

Article

Not peer-reviewed version

Development of a Three-Dimensional Inversion Program for Magnetotellurics Using Finite Volume and Adjoint State Method

[Atsushi Suzuki](#) *

Posted Date: 30 August 2024

doi: 10.20944/preprints202408.2236.v1

Keywords: Magnetotellurics; inversion; Finite volume method; Adjoint state method



Preprints.org is a free multidiscipline platform providing preprint service that is dedicated to making early versions of research outputs permanently available and citable. Preprints posted at Preprints.org appear in Web of Science, Crossref, Google Scholar, Scilit, Europe PMC.

Copyright: This is an open access article distributed under the Creative Commons Attribution License which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Article

Development of a Three-Dimensional Inversion Program for Magnetotellurics Using Finite Volume and Adjoint State Method

Atsushi Suzuki

JX Metals Resources Exploration & Development Co., Ltd., Tokyo, Japan; 1-9, Kanda Nishikicho, Chiyoda-Ku, Tokyo 101-0054, Japan; suzuki@tankai.co.jp

Abstract: The finite volume method (FVM) is a kind of numerical analysis method that can take into account topography and apply local subdivision cells as the finite element method (FEM). The adjoint state method (ASM) can easily calculate the gradient vector of the objective function for the inversion. The author developed a program that applied these methods to 3-D Magnetotelluric (MT) inversion and verified it using test models. As a result, it was confirmed that the program can perform analysis properly and efficiently considering topography and subsurface resistivity heterogeneity.

Keywords: magnetotellurics; inversion; finite volume method; adjoint state method

Introduction

The magnetotelluric (MT) method is a geophysical exploration method to estimate the resistivity structure in the subsurface and is currently used in, for example, geothermal exploration and oil and gas field exploration (Takakura 2014; Key, Constable, and Weiss 2006). In the MT method, 3D analysis is now often conducted to estimate the resistivity structure in the subsurface from the acquired data, and the inversion programs were developed for this purpose (Sasaki, 2004; Siripunvaraporn et al., 2005; Kelbert et al., 2014; Usui, 2015; Usui et al., 2024). In addition, a comparison of them was also performed (Uchida and Yamaya, 2023).

In the 3D MT analysis, forward modeling is performed to analyze the propagation of electromagnetic waves into the subsurface. In general, this calculation needs to be solved analytically. Hence, some numerical analysis is necessary. For this purpose, the finite difference method (FDM), FEM, and FVM are often used (e.g., Sasaki, 2004; Siripunvaraporn et al., 2005; Nam et al., 2007; Kelbert et al., 2014; Jahandari and Farquharson, 2015; Usui, 2015; Zhu et al., 2022; Usui et al., 2024). Although FDM can be a simple discretization method, it is not easy to apply the unstructured grid. On the other hand, FEM and FVM can treat the unstructured grid and calculate electromagnetic response (Jahandari & Farquharson, 2015; Nam et al., 2007; Usui, 2015; Zhu et al., 2022; Usui et al., 2024).

Further, inversion is also performed to estimate the 3D resistivity model in the subsurface from actual data. This process needs to update the current model based on the result obtained from the forward modeling. MT inversion often uses the Jacobi matrix of the objective function to update model parameters in inversion (e.g., Sasaki, 2004; Siripunvaraporn et al., 2005; Kelbert et al., 2014; Usui, 2015; Usui et al., 2024). However, the computational cost of acquiring the Jacobi matrix is enormous (Plessix 2006). Uchida (2022) analyzed actual data using various mesh size models and found that some require a large portion of the overall computation time to calculate the Jacobi matrix.

ASM is an efficient way of finding the gradient vector rather than calculating the Jacobi matrix from a point of computational cost (Plessix 2006). The advantage of this method is that the computationally expensive part is only one solution of the adjoint equation, whose scale is the same as the forward modeling, regardless of the number of model parameters or data.

The author developed a program that applied the FVM and ASM to the 3D MT inversion to achieve efficient analysis considering the unstructured grid. This paper reviews and discusses the methodology and results of some example models.

Method

Forward Modelling

In general, the MT analysis combines Maxwell's equations in the frequency domain to form an equation of only the electric or magnetic field, which is then discretized and solved to obtain the electric or magnetic field.

Some discretization methods exist including the finite difference method (FDM) (e.g., Kelbert et al., 2014; Sasaki, 2004; Siripunvaraporn et al., 2005), FEM (e.g., Mitsuhashi & Uchida, 2004; Usui, 2015; Usui et al., 2024; Zhu et al., 2022) and FVM (e.g., Jahandari & Farquharson, 2015).

This program adopts the hexahedral cell-centered finite volume method as the forward modeling. The advantages of this method are

- 1) the unknowns are placed at the cell center and discretized on the cell surface using a simple method similar to FDM.
- 2) Unstructured meshes can be applied as FEM, allowing calculations with topography and local subdivision.

