An inverse problem of reconstructing the time-dependent coefficient in a one-dimensional hyperbolic equation

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Abstract

In this paper, for the first time the inverse problem of reconstructing the time-dependent potential (TDP) and displacement distribution in the hyperbolic problem with periodic boundary conditions (BCs) and nonlocal initial supplemented by over-determination measurement is numerically investigated. Though the inverse problem under consideration is ill-posed by being unstable to noise in the input data, it has a unique solution. The Crank–Nicolson finite difference method (CN-FDM) along with the Tikhonov regularization (TR) is applied for calculating an accurate and stable numerical solution. The programming language MATLAB built-in lsqnonlin is used to solve the obtained nonlinear minimization problem. The simulated noisy input data can be inverted by both analytical and numerically simulated. The obtained results show that they are accurate and stable. The stability analysis is performed by using Fourier series.

Keywords: Hyperbolic equation; Inverse problem; Periodic boundary; Integral boundary; Tikhonov regularization; Optimization

1 Introduction

The reconstruction of the unknown coefficients in the inverse problem of the hyperbolic problem has various applications in science and engineering. In the last few decades, various authors have reconstructed the unknown coefficients in the inverse problem of the hyperbolic wave equations. For example, Bakushinsky and Leonov [2] recovered the space-dependent coefficient from integral data, while authors of [5, 24, 25] determined the time-dependent source coefficients. Cannon and Dunninger [7] reconstructed a force function from over-specified data. Further, Cannon and DuChateau [6] determined both time- and space-dependent coefficients. Stefanov and Uhlmann [27] recovered a source term in anisotropic media. Additionally, Bellassoued and Yamamoto [3] studied the inverse problem to determine the unknown term in the hyperbolic model with variable terms. Boumenir and Tuan [4] showed the process for reconstructing the unknown coefficient in the inverse problem of the wave equation from a finite number of special lateral measurements. Yamamoto [31] considered an inverse problem for identifying space-dependent function from Neumann BCs and showed a TR and reconstruction formula. Yang et al.
[32] recovered heat source term in the inverse problem of heat conduction equation using Tikhonov regularization. Huntul [12] identified the unknown time-dependent coefficient in the third-order equation from nonlocal integral observation.

Recently, the inverse problems of the wave equations for recovering time-dependent potential from over-determination integral condition have been investigated by Tekin [29] while the time-dependent force function has been studied by Hussein and Lesnic [19]. The authors of [13–15, 17] studied the inverse problems for identifying the time-wise potential terms in third and fourth-order equations. Huntul et al. [18] investigated an inverse problem to reconstruct the time-wise potential terms in a wave equation as an over-determination condition. In [10, 11, 20], authors recovered the time- and space-dependent source functions. Huntul and Tamsir [16] investigated an inverse problem to recover a time-wise heat source from the integral condition. Still, the inverse problem of reconstructing the time-wise potential coefficient numerically for the hyperbolic problems with integral and periodic BCs is inadequate in the literature.

In this paper, the time-dependent potential is reconstructed for the first time numerically in a one-dimensional hyperbolic problem with periodic and integral BCs from the over-determination estimation. Azizbayov [1] has already proved that this problem is locally uniquely solvable, but no numerical realization has been carried out till now, which is the main contribution of proposed work.

The paper is organized as follows: The research problem is stated in Sect. 2. Sect. 3 discretizes the direct problem using CN-FDM. Stability analysis is presented in Sect. 4, while the minimization technique is given in Sect. 5. Some numerical results are tabulated in Sect. 6, while the conclusions are highlighted in Sect. 7.

2 Formulation of the inverse problem

We consider an inverse problem of reconstructing an unknown time-dependent potential coefficient \( \alpha(t) \) in the one-dimensional hyperbolic equation

\[
\frac{\partial^2 z}{\partial t^2} - \frac{\partial^2 z}{\partial r^2} = \alpha(t)z + s(r, t), \quad (r, t) \in D_T := (0, 1) \times (0, T),
\]

where \( z = z(r, t) \) is an unknown displacement, \( s(r, t) \) is a known source term which is as distributed force. The nonlocal initial conditions (ICds)

\[
z(r, 0) = \int_0^T P_1(t)z(r, t) \, dt + \xi(r), \quad z_t(r, 0) = \int_0^T P_2(t)z(r, t) \, dt + \zeta(r), \quad r \in [0, 1],
\]

the periodic BCs

\[
z(0, t) = z(1, t), \quad t \in [0, T],
\]

the integral boundary condition

\[
\int_0^1 z(r, t) \, dr = 0, \quad t \in [0, T],
\]

and the over-determination condition

\[
z(R_0, t) = q(t), \quad t \in [0, T],
\]
where \( R_0 \in (0,1) \) is some fixed point, the functions \( P_1(t), P_2(t), \xi(r), \zeta(r), \) and \( q(t) \) are given.

