



RESEARCH ARTICLE - MATHEMATICS

## Mixed Crank-Nicolson and Galerkin Methods for Solving Nonlinear Hyperbolic Partial Differential Equation

M. A. Jawad<sup>1\*</sup>, W. A. Ibrahim<sup>2</sup>, S. J. M. Al-Qaisi<sup>3</sup>

<sup>1\*</sup> Department of Mathematics, College of Basic Education, Diyala University, Diyala –Iraq

<sup>2</sup> Department of Mathematics, Almuqdad College of Education, University of Diyala, Baqubah, Iraq

<sup>3</sup> Diyala Directorate of Education, Diyala-Iraq

\* Corresponding author E-mail: [basicmathte1@uodiyala.edu.iq](mailto:basicmathte1@uodiyala.edu.iq)

Article Info.	Abstract
<p><i>Article history:</i></p> <p>Received 13 January 2024</p> <p>Accepted 21 April 2024</p> <p>Publishing 30 September 2024</p>	<p>In this work, the approximation solution for nonlinear hyperbolic partial differential equation (NLHPDE) is obtained by using the mixed Crank-Nicolson (CN) scheme and the Galerkin Method (GM) and it is symbolized by (MCNGM). At first the CN is utilized for the variable of time to obtain the discrete weak form for the NLHPDE, and then the GM is utilized which reduces the DWF into the Galerkin nonlinear algebraic system (GNLAS) at each step of time. Through utilizing the predictor-corrector techniques which are symbolized by the obtained GNLAS is transformed into Galerkin linear algebraic system (GLAS) which is solved by applying the Cholesky method. The convergence of the method is studied. Some examples are given to illustrate the efficiency and the accuracy for the proposed method.</p>

This is an open-access article under the CC BY 4.0 license (<http://creativecommons.org/licenses/by/4.0/>)

*The official journal published by the College of Education at Mustansiriya University*

**Keywords:** Nonlinear Hyperbolic Partial Differential Equation, method of Galerkin finite elements, Crank-Nicolson Scheme, Convergent.

### 1. Introduction

As it is known many natural phenomena usually are described by mathematical problems represent by nonlinear PDEs (NLPDEs) in general and by NLHPDEs in particular. Of course there is a need to solve such problem often; but in fact solving them analytically is difficult if it is not impossible. Therefore, numerical and approximate methods became an urgent need to find the approximate solution of such mathematical problems.

In the previous decades many approximate and numerical methods were used for solving the PDEs in general; like the finite difference method (FDM) [1][2][3][4], method of lines [5][6], meshless method [7][8], and many other methods. The CN and GM have been used in many applications like: digital image processing [9], reactors [10], power system [11] and in groundwater [12]. In the recent years many other methods have been introduced to solve such problems, more precisely in 2018, GM has been used to solve the hyperbolic partial differential equation [13], in 2019, the MCNGM has been applied to solve NLPDEs of parabolic type [14], while in 2020 [15], used the CN with heredity and its program implementation for solving PDEs, also in 2020 the CN with the FDM analyzed and proposed for a NLPDEs of integral type [16], also in 2020 the CN was applied to a random component heat equation [17], in 2021 the Galerkin-Implicit Methods [18], also in 2021 MCNGM are used for nonlinear time fractional parabolic problems [19], in 2022 the mixed method of homotopy perturbation and the variation iteration method [20] used to solve NLHPDEs, also in 2022 the time fractional telegraph equation is solved numerically based on the CN [21], also in 2022 the MCNGM are used for Maxwell's equations [22] and later in 2022 CN is used to solve uncertain heat equation [23].

In this work, the MCNGM is used to find the approximation solution of the NLHPDE. In the proposed method, at first the CN scheme is utilized to obtain the discrete weak form, and then the GM is utilized to get the GNLAS at each step of time. Then the predictor (PrT)- corrector (CoT) techniques (PrCoT) is applied to solve this GNLAS, in the PrCoT, the GNLAS transforms to a GLAS, which is solved by using the cholesky method. The convergence of the method (the PrCoT) is proved. Some examples are given to illustrate the efficiency and the accuracy for the method. Also, the numerical results are given.

## 2. Basic Concepts

Definition1 [24]: A point  $n^* \in Q \subset R^2$  is called a fixed point (FP) of the function  $Q \rightarrow R^2$  if  $y(n^*) = n^*$ .

Definition2 [25]: A function  $y: Q \subset R^2 \rightarrow R^2$  is called contractive on  $P$  if  $\forall p_1, p_2 \in Q: \|y(p_2) - y(p_1)\| \leq \alpha \|p_1 - p_2\|$ , where  $\alpha \in (0,1)$ .

Theorem1 [26]: A contractive function  $y$  on a complete normed space  $Q$  has a unique FP  $n^*$  in  $Q$ .

Theorem2 [27]: Let  $\{v_n\}$  be a bounded sequence in the space  $L^\infty(Q)$ . Then, there exists a subsequence  $\{\theta\}$  and a function  $v_0 \in L^\infty(Q)$  s.t,  $v_{\theta} \rightarrow v_0$  in  $L^\infty(Q)$ .

