A NSFD method for the singularly perturbed Burgers-Huxley equation

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This article focuses on a numerical solution of the singularly perturbed Burgers-Huxley equation. The simultaneous presence of a singular perturbation parameter and the nonlinearity raise the challenge of finding a reliable and efficient numerical solution for this equation via the classical numerical methods. To overcome this challenge, a nonstandard finite difference (NSFD) scheme is developed in the following manner. The time variable is discretized using the backward Euler method. This gives rise to a system of nonlinear ordinary differential equations which are then dealt with using the concept of nonlocal approximation. Through a rigorous error analysis, the proposed scheme has been shown to be parameter-uniform convergent. Simulations conducted on two numerical examples confirm the theoretical result. A comparison with other methods in terms of accuracy and computational cost reveals the superiority of the proposed scheme.

KEYWORDS

Burgers-Huxley equation, singularly perturbed problem, nonlinear equation, nonstandard finite difference methods, parameter-uniform convergence, error analysis, error bounds

1. Introduction

The time-dependent singularly perturbed Burgers-Huxley equation models a large class of physical phenomena such as the interaction between convection effect, reaction mechanism and diffusion transport. In one dimensional space this equation has the form [1, 2]

$$\begin{cases} u_t - \varepsilon u_{xx} + a(u)u_x + b(u)u = 0, \ (x,t) \in D \equiv \Omega \times (0,T] \equiv (0,1) \times (0,T], \\ u(x,0) = u_0(x), \ \ x \in \Omega, \\ u(0,t) = h_1(t), \ \ u(1,t) = h_2(t), \ \ t \in [0,T], \end{cases} \quad (1.1)$$

where

$$a(u) = \alpha u, \quad b(u) = -\beta(1-u)(u-\gamma),$$

$\alpha \geq 0$, $\beta \geq 0$, $\gamma \in (0,1)$ are real parameters and $u_0(x)$, $h_1(t)$, and $h_2(t)$ are sufficiently smooth prescribed functions. Here, $0 < \varepsilon \ll 1$ denotes the perturbation parameter.

The field of numerical methods for singularly perturbed problems has flourished significantly over the last couple of decades. The studies focused on the design and analysis of numerical methods that were parameter-uniformly convergent. In other words, research was mostly motivated by the challenge posed by the presence of the singular perturbation parameter. A vast majority of researchers discussed linear problems. Nonlinear singularly perturbed equations, and in particular Equation (1.1), received little attention from scholars in this research area. Nonlinearity constitutes an another layer of difficulties of singularly perturbed problems.

In general, the solution of the singularly perturbed Burgers-Huxley problem (1.1) with arbitrary initial and boundary conditions (IBCs) cannot be expressed in terms of a finite number of elementary functions [2, 3]. Thus, several scholars seek approximate analytic solutions using

different transformation techniques [4–6]. However, these techniques still hold for only some specific parameters and initial boundary conditions. Hence, for general cases, there is a need to develop numerical methods to obtain an approximate solution of the problem.

For the last two decades, various numerical methods have been studied to solve the Burgers-Huxley equations of type (1.1) for $\varepsilon = 1$ and specific IBCs in the framework of NSFD methods. For example, A.R. Appadu *et al.* [7] developed two novel nonstandard finite difference schemes, and explicit exponential finite difference method and a fully implicit exponential finite difference method.

For the singularly perturbed case, that is when $0 < \varepsilon \ll 1$, various parameter-uniformly convergent methods have been developed to solve (1.1) [1, 2, 8, 9]. It is to be noted that all these works used the Newton's quasilinearization to deal with the nonlinearity.

In this article, a nonstandard finite difference method for the singularly perturbed case is proposed. To deal with the nonlinearity, rather than using quasilinearization, for the first time, nonlinear terms are approximated in a nonlocal manner following one of Micken's rules [10]. The resulting method preserves the properties of the continuous solution and provides reliable numerical results. The method is proved to be first order parameter-uniform convergent in time and space.

The rest of this article is structured in the following manner. In the next section, we study a priori estimates of the analytic solution to the problem. Section 3 is about the proposed NSFD method and its parameter-uniform convergence. Section 4 deals with the implementation of the method to confirm the theoretical results and compare with other methods. Section 5 concludes the present work and provides a direction for future work.

2. Some analytical results: A priori estimates

Throughout this article, we assume that u_0, h_1, and h_2 are sufficiently smooth functions and that a and b satisfy

$$a(u) \geq p > 0, \quad b(u) \geq q \geq 0. \qquad (2.1)$$

Under these assumptions the problem (1.1) has a unique solution $u(x,t) \in C^{2,1}(\bar{D})$ which exhibits a boundary layer near $x = 1$ [1].

In the rest of the paper, C denotes a generic constant independent of the parameters and mesh sizes, and $\|.\|_D$ denotes the maximum norm over D. For the solution $u(x,t)$ of (1.1), we have the following bounds.

Lemma 2.1. *[Uniform stability estimate for continuous problem]* *Let $u(x,t)$ be the exact solution of (1.1) on $\bar{D}$, then*

$$\|u\|_{\bar{D}} \leq C\|u_0\|_{D_i} + \|u\|_{\partial D},$$

where $D_i = \{(x,t) : t = 0, x \in [0,1]\}, \partial D = D_i \cup D_L \cup D_R, D_L = \{(x,t) : x = 0, t \in [0,T]\}$, and $D_R = \{(x,t) : x = 1, t \in [0,T]\}$.

Proof: The proof can be seen in [9].   □

Remark 2.2. From our assumption in (2.1) and Lemma 2.1, one can observe that the solution $u(x,t)$ satisfies

$$\frac{p}{\alpha} \leq u(x,t) \leq C\|u_0\|_{D_i} + \|u\|_{\partial D}, \ (x,t) \in \bar{D}. \qquad (2.2)$$

3. Proposed scheme and its convergence analysis

This section is dedicated to the construction of a new scheme and to the analysis of its parameter-uniform convergence. The first step is to discretize the solution domain $\bar{D}$ and provide the definition of the NSFD methods given in [11] (a revised version of [10]).

Let M and N be positive integers and $\bar{D}^{M,N} = \bar{\Omega}^M \times [0,T]^N$ be uniform grid discretization of the solution domain $\bar{D} = \bar{\Omega} \times [0,T]$ such that

$$\bar{\Omega}^M = \{x_m : x_m = mh, \ m = 0, 1, \cdots, M\}$$

and

$$[0,T]^N = \{t_n : t_n = n\Delta t, \ n = 0, 1, \cdots, N\},$$

where $h = 1/M$ and $\Delta t = T/N$ are spatial and temporal step sizes, respectively.