The numerical solution of the inverse problem hyperbolic wave equation (1)–(5) is written as \( \{\alpha(t), z(r,t)\} \) such that \( \alpha(t) \in C[0,T] \) and \( z(r,t) \in C^2(\partial \mathcal{D}_T) \). Along with inverse problem (1)–(5), the following auxiliary inverse problem is considered. It is needed to find \( \{\alpha(t), z(r,t)\} \in C[0,T] \times C^2(\partial \mathcal{D}_T) \) from (1)–(3), and

\[
\begin{align*}
\xi_t(0, t) &= z(1,t), \quad t \in [0, T], \\
q''(t) - z_{rr}(R_0, t) &= \alpha(t)q(t) + s(R_0, t), \quad t \in [0, T].
\end{align*}
\]

The upcoming theorems are taken from [1], and they read as follows.

**Theorem 1** Suppose that \( \xi(r), \zeta(r) \in C[0,1], P_1(t) \in C[0,T], i = 1,2, q(t) \in C^2[0,T], q(t) \neq 0, s(r,t) \in C(\partial \mathcal{D}_T), \int_0^1 s(r,t) \, dr = 0, \quad t \in [0, T], \) and the consistency conditions, given below, are satisfied:

\[
\begin{align*}
\int_0^1 \xi(r) \, dr &= 0, \quad \int_0^1 \zeta(r) \, dr = 0, \\
q(0) &= \int_0^T q(t)P_1(t) \, dt + \xi(R_0), \quad q'(0) = \int_0^T q(t)P_2(t) \, dt + \zeta(R_0).
\end{align*}
\]

Then the following arguments are true:

1. Each classical solution \( \{\alpha(t), z(r,t)\} \) of (1)–(5) is a solution of (1)–(3), (6), and (7);
2. Each solution \( \{\alpha(t), z(r,t)\} \) of (1)–(3), (6), and (7) by virtue of

\[
\left(T \left\| P_2(t) \right\|_{C[0,T]} + T \left\| P_1(t) \right\|_{C[0,T]} + \frac{T}{2} \left\| \alpha(t) \right\|_{C[0,T]} \right) < 1
\]

is a classical solution of (1)–(5).

We impose the below conditions to the functions \( \xi, \zeta, s, P_1, P_2, \) and \( q \) [1]:

(A1) \( \xi'''(r) \in L_2(0, 1), \xi(r) \in C^2[0,1], \xi(0) = \xi(1), \xi'(0) = \xi'(1), \xi''(0) = \xi''(1) \);

(A2) \( \zeta''(r) \in L_2(0, 1), \zeta(r) \in C^1[0,1], \zeta(0) = \zeta(1), \zeta'(0) = \zeta'(1) \);

(A3) \( s(r,t), s_r(r,t) \in C(\partial \mathcal{D}_T), s_r(0,t) \in L_2(\partial \mathcal{D}_T), s(0,t) = s(1,t), s_r(0,t) = s_r(1,t), t \in [0, T] \);

(A4) \( P_1(t), P_2(t) \in C[0,T], q(t) \in C^2[0,T], q(t) \neq 0, t \in [0, T] \).