In MT forward modeling, the second-order Maxwell's equations in the form of the electric $\mathbf{E}$ or the magnetic $\mathbf{H}$ fields are often solved (Siripunvaraporn, Egbert, and Lenbury 2002). The H-forming equation (1) is solved in this program (Figure 1).

$$\nabla \times \rho \nabla \times \mathbf{H} = -i\omega\mu_0\mathbf{H}, \quad (1)$$

where ω is the angular frequency, μ_0 is the magnetic permeability of the vacuum, and ρ is specific resistance. In FVM, (1) is discretized after volume integration of both sides of Eq. (1), and the left-hand side is converted to surface integration by integral formula.

$$\sum \mathbf{n} \times \rho (\nabla \times \mathbf{H}) \Delta S = -i\omega\mu_0\mathbf{H} \Delta V, \quad (2)$$

where $\mathbf{n}$ is the normal vector of each cell surface, ΔS is the Area of each cell surface, Δ and V is the Cell volume. Note that (2) is valid with arbitrary geometry; therefore, if we can properly approximate $\nabla \times \mathbf{H}$ term on each cell surface, it is possible to use unstructured meshes and local mesh refinement as FEM in Usui et al. (2024). For the air layer, (2) should be modified because of poor convergence. They can be achieved by converting and discretizing (1), assuming ρ is constant in the air layer, and applying the condition $\nabla \cdot \mathbf{H} = 0$ as follows:

$$\sum \mathbf{n} \cdot \nabla \mathbf{H} \Delta S = -i\omega\mu_0\sigma\mathbf{H} \Delta V, \quad (3)$$

where $\sigma = 1/\rho$. (3) is a general Poisson equation, which should contribute to the computational stability in the air layer. This program solves (2) and (3) in the subsurface and the air layers, respectively. In (2), ρ on the cell surfaces are needed. This program uses the average as Taylor and Weaver (1976) and Mackie and Madden (1993). For computational stability, a transition layer is provided between the air and subsurface. In this layer, the air resistivity is $10^6 \Omega \cdot m$ on the surfaces between the air and subsurface (Figure 2).

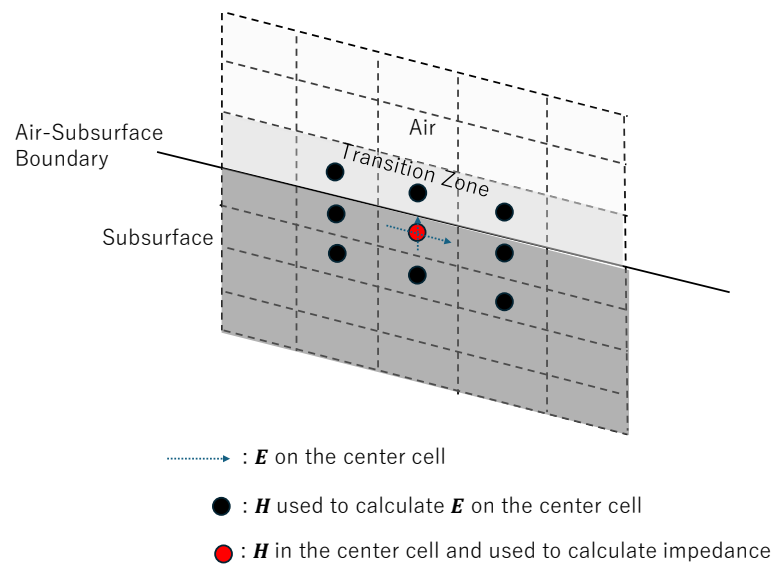


Figure 1. Schematic cartoon that indicates how to calculate $\partial \mathbf{H} / \partial \mathbf{e}_i$ on the surfaces of cells and the locations of electric field. $\mathbf{e}_1$ and $\mathbf{e}_2$ are parallel to the surface $\mathbf{S}$. Note that this calculation can be applied to not only the structured grid but also the unstructured and non-conforming grid.

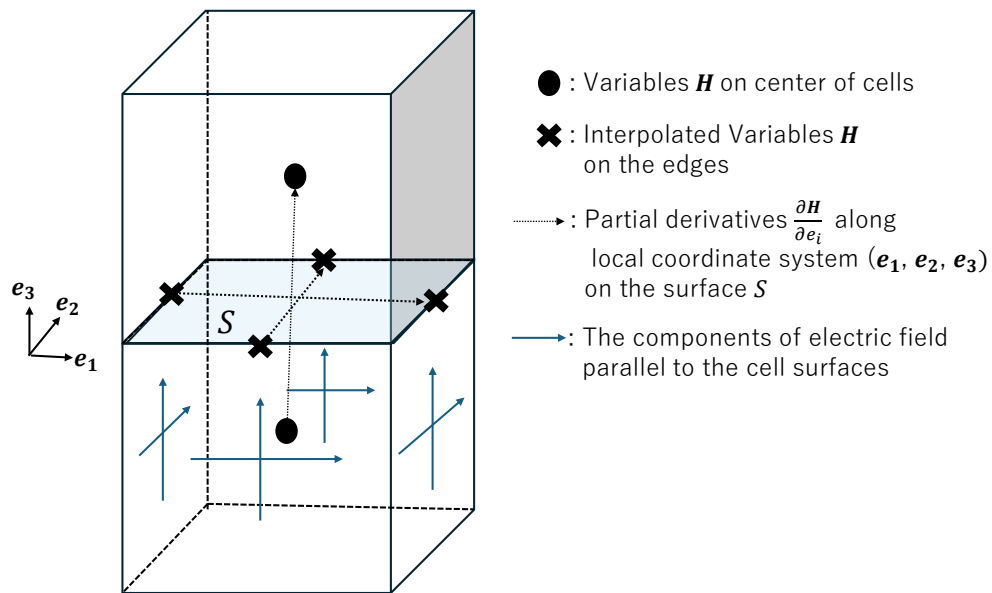


Figure 2. Schematic cartoon of a 2D case that indicates where electromagnetic fields are calculated. $\mathbf{H}$ denotes magnetic field.