Definition 3.1. A discrete scheme to determine approximate solutions u_m^n to the solution $u(x,t)$ of the problem (1.1) is called a NSFD method if at least one of the following conditions is satisfied [10]:

1. The classical denominator h or h^2 of the discrete first or second order derivative is replaced by a nonnegative function ψ such that

$$\psi(h) = h + O(h) \ \text{ or } \ \psi^2(h) = h^2 + O(h^2).$$

For example, denominator functions that satisfy the above conditions are

$$\psi(h) = h, \ \sin(h), \ \frac{e^{\beta h} - 1}{\beta}, \ \frac{h\varepsilon}{a}\left(\exp(\frac{ah}{\varepsilon}) - 1\right) \qquad (3.1)$$

and so on.

2. Nonlinear terms that occur in the differential equation are approximated in a nonlocal way. For example,

$$u_m^{n+1} \approx u_m^n, \ (u_m^n)^2 \approx u_m^n v_{m+1}^n, \ (u_m^n)^2 \approx u_m^n \frac{u_{m-1}^n + u_m^n + v_{m+1}^n}{3},$$
$$(u_m^{n+1})^3 \approx 2(u_m^n)^3 - (u_m^n)^2 u_m^{n+1}, \ (u_m^n)^3 \approx u_{m-1}^n u_m^n u_{m+1}^n, \quad (3.2)$$

and so on.

Definition 3.2. Assume that the solution of (1.1) satisfies some property P. The difference equation of (1.1) in u_m^n is called (qualitatively) stable with respect to P if, for any values of the mesh sizes Δt and h, solution of the difference equation replicates the property P.

Remark 3.3. In [10], Mickens set five rules for the constructions of the finite difference models that can replicate the properties of the exact solution. Definition 3.1 satisfies only two of these rules.

The remaining rules are expressed in terms of definition 3.2. For example, the schemes under consideration in this paper is qualitatively stable with respect to the order of the differential equation, they do satisfy positivity and uniform boundedness.

To construct the scheme, first we semidiscretize the problem (1.1) in time direction and then in space direction.

3.1. Semidiscrete scheme

Denote $u^n(x)$ as the approximation of $u(x, t_n)$ at time level t_n, $0 \leq n \leq N$. Now, we apply backward Euler finite difference method to discretize the continuous problem (1.1) in the temporal direction and obtain the following semidiscrete scheme:

$$\begin{cases} u^0(x) = u(x, 0) = u_0(x), \\ (I + \Delta t \mathcal{L}_\varepsilon^N) u^{n+1}(x) = u^n(x), \ x \in \Omega, \\ u^{n+1}(0) = h_1(t_{n+1}), \ u^{n+1}(1) = h_2(t_{n+1}), \quad n = 0, 1, \cdots, N-1, \end{cases} \tag{3.3}$$

where

$$\mathcal{L}_\varepsilon^N \equiv -\varepsilon \frac{d^2}{dx^2} + a(u^n) \frac{d}{dx} + b(u^n).$$

Here the nonlinearity $a(u)u_x$ and $b(u)u$ approximated in a nonlocal way as

$$a(u^{n+1}) \frac{du^{n+1}}{dx} \approx a(u^n) \frac{du^{n+1}}{dx} \quad \text{and} \quad b(u^{n+1})u^{n+1} \approx b(u^n)u^{n+1}.$$

and thus this approximation satisfies the condition (2) in definition 3.1.

3.2. Convergence analysis for the semidiscrete scheme

The local truncation error e_{n+1} of the semidiscrete scheme (3.3) is given by $e_{n+1} = u(x, t_{n+1}) - \hat{u}^{n+1}(x)$, where $\hat{u}^{n+1}(x)$ is the solution of

$$\begin{cases} (I + \Delta t \mathcal{L}_\varepsilon^N) \hat{u}^{n+1} = u(x, t_n), \ x \in \Omega, \\ \hat{u}^{n+1}(0) = h_1(t_{n+1}), \ \hat{u}^{n+1}(1) = h_2(t_{n+1}), \quad n = 0, 1, \cdots, N-1. \end{cases} \tag{3.4}$$

That is, $\hat{u}^{n+1}(x)$ is the solution obtained after one step of semidiscrete scheme (3.3) by taking the exact value $u(x, t_n)$ instead of $u^n(x)$ as the starting data.

In order to analyze the uniform convergence of the solution $u^n(x)$ of (3.3) to the exact solution $u(x, t_n)$, we will perform the stability analysis and establish the consistency result. First, let us consider the semidiscrete maximum principle for the operator $I + \Delta t \mathcal{L}_\varepsilon^N$.

Lemma 3.4. Let $\Phi^{n+1}(x)$ be a function such that $\Phi^{n+1}(0) \geq 0$, $\Phi^{n+1}(1) \geq 0$ and $(I + \Delta t \mathcal{L}_\varepsilon^N)\Phi^{n+1}(x) \geq 0$ for all $x \in \Omega$. Then $\Phi^{n+1}(x) \geq 0$ for all $x \in \bar{\Omega}$.

Proof: Suppose there is $x^* \in \bar{\Omega}$ such that $\Phi^{n+1}(x^*) = \min_{x \in \bar{\Omega}} \Phi^{n+1}(x) < 0$. From the given hypothesis and second derivative test, we have $x^* \neq 0$, $x^* \neq 1$, $\Phi_x^{n+1}(x^*) = 0$ and $\Phi_{xx}^{n+1}(x^*) > 0$. Then, from (3.3) we have

$$\begin{aligned} (I + \Delta t \mathcal{L}_\varepsilon^N)\Phi^{n+1}(x^*) &= \Phi^{n+1}(x^*) + \Delta t \left(-\varepsilon \Phi_{xx}^{n+1} + a(\Phi^n)\Phi_x^{n+1} \right. \\ &\quad \left. + b(\Phi^n)\Phi^{n+1} \right)(x^*) \\ &= \Phi^{n+1}(x^*) + \Delta t \left(-\varepsilon \Phi_{xx}^{n+1}(x^*) \right. \\ &\quad \left. + a(\Phi^n(x^*))\Phi_x^{n+1}(x^*) + b(\Phi^n(x^*))\Phi^{n+1}(x^*) \right) \\ &= \Phi^{n+1}(x^*) + \Delta t \left(-\varepsilon \Phi_{xx}^{n+1}(x^*) \right. \\ &\quad \left. + b(\Phi^n(x^*))\Phi^{n+1}(x^*) \right). \end{aligned}$$