## 3. Description of the NLHPDE

Let  $S = [0, T]$ , with  $0 < T < \infty$ ,  $Q \subset R^2$  be a bounded region with boundary  $\partial Q$ ,  $\vartheta = Q \times S$ ,  $E = \partial Q \times S$ . Then, the NLHPDE is given by:

$$X_{tt} - \Delta X + X = g(\vec{r}, t, X), \text{ in } \vartheta \tag{1}$$

$$X(\vec{r}, 0) = X^0(\vec{r}), \text{ in } Q \tag{2}$$

$$X_t(\vec{r}, 0) = X^1(\vec{r}), \text{ in } Q \tag{3}$$

$$X(\vec{r}, t) = 0, \text{ on } E \tag{4}$$

where  $X = X(\vec{r}, t) \in H_0^2(Q)$ ,  $\Delta X = \sum_{i=1}^2 \frac{\partial^2 X}{\partial r_i^2}$ ,  $\vec{r} = (r_1, r_2) \in R^2$ ,  $0 < r_1, r_2 < 1$ , and  $g \in L^2(Q)$ .

Now, let  $V = H_0^1(Q) = \{\sigma: \sigma \in H^1(Q), \sigma = 0 \text{ on } \partial Q\}$  and  $X_t = p$ . Then, the weak form of (1-4) is:

$$\langle X_{tt}, \sigma \rangle + (\nabla X, \nabla \sigma) + (X, \sigma) = (g(X), \sigma), \forall \sigma \in V \tag{5}$$

$$(X(0), \sigma) = (X^0, \sigma) \text{ in } Q, X^0 \in V \tag{6}$$

$$(p(0), \sigma) = (X^1, \sigma) \text{ in } Q, X^1 \in L^2(\vartheta) \tag{7}$$

The following assumptions are necessary to study both the existence and the convergence of the solution.

### Assumptions (ASM):

(1) Let  $q_1$  and  $q_2$  be a nonnegative constants that satisfy the following:

$$a) |(\nabla X, \nabla \sigma)| \leq q_1 \|\nabla X\|_1 \|\nabla \sigma\|_1, \forall X, \sigma \in V$$

$$b) (\nabla X, \nabla X) \geq q_2 \|\nabla X\|_1^2, \forall X \in V$$

(2)  $g$  is continuous w.r.t  $X_j^n$  and defined on  $Q \times R$ , and satisfies:

$$a) |g(\vec{r}, t_j^n, X_j^n)| \leq e(\vec{r}, t) + \gamma |X_j^n| \text{ where, } \gamma > 0, X_j^n \in Q \text{ and } e \in L^2(\vartheta).$$

$$b) |g(\vec{r}, t_j^n, X_j^n) - g(\vec{r}, t_j^n, X_i^n)| \leq L |X_j^n - X_i^n|, \text{ where } L \text{ is a Lipchitz constant and } X_j^n, X_i^n \in Q.$$

## 4. Discretization of the Problem:

In this section, the method of GM is applied for discretizing ((5)-(7)). Let  $\vartheta$  be divided into sub regions  $\vartheta_{ij} = Q_i^n \times S_j^n$ , and let  $\{Q_i^n\}_{i=1}^{N(n)}$  be a triangulation of  $\bar{Q}$  and  $\{S_j^n\}_{j=0}$  be a subdivision of the interval  $S$  into  $Y(n)$  intervals. Then,  $S_j = S_j^n := [t_j^n, t_{j+1}^n]$  has the same length  $\Delta t = \frac{T}{Y}$ . Also, let  $V_n \subset V = H_0^1(Q)$  be the space of piecewise affine functions in  $Q$ .

Now, the following weak form is obtained from using CN formula and is given  $\forall \sigma \in V_n$  as:

$$(p_{j+1}^n - p_j^n, \sigma) + \Delta t \left( \nabla X_{\frac{1}{2}j}^n, \nabla \sigma \right) + \Delta t \left( X_{\frac{1}{2}j}^n, \sigma \right) = \Delta t \left( g(t_j^n, X_{\frac{1}{2}j}^n), \sigma \right) \quad (8)$$

Since,  $\Delta t(p_{j+1}^n) = X_{j+1}^n - X_j^n$  then, (8) becomes

$$\begin{aligned} (X_{j+1}^n - X_j^n, \sigma) + (\Delta t)^2 \left( \nabla X_{\frac{1}{2}j}^n, \nabla \sigma \right) + (\Delta t)^2 \left( X_{\frac{1}{2}j}^n, \sigma \right) = \\ (\Delta t)(p_j^n, \sigma) + (\Delta t)^2 \left( g(t_j^n, X_{\frac{1}{2}j}^n), \sigma \right) \end{aligned} \quad (9)$$

Where  $X_{\frac{1}{2}j}^n = \frac{1}{2}(X_{j+1}^n + X_j^n)$

$$(X(0), \sigma) = (X^0, \sigma), \text{ in } Q \quad (10)$$

$$(p(0), \sigma) = (X^1, \sigma), \text{ in } Q \quad (11)$$

where  $X^0 \in V, X^1 \in L^2(Q), X_j^n = X^n(r, t_j^n)$ , and  $p_j^n = p^n(\vec{r}, t_j^n) \in V_n, \forall j = 0, 1, \dots, Y - 1$ .

