.01$	$\Delta t = \Delta x = 0.004$	$\Delta t = \Delta x = 0.002$
$K=3$		$L_{r_{\infty}} = 3.2864e-5$	$L_{r_{\infty}} = 5.5480e-6$	$L_{r_{\infty}} = 2.0943e-6$

t=0.5	K=4	$L_{r_{\infty}} = 1.4554e-4$	$L_{r_{\infty}} = 1.8431e-5$	$L_{r_{\infty}} = 4.5961e-6$
t=1	K=3	$L_{r_{\infty}} = 8.9482e-5$	$L_{r_{\infty}} = 1.0789e-5$	$L_{r_{\infty}} = 2.2481e-6$
	K=4	$L_{r_{\infty}} = 3.0119e-4$	$L_{r_{r_{\infty}}} = 4.1730e-5$	$L_{r_{\infty}} = 1.0138e-5$

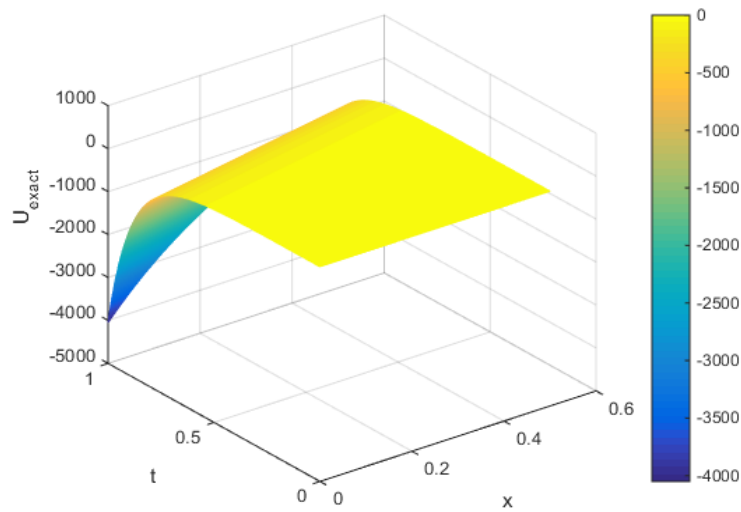


Figure 2. Function U with $k = 3$ in Example 2.

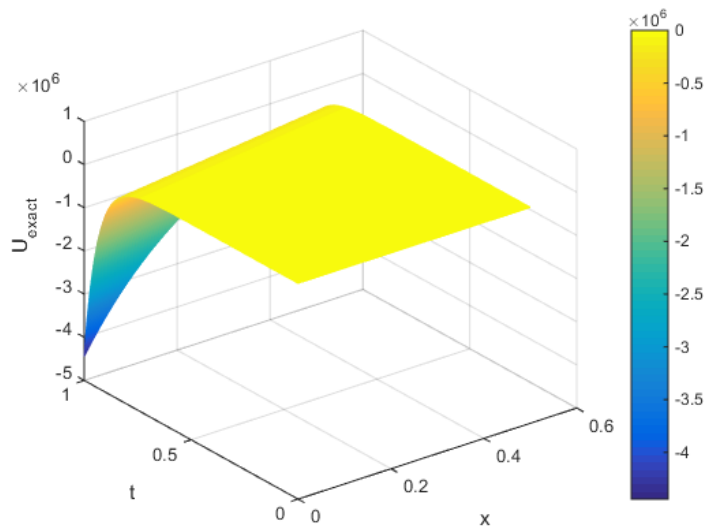


Figure 3. Function U with $k = 4$ in Example 2.

Example 3. We know from [23] that the exact solution of equation (1) is

$$U(x, t) = a_0 \exp(-kx + \int k^2 + \beta dt) \quad (8)$$

where $\delta = 1$, a_0 , k , and β are arbitrary, and $\alpha = \frac{\beta}{k}$. Set $a_0 = 2$. Consider equation (1) for $x \in [0,1]$, with initial and boundary conditions as in (8). Table 3 shows the maximum relative errors at $t = 0.5$ and $t = 1$ with $\beta = 6t^5$ and various values of $k, \Delta t$, and Δx . In Figure 4 and Figure 5, the function U introduced in relation (8) is plotted with $a_0 = 2$, $\beta = 6t^5$, and $k = 2, 3$, respectively.

Table 3. The maximum relative errors at $t = 0.5$ and $t = 1$ with $\beta = 6t^5$ and various values of $k, \Delta t$, and Δx , and $\alpha = \frac{\beta}{k}$ in Example 3.