Assumption (2.1) leads to $(I + \Delta t \mathcal{L}_\varepsilon^N)\Phi^{n+1}(x^*) < 0$, which contradicts the given assumption in Lemma 3.4 and thus $\Phi^{n+1}(x) \geq 0$ for all $x \in \bar{\Omega}$. $\square$

This maximum principle leads to the following stability result

$$\|(I + \Delta t \mathcal{L}_\varepsilon^N)^{-1}\| \leq C. \tag{3.5}$$

Lemma 3.5. *(Local error estimate)*
Estimate of the local error e_{n+1} is given by

$$\|e_{n+1}\| \leq C\Delta t^2. \tag{3.6}$$

Proof: See [2]. $\square$

The global error E_{n+1} associated to the semidiscrete scheme (3.3) at $(n + 1)$-th time level is given by $E_{n+1} = u(x, t_{n+1}) - u^{n+1}(x) = \sum_{i=1}^{n+1} e_i$. Using the local error estimate (Lemma 3.5) and triangular inequality the following global error estimate follows.

Theorem 3.6. *(Global error estimate)*
The global error E_{n+1} of the time discretisation at the $n + 1$ time step satisfies

$$\|E_{n+1}\| \leq C\Delta t, \quad n\Delta t \leq T. \tag{3.7}$$

Therefore, the semidiscrete scheme (3.3) is a first order uniformly convergent in the time direction.

The following lemma provides the asymptotic estimates of the exact solution u^{n+1} of (3.3) and its derivatives. These estimates will be used in the convergence analysis of the fully discrete scheme

Lemma 3.7. *If $u^{n+1}(x)$ is the solution of the problem (3.3), then there exists a constant C such that*

$$\|\frac{\partial^i u^{n+1}}{\partial x^i}\|_{\bar{\Omega}} \leq C \left(1 + \varepsilon^{-i} \exp(\frac{-p(1-x)}{\varepsilon}) \right), \ 0 \leq i \leq 4, \ \forall x \in \bar{\Omega}. \tag{3.8}$$

Proof: See the proof of Lemma 4.1 in [12] $\square$

3.3. The spatial discretisation

In this subsection, we totally discretise the semidiscrete scheme (3.3) on a uniform mesh $\bar{\Omega}^M$. Let the approximation of $u^n(x)$ at x_m be denoted by u_m^n, $0 \leq m \leq M$. Similarly, let the approximations of $a(u_m^n)$ and $b(u_m^n)$ be denoted respectively by a_m^n and b_m^n.

Using the upwind finite difference scheme and nonstandard finite difference methodology of Mickens [13], the semidiscrete problem (3.3) can be fully-discretised as

$$\begin{cases} (I + \Delta t \mathcal{L}_\varepsilon^{M,N})u_m^{n+1} = u^n(x_m), \ 1 \leq m \leq M-1, \\ u_0^{n+1} = h_1(t_{n+1}), \ u_M^{n+1} = h_2(t_{n+1}), \end{cases} \tag{3.9}$$

where

$$\mathcal{L}_\varepsilon^{M,N} u_m^{n+1} \equiv -\varepsilon \left[\frac{u_{m+1}^{n+1} - 2u_m^{n+1} + u_{m-1}^{n+1}}{(\psi_m^n)^2} \right] + a_m^n \left[\frac{u_m^{n+1} - u_{m-1}^{n+1}}{h} \right] + b_m^n u_m^{n+1},$$

FIGURE 1
Numerical solutions of Example 4.1 using the proposed method for $T = 1$, $M = 64$, $N = 40$ and $\varepsilon = 10^{-4}$. A layer is observed near $x = 1$.
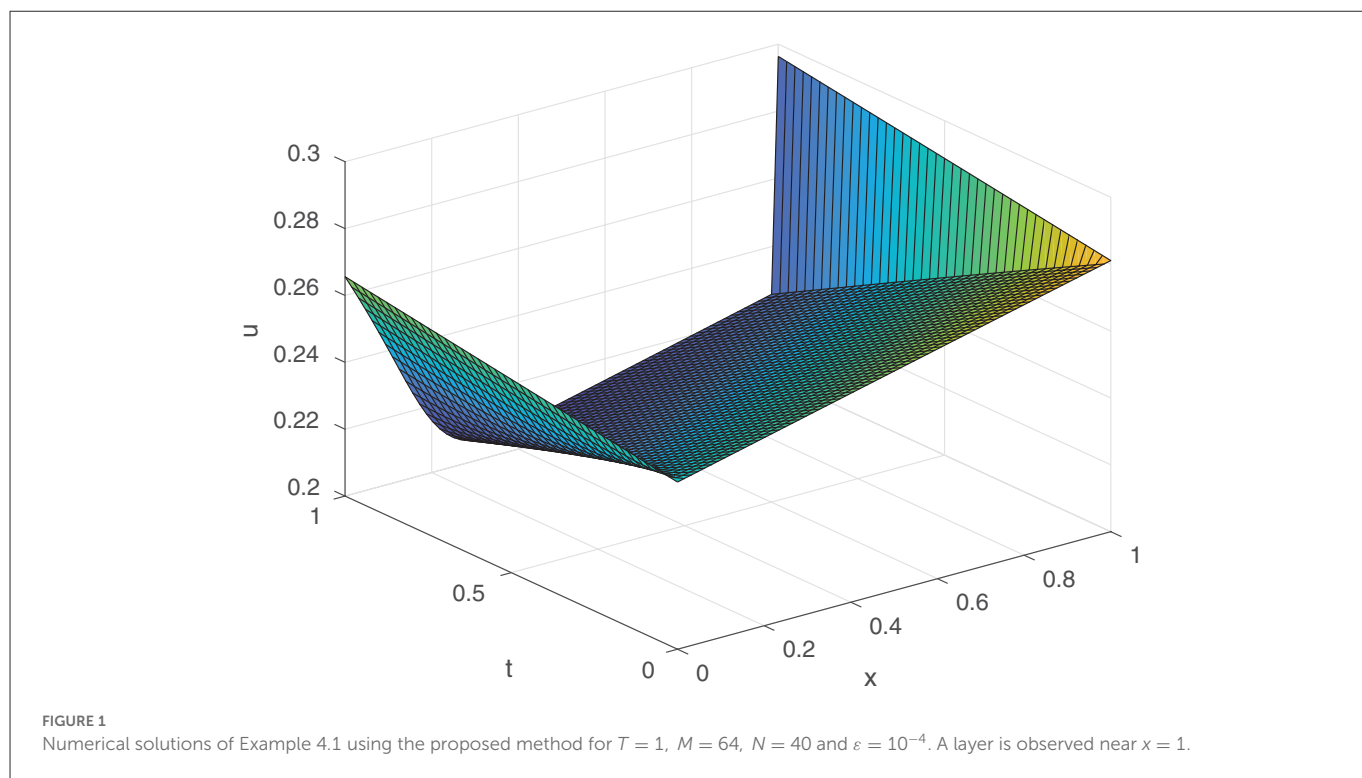
TABLE 1 Maximum absolute errors for Example 4.1 for $M = 32$ at the number of intervals N.