### 5. The Approximation Solution of the NLHPDE:

In this part, the MCNGM is used to find the APSOL  $\bar{X}^n = (X_0^n, X_1^n, \dots, X_Y^n)$  for the DWF (8-11) through the following steps:

1- From the basis of  $V_n$  and by using the MCNGM, let  $\bar{X}^n(\vec{r}, t_j^n)$  (with  $\bar{X}_t^n(\vec{r}, t_j^n) = \bar{p}(\vec{r}, t_j^n)$ )

be an APSOL of ((8)-(11)) s.t:

$$\bar{X}^n(\vec{r}, t_j^n) = \sum_{k=1}^N f_k^j \sigma_i \text{ and } \bar{p}^n(\vec{r}, t_j^n) = \sum_{k=1}^N w_k^j \sigma_i, \forall \mu_i \in V_n,$$

where  $f_k^j = f_k(t_j^n)$ , and  $w_k^j = w_k(t_j^n)$  are unknown to be determined,  $\forall j = 0, 1, \dots, Y - 1$ .

2-Using the APSOL in ((8)-(11)),  $\forall j = 0, 1, \dots, Y - 1$ , given:

$$\begin{aligned} \left( \left( 1 + \frac{1}{2}(\Delta t)^2 \right) D + \frac{1}{2}(\Delta t)^2 Z \right) F^{j+1} = \\ \left( \left( 1 - \frac{1}{2}(\Delta t)^2 \right) D - \frac{1}{2}(\Delta t)^2 Z \right) F^j + (\Delta t)DW^j + (\Delta t)^2 \vec{U} \end{aligned} \quad (12)$$

$$W^{j+1} = \frac{1}{\Delta t} (F^{j+1} - F^j) \quad (13)$$

$$DF^0 = \vec{U}^0 \quad (14)$$

$$DW^0 = \vec{U}^1 \quad (15)$$

Where  $D = (d_{ik})_{N \times N}$ ,  $d_{ik} = (\sigma_k, \sigma_i)$ ,  $Z = (z_{ik})_{N \times N}$ ,  $z_{ik} = (\nabla \sigma_k, \nabla \sigma_i)$ ,  $U = \left( g \left( t_j^n, X_{\frac{1}{2}j}^n \right), \sigma_i \right)$

$F_{N \times 1}^j = (f_1^j, f_2^j, \dots, f_N^j)^T$ ,  $W_{N \times 1}^j = (w_1^j, w_2^j, \dots, w_N^j)^T$ ,  $\vec{U}^0 = (u_i^0)_{N \times 1}$ ,  $u_i^0 = (X^0, \sigma_i)$ , and

$\vec{U}^1 = (u_i^1)_{N \times 1}$  with  $u_i^1 = (X^1, \sigma_i)$ ,  $\forall i, k = 1, 2, \dots, N$ .

3-The GNLAS ((12)-(15)) has a unique solution. To solve it, first and from solving (14) and

(15) respectively, the solutions  $F^0$  and  $W^0$  are found, then (12) is solved by using the PrCoT (for each  $j = 0, 1, \dots, Y - 1$ ) as:

In the PrT, suppose that  $F^{j+1} = F^j$  in the elements of  $\vec{U}$  (in the RHS), then it convert to a GLAS, which is solved to get  $F^{j+1}$ , then resolve (12) with setting  $\bar{F}^{j+1} = F^{j+1}$  (in the elements of  $\vec{U}$ ) to get the corrector solution (CoS)  $F^{j+1}$ , finally  $W^{j+1}$  is obtained from setting  $F^{j+1}$  in (13); of course this technique can be repeated for more than one time. This PrCoT can be expressed as:

$$\begin{aligned} & \left( X_{j+1}^{(l+1)}, \sigma_i \right) + (\Delta t)^2 \left( \nabla X_{\frac{1}{2}j}^{(l+1)}, \nabla \sigma_i \right) + (\Delta t)^2 \left( X_{\frac{1}{2}j}^{(l+1)}, \sigma_i \right) = \\ & \left( X_j^n, \sigma_i \right) + (\Delta t) \left( p_j^n, \sigma_i \right) + (\Delta t)^2 \left( g \left( t_j^n, X_{\frac{1}{2}j}^{(l)} \right), \sigma_i \right) \end{aligned} \quad (16)$$

$$p_{j+1}^{(l+1)} = \frac{(X_{j+1}^{(l+1)} - X_j^n)}{\Delta t} \quad (17)$$

Equation (17) shows that the iterative method depends only on  $X_{j+1}^{(l+1)}$ , where  $l$  represents the number of iterations.

**Theorem:** For “sufficiently small”  $\Delta t$  and for any fixed  $j$  ( $0 \leq j \leq Y - 1$ ), the DWF ((8)-(11)), has a unique solution  $X^n = (X_0^n, X_1^n, \dots, X_N^n)$  and the sequence of the CoS converges in  $R$ .