**Theorem 2** Let conditions (A1)–(A4) be satisfied, and suppose that

\[
\left\{ (A_1(T) + A_2(T) + 2)(B_1(T) + B_2(T)) + C_1(T) + C_2(T) \right\}(A_1(T) + A_2(T) + 2) < 1,
\]

where

\[
\begin{align*}
A_1(T) &= T \left\| \xi(r) \right\|_{L_2(0,1)} + T \sqrt{T} \left\| s(r,t) \right\|_{L_2(\partial \mathcal{D}_T)} + \left\| \xi(r) \right\|_{L_2(0,1)} + 2 \sqrt{6} \left\| \xi'''(r) \right\|_{L_2(0,1)} \\
&\quad + 2 \sqrt{6} T \left\| s_r(r,t) \right\|_{L_2(\partial \mathcal{D}_T)} + 2 \sqrt{6} \left\| \zeta''(r) \right\|_{L_2(0,1)},
\end{align*}
\]
Let the conditions
\[
\int_0^1 \xi(r) \, dr = 0, \quad \int_0^1 \zeta(r) \, dr = 0, \quad \int_0^1 s(r, t) \, dr = 0, \quad t \in [0, T],
\]
\[
q(0) = \int_0^T P_1(t)q(t) \, dt + \xi(R_0), \quad q'(0) = \int_0^T P_2(t)q(t) \, dt + \zeta(R_0),
\]
\[
T\|P_2(t)\|_{C[0,T]} + T\|P_1(t)\|_{C[0,T]} + \frac{T}{2} (A_1(T) + A_2(T) + 2) < 1,
\]
and all the assumptions of Theorem 2 be satisfied. Then (1)–(5) has a unique solution in \(K = K_R\).
Equation (11) yields

\[-A_{i,j+1} + (1 + 2A - B_{j+1})z_{i,j+1} - A_{i+1,j+1} = A_{i,j} + (2 + 2A + B_j)z_{i,j} + A_{i+1,j} - z_{i,j-1}
\]

\[+ \frac{(\Delta t)^2}{2}(s_{i,j+1} + s_{i,j}), \quad i = 1(1)M, j = 1(1)N, \tag{12}\]

where

\[A = \frac{(\Delta t)^2}{2(\Delta r)^2}; \quad B_j = \frac{(\Delta t)^2}{2}a_j.\]

The discretization of nonlocal ICs (2) is

\[z_{i,0} = \frac{\Delta t}{3} \left( P_{i,0}^1 z_{i,0} + P_{i,N}^1 z_{i,N} + \sum_{l=1,3}^{N-1} P_{i,l}^1 z_{i,l} + \sum_{l=2,4}^{N-2} P_{i,l}^1 z_{i,l} \right) + \xi (r_i), \quad i = 0(1)M, \tag{13}\]

\[\frac{z_{i,1} - z_{i,-1}}{2(\Delta t)} = \frac{\Delta t}{3} \left( P_{i,0}^2 z_{i,0} + P_{i,N}^2 z_{i,N} + \sum_{l=1,3}^{N-1} P_{i,l}^2 z_{i,l} + \sum_{l=2,4}^{N-2} P_{i,l}^2 z_{i,l} \right) + \zeta (r_i), \quad i = 0(1)M, \tag{14}\]

and the periodic BC (3) is

\[z(0, t) = z_{0j} = z(1, t) = z_{M,j}. \tag{15}\]

Finally, discretization of integral BC (4) is given as follows:

\[\int_0^1 z(r, t) \, dr = \frac{\Delta r}{3} \left( z_{0j} + z_{M,j} + \sum_{l=1,3}^{M-1} z_{l,j} + \sum_{l=2,4}^{M-2} z_{l,j} \right) = 0, \quad j = 0(1)N. \tag{16}\]

For \(i = 0\) and \(M\), from equations (15) and (16), we get

\[2z_{0j} + 4 \sum_{l=1,3}^{M-1} z_{l,j} + 2 \sum_{l=2,4}^{M-2} z_{l,j} = 0, \quad i = 0, j = 0(1)N, \tag{17}\]

and

\[4 \sum_{l=1,3}^{M-1} z_{l,j} + 2 \sum_{l=2,4}^{M-2} z_{l,j} + 2z_{M,j} = 0, \quad i = M, j = 0(1)N. \tag{18}\]
The above equations (12), (17), and (18) can be reformulated and converted at time $t_{j+1}$ into the $(M+1) \times (M+1)$ system:

$$
\begin{bmatrix}
2 & 4 & 2 & 4 & 2 & \cdots & 4 & 0 \\
-\hat{A} & \hat{B} & -A & 0 & 0 & \cdots & 0 & 0 \\
0 & -\hat{A} & \hat{B} & -A & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\
0 & 0 & \cdots & -A & \hat{B} & -A & 0 & 0 \\
0 & 0 & \cdots & 0 & 0 & -A & \hat{B} & -A \\
0 & 0 & \cdots & 0 & 0 & \cdots & \cdots & \cdots \\
0 & 4 & \cdots & 4 & 2 & \cdots & 4 & 2 \\
\end{bmatrix}
\begin{bmatrix}
z_{0,j+1} \\
z_{1,j+1} \\
z_{2,j+1} \\
\vdots \\
z_{M-2,j+1} \\
z_{M-1,j+1} \\
z_{M,j+1} \\
\end{bmatrix}
= 
\begin{bmatrix}
R_{0,j} \\
R_{1,j} \\
R_{2,j} \\
\vdots \\
R_{M-2,j} \\
R_{M-1,j} \\
R_{M,j} \\
\end{bmatrix},
$$

(19)

where

$$
\hat{B} = 1 + 2A - B_{j+1}, \quad R_{0,j} = R_{M,j} = 0,
$$

$$
R_{i,j} = A z_{i-1,j} + (2 + 2A + B_j) z_{i,j} + A z_{i+1,j} - z_{i,j-1} + \frac{(\Delta t)^2}{2} (s_{i,j+1} + s_{i,j}), \quad i = 1(1) M - 1, j = 1(1) N.
$$

At time $t_1$, using (14) in (12), we get

$$
-A z_{j-1,1} + (2 + 2A - B_1) z_{j,1} - A z_{j+1,1} = A z_{j-1,0} + (2 + 2A + B_0) z_{j,0} + A z_{j+1,0} + 2(\Delta t)\zeta (r_i)
$$

$$
+ \frac{k^2}{3} \left( P^2_{1,0} N + P^2_{1,N} N + \sum_{l=1,3}^{N-1} P^2_{2,l} z_{l,1} + \sum_{l=2,4}^{N-2} P^2_{2,l} z_{l,1} \right)
$$

$$
+ \frac{(\Delta t)^2}{2} (s_{i,1} + s_{i,0}), \quad i = 1(1) M - 1.
$$

(20)

Now, at time $t_1$, equations (17), (18), and (20) can be reformulated into the $(M+1) \times (M+1)$ system:

$$
\begin{bmatrix}
2 & 4 & 2 & 4 & 2 & \cdots & 4 & 0 \\
-\hat{B} & -A & 0 & 0 & \cdots & 0 & 0 \\
0 & -\hat{B} & -A & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\
0 & 0 & \cdots & -A & \hat{B} & -A & 0 & 0 \\
0 & 0 & \cdots & 0 & 0 & -A & \hat{B} & -A \\
0 & 0 & \cdots & 0 & 0 & \cdots & \cdots & \cdots \\
0 & 4 & \cdots & 4 & 2 & \cdots & 4 & 2 \\
\end{bmatrix}
\begin{bmatrix}
z_{0,1} \\
z_{1,1} \\
z_{2,1} \\
\vdots \\
z_{M-2,1} \\
z_{M-1,1} \\
z_{M,1} \\
\end{bmatrix}
= 
\begin{bmatrix}
R_{0,0} \\
R_{1,0} \\
R_{2,0} \\
\vdots \\
R_{M-2,0} \\
R_{M-1,0} \\
R_{M,0} \\
\end{bmatrix},
$$

(21)

where

$$
\hat{B} = 2 + 2A - B_1, \quad R_{0,0} = R_{M,0} = 0,
$$

$$
R_{i,0} = A z_{j-1,0} + (2 + 2A + B_0) z_{j,0} + A z_{j+1,0} + 2(\Delta t)\zeta (r_i)
$$
\[ + \frac{(\Delta t)^2}{3} \left( p_{2z_1}^0 + p_{2z_N}^N + \sum_{l=1,3}^{N-1} p_{2z_{l,l}}^l + \sum_{l=2,4}^{N-2} p_{2z_{l,l}}^l \right) \]
\[ + \frac{(\Delta t)^2}{2} (s_{i,1} + s_{i,0}), \quad i = 1(1)M - 1. \]

4 Stability analysis

The von Neumann stability [8, 9, 22, 23, 28, 30] is carried out for the discretized system of hyperbolic wave equation.