$\varepsilon \downarrow$	$N = 20$	40	80	160	320	640
1.00e+00	3.15e-04	1.82e-04	9.81e-05	5.10e-05	2.60e-05	1.31e-05
	0.79	0.89	0.94	0.97	0.99	
1.00e-02	8.31e-04	4.31e-04	2.20e-04	1.11e-04	5.58e-05	2.80e-05
	0.95	0.97	0.99	0.99	1.00	
1.00e-04	1.06e-03	5.61e-04	2.90e-04	1.47e-04	7.43e-05	3.73e-05
	0.91	0.95	0.98	0.99	0.99	
1.00e-06	1.06e-03	5.61e-04	2.90e-04	1.47e-04	7.43e-05	3.73e-05
	0.91	0.95	0.98	0.99	0.99	
1.00e-08	1.06e-03	5.61e-04	2.90e-04	1.47e-04	7.43e-05	3.73e-05
	0.91	0.95	0.98	0.99	0.99	
1.00e-10	1.06e-03	5.61e-04	2.90e-04	1.47e-04	7.43e-05	3.73e-05
	0.91	0.95	0.98	0.99	0.99	
1.00e-12	1.06e-03	5.61e-04	2.90e-04	1.47e-04	7.43e-05	3.73e-05
	0.91	0.95	0.98	0.99	0.99	

and the denominator function ψ_m^n is

$$(\psi_m^n)^2(h, \varepsilon) = \frac{h\varepsilon}{a_m^n}\left(\exp(\frac{a_m^n h}{\varepsilon}) - 1\right).$$

The denominator function is derived systematically to replicate the dissipativity properties of the solution of the differential equations. Interested readers may refer to [14] for details. Observe

that

$$(\psi_m^n)^2(h, \varepsilon) = h^2 + O(\frac{h^3}{\varepsilon}),$$

and thus this function satisfies the condition (1) in definition 3.1.
The method in Eq. (3.9) is a linear nonstandard finite difference (LNSFD) and can be written as:

$$E_m^{n+1} u_{m-1}^{n+1} + F_m^{n+1} u_m^{n+1} + G_m^{n+1} u_{m+1}^{n+1} = H_m^{n+1}, \quad 1 \le m \le M - 1,$$

$$(3.10)$$

TABLE 2 Maximum absolute errors for Example 4.1 for $N = 10$ at the number of intervals M.

$\varepsilon \downarrow$	$M = 32$	64	128	256	512	1,024
1.00e+00	5.579e-05	2.989e-05	1.545e-05	7.854e-06	3.959e-06	1.988e-06
	0.90	0.95	0.98	0.99	0.99	
1.00e-02	1.271e-03	1.622e-03	1.144e-03	6.693e-04	3.596e-04	1.861e-04
	-0.35	0.50	0.77	0.90	0.95	
1.00e-04	2.064e-03	1.229e-03	6.846e-04	8.908e-04	3.387e-03	2.938e-03
	0.75	0.84	-0.38	-1.93	0.20	
1.00e-06	2.064e-03	1.229e-03	6.815e-04	3.624e-04	1.872e-04	9.526e-05
	0.75	0.85	0.91	0.95	0.98	
1.00e-08	2.064e-03	1.229e-03	6.815e-04	3.624e-04	1.872e-04	9.526e-05
	0.75	0.85	0.91	0.95	0.98	
1.00e-10	2.064e-03	1.229e-03	6.815e-04	3.624e-04	1.872e-04	9.526e-05
	0.75	0.85	0.91	0.95	0.98	
1.00e-12	2.064e-03	1.229e-03	6.815e-04	3.624e-04	1.872e-04	9.526e-05
	0.75	0.85	0.91	0.95	0.98	

TABLE 3 Comparison of absolute errors for Example 4.1 using the proposed scheme (LNSFD) with results of [7, 19] and NNSFD for $\varepsilon = 1$, $\alpha = \beta = 1$, $\gamma = 0.001$ at some values of x and t.

t	x	LNSFD	NNSFD	VIM[19]	NSFD[7]
0.05	0.1	7.5615e-09	7.5595e-09	1.87405-08	8.13470-09
	0.5	1.6945e-08	1.6945e-08	1.87405-08	1.78493-08
	0.9	7.5601e-09	7.5601e-09	1.87405-08	8.13524-09
0.1	0.1	1.1125e-08	1.1123e-08	3.74813-08	1.19758-08
	0.5	2.8280e-08	2.8280e-08	1.37481-08	3.02147-08
	0.9	1.1124e-08	1.1124e-08	3.74813-08	1.19770-08
1	0.1	1.6853e-08	1.6850e-08	3.74812-08	1.86370-08
	0.5	4.6811e-08	4.6809e-08	3.74813-08	5.17712-08
	0.9	1.6853e-08	1.6853e-08	3.74813-08	1.86393-08

TABLE 4 CPU time for Example 4.1.

	LNSFD	NNSFD
CPU Time (Sec)	0.010792	6.777089

where

$$E_m^{n+1} = -\Delta t\left(\frac{\varepsilon}{(\psi_m^n)^2} + \frac{a_m^n}{h}\right), \qquad (3.11)$$

$$F_m^{n+1} = 1 + \Delta t\left(\frac{2\varepsilon}{(\psi_m^n)^2} + \frac{a_m^n}{h} + b_m^n\right), \qquad (3.12)$$

$$G_m^{n+1} = -\frac{\varepsilon\,\Delta t}{(\psi_m^n)^2}, \qquad (3.13)$$

$$H_m^{n+1} = u_m^n.$$

Equation (3.10) leads to the tridiagonal system which can be solved with Thomas Algorithm [15, 16].