**Proof:** Let  $X^{(l+1)} = (X_0^{(l+1)}, X_1^{(l+1)}, \dots, X_N^{(l+1)})$ , and  $\bar{X}^{(l+1)} = (\bar{X}_0^{(l+1)}, \bar{X}_1^{(l+1)}, \dots, \bar{X}_N^{(l+1)})$ , where  $X^{(l+1)}$  and  $\bar{X}^{(l+1)}$  are the solutions of equation (16). Hence

$$\begin{aligned} & \left( X_{j+1}^{(l+1)}, \sigma_i \right) + (\Delta t)^2 \left( \nabla X_{\frac{1}{2}j}^{(l+1)}, \nabla \sigma_i \right) + (\Delta t)^2 \left( X_{\frac{1}{2}j}^{(l+1)}, \sigma_i \right) = \left( X_j^n, \sigma_i \right) + \Delta t \left( p_j^n, \sigma_i \right) + \\ & (\Delta t)^2 \left( g \left( t_j^n, X_{\frac{1}{2}j}^{(l)} \right), \sigma_i \right) \end{aligned} \quad (18)$$

and

$$\begin{aligned} & \left( \bar{X}_{j+1}^{(l+1)}, \sigma_i \right) + (\Delta t)^2 \left( \nabla \bar{X}_{\frac{1}{2}j}^{(l+1)}, \nabla \sigma_i \right) + (\Delta t)^2 \left( \bar{X}_{\frac{1}{2}j}^{(l+1)}, \sigma_i \right) = \\ & \left( X_j^n, \sigma_i \right) + \Delta t \left( p_j^n, \sigma_i \right) + (\Delta t)^2 \left( g \left( t_j^n, \bar{X}_{\frac{1}{2}j}^{(l)} \right), \sigma_i \right) \end{aligned} \quad (19)$$

Subtracting (19) from (18) then setting  $\sigma_i = (\bar{X}_{j+1}^{(l+1)} - X_{j+1}^{(l+1)})$ , it yields to

$$\begin{aligned} & \left( \bar{X}_{j+1}^{(l+1)} - X_{j+1}^{(l+1)}, \bar{X}_{j+1}^{(l+1)} - X_{j+1}^{(l+1)} \right) + (\Delta t)^2 \left( \nabla \bar{X}_{\frac{1}{2}j}^{(l+1)} - \nabla X_{\frac{1}{2}j}^{(l+1)}, \nabla \bar{X}_{j+1}^{(l+1)} - \nabla X_{j+1}^{(l+1)} \right) \\ & + (\Delta t)^2 \left( \bar{X}_{\frac{1}{2}j}^{(l+1)} - X_{\frac{1}{2}j}^{(l+1)}, \bar{X}_{j+1}^{(l+1)} - X_{j+1}^{(l+1)} \right) = \\ & (\Delta t)^2 \left( g \left( t_j^n, \bar{X}_{\frac{1}{2}j}^{(l)} \right) - g \left( t_j^n, X_{\frac{1}{2}j}^{(l)} \right), \bar{X}_{j+1}^{(l+1)} - X_{j+1}^{(l+1)} \right) \end{aligned} \quad (20)$$

From ASM (1 - b) the  $2^d$  and  $3^d$  terms in the LHS of (20) are nonnegative, and from ASM (2- b) on  $g$  in RHS of (20), and then by applying the Cauchy Schwarz inequality to obtain

$$\left\| \bar{X}_{j+1}^{(l+1)} - X_{j+1}^{(l+1)} \right\|_0 \leq \xi \left\| \bar{X}_{\frac{1}{2}j}^{(l)} - X_{\frac{1}{2}j}^{(l)} \right\|_0, \xi = \frac{(\Delta t)^2 L}{(1 + (\Delta t)^2)} \quad (21)$$

But  $\xi < 1$ , (for “sufficiently small”  $\Delta t$ ), Which leads to that  $g$  is contractive. Also since  $\{X^{(l)}\} \in R$  for each  $l$ , that  $(\bar{X}^{(l+1)}) = X^{(l+1)} \in R$  for each  $l$ , i. e  $g(X) \in R$ , and by theorem 2 we get that  $\{X^{(l)}\}$  is converged to a point in  $R$ .

## 6. Numerical Examples

In this section, some numerical examples are carried out to show the efficiency and the accuracy for the proposed method.

**Example 1:** Consider the following NLHPDE:

$$X_{tt} - \Delta X + X = g(\vec{r}, t, X), \quad \vec{r} = (r_1, r_2), \vartheta = Q \times S, Q = (0,1) \times (0,1), S = [0,1]$$

$$X(\vec{r}, 0) = r_1 r_2 - r_1^2 r_2 - r_1 r_2 (r_2 - r_1 r_2), \text{ in } Q$$

$$X_t(\vec{r}, 0) = X^1(\vec{r}), \text{ in } Q$$

$$X(\vec{r}, t) = 0, \text{ on } E = \partial Q \times S$$

$$\text{Where, } g(\vec{r}, t, X) = e^{-2t} \left[ (r_1 r_2 - r_1^2 r_2 - r_1 r_2 (r_2 - r_1 r_2)) \left( 5 - \sin \left( (r_1 r_2 - r_1^2 r_2 - r_1 r_2 (r_2 - r_1 r_2)) e^{-2t} \right) \right) + 2(r_1(1 - r_1) + r_2(1 - r_2)) \right].$$

Where the exact solution of the problem is

$$X(\vec{r}, t) = (r_1 r_2 - r_1^2 r_2 - r_1 r_2 (r_2 - r_1 r_2)) e^{-2t}$$

The MCNGM was used to solve this problem with D=9, Y=20 and T=1, the numerical results are given at  $\hat{t} = 0.5$  in the Table (1) and are shown in Figure (1).