To discretize $\nabla \mathbf{H}$ in the air layer and $\nabla \times \mathbf{H}$ in the subsurface layer, $\partial \mathbf{H} / \partial x_i$ on the cell surfaces, where i is 1, 2, or 3 and x_i is a global coordinate system, should be calculated. This can be done by converting the spatial partial differentiation from the local coordinate system ($\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$) to the global coordinate system, as follows:

$$\begin{pmatrix} \frac{\partial \mathbf{H}}{\partial x_1} \\ \frac{\partial \mathbf{H}}{\partial x_2} \\ \frac{\partial \mathbf{H}}{\partial x_3} \end{pmatrix} = \begin{pmatrix} \mathbf{e}_1 \cdot \mathbf{e}_{x_1} & \mathbf{e}_1 \cdot \mathbf{e}_{x_2} & \mathbf{e}_1 \cdot \mathbf{e}_{x_3} \\ \mathbf{e}_2 \cdot \mathbf{e}_{x_1} & \mathbf{e}_2 \cdot \mathbf{e}_{x_2} & \mathbf{e}_2 \cdot \mathbf{e}_{x_3} \\ \mathbf{e}_3 \cdot \mathbf{e}_{x_1} & \mathbf{e}_3 \cdot \mathbf{e}_{x_2} & \mathbf{e}_3 \cdot \mathbf{e}_{x_3} \end{pmatrix}^{-1} \begin{pmatrix} \frac{\partial \mathbf{H}}{\partial e_1} \\ \frac{\partial \mathbf{H}}{\partial e_2} \\ \frac{\partial \mathbf{H}}{\partial e_3} \end{pmatrix} \quad (4)$$

$\frac{\partial \mathbf{H}}{\partial e_i}$ ($i = 1, 2, 3$) were obtained by interpolation from the surrounding cells.

Boundary conditions on XZ and YZ planes are $\partial \mathbf{H} / \partial \mathbf{n}$, where $\mathbf{n}$ is the normal vector. On XY of the subsurface side, $\mathbf{H} = \mathbf{0}$ is adopted. On XY of the air layer side, the boundary conditions are source terms of the modeling. The author set $H_x = 1$ and $H_y = 1$ in twice forward modeling to obtain response functions, respectively (e.g., Siripunvaraporn et al., 2002).

Through these operations, (2) and (3) are discretized, and the linear equation $A\mathbf{x} = \mathbf{b}$ (A : coefficient matrix, $\mathbf{x}$: unknown magnetic field vector, $\mathbf{b}$: source term) can be assembled. We can obtain $\mathbf{H}$ by solving this equation.

To solve the linear equation, this program adopts BiCGSafe (Fujiwara, Yoshida, and Fujino 2006) with block incomplete LU decomposition as a preconditioner as Toh et al. (2000). We set the convergence condition using the maximum change of the solution from the previous iteration $\Delta \mathbf{x}_i^{max} = \max(\mathbf{x}_i - \mathbf{x}_{i-1})$, where i is iteration number and the residual $\mathbf{r}_i = A\mathbf{x}_i - \mathbf{b}$.

$$\min \left(\frac{|\mathbf{r}_i|}{|\mathbf{r}_0|}, \frac{|\Delta \mathbf{x}_i^{max}|}{|\mathbf{x}_i^{max}|} \right) < \epsilon, \quad (5)$$

where ϵ is tolerance.

The divergence correction for subsurface layers is applied as Dong and Egbert (2019), adding a term $\lambda \nabla(\nabla \cdot \mathbf{H})$, where λ is the scaling factor, to equation (1). In this program, λ is set to $\rho \times \min(1, k\omega)$, where k is the safety factor equal to 0.1 because the large value of λ resulted in poor accuracy solutions, especially in long periods.

Once magnetic field $\mathbf{H}$ is obtained, the electric field $\mathbf{E}$ is necessary at the cell center to calculate the impedance tensor. The following equation is used for this calculation.

$$\mathbf{E} = \rho(\nabla \times \mathbf{H}) \quad (6)$$

Based on equation (6), the electric fields of two directions parallel to cell surfaces are calculated on each cell surface (Figure 1).

$$\mathbf{E}^{center} \cdot \mathbf{p}_{ij} = \mathbf{E}_{ij}^{parallel}, \quad (7)$$

where $\mathbf{E}^{center}$ is the required electric field at the cell center, $\mathbf{E}_{ij}^{parallel}$ is the electric field on the cell surface calculated from equation (6), $\mathbf{p}$ is the unit vector parallel to the cell surface, i is unit vector directions (1 or 2), and j is cell surfaces number (from 1 to 4). The component of the electric field orthogonal to cell surfaces is excluded because it is not continuous on cell surfaces. From equation (7), eight equations associated with $\mathbf{E}^{center}$ in each cell can be obtained. Therefore, $\mathbf{E}^{center}$

can be calculated using the least squares method. Because $\mathbf{E}^{center}$ has three components. Using $\mathbf{H}$ and $\mathbf{E}^{center}$, The impedance tensor and magnetic transfer function can be obtained in the same way as in Usui (2015).

Since this program uses the cell-centered FVM, it is necessary to use the surrounding cell-centered magnetic field to obtain the electric field at the cell surface. This program calculates the impedance tensor and magnetic transfer function on the earth's surface using the electromagnetic fields in the transition zone and near-surface cells, as shown in Figure 2, which is a 2D case.

For the calculation of an impedance tensor with distortion, the method adopted by Ogawa (2002) is used.

$$\mathbf{Z}_{calc} = \mathbf{D}\mathbf{Z}, \quad (8)$$

where $\mathbf{Z}_{calc}$ is the calculated impedance tensor affected distortion, $\mathbf{D}$ is the distortion tensor, and $\mathbf{Z}$ is the originally calculated impedance tensor.