Remark 3.8. The developed scheme is a linear difference equation even if the original equation (1.1) is nonlinear. This is one of the feature of NSFD.

3.4. Convergence analysis of the space discretisation

The difference operator $I + \Delta t\mathcal{L}_\varepsilon^{M,N}$ of (3.9) satisfies the following maximum principle. Hence (3.9) has a unique discrete solution u_m^n.
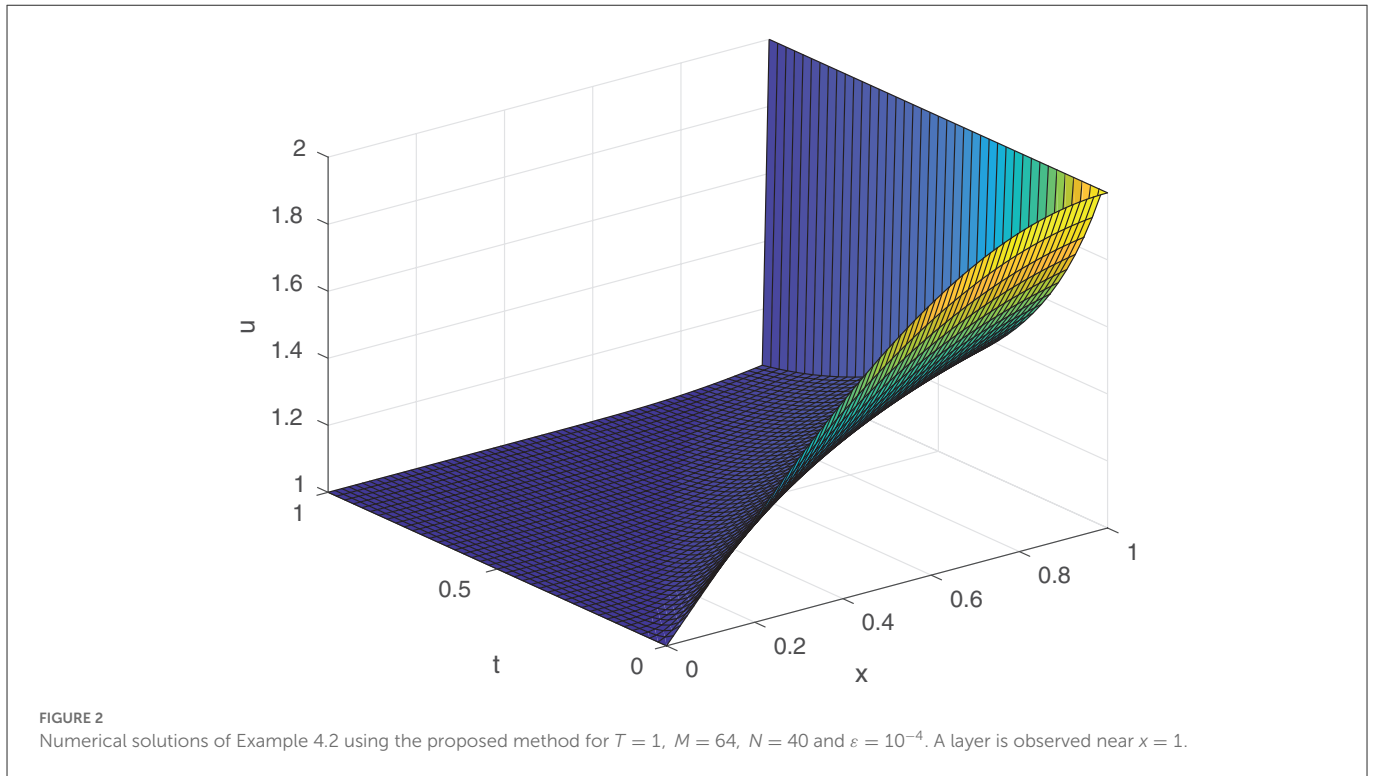
Lemma 3.9. *Let* Φ_m^{n+1}, $m = 0, 1, \cdots, M$ *be fully discrete mesh functions. If* $\Phi_0^{n+1} \geq 0$, $\Phi_M^{n+1} \geq 0$ *and* $(I + \Delta t\mathcal{L}_\varepsilon^{M,N})\Phi_m^{n+1} \geq 0$ *for* $1 \leq m \leq M - 1$ *then* $\Phi_m^{n+1} \geq 0$ *for* $0 \leq m \leq M$.

Proof: Let us proof by contradiction. Assume there is m^* such that $\Phi_{m^*}^{n+1} = \min\limits_{0\leq m\leq M} \Phi_m^{n+1} < 0$. Thus, we have, $m^* \notin \{0, M\}$, $\Phi_{m^*+1}^{n+1} - \Phi_{m^*}^{n+1} \geq 0$, and $\Phi_{m^*-1}^{n+1} - \Phi_{m^*}^{n+1} \geq 0$. Now, from (3.9) we have

$$(I + \Delta t\mathcal{L}_\varepsilon^{M,N})\Phi_{m^*}^{n+1} = \Phi_{m^*}^{n+1} + \Delta t\left[-\varepsilon\left(\frac{\Phi_{m^*+1}^{n+1} - 2\Phi_{m^*}^{n+1} + \Phi_{m^*-1}^{n+1}}{(\psi_{m^*}^n)^2}\right)\right.$$
$$\left. + a_{m^*}^n\left(\frac{\Phi_{m^*}^{n+1} - \Phi_{m^*-1}^{n+1}}{h}\right) + b_{m^*}^n\Phi_{m^*}^{n+1}\right]$$
$$= -\varepsilon\Delta t\left(\frac{\Phi_{m^*+1}^{n+1} - \Phi_{m^*}^{n+1} + \Phi_{m^*-1}^{n-1} - \Phi_{m^*}^{n+1}}{(\psi_{m^*}^n)^2}\right)$$
$$+ a_{m^*}^n\Delta t\left(\frac{\Phi_{m^*}^{n+1} - \Phi_{m^*-1}^{n+1}}{h}\right)$$
$$+ \left(1 + \Delta t b_{m^*}^n\right)\Phi_{m^*}^{n+1} < 0,$$

which contradicts the given assumption $(I + \Delta t\mathcal{L}_\varepsilon^{M,N})\Phi_m^{n+1} \geq 0$ for $1 \leq m \leq M - 1$ and hence $\Phi_m^{n+1} \geq 0$ for $0 \leq m \leq M$. □

This leads immediately to the following stability result, analogous to the continuous result.