**Theorem 5** Show that the proposed numerical scheme is stable.

**Proof** Taking \( s(r, t) = 0 \) and assuming local constant \( \alpha_j = \alpha_{j+1} = \bar{b}_1 \) for known level in (12), we get

\[ -A z_{i-1,j+1} + (1 + 2A - B)z_{i,j+1} - A z_{i+1,j+1} = A z_{i-1,j} + (2 + 2A + B)z_{i,j} + A z_{i+1,j} - z_{i,j-1}, \]

(22)

where

\[ B = \frac{(\Delta t)^2}{2} \bar{b}_1. \]

Now we consider one Fourier mode out of the full solution \( z_{i,j} = 2^i e^{i\omega} \) as trial solutions at a given point \( r_i \), where \( \phi = \theta h \) is the phase angle, \( i \) is the node number, and \( \omega = \sqrt{-1} \). Using trial solutions in the above equation and simplifying the terms, we get

\[ (-2A \cos \phi + (1 + 2A - B))\hat{z}^2 - (2A \cos \phi + (2 + 2A + B))\hat{z} + 1 = 0, \]

(23)

which can be written as

\[ \beta_1 \hat{z}^2 - \beta_2 \hat{z} + \beta_3 = 0, \]

(24)

where

\[ \beta_1 = -2A \cos \phi + (1 + 2A - B), \quad \beta_2 = 2A \cos \phi + (2 + 2A + B), \quad \beta_3 = 1. \]

Under the transformation \( \hat{z} = \frac{1 + \rho}{1 - \rho} \) in equation (24), we get

\[ (\beta_1 + \beta_2 + \beta_3) \rho^2 + 2(\beta_1 - \beta_3) \rho + (\beta_1 - \beta_2 + \beta_3) = 0. \]

(25)

The discretized system (22) will be stable if

\[ \beta_1 + \beta_2 + \beta_3 \geq 0, \quad \beta_1 - \beta_3 \geq 0, \quad \beta_1 - \beta_2 + \beta_3 \geq 0. \]

(26)

Using the values of \( \beta_1, \beta_2, \beta_3 \) and simplifying the terms, we get

\[ \beta_1 + \beta_2 + \beta_3 = \frac{2(\Delta t)^2}{(\Delta t)^2} + 3, \]

(27)
\[ \beta_1 - \beta_3 = \frac{(\Delta t)^2}{(\Delta r)^2} \left( 2 \sin^2 \left( \frac{\phi}{2} \right) - (\Delta r)^2 \bar{b}_1 \right) , \]  
\[ \beta_1 - \beta_2 + \beta_3 = 2 + \frac{2(\Delta t)^2}{(\Delta r)^2} \left( 2 \sin^2 \left( \frac{\phi}{2} \right) - (\Delta r)^2 \bar{b}_1 - 1 \right) . \]  

(28)  

(29)

It is clear from (27) that \( \beta_1 + \beta_2 + \beta_3 \geq 0 \). From (28) and (29), we get \( \beta_1 - \beta_3 \geq 0 \) and \( \beta_1 - \beta_2 + \beta_3 \geq 0 \) if \( (\Delta r)^2 \leq \frac{1}{\bar{b}_1} \left( 2 \sin^2 \frac{\phi}{2} - 1 \right) \), and the discretized system of (1) will be stable. \( \square \)

5 Numerical solution of the inverse problem

In this section, we want to find accurate and stable identification of \( z(r, t) \) and \( \alpha(t) \) which satisfies the nonlinear and ill-posed inverse problem (1)–(5). This is achieved by minimizing the objective function

\[ F(\alpha) = \| z(R_0, t) - q(t) \|_{L^2[0, T]}^2 , \]  

(30)

where \( z \) solves the forward problem (1)–(4) for given \( \alpha(t) \). The discrete form of equation (30) is

\[ F(\alpha) = \sum_{j=1}^{N} [w(R_0, t_j) - q(t_j)]^2 . \]  

(31)

The objective function \( F \) (31) is minimized by the MATLAB subroutine \texttt{lsqnonlin} [21].

6 Numerical results and discussion

The accuracy is measured by RMS:

\[ \text{RMS}(\alpha) = \left[ \frac{T}{N} \sum_{j=1}^{N} \left( \alpha_{\text{numerical}}(t_j) - \alpha_{\text{exact}}(t_j) \right)^2 \right]^{1/2} , \]  

(32)

\[ \text{RMS}(q) = \left[ \frac{T}{N} \sum_{j=1}^{N} \left( q_{\text{numerical}}(t_j) - q_{\text{exact}}(t_j) \right)^2 \right]^{1/2} . \]  

(33)

Here, for simplicity, we take \( T = 1 \), and \( 10^2 \) and \( -10^2 \) as upper and lower bounds, respectively, for coefficient \( \alpha(t) \).