Note that C++ language is used to implement this program, and the library Eigen (Jacobi, Guennebaud, and other contributor 2024) is used for the internal matrix operations. Parallelization is implemented by OpenMP. The linear equations $\mathbf{A}\mathbf{x} = \mathbf{b}$ at each frequency are solved on each thread.

Inversion

In this program, ASM is implemented for MT inversion. The details are described in Plessix (2006).

In MT inversion, an objective function $F = F(\mathbf{m})$, where $\mathbf{m} \in \mathbb{R}^{N_{model}}$ is the vector of model parameter, and N_{model} is the number of model parameters, should be minimized. In ASM, the objective function F is considered as whose independent variables are $\mathbf{m}$ and $\mathbf{x} \in \mathbb{C}^{3N_{cell}}$, where $\mathbf{x}$ is the solution of linear equation $\mathbf{A}\mathbf{x} = \mathbf{b}$ in forward modeling and N_{cell} is the total number of computation cells, that is,

$$F = F(\mathbf{m}, \mathbf{x}) \quad (9)$$

Next, the adjoint equation is solved about λ .

$$\mathbf{A}^* \lambda = \left[\frac{\partial F(\mathbf{m}, \mathbf{x})^*}{\partial \mathbf{x}} \right]^T \quad (10)$$

Note that (10) is the same scale equation as $\mathbf{A}\mathbf{x} = \mathbf{b}$. Hence, the computational cost is roughly the same as solving $\mathbf{A}\mathbf{x} = \mathbf{b}$. The gradient vector of the objective function can be obtained using λ as follows:

$$\frac{\partial F(\mathbf{m})}{\partial m_i} = \frac{\partial F(\mathbf{m}, \mathbf{x})}{\partial m_i} - \text{Re} \left(\lambda^* \frac{\partial A}{\partial m_i} \mathbf{x} \right), \quad (11)$$

where i is from 1 to N_{model} . Note that $\partial F(\mathbf{m}, \mathbf{x})/\partial m_i$, and $\partial F(\mathbf{m}, \mathbf{x})^*/\partial \mathbf{x}$ are calculated by automatic differentiation using library kv (Kashiwagi 2024).

The objective function F is set as below in this program.

$$F(\mathbf{m}) = |\mathbf{W}\mathbf{d}_{obs} - \mathbf{W}\mathbf{d}_{calc}|^2 + \alpha^2 |R \log(\mathbf{m})|^2 + \beta^2 \sum_{n=1}^{N_{site}} \sum_{i=1}^2 \sum_{j=1}^2 (D_{ij}^n - I_{ij}), \quad (12)$$

where $\mathbf{d}_{obs}$ is the observed data vector, $\mathbf{d}_{calc}$ is calculated data vector, $\mathbf{W}$ is weight matrix, R is smoothness matrix for constraint term, which is necessary in ill-posed MT inversion, N_{site} is the

total number of observation sites where impedance tensor is obtained, $D^n \in \mathbb{C}^{2 \times 2}$ is distortion tensor in each observation site n as described in Ogawa (2002), I is identity matrix, α^2 and β^2 are trade-off parameters.

To minimize the objective function F with the gradient vector, Adam (Kingma and Ba, 2014) is implemented, which can be switched easily because the library OptimLib (O'Hara 2024) is used internally.

Numerical Examples

Forward Modelling Considering Topography.

To evaluate if the developed program can consider topography, the trapezoidal hill model, which is often used to validate the effect of topography (e.g., Nam et al., 2007; Usui, 2015; Zhu et al., 2022, Figure 3) was adopted.

The minimum size of cells was 25 m $\times$ 25 m $\times$ 10 m and the total number of cells was 519,456. The resistivity in the air and the subsurface was 10^6 and 100 Ω m respectively. The calculated frequency was 2 Hz. These conditions are the same as previous works (Nam et al., 2007; Usui, 2015; Zhu et al., 2022). The computer we used was Precision 3630 Tower (CPU: Intel® Core™ i7-9700 CPU 3.00GHz, RAM:16GB).

The results are shown in Figure 4. The results are compared to Zhu et al. (2022). Except for some points, these results are mostly consistent with each other. The gaps between them would be caused by the difference in the discretization method, as which Zhu et al. (2022) adopted FEM, and the size of cells. However, since the differences are quite small, it would not influence when the inversion is performed.

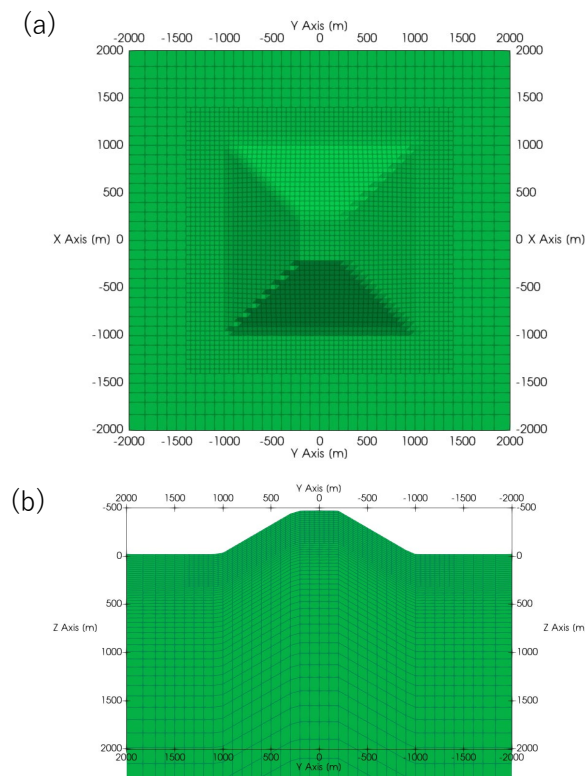


Figure 3. (a) XY and (b) YZ plane with used mesh of trapezoidal model in Nam et al. (2007).