FIGURE 2
Numerical solutions of Example 4.2 using the proposed method for $T = 1$, $M = 64$, $N = 40$ and $\varepsilon = 10^{-4}$. A layer is observed near $x = 1$.

Lemma 3.10. *[Uniform stability estimate for discrete problem]*
Let Φ_m^{n+1}, $0 \leq m \leq M$, be any mesh functions such that $\Phi_0^{n+1} = \Phi_M^{n+1} = 0$, then

$$|\Phi_m^{n+1}| \leq \frac{1}{1 + q\Delta t} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}|, \ 0 \leq m \leq M.$$

Proof: Define

$$[\Psi^\pm]_m^{n+1} = \frac{1}{1 + q\Delta t} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}| \pm \Phi_m^{n+1}$$

which implies

$$[\Psi^\pm]_0^{n+1} = \frac{1}{1 + q\Delta t} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}| \pm \Phi_0^{n+1}$$
$$= \frac{1}{1 + q\Delta t} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}| \geq 0,$$

$$[\Psi^\pm]_M^{n+1} = \frac{1}{1 + q\Delta t} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}| \pm \Phi_M^{n+1}$$
$$= \frac{1}{1 + q\Delta t} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}| \geq 0,$$

and, for $m = 1, 2, \cdots, M - 1$,

$$(I + \Delta t \mathcal{L}_\varepsilon^{M,N})[\Psi^\pm]_m^{n+1} = (I + \Delta t \mathcal{L}_\varepsilon^{M,N})$$
$$\left(\frac{1}{1 + q\Delta t} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}| \pm \Phi_m^{n+1} \right)$$
$$= \frac{1 + \Delta t b_m^n}{1 + q\Delta t} \max_{0 \leq m \leq M} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_m^{n+1}| \pm (I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_m^{n+1}.$$

Since $b^n(x_m) \geq q$, one has $(I + \Delta t \mathcal{L}_\varepsilon^{M,N})[\Psi^\pm]_m^{n+1} \geq 0$. Thus, by discrete maximum principle given in Lemma 3.9, one obtains

$[\Psi^\pm]_m^{n+1} \geq 0$. This gives the required result

$$|\Phi_m^{n+1}| \leq \frac{1}{q} \max_{1 \leq i \leq M-1} |(I + \Delta t \mathcal{L}_\varepsilon^{M,N})\Phi_i^{n+1}|, \ 0 \leq m \leq M.$$

$\square$

Remark 3.11. Lemmas 3.9 and 3.10 show the developed nonstandard finite difference method replicates the positivity and boundedness of the solution, respectively. Hence, the proposed method is qualitatively stable.

Theorem 3.12. *(Error estimate in the spatial direction)*
Let $u^{n+1}(x)$ be the solution of the semidiscrete problem (3.4) and u_m^{n+1} be the solution of the full discretisation (3.9). Then, the error estimate is given by

$$\left| u^{n+1}(x_m) - u_m^{n+1} \right| \leq Ch, \ 0 \leq m \leq M.$$

Proof: The truncation error of the complete discretization (3.9) is given by

$$(I + \Delta t \mathcal{L}_\varepsilon^{M,N}) \left(u^{n+1}(x_m) - u_m^{n+1} \right) \tag{3.14}$$
$$= \Delta t \left[-\varepsilon (u^{n+1})''(x_m) + a^n(x_m)(u^{n+1})'(x_m) + b^n(x_m)u^{n+1}(x_m) \right]$$
$$- \Delta t \left[-\varepsilon \frac{h^2}{(\psi_m^n)^2} D^+ D^- u^{n+1}(x_m) + a^n(x_m) D^- u^{n+1}(x_m) + b^n(x_m)u^{n+1}(x_m) \right]$$
$$= \varepsilon \Delta t \left(\frac{h^2}{(\psi_m^n)^2} D^+ D^- - \frac{d^2}{dx^2} \right) u^{n+1}(x_m) + a^n(x_m)\Delta t \left(\frac{d}{dx} - D^- \right) u^{n+1}(x_m), \tag{3.15}$$

where

$$D^- u^{n+1}(x_m) = \frac{u^{n+1}(x_m) - u^{n+1}(x_{m-1})}{h},$$

TABLE 5 Maximum absolute errors for Example 4.2 for $M = 32$ at the number of intervals N.

$\varepsilon \downarrow$	$N = 20$	40	80	160	320	640
1.00e+00	8.75e-03	5.46e-03	3.09e-03	1.64e-03	9.86e-04	8.23e-04
	0.68	0.82	0.91	0.73	0.26	
1.00e-02	1.09e-02	6.28e-03	3.54e-03	1.89e-03	9.83e-04	5.01e-04
	0.79	0.83	0.90	0.95	0.97	
1.00e-04	1.08e-02	5.91e-03	3.33e-03	1.78e-03	9.27e-04	4.73e-04
	0.87	0.83	0.90	0.94	0.97	
1.00e-06	1.08e-02	5.91e-03	3.33e-03	1.78e-03	9.27e-04	4.73e-04
	0.87	0.83	0.90	0.94	0.97	
1.00e-08	1.08e-02	5.91e-03	3.33e-03	1.78e-03	9.27e-04	4.73e-04
	0.87	0.83	0.90	0.94	0.97	
1.00e-10	1.08e-02	5.91e-03	3.33e-03	1.78e-03	9.27e-04	4.73e-04
	0.87	0.83	0.90	0.94	0.97	
1.00e-12	1.08e-02	5.91e-03	3.33e-03	1.78e-03	9.27e-04	4.73e-04
	0.87	0.83	0.90	0.94	0.97	

TABLE 6 Maximum absolute errors for Example 4.2 for $N = 10$ at the number of intervals M.