The inverse problem (1)–(5) is solved with exact and noisy measurement data (5). We numerically simulated the perturbed data as follows:

\[ q'(t_j) = q(t_j) + \epsilon_j, \quad j = 0(1)N, \]  

(34)

where \( \epsilon_j \) are random variables with mean zero and standard deviation \( \sigma \) given by

\[ \sigma = p \times \max_{t \in [0, T]} |q(t)|, \]  

(35)

where \( p \) denotes the percentage of noise. In the case of noisy data (34), we replace \( q(t_j) \) with \( q'(t_j) \) in (31).
6.1 Test 1

The proposed problem (1)–(5) is considered with some smooth potential term

$$\alpha(t) = -1 - \pi t, \quad t \in [0,1],$$  \hfill (36)

the exact solution

$$z(r, t) = \exp(-1 - t) \cos(2\pi r), \quad (r, t) \in \overline{D_T},$$  \hfill (37)

and the rest of the data are as follows:

$$\int_0^1 z(r, t) \, dr = 0, \quad P_1(t) = 0, \quad P_2(t) = 0, \quad R_0 = 0.5,$$

$$\xi(r) = \exp(-1) \cos(2\pi r),$$

$$\zeta(r) = -\exp(-1) \cos(2\pi r),$$

$$s(r, t) = \exp(-1 - t)(2 + 4\pi^2 + \pi t) \cos(2\pi r).$$  \hfill (38)

The periodic BCs

$$z(0, t) = \exp(-1 - t) = z(1, t), \quad t \in [0,1],$$  \hfill (39)

and the over-determination condition

$$q(t) = z(R_0, t) = -\exp(-1 - t), \quad t \in [0,1].$$  \hfill (40)

It is observed that Theorem 3 guarantees the uniqueness of the solution because their conditions have been fulfilled. The accuracy of problem (1)–(4) has been assessed with data (37) and (38) when $\alpha(t)$ is given by (36). Fig. 1 depicts the approximate displacement measurement in (5) in comparison to the analytical solution (40) obtained by using the Crank–Nicolson FDM with $M = N \in \{10, 20, 40, 80\}$. The exact (37) and approximate solutions for the displacement $z(r, t)$ are illustrated in Fig. 2. The absolute errors between
those solutions are also included and it can be seen that these errors decrease as the CN-FDM grid becomes finer. A good agreement observed between the analytical (40) and the approximate $q(t)$ solutions as the mesh size decreases, see Table 1.

Next, we fix $\Delta r = \Delta t = 0.025$, and the time-dependent potential coefficient $\alpha(t)$ determination is started, when the value of $p$ equal zero in equation (5), as put in (35). The
Table 1 The RMS error norm for direct problem

| $M=N$ | 10  | 20  | 40  | 80  |
|-------|-----|-----|-----|-----|
| RMS$(q)$ | 8.5E–3 | 2.2E–3 | 5.3E–4 | 1.1E–4 |

Figure 3 (a) $F(31)$, (b) the exact (36) and approximate $\alpha(t)$ with $p=0$ (Test 1)

initial guess (IG) for $\alpha$ is taken as follows:

$$\alpha^0(t_j) = \alpha(0) = -1, \quad j = 1(1)N.$$  \hspace{1cm} (41)

Fig. 3(a) is depicted by the objective function $F(31)$, and a monotonically decreasing convergence is observed by making 15 time repetitions for attaining a less order tolerance $O(10^{-30})$. Numerical outcomes for the potential $\alpha(t)$ are observed in Fig. 3(b) and also obtaining with $\text{RMS}(\alpha) = 0.0868$. The perturbed data has been used to examine the stability of the approximate solution. The noise $p \in \{\frac{1}{100} \%, \frac{1}{10} \%\}$ is included to simulate the input noisy data via equation (34) for $q(t)$. The potential term $\alpha(t)$ is depicted in Figs. 4 and 5. From Figs. 4(a) and 5(a) it can be observed that as $p$ is increased the approximate results for $\alpha(t)$ start to build up oscillations. To retrieve stability, we penalize $F(30)$ by adding $\beta \|\alpha(t)\|_{L^2[0,T]}^2$ to it, where $\beta > 0$ is Tikhonov’s regularization parameter. Then a discretized form of Tikhonov functional is

$$F_\beta(\alpha) = F(\alpha) + \beta \sum_{j=1}^{N} \alpha_j^2.$$ \hspace{1cm} (42)