		$\Delta t = \Delta x = 0.01$	$\Delta t = \Delta x = 0.004$	$\Delta t = \Delta x = 0.002$
t=0.5	K=2	$L_{r_{\infty}} = 9.2685e-5$	$L_{r_{\infty}} = 1.3769e-5$	$L_{r_{\infty}} = 3.0017e-6$
	K=3	$L_{r_{\infty}} = 6.2437e-4$	$L_{r_{\infty}} = 9.9817e-5$	$L_{r_{\infty}} = 2.4877e-5$
t=1	K=2	$L_{r_{\infty}} = 1.8225e-4$	$L_{r_{\infty}} = 2.9006e-5$	$L_{r_{\infty}} = 7.2472e-6$
	K=3	$L_{r_{\infty}} = 8.3802e-4$	$L_{r_{\infty}} = 1.2849e-4$	$L_{r_{\infty}} = 2.9344e-5$

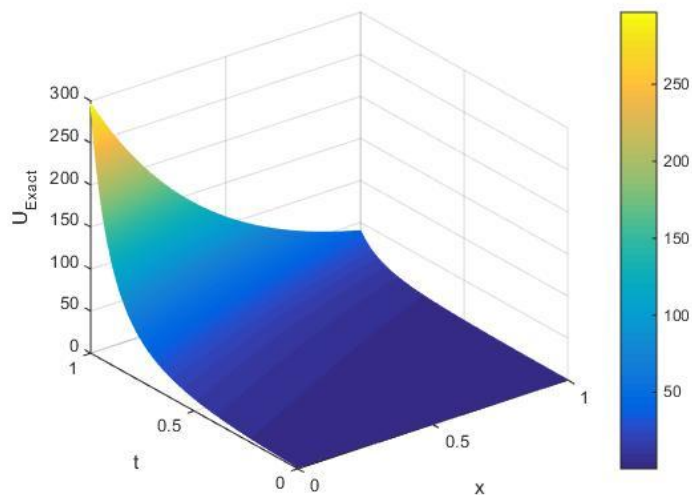


Figure 4. Function U with $a_0 = 2$, $\beta = 6t^5$ and $k = 2$ in Example 3.

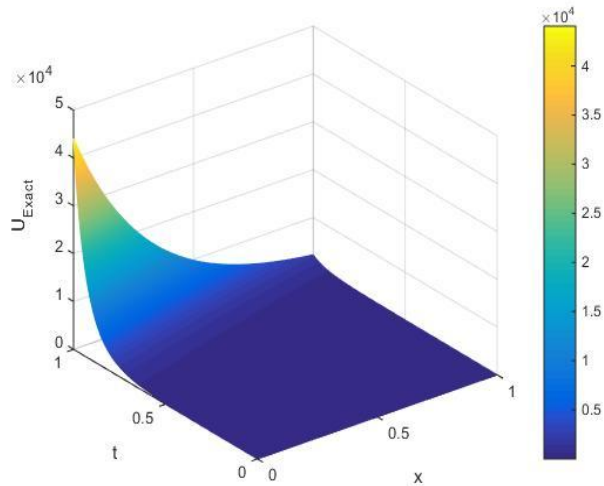


Figure 5. Function U with $\alpha_0 = 2$, $\beta = 6t^5$ and $k = 3$ in Example 3.

Figure 1 shows that the changes of the function U are not large for $x \in [0,1]$ and $t \in [0,1]$. Figures 2, 3 show that the changes of the function U are large for $x \in [0,.5]$ and $t \in [0,1]$. Figures 4 and 5 show that the changes of the function U are large for $x \in [0,1]$ and $t \in [0,1]$. The numerical errors in Tables 1, 2, and 3 confirm that the proposed method is accurate enough. As far as the author is aware, many researchers have not tested their proposed methods on the Burger–Fisher equation that the changes of its solution function are large. The Crank–Nicholson method with the JFNK method seems to be a powerful tool for approximating the Burger–Fisher equation that the changes of its solution function are large.

Conclusion

In this paper, an implicit finite difference scheme based on the Crank–Nicolson method was used to discretize the Burger–Fisher equation with initial and boundary conditions. The JFNK method was applied for solving the system of nonlinear equations. Despite the big changes in the solution functions, numerical tests confirmed the accuracy of the proposed numerical method.

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