$\varepsilon \downarrow$	$M = 32$	64	128	256	512	1,024
1.00e+00	6.399e-03	3.542e-03	1.861e-03	9.533e-04	4.824e-04	2.427e-04
	0.85	0.93	0.96	0.98	0.99	
1.00e-02	9.602e-02	4.611e-02	4.394e-02	5.108e-02	3.545e-02	2.077e-02
	1.06	0.07	-0.22	0.53	0.77	
1.00e-04	4.147e-03	2.242e-03	1.167e-03	5.956e-04	3.294e-04	7.375e-03
	0.89	0.94	0.97	0.85	-4.48	
1.00e-06	4.147e-03	2.242e-03	1.167e-03	5.956e-04	3.009e-04	1.513e-04
	0.89	0.94	0.97	0.98	0.99	
1.00e-08	4.147e-03	2.242e-03	1.167e-03	5.956e-04	3.009e-04	1.513e-04
	0.89	0.94	0.97	0.98	0.99	
1.00e-10	4.147e-03	2.242e-03	1.167e-03	5.956e-04	3.009e-04	1.513e-04
	0.89	0.94	0.97	0.98	0.99	
1.00e-12	4.147e-03	2.242e-03	1.167e-03	5.956e-04	3.009e-04	1.513e-04
	0.89	0.94	0.97	0.98	0.99	

$$D^+ u^{n+1}(x_m) = \frac{u^{n+1}(x_{m+1}) - u^{n+1}(x_m)}{h},$$

$$D^+ D^- u^{n+1}(x_m) = \frac{u^{n+1}(x_{m+1}) - 2u^{n+1}(x_m) + u^{n+1}(x_{m-1})}{h^2}.$$

Applying absolute values and using triangle inequalities in (3.15) leads to

$$\left| (I + \Delta t \mathcal{L}_\varepsilon^{M,N})(u^{n+1}(x_m) - u_m^{n+1}) \right|$$
$$\leq \varepsilon \Delta t \left| (\frac{h^2}{(\psi_m^n)^2} - 1) D^+ D^- u^{n+1}(x_m) \right|$$
$$+ \varepsilon \Delta t \left| (D^+ D^- - \frac{d^2}{dx^2}) u^{n+1}(x_m) \right|$$
$$+ a^n(x_m) \Delta t \left| (\frac{d}{dx} - D^-) u^{n+1}(x_m) \right|. \tag{3.16}$$

From the use the Taylor series expansions of some terms in the above equation, on obtains

$$\frac{h^2}{(\psi_m^n)^2} - 1 = -\frac{a^n(\eta_1)h}{2\varepsilon + ha(\eta_1)}, \tag{3.17}$$

$$D^+ D^- u^{n+1}(x_m) = \frac{d^2 u^{n+1}}{dx^2}(\eta_2), \tag{3.18}$$

$$(D^+ D^- - \frac{d^2}{dx^2}) u^{n+1}(x_m) = \frac{h^2}{16} \frac{d^4 u^{n+1}}{dx^4}(\eta_3), \tag{3.19}$$

$$(\frac{d}{dx} - D^-) u^{n+1}(x_m) = \frac{h}{2} \frac{d^2 u^{n+1}}{dx^2}(\eta_4), \tag{3.20}$$

for some η_i such that $x_{m-1} \leq \eta_i \leq x_{m+1}$, $i = 1, 2, 3, 4$.

Substituting equations (3.17)–(3.20) in (3.16) and from the boundedness of $a^n(x)$ in (3.8), one has

$$\left|(I + \Delta t \mathcal{L}_\varepsilon^{M,N})(u^{n+1}(x_m) - u_m^{n+1})\right| \leq \varepsilon \Delta t \frac{a^n(\eta_1)h}{2\varepsilon + ha^n(\eta_1)}\left|\frac{d^2 u^{n+1}}{dx^2}(\eta_2)\right|$$
$$+ \varepsilon \Delta t \frac{h^2}{16}\left|\frac{d^4 u^{n+1}}{dx^4}(\eta_3)\right|$$
$$+ a^n(x_m)\Delta t \frac{h}{2}\left|\frac{d^2 u^{n+1}}{dx^2}(\eta_4)\right|, \tag{3.21}$$

and this gives

$$\left|(I + \Delta t \mathcal{L}^{M,N})(u^{n+1}(x_m) - u_m^{n+1})\right|$$
$$\leq Ch\left(1 + \varepsilon^{-2}\exp\left(\frac{-p(1 - x_m)}{\varepsilon}\right)\right)$$
$$+ Ch^2\left(\varepsilon + \varepsilon^{-3}\exp\left(\frac{-p(1 - x_m)}{\varepsilon}\right)\right)$$
$$\leq Ch\left(1 + \varepsilon^{-3}\exp\left(\frac{-p(1 - x_m)}{\varepsilon}\right)\right), \text{ for } m = 1, 2, \cdots, M - 1.$$

Applying Lemma 7 of [17] gives

$$\lim_{\varepsilon \to 0}\left|(I + \Delta t \mathcal{L}^{M,N})(u^{n+1}(x_m) - u_m^{n+1})\right| \leq Ch \tag{3.22}$$

which, upon use of Lemma 3.10, leads to

$$\lim_{\varepsilon \to 0}|u^{n+1}(x_m) - u_m^{n+1}| \leq Ch. \tag{3.23}$$

$\square$

Combining the schemes (3.3) and (3.9), one obtains the following fully-discrete scheme on the mesh $\bar{\Omega}^M \times [0, T]^N$.

$$\begin{cases} (I + \Delta t \mathcal{L}_\varepsilon^{M,N})u_m^{n+1} = u_m^n, \ 1 \leq m \leq M - 1, \ 0 \leq n \leq N - 1, \\ u^{n+1}(0) = h_1(t_{n+1}), \ u^{n+1}(1) = h_2(t_{n+1}), \ 0 \leq n \leq N - 1. \end{cases} \tag{3.24}$$

The temporal and spatial error estimates in Theorem 3.6 and 3.12, respectively, give the the following main result of this paper.

Theorem 3.13. *If $u(x, t)$ is the exact solution of the continuous problem (1.1) and u_m^n is the solution of the fully-discrete scheme (3.24), then*

$$|u(x_m, t_n) - u_m^n| \leq C(\Delta t + h), \ 0 \leq m \leq M, 0 \leq n \leq N.$$

Therefore, the presented discrete scheme is ε-uniform convergent of order one both in time and space.