Table1- Comparison between exact and approximation solutions

$r_1$	$r_2$	Exact	Approximation	Absolute error	$r_1$	$r_2$	Exact	Approximation	Absolute error
0.1	0.1	0.0030	0.0026	0.0004	0.5	0.5	0.0230	0.0192	0.0038
0.2	0.1	0.0053	0.0046	0.0007	0.6	0.5	0.0221	0.0185	0.0036
0.3	0.1	0.0070	0.0061	0.0009	0.7	0.5	0.0193	0.0163	0.0030
0.4	0.1	0.0079	0.0069	0.0010	0.8	0.5	0.0147	0.0125	0.0022
0.5	0.1	0.0083	0.0071	0.0012	0.9	0.5	0.0083	0.0071	0.0012
0.6	0.1	0.0079	0.0067	0.0012	0.1	0.6	0.0079	0.0067	0.0012
0.7	0.1	0.0070	0.0059	0.0011	0.2	0.6	0.0141	0.0119	0.0022
0.8	0.1	0.0053	0.0045	0.0008	0.3	0.6	0.0185	0.0156	0.0029
0.9	0.1	0.0030	0.0026	0.0004	0.4	0.6	0.0212	0.0177	0.0035
0.1	0.2	0.0053	0.0046	0.0007	0.5	0.6	0.0221	0.0185	0.0036

0.2	0.2	0.0094	0.0082	0.0012	0.6	0.6	0.0212	0.0178	0.0034
0.3	0.2	0.0124	0.0107	0.0017	0.7	0.6	0.0185	0.0158	0.0027
0.4	0.2	0.0141	0.0121	0.0020	0.8	0.6	0.0141	0.0121	0.0020
0.5	0.2	0.0147	0.0125	0.0022	0.9	0.6	0.0079	0.0069	0.0010
0.6	0.2	0.0141	0.0119	0.0022	0.1	0.7	0.0070	0.0059	0.0011
0.7	0.2	0.0124	0.0105	0.0019	0.2	0.7	0.0124	0.0105	0.0019
0.8	0.2	0.0094	0.0080	0.0014	0.3	0.7	0.0162	0.0137	0.0025
0.9	0.2	0.0053	0.0045	0.0008	0.4	0.7	0.0185	0.0156	0.0029
0.1	0.3	0.0070	0.0061	0.0009	0.5	0.7	0.0193	0.0163	0.0030
0.2	0.3	0.0124	0.0107	0.0017	0.6	0.7	0.0185	0.0158	0.0027
0.3	0.3	0.0162	0.0139	0.0023	0.7	0.7	0.0162	0.0139	0.0023
0.4	0.3	0.0185	0.0158	0.0027	0.8	0.7	0.0124	0.0107	0.0017
0.5	0.3	0.0193	0.0163	0.0030	0.9	0.7	0.0070	0.0061	0.0009
0.6	0.3	0.0185	0.0156	0.0029	0.1	0.8	0.0053	0.0045	0.0008
0.7	0.3	0.0162	0.0137	0.0025	0.2	0.8	0.0094	0.0080	0.0014
0.8	0.3	0.0124	0.0105	0.0019	0.3	0.8	0.0124	0.0105	0.0019
0.9	0.3	0.0070	0.0059	0.0011	0.4	0.8	0.0141	0.0119	0.0022
0.1	0.4	0.0079	0.0069	0.0010	0.5	0.8	0.0147	0.0125	0.0022
0.2	0.4	0.0141	0.0121	0.0020	0.6	0.8	0.0141	0.0121	0.0020

0.3	0.4	0.0185	0.0158	0.0027	0.7	0.8	0.0124	0.0107	0.0017
0.4	0.4	0.0212	0.0178	0.0034	0.8	0.8	0.0094	0.0082	0.0012
0.5	0.4	0.0221	0.0185	0.0036	0.9	0.8	0.0053	0.0046	0.0007
0.6	0.4	0.0212	0.0177	0.0035	0.1	0.9	0.0030	0.0026	0.0004
0.7	0.4	0.0185	0.0156	0.0029	0.2	0.9	0.0053	0.0045	0.0008
0.8	0.4	0.0141	0.0119	0.0022	0.3	0.9	0.0070	0.0059	0.0011
0.9	0.4	0.0079	0.0067	0.0012	0.4	0.9	0.0079	0.0067	0.0012
0.1	0.5	0.0083	0.0071	0.0012	0.5	0.9	0.0083	0.0071	0.0012
0.2	0.5	0.0147	0.0125	0.0022	0.6	0.9	0.0079	0.0069	0.0010
0.3	0.5	0.0193	0.0163	0.0030	0.7	0.9	0.0070	0.0061	0.0009
0.4	0.5	0.0221	0.0185	0.0036	0.8	0.9	0.0053	0.0046	0.0007
					0.9	0.9	0.0030	0.0026	0.0004

