WAVE TOMOGRAPHY IN MEDICINE: THEORY AND NUMERICAL METHODS

M. A. Shishlenin, N. S. Novikov, S. I. Kabanikhin

Sobolev Institute of Mathematics, Novosibirsk, Russia

Tomography has been developing intensively over the past two decades. The development of ultrasound early diagnosis methods for soft tissue malignancies is one of the key problems in medicine, and the absence of exposure to ionized radiation has become an urgent trend in medicine, which makes ultrasound tomography very promising.

Currently, ultrasound tomographs are being developed by various research groups around the world [1-5]. This work, which has been going on for more than 10 years, has reached the pilot production stage. However, there is still no commercially available ultrasound scanner.

One of the serious problems of ultrasound tomography is the development of effective and stable methods for solving nonlinear coefficient inverse problems.

The most adequate model is a three-dimensional inverse problem in which the speed of wave propagation, the density of the medium, and other acoustic parameters must be reconstructed from data recorded by detectors located at the boundary of the region.

Consider the following two-dimensional direct problem for a hyperbolic system of equations [6]

. .

$$\frac{\partial u}{\partial t} + \frac{1}{\rho} \frac{\partial p}{\partial x} = 0, \quad \frac{\partial v}{\partial t} + \frac{1}{\rho} \frac{\partial p}{\partial y} = 0,$$
$$\frac{\partial p}{\partial t} + \rho c^2 \left(\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y}\right) = \Theta_{\Omega}(x, y) I(t)$$

with zero initial data and non-reflective boundary conditions.

Here u(x, y, t) is the velocity in the x direction, v(x, y, t) is the velocity in the y direction, p(x, y, t) is the pressure, c(x, y) is the velocity of wave propagation in the medium, $\rho(x, y)$ is pressure, $\Omega = (x, y) \in [0:L] \times [0:L]$ is domain, I(t) probing source, $\Theta_{\Omega}(x, y)$ is localization of sources.

Inverse problem is to find coefficients c(x, y) and $\rho(x, y)$ by known the additional information [7]

$$f_i(t) = p(x_i, y_i, t), \ i = 1 \dots N.$$

Here N is the number of receivers.

The solution of the inverse problem is reduced to the task of minimizing the target functional by the accelerated gradient method [4]. Formulas for the gradient of the functional are obtained using solutions to the corresponding conjugate problems. The results of numerical calculations and a comparative analysis of numerical algorithms, including neural network approaches, are presented [8].

The work was carried out with the financial support of the Russian Science Foundation project 24-41-04004 "Identification and research of mathematical models in science and industry – regularization and machine learning".

References

1. **Natterer, F.** Acoustic mammography in the time domain [Text] : technical report preprints // Institute for Numeric and Applied Mathematics, University of Munster. – 2008. – P. 7.

2. **Zhao, W.** Simulation-to-real generalization for deep-learning-based refraction-corrected ultrasound tomography image reconstruction [Text] / W. Zhao, Y. Fan, H. Wang et al. // Physics in Medicine and Biology. – 2023. – Vol. 68, No. 3. – P. 035016.

3. Natterer, F. Possibilities and limitations of time domain wave equation imaging [Text] // Contemporary Mathematics. – 2011. – Vol. 559 (Providence, RI: American Mathematical Society). – P. 151–162.

4. **Duric**, N. Breast ultrasound tomography: bridging the gap to clinical practice [Text] / N. Duric, P. Littrup, C. Li et al. // Proc. SPIE. - 2012. – P. 8320.

5. Wiskin, J. Quantitative assessment of breast density using transmission ultrasound tomography [Text] /

J. Wiskin, B. Malik, R. Natesan, M. Lenox // Medical Physics. - 2019b. - Vol. 46, No. 6. - P. 2610-2620.

6. **Kabanikhin, S. I.** Numerics of acoustical 2D tomography based on the conservation laws [Text] / S. I. Kabanikhin, D. V. Klyuchinskiy, N. S. Novikov, M. A. Shishlenin // Journal of Inverse and Ill-Posed Problems. – 2020. – Vol. 8, No. 2. – P. 287–297.

7. Klyuchinskiy, D. V. CPU-time and RAM memory optimization for solving dynamic inverse problems using gradient-based approach [Text] / D. V. Klyuchinskiy, N. S. Novikov, M. A. Shishlenin // Journal of Computational Physics. – 2021. Vol. 439, No. 110374.

8. **Prikhodko, A.** Encoder neural network in 2D acoustic tomography [Text] / A. Prikhodko, M. A. Shishlenin, N. S. Novikov, D. V. Klyuchinskiy // Applied and Computational Mathematics. – 2024. – Vol. 23, No. 1. – P. 83–98.