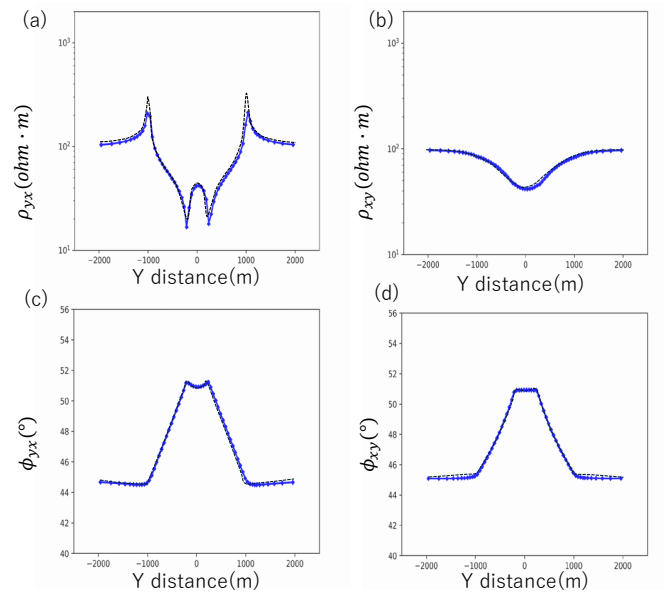


Figure 4. (a) Apparent resistivity of YX mode and (b) XY mode. (c) The phase of YX mode and (d) ZY mode of 2 Hz. Blue lines and diamonds denote our results. Black dashed lines represent the result in Zhu et al. (2022).

Forward Modelling Considering Resistivity Heterogeneity

As a test case for heterogeneity of subsurface resistivity, the model in Mackie et al. (1993) was adopted (Figure 5). The total number of cells was 370,712, with minimum size of cells 500 m × 500 m × 200 m. The resistivity in the air and the subsurface was $10^6 \Omega \cdot m$. The compared frequency was 10 and 1000 s. The computer used is the same one used in the Trapezoidal hill model.

The results of Apparent resistivity and phase in cross sections along the Y direction at X=0 m for periods of 10 and 1000 s are compared with Mackie et al. (1993) in Figure 6 and Figure 7. While there are some small differences, the results are in good agreement with those of Mackie et al. (1993).

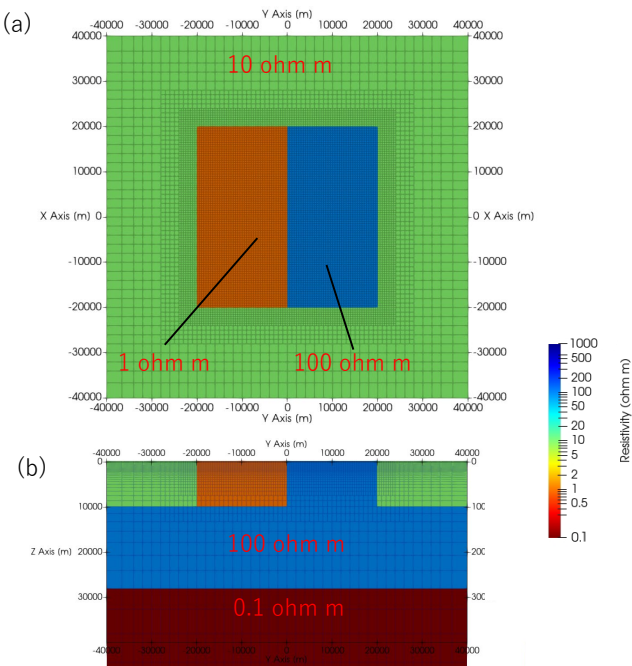


Figure 5. (a) XY and (b) YZ plane of the resistivity structure model in our forward calculation after Mackie et al. (1993). Transparent black lines denote cell edges.

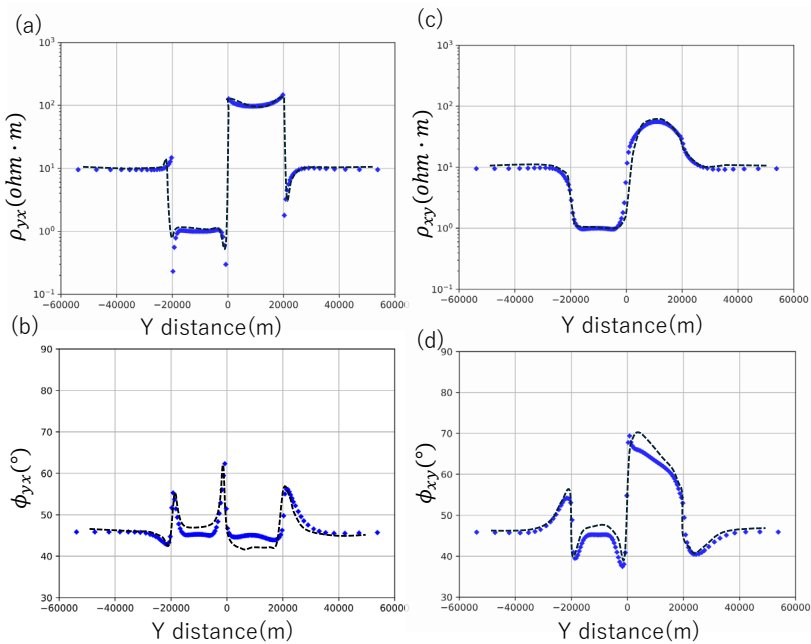


Figure 6. (a) Apparent resistivity of YX mode and (b) XY mode. (c) The phase of YX mode and (d) ZY mode of 10 s. Blue diamonds denote our results. Black dashed lines represent the result in Mackie et al. (1993).