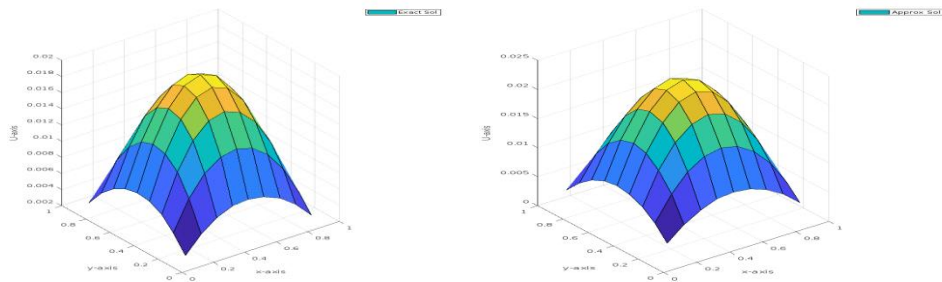


Figure1: shows the exact and the approximation solution

**Example 2:** Consider the following NLHPDE:

$$X_{tt} - \Delta X + X = g(\vec{r}, t, X), \quad \vec{r} = (r_1, r_2), \vartheta = Q \times S, Q = (0,1) \times (0,1), S = [0,1]$$

$$X(\vec{r}, 0) = r_1^2 r_2 \sin(r_1 r_2 - r_1 - r_2 + 1), \text{ in } Q$$

$$X_t(\vec{r}, 0) = X^1(\vec{r}), \text{ in } Q$$

$$X(\vec{r}, t) = 0, \text{ on } E = \partial Q \times S$$

Where,  $g(\vec{r}, t, X) = e^{-2t}[r_2 \sin(r_1 r_2 - r_1 - r_2 + 1) (5r_1^2 - 2 - r_1^2 \sin(r_1^2 r_2 \sin(r_1 r_2 - r_1 - r_2 + 1) e^{-2t}) + (r_1^4 - 2r_1^3 + r_1^2) + (r_1^2 r_2^2 - 2r_1^2 r_2 + r_1^2)) - 2r_1 \cos(r_1 r_2 - r_1 - r_2 + 1)(r_1(r_1 - 1) + 2r_2(r_2 - 1))]$ .

Where the exact solution of this problem is  $X(\vec{r}, t) = r_1^2 r_2 \sin(r_1 r_2 - r_1 - r_2 + 1) e^{-2t}$

The MCNGM was used to solve this problem with  $D=9$ ,  $Y=20$  and  $T=1$ , the numerical results are given at  $\hat{t} = 0.5$  in the Table (2) and are shown in Figure (2).

Table2: Comparison between exact and approximation solutions

$r_1$	$r_2$	exact	approximation	Absolute error	$r_1$	$r_2$	exact	approximation	Absolute error
0.1	0.1	0.0003	0.0002	0.0001	0.5	0.5	0.0114	0.0095	0.0019
0.2	0.1	0.0010	0.0009	0.0001	0.6	0.5	0.0132	0.0109	0.0023
0.3	0.1	0.0020	0.0017	0.0003	0.7	0.5	0.0135	0.0112	0.0023
0.4	0.1	0.0030	0.0026	0.0004	0.8	0.5	0.0118	0.0099	0.0019
0.5	0.1	0.0040	0.0034	0.0006	0.9	0.5	0.0074	0.0063	0.0011
0.6	0.1	0.0047	0.0039	0.0008	0.1	0.6	0.0008	0.0007	0.0001
0.7	0.1	0.0048	0.0040	0.0008	0.2	0.6	0.0028	0.0025	0.0003
0.8	0.1	0.0042	0.0036	0.0006	0.3	0.6	0.0055	0.0048	0.0007
0.9	0.1	0.0027	0.0023	0.0004	0.4	0.6	0.0084	0.0071	0.0013
0.1	0.2	0.0005	0.0004	0.0001	0.5	0.6	0.0110	0.0092	0.0018
0.2	0.2	0.0018	0.0015	0.0003	0.6	0.6	0.0127	0.0106	0.0021
0.3	0.2	0.0035	0.0031	0.0004	0.7	0.6	0.0129	0.0109	0.0020
0.4	0.2	0.0054	0.0047	0.0007	0.8	0.6	0.0113	0.0096	0.0017
0.5	0.2	0.0072	0.0061	0.0011	0.9	0.6	0.0071	0.0062	0.0009
0.6	0.2	0.0083	0.0070	0.0013	0.1	0.7	0.0007	0.0006	0.0001
0.7	0.2	0.0086	0.0072	0.0014	0.2	0.7	0.0024	0.0022	0.0002