For $p \in \{0.01\%, 0.1\%\}$ noise, Figs. 4(b) and 5(b) show the analytical (36) and the approximate $\alpha(t)$ achieved by minimizing $F_\beta$ (42) for various $\beta$. The $\text{RMS}(\alpha)$ values are $\{3.1204, 0.0790, 0.0776\}$ for the value of $p = \frac{1}{100} \%$, and $\{34.2297, 0.1573, 0.1100\}$ for the value of $p = \frac{1}{10} \%$, respectively, with $\beta \in \{0, 10^{-10}, 10^{-9}\}$ and $\beta \in \{0, 10^{-9}, 10^{-8}\}$. It is noticed that the approximate $\alpha(t)$ obtained with $\beta = 0$ demonstrates instability; however, on inclusion of regularization with $\beta = 10^{-10}$ to $10^{-8}$, a stable solution is obtained which is consistent in accuracy with $p = 0.01\%$ and $0.1\%$ violating the input data (34). The absolute error norms between the exact (37) and approximate solutions with $\beta = 0, 10^{-9}$
and $10^{-8}$ are illustrated in Fig. 6. It is observed that the displacement $z(r,t)$ component is accurate and stable when penalty term $\beta > 0$ is added as in (42) to stabilize the solution.

### 6.2 Test 2

The smooth time-wise potential $\alpha(t)$ given by (36) has been recovered in Test 1. Now, consider a nonlinear numerical test problem which is given by

\[
\alpha(t) = -\exp(t) - \pi \cos^2(3\pi t), \quad t \in [0,1],
\]

\[
s(r,t) = \exp(-1-t)(1 + \exp(t) + 4\pi^2 + \pi \cos^2(3\pi t)) \cos(2\pi r).
\]
Figure 6 The absolute errors between the analytical (37) and approximate $z(r,t)$ with $\beta$: (a) 0, (b) $10^{-9}$, and (c) $10^{-8}$ for $p = 0.1\%$ (Test 1)

Since the conditions of Theorem 3 hold, the solution uniqueness is guaranteed. The IG for $q(t)$ is considered as follows:

$$\alpha^j(t_j) = \alpha(0) = -1 - \pi, \quad j = 1(1)N.$$  \hspace{1cm} (45)

We take fixed values $M = N = 40$ as in the previous test problem, and value of $p = 0$ in $q(t)$, as also in equation (35). Fig. 7(a) is depicted by the objective function $F$ (31), and a monotonically decreasing convergence is observed by making 15 time repetitions for...
attaining a less order tolerance $O(10^{-30})$. Numerical outcomes for the potential $\alpha(t)$ are observed in Fig. 7(b) and also obtaining with RMS($\alpha$) = 0.0872.

The $F_{\beta}$ (31) is depicted in Fig. 7(a), where a monotonically decreasing convergence is achieved in 15 iterations for achieving $O(10^{-30})$. Fig. 7(b) depicts the exact (43) and approximate $\alpha(t)$, obtaining with RMS($\alpha$) = 0.0872.

Next, we take $p \in \{1\% , 10\% \}$ to the measured data $q(t)$, as also in equation (35), for checking the stability of problem. The regularized function $F_{\beta}$ (42) is depicted in Fig. 8 with various parameters. The approximate outcomes of potential $\alpha(t)$ are illustrated in Figs. 9 and 10. As in the previous test problem, it is observed that the numerical solutions for $\beta = 0$ tabulated in Figs. 9(a) and 10(a) are unstable due to high oscillations and unbounded, obtaining with RMS($\alpha$) = 3.3676 for the value of $p = \frac{1}{100}$% and RMS($\alpha$) = 33.9465 for the value of $p = \frac{1}{10}$%. However, the addition of some restrictions on $\beta > 0$ in $F_{\beta}$ (42) improves the stability of the numerical solutions, as drawn further in Figs. 9(b) and 10(b), respectively. It is observed that the addition of $\beta \in \{10^{-12}, 10^{-11}, 10^{-10}\}$ for $\frac{1}{100}$% and
\[ \beta \in \{10^{-11}, 10^{-10}, 10^{-9}\} \text{ for } \frac{1}{10} \% \text{ provides a perfect and stable approximate solution for } \alpha(t), \]

getting with $\text{RMS}(\alpha) \in \{0.2368, 0.1245, 0.2728\}$ and $\text{RMS}(\alpha) \in \{0.8454, 0.5282, 0.6936\}$.