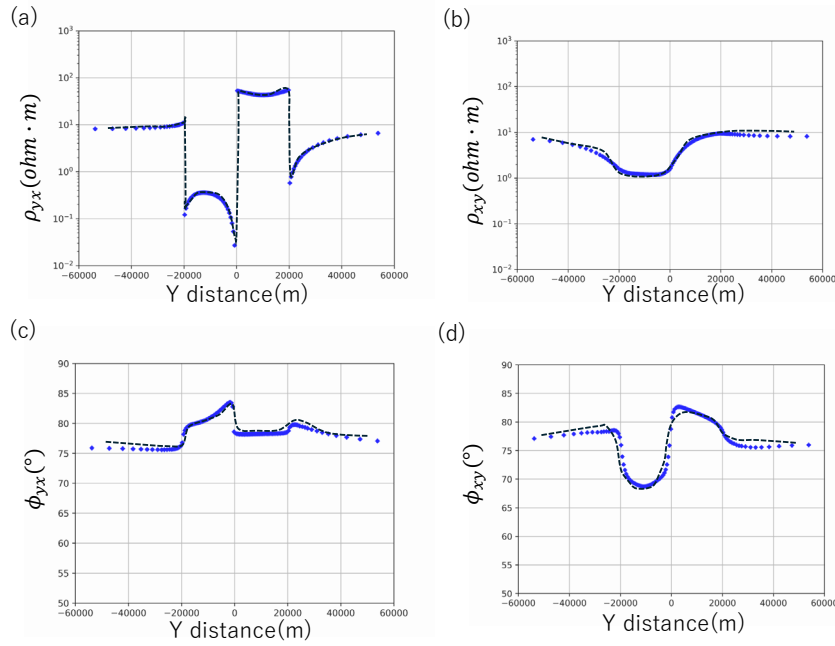


Figure 7. (a) Apparent resistivity of YX mode and (b) XY mode. (c) The phase of YX mode and (d) ZY mode of 1000 s. Blue diamonds denote our results. Black dashed lines represent the result in Mackie et al. (1993).

Inversion for Dublin Secret Model 2

The author performed inversion using Dublin Secret Model 2 (DSM2) in Miensoyust et al. (2013). This model was used as a comparison between different programs. The synthetic observed data of impedance tensor are openly available in Miensoyust et al. (2013). The author used them as observed data.

The total number of cells was 223,784. The number of cells that were inverted as model parameters is 187,172. The number of degrees of freedom was 671,352. The minimum size of cells was 2000 m × 2000 m × 100 m. The initial resistivity was 100 Ω m. The total number of calculated periods was 10, from 0.025 to 1000 s. The author initially set α^2 to 100 and lowered it after the convergence criteria were achieved. The criteria were that the maximum change of model parameters or the change of objective function from the previous iteration was below 0.3 %. For simplicity, β^2 was set to the same value as α^2 . The computer for this calculation was Dell XPS 8950 (CPU: Intel® Core™ i7-12700K 3.60 GHz, RAM:64GB).

The results of the inverted resistivity structure compared to the true model are shown in Figure 8. The author adopted the model when α^2 was 0.1, judging from the L-curved figure between RMS and model roughness (Figure 9) as in Usui et al. (2024). The resistive and conductive anomalies shallower than 10 km are inverted. While the conductive anomaly is connected to the one deeper than 30 km, this would be caused by low sensitivity under the conductive anomaly, as mentioned in (Usui 2015). In the conductive layer deeper than 30 km, The value of our result is a little high. This may be improved when data in longer periods are used.

Considering that the synthetic data is distorted according to Groom and Bailey (1989) and has 5% random noise of maximum impedance tensor value (Miensoyust et al. 2013), the results of our program seem in agreement with the true model. A comparison of the results of Miensoyust et al. (2013) would also support the agreement.

The total calculation time was about 20 hours. Usui (2015) calculated the same model in about 15 hours, with a number of degrees of freedom of 687,599, roughly the same as our calculation. Our calculation was performed without MPI parallelization, while one in Usui (2015) did with it. Nevertheless, we could obtain the results at a comparable time as Usui (2015), although the conditions

of computation, such as the computers and the total number of calculated frequencies, were different. Further, our calculation was conducted on an ordinary personal computer. They would suggest the efficient performance of our program.