0.8	0.2	0.0075	0.0063	0.0012	0.3	0.7	0.0048	0.0042	0.0006
0.9	0.2	0.0048	0.0040	0.0008	0.4	0.7	0.0074	0.0063	0.0011
0.1	0.3	0.0007	0.0006	0.0001	0.5	0.7	0.0096	0.0081	0.0015
0.2	0.3	0.0023	0.0021	0.0002	0.6	0.7	0.0111	0.0094	0.0017
0.3	0.3	0.0047	0.0041	0.0006	0.7	0.7	0.0113	0.0097	0.0016
0.4	0.3	0.0072	0.0062	0.0010	0.8	0.7	0.0099	0.0085	0.0014
0.5	0.3	0.0095	0.0080	0.0015	0.9	0.7	0.0063	0.0054	0.0009
0.6	0.3	0.0110	0.0092	0.0018	0.1	0.8	0.0005	0.0005	0.0000
0.7	0.3	0.0113	0.0094	0.0019	0.2	0.8	0.0019	0.0017	0.0002
0.8	0.3	0.0099	0.0083	0.0016	0.3	0.8	0.0037	0.0033	0.0004
0.9	0.3	0.0063	0.0053	0.0010	0.4	0.8	0.0056	0.0048	0.0008
0.1	0.4	0.0008	0.0007	0.0001	0.5	0.8	0.0073	0.0062	0.0011
0.2	0.4	0.0027	0.0024	0.0003	0.6	0.8	0.0085	0.0072	0.0013
0.3	0.4	0.0054	0.0047	0.0007	0.7	0.8	0.0086	0.0075	0.0011
0.4	0.4	0.0083	0.0071	0.0012	0.8	0.8	0.0075	0.0066	0.0009
0.5	0.4	0.0109	0.0091	0.0018	0.9	0.8	0.0048	0.0042	0.0006
0.6	0.4	0.0126	0.0105	0.0021	0.1	0.9	0.0003	0.0003	0.0000
0.7	0.4	0.0129	0.0107	0.0022	0.2	0.9	0.0011	0.0010	0.0001
0.8	0.4	0.0113	0.0094	0.0019	0.3	0.9	0.0021	0.0018	0.0003
0.9	0.4	0.0071	0.0060	0.0011	0.4	0.9	0.0032	0.0027	0.0005
0.1	0.5	0.0008	0.0007	0.0001	0.5	0.9	0.0041	0.0035	0.0006
0.2	0.5	0.0029	0.0025	0.0004	0.6	0.9	0.0048	0.0041	0.0007
0.3	0.5	0.0057	0.0049	0.0008	0.7	0.9	0.0049	0.0042	0.0007
0.4	0.5	0.0087	0.0074	0.0013	0.8	0.9	0.0042	0.0037	0.0006
					0.9	0.9	0.0027	0.0024	0.0003

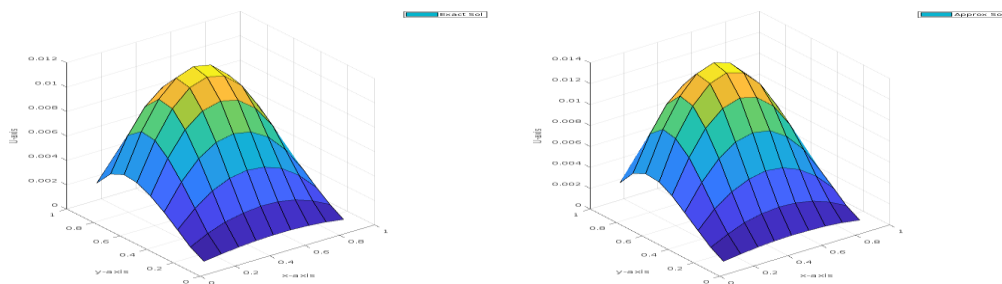


Figure2: shows the exact and the approximation solutions

## 7. Conclusions

From the solutions for the above given examples, one can conclude that:

The MCNGM was used successfully to find the APSOL of the NLHPDES. The uniqueness of the APSOL for the weak form (which was obtained from the PrCrT) was proved. The GFEM was applied easily and the elements in the GNAS are in analytic form (exact) comparing with other methods that the elements are in approximate or in a full discrete form. The cholesky method which used inside the PrCrT was very efficient and fast to solve the GLAS because it saves a lot of calculations. The approximation solution for the two examples illustrate the accuracy and the efficiency of the proposed method. It is important to mention here that the approximate vector solution are given at the value of  $\hat{t} = 0.5$  to brief the size of the paper, in fact same results with same accuracy were obtained at any value of  $\hat{t}$  provided this value belong to  $S$ .

## 8. References

- [1] Abd Al-Haq A. J. M. (2017). "Numerical Methods for Solving Hyperbolic Type problems". This thesis was defended successfully on 92(3).
- [2] Gavete L., Ureña F., Benito J. J., García A., Ureña M. and Salet E. (2017). "Solving second order non-linear elliptic partial differential equations using generalized finite difference method". *Journal of Computational and Applied Mathematics*, 318, 378-387.
- [3] Ureña F., Gavete L., Garcia A., Benito J. J. and Vargas A. M. (2019). "Solving second order non-linear parabolic PDEs using generalized finite difference method (GFDM)". *Journal of Computational and Applied Mathematics*, 354, 221-241.
- [4] Gavete L., Benito J. J. and Ureña F. (2016). "Generalized finite differences for solving 3D elliptic and parabolic equations". *Applied Mathematical Modelling*, 40(2), 955-965.
- [5] Hyman J. M. (2018). "Method of Lines Solution of Partial Differential Equations (Classic Reprint)", Forgotten Books.
- [6] Gao X. W., Zhu Y. M. and Pan T. (2023). "Finite line method for solving high-order partial differential equations in science and engineering". *Partial Differential Equations in Applied Mathematics*, 7, 100477.