The comparison between analytical (37) and approximate $z(r, t)$ in terms of absolute errors with and without descriptor parameters are depicted in Fig. 11, where $\beta > 0$ has the effect of reducing the unstable performance of the restored displacement. Refer to Table 2 for additional information on the RMS($\alpha$) values, the minimum value of $F$ or $F_\beta$ at final
Table 2  The RMS values (32) for $p \in \{0, 0.01\%, 0.1\%\}$ and $\beta \in \{10^{-12}, 10^{-11}, 10^{-10}, 10^{-9}, 10^{-8}, 10^{-7}\}$ for Tests 1 and 2

| $p$   | $\beta$ | RMS($\alpha$) | Minimum value of $F$ or $F_\beta$ | $p$   | $\beta$ | RMS($\alpha$) | Minimum value of $F$ or $F_\beta$ |
|-------|---------|----------------|-----------------------------------|-------|---------|----------------|-----------------------------------|
| 0     | 0       | 0.0868         | $F = 2.4E-30$                     | 0     | 0       | 0.0872         | $F = 8.1E-30$                     |
| 0.01% | 0       | 3.4204         | $F = 1.3E-29$                     | 0.01% | 0       | 3.3676         | $F = 8.5E-30$                     |
| $10^{-10}$ | 0.0790 | $F_\beta = 1.1E-7$ | $10^{-12}$ | 0.2368 | $F_\beta = 7.1E-8$ | $10^{-11}$ | 0.1245 | $F_\beta = 2.3E-7$ |
| $10^{-9}$ | 0.0776 | $F_\beta = 4.2E-7$ | $10^{-10}$ | 0.2728 | $F_\beta = 1.5E-6$ | $10^{-8}$ | 0.0982 | $F_\beta = 3.2E-6$ |
| $10^{-7}$ | 0.0982 | $F_\beta = 3.2E-6$ | $10^{-9}$ | 0.0982 | $F_\beta = 1.1E-5$ | $10^{-10}$ | 0.5282 | $F_\beta = 2.3E-7$ |
| 0.1%  | 0       | 34.2297        | $F = 1.3E-29$                     | 0.1%  | 0       | 33.9465        | $F = 2.1E-29$                     |
| $10^{-9}$ | 0.1573 | $F_\beta = 8.4E-6$ | $10^{-11}$ | 0.8454 | $F_\beta = 6.7E-6$ | $10^{-8}$ | 0.1100 | $F_\beta = 1.1E-5$ |
| $10^{-7}$ | 0.1701 | $F_\beta = 3.6E-5$ | $10^{-9}$ | 0.6936 | $F_\beta = 1.6E-5$ | $10^{-10}$ | 0.5282 | $F_\beta = 8.6E-6$ |

iteration without and with regularization for both numerical test problems. For the term $\alpha(t)$, the same results may be derived regarding the stable reconstruction.

7 Concluding remarks

For the first time, the reconstruction of a potential coefficient $\alpha(t)$ and the displacement distribution $z(r, t)$ from the over-determination condition in the hyperbolic equation using nonlocal ICs and periodic BCs have been computationally addressed. The CN-FDM was applied to solve the direct problem. The stability analysis of the discretized system of the wave equation has been discussed using the von Neumann method. The TR was used to overcome the instability due to the ill-posed issue. For both numerical test problems, the RMS values for noise without and with regularization were compared. It has been noticed that on the inclusion of regularization with $\beta = 10^{-12}$ to $\beta = 10^{-7}$, a stable solution is obtained which is consistent in accuracy with 0.01% and 0.1% noise. Finally, future study into the extension of the suggested approach for determining the time-wise potential coefficient in a two-dimensional wave equation is an intriguing prospect.

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