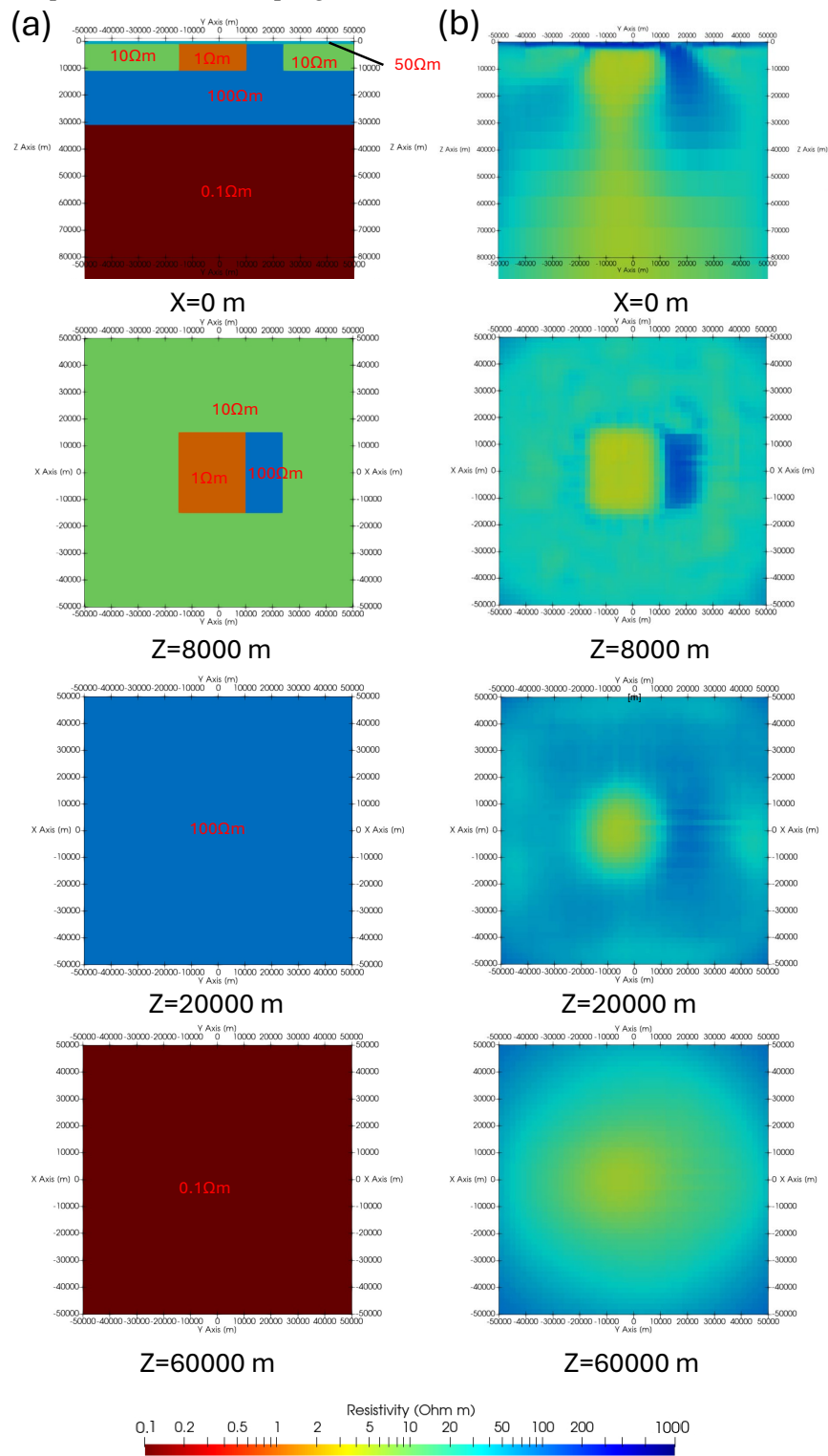


Figure 8. Each cross-section of the true DSM2 model in Miensoyust et al. (2013) and our results. (a) True Model and (b) Our inversion results.

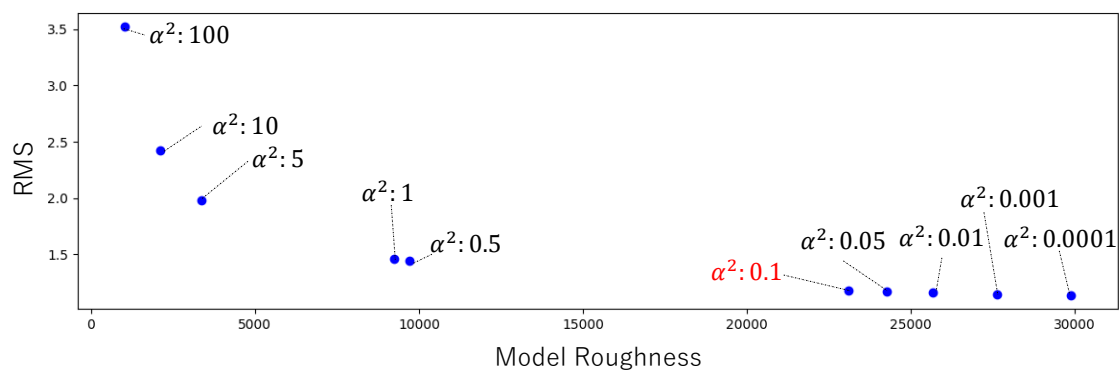


Figure 9. The L-curve picture of our inversion results. The vertical axis and horizontal axis are RMS and model roughness in each α^2 , respectively. We chose $\alpha^2 = 0.1$ (red one) as our final result.

Conclusion

The author developed a program for 3D MT inversion. For forward modeling and inversion, FVM, in which topography can be considered smoothly, and ASM, in which the gradient vector of the objective function can be efficiently obtained, were adopted, respectively.

As test models for forward modeling, the trapezoidal hill model in Nam et al. (2007) and a model that contains resistivity heterogeneity in Mackie et al. (1993) were calculated. The differences between our results and the previous works were slight. Therefore, the program can calculate the wave propagation correctly.

For the test model of inversion, the author adopted DSM2. The results were in good agreement with the true model and the ones in previous works. The calculation time was also comparable to the previous work with an ordinary personal computer. As a future work, in addition to OpenMP parallelization, MPI parallelization would be necessary for our program to perform more efficient calculations.