- [7] Hajiketabi M. and Abbasbandy S. (2018). "The combination of meshless method based on radial basis functions with a geometric numerical integration method for solving partial differential equations: Application to the heat equation". *Engineering Analysis with Boundary Elements*, 87, 36-46.
- [8] Yang J., Hu H., Koutsawa Y. and Potier-Ferry M. (2017). "Taylor meshless method for solving non-linear partial differential equations". *Journal of Computational Physics*, 348, 385-400.
- [9] Raju I., Banu M. S., Mim S. A., Hossain S. and Saha H. K. (2022). "A Case study on simulation of heat equation by Crank-Nicolson Method in Accordance with digital image processing". *International Journal of Scientific & Engineering*,13(1).
- [10] Abdulwahab and Olanipekun S. (2013). "Application of Crank-Nicolson Finite-Difference Method to the Solution of the Dynamic Model of a Reactor". *International Journal of Advanced Scientific and Technical Research*,6(3).
- [11] Xia B, Wu H,Qiu Y. and Song Y. (2018). "A Galerkin Method-based Polynomial Approximation for Parametric Problems in Power System Transient Analysis", *journal Of IEEE Pes Transactions On Power Systems*.
- [12] Samarinas N., Tzimopoulos C. and Evangelides C. (2022). "Inefficient method to solve the fuzzy Crank-Nicolson scheme with application to the groundwater flow problem". *journal of hydro informatics*, 24(3).
- [13] Secer A. (2018)."Sinc-Galerkin method for solving hyperbolic partial differential equations". *An International Journal of Optimization and Control: Theories & Applications*, 8(2), pp.250-258.
- [14] Al-Hawasy J. A. and Jawad M. A. (2019). "Approximation Solution of Nonlinear Parabolic Partial Differential Equation via Mixed GalerkinFinite Elements Method with the Crank-Nicolson Scheme". *Iraqi Journal of Science*,60(2), pp:353-36.
- [15] Gorbova T.V., Pimenov V.G., and Solodushkin S.I. (2020). "Crank–Nicolson Numerical Algorithm for Nonlinear Partial Differential Equation with Heredity and Its Program Implementation". *Springer Nature Switzerland AG*.
- [16] Zheng X., Qiu W. and Hongbin C. (2020). "A Crank–Nicolson-type finite-difference scheme, and itsalgorithm implementation for a nonlinear partial integro-differential equation arising from viscoelasticity". *Computational and Applied Mathematics* , 39-295.
- [17] Anaç H., Merdan M. and Kesemen T. (2020). "Application Of Crank-Nicolson Method To A Random Component Heat Equation". *Journal of Engendering & Natural Sciences*,38 (1), pp.475-480.
- [18] Al-Hawasy J. A. A. and Mansour N. F. (2021). "The Galerkin-Implicit Method for Solving Nonlinear Variable Coefficients Hyperbolic Boundary Value Problem". *Iraqi Journal of Science*, 62(11),pp: 3997-4005.
- [19] Li L., She M. and Niu Y. (2021). "Fractional Crank-Nicolson-Galerkin Finite Element Methods for Nonlinear Time Fractional Parabolic Problems with Time Delay" *Journal of Function Spaces Volume 2021*, Article ID 9981211,<https://doi.org/10.1155/2021/9981211>.

- [20] Abdulazeez S. T., Modanli M. and Husien A. M. (2022). "Numerical Scheme Methods For Solving Nonlinear Pseudo-Hyperbolic Partial Differential Equations". *Journal of Applied Mathematics and Computational Mechanics*, 21(4), 5-15.
- [21] Hajinezhad H. and Soheili A. R. (2022). "A numerical approximation for the solution of a time-fractional telegraph equation based on the Crank–Nicolson method". *Iranian Journal of Numerical Analysis and Optimization*, 12(3), pp 607–628.
- [22] Zeng Y. and Luo Z. (2022). "The Crank-Nicolson Mixed finite element Method for the improved system of time-domain Maxwell's equations". *Journal of Applied Mathematics and computer*, 433(c).
- [23] Liu J. and Hao Y. (2022). "Crank-Nicolson for solving uncertain heat equation". *Soft comput*, 26(3).
- [24] pata and vittorino (2019). "Fixed Point theorems and applications". Springer international publishing.
- [25] Berinde V. and Păcurar M. (2020). "Approximating fixed points of enriched contractions in Banach spaces". *Journal of Fixed Point Theory and Applications*, 22, 1-10.
- [26] Suhas P. and Uttam D. (2005). "Random Fixed Point Theorems for Contraction Mappings, in MetricSpace". *International Journal of Science and Research (IJSR)* ISSN: 2319-7064.
- [27] Al-Hawasy J. A. and Jasim D. K. (2020). "The Continuous Classical Optimal Control Problems for Triple Elliptic Partial Differential Equations". *IHJPAS.*, 33(1), 143-151, DOI:10.30526/33.1.2380.