4. Numerical implementation

In this section, two test examples are provided to demonstrate the efficiency and applicability of the proposed numerical method. The first example is taken from a recent article [7] for $\varepsilon = 1$, and we modified it by multiplying the highest derivative term by ε to make the problem singularly perturbed. It is the first time to consider this example for the singularly perturbed case. The second example is taken from [2] for different initial and boundary conditions to satisfy our assumption (2.1). All the computations are carried out by Intel Coreâ i5-4210M CPU @2.60$GHz \times 4$ with MATLAB 2017.

Example 4.1. Consider the following singularly perturbed Burgers-Huxley problem:

$$\begin{cases} \frac{\partial u}{\partial t} - \varepsilon \frac{\partial^2 u}{\partial x^2} + \alpha u \frac{\partial u}{\partial x} - \beta(1 - u)(u - \gamma)u = 0, \ (x, t) \in (0, 1) \times (0, T], \\ u(x, 0) = \frac{\gamma}{2} + \frac{\gamma}{2}\tanh(A_1 x), \ 0 \leq x \leq 1, \\ u(0, t) = \frac{\gamma}{2} + \frac{\gamma}{2}\tanh(-A_1 A_2 t), \ u(1, t) = \frac{\gamma}{2} + \frac{\gamma}{2}\tanh(A_1(1 - A_2 t)), \\ \qquad 0 \leq t \leq T, \end{cases}$$

where

$$A_1 = \frac{-\alpha + \sqrt{\alpha^2 + 8\beta}}{8}, \ A_2 = \frac{\gamma \alpha}{2} - \frac{(2 - \gamma)(-\alpha + \sqrt{\alpha^2 + 8\beta})}{8}.$$

For $\varepsilon = 10^{-4}, \alpha = \beta = 1$ and $\gamma = 0.5$, we plotted the numerical solution of Example 4.1 using the proposed method in Figure 1. As we see, the method resolves the boundary layer at $x = 1$. Thus, our proposed method replicates the property of the continuous solution.

Since the exact solution of this problem is not known, for each ε, the maximum pointwise error is calculated using double mesh principle [18] given by

$$E_\varepsilon^{M,N} = \max_{(x_m, t_n) \in D^{M,N}}|u_{m;M}^{n;N} - u_{2m;2M}^{2n;2N}|,$$

where $u_{m;M}^{n;N}$ and $u_{2m;2M}^{2n;2N}$ are approximate solutions to problem (3.9) on $D^{M,N}$ and $D^{2M,2N}$, respectively. The corresponding order of convergence is given by

$$R_\varepsilon^{M,N} = \log_2\left(\frac{E_\varepsilon^{M,N}}{E_\varepsilon^{2M,2N}}\right).$$

In Tables 1, 2, we compute the maximum absolute errors and the corresponding order of convergences using the proposed method for $\alpha = \beta = 1, \gamma = 0.5$ and various values of ε with fixed values of M and N, respectively. From these results, we can observe that the method is ε-uniform convergent of order one both in time and space. This confirms the theoretical error results.

Table 3 compares absolute errors at some values of x and t using the proposed method and methods in [7, 19] and the nonlinear nonstandard finite difference (NNSFD) method given by

$$\frac{u_m^{n+1} - u_m^n}{\phi(\Delta t, \varepsilon)} - \varepsilon \frac{u_{m+1}^{n+1} - 2u_m^{n+1} + u_{m-1}^{n+1}}{(\psi_m^{n+1})^2(h, \varepsilon)} + a_m^{n+1}\frac{u_m^{n+1} - u_{m-1}^{n+1}}{h}$$
$$+ b_m^{n+1}u_m^{n+1} = 0, \tag{4.1}$$

where the denominator functions

$$\phi(\Delta t, \varepsilon) = \frac{e^{\varepsilon \Delta t} - 1}{\varepsilon}, \ (\psi_m^{n+1})^2(h, \varepsilon) = \frac{h\varepsilon}{a_m^{n+1}}\left(\exp(\frac{a_m^{n+1}h}{\varepsilon}) - 1\right).$$

The nonlinear system (4.1) can be easily solved using Newton's method (see in [7]). The solution at the previous time-step is taken as the initial estimate. The methods considered in [7] and [19] are NSFD methods and variational iteration method (VIM), respectively. From the table, we observe that our proposed method gives more accurate results.

Also, in Table 4, we compare the CPU times required to compute the solutions of Example 4.1 for $M = 10, N = 2000, \alpha = \beta = 1, \gamma = 0.001, T = 10$ using the LNSFD and NNSFD methods. As one can observe in the table, the CPU time for NNSFD is larger than the LNSFD.

Example 4.2. Consider the following SPBHE [2]:

$$\begin{cases} \frac{\partial u}{\partial t} - \varepsilon \frac{\partial^2 u}{\partial x^2} + u\frac{\partial u}{\partial x} - (1-u)(u-0.5)u = 0, \quad (x,t) \in (0,1) \times (0,1], \\ u(x,0) = 1 + \sin(\pi x/2), \ 0 < x < 1, \\ u(0,t) = 1, \ u(1,t) = 2, \ 0 \le t \le 1. \end{cases}$$

5. Conclusion

In this paper, a nonstandard finite difference method for a singularly perturbed Burgers-Huxley equation has been developed. First, the backward-Euler scheme was applied to discretize the problem (1.1) with respect to time derivative and the upwind nonstandard finite difference scheme on uniform mesh to approximate the spatial derivative. Then, the presented method was proved to be first-order convergent in both the spatial and temporal variables. Numerical results are given in Figures 1, 2 and in Tables 1–6 for two test examples to confirm the theoretical results and to compare with recent results. It has been observed from these figures and tables that the numerical results are in agreement with the theoretical findings. For $\varepsilon = 1$, in Table 3, comparisons with the NNSFD and the VIM of [19] and the NSFD method of [7] reveal that the proposed NSFD method gives more accurate results. In addition, the present method is also applicable when $0 < \varepsilon < 1$. Thus, in all cases, the present method produces more accurate results than the existing schemes.

For future work, one may proceed with a higher order scheme for the problem under consideration for example by using the Crank-Nicolson method or those presented in [20, 21]. We are currently working in this direction. Also to note is that higher order parameter-uniform numerical methods for Burgers-Huxley equations are quasi-absent. The design of direct higher order parameter-uniform convergent methods is thus an interesting direction to explore.

Data availability statement

The original contributions presented in the study are included in the article/supplementary material, further inquiries can be directed to the corresponding author.

Author contributions

ED constructed the scheme and implemented the MATLAB code under JM's supervision. All authors participated in the analysis of the theoretical results and in the writing of the manuscript.

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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