Acknowledgement: The authors report there are no competing interests to declare. The data supporting the findings of this study is available from the corresponding author, upon reasonable request.

References

1. Dong, Hao, and Gary D. Egbert. 2019. "Divergence-Free Solutions to Electromagnetic Forward and Adjoint Problems: A Regularization Approach." *Geophysical Journal International* 216 (2): 906–18. <https://doi.org/10.1093/gji/ggy462>.
2. Fujiwara, Maki, Masahiro Yoshida, and Seiji Fujino. 2006. "BiCGSafe Method with Crout Version of ILU Decomposition Giving Triple Safe Keys for Convergence." *IPSI Transactions on Advanced Computing Systems* 47 (7): 52–60.
3. Groom, Ross W, and Richard C Bailey. 1989. "Decomposition of Magnetotelluric Impedance Tensors in the Presence of Local Three-Dimensional Galvanic Distortion." *JOURNAL OF GEOPHYSICAL RESEARCH* 94 (B2): 1913–25.
4. Jacobi, Benoît, Gaël Guennebaud, and other contributor. 2024. "Eigen.", <https://eigen.tuxfamily.org/>
5. Jahandari, H., and C. G. Farquharson. 2015. "Finite-Volume Modelling of Geophysical Electromagnetic Data on Unstructured Grids Using Potentials." *Geophysical Journal International* 202 (3): 1859–76. <https://doi.org/10.1093/gji/ggv257>.
6. Kashiwagi, Masahide. 2024. "Kv - a C++ Library for Verified Numerical Computation.", <http://verifiedby.me/kv/>
7. Kelbert, Anna, Naser Meqbel, Gary D. Egbert, and Kush Tandon. 2014. "ModEM: A Modular System for Inversion of Electromagnetic Geophysical Data." *Computers and Geosciences* 66 (May):40–53. <https://doi.org/10.1016/j.cageo.2014.01.010>.
8. Key, Kerry W., Steven C. Constable, and Chester J. Weiss. 2006. "Mapping 3D Salt Using the 2D Marine Magnetotelluric Method: Case Study from Gemini Prospect, Gulf of Mexico." *Geophysics* 71 (1). <https://doi.org/10.1190/1.2168007>.

9. Kingma, Diederik P., and Jimmy Ba. 2014. "Adam: A Method for Stochastic Optimization," December. <http://arxiv.org/abs/1412.6980>.
10. Mackie, Randall L, Theodore R Madden, and Wannamaker Philip E. 1993. "Three-Dimensional Magnetotelluric Modeling Using Difference Equations - Theory and Comparisons to Integral Equation Solutions." *GEOPHYSICS*. Vol. 58. <http://segdl.org/>.
11. Miensoopust, Marion P., Pilar Queralt, Alan G. Jones, Dmitry Avdeev, Anna Avdeeva, Ralph Uwe Börner, David Bosch, et al. 2013. "Magnetotelluric 3-D Inversion-a Review of Two Successful Workshops on Forward and Inversion Code Testing and Comparison." *Geophysical Journal International* 193 (3): 1216–38. <https://doi.org/10.1093/gji/ggt066>.
12. Mitsuhashi, Yuji, and Toshihiro Uchida. 2004. "3D Magnetotelluric Modeling Using the T-Ω Finite-Element Method." *Geophysics* 69 (1): 108–19. <https://doi.org/10.1190/1.1649380>.
13. Nam, Myung Jin, Hee Joon Kim, Yoonho Song, Tae Jong Lee, Jeong-Sul Son, and Jung Hee Suh. 2007. "3D Magnetotelluric Modelling Including Surface Topography." *Geophysical Prospects* 55 (2): 277–87. <https://doi.org/10.1111/j.1365-2478.2007.00614.x>.
14. Ogawa, Yasuo. 2002. "ON TWO-DIMENSIONAL MODELING OF MAGNETOTELLURIC FIELD DATA." *Surveys in Geophysics* 23:251–73.
15. O'Hara, Keith. 2024. "OptimLib.", <https://www.kthohr.com/optimlib.html>
16. Plessix, Rene Edouard. 2006. "A Review of the Adjoint-State Method for Computing the Gradient of a Functional with Geophysical Applications." *Geophysical Journal International*. <https://doi.org/10.1111/j.1365-246X.2006.02978.x>.
17. Sasaki, Yutaka. 2004. "Three-Dimensional Inversion of Static-Shifted Magnetotelluric Data." *Earth Planets Space*. Vol. 56.
18. Siripunvaraporn, Weerachai, Gary Egbert, and Yongwimon Lenbury. 2002. "Numerical Accuracy of Magnetotelluric Modeling: A Comparison of Finite Difference Approximations." *LETTER Earth Planets Space*. Vol. 54.
19. Siripunvaraporn, Weerachai, Gary Egbert, Yongwimon Lenbury, and Makoto Uyeshima. 2005. "Three-Dimensional Magnetotelluric Inversion: Data-Space Method." *Physics of the Earth and Planetary Interiors* 150 (1-3 SPEC. ISS.): 3–14. <https://doi.org/10.1016/j.pepi.2004.08.023>.
20. Takakura, Shinichi. 2014. "Estimation of a Regional Geothermal System by the Electromagnetic Exploration." *BUTSURI-TANSA* 67 (3): 195–203.
21. Toh, Hiroaki, Adam SCHULTZ, and Makoto Uyeshima. 2000. "Electromagnetic Induction in Fully Heterogeneous Spheres Using the Staggered Grid Finite Difference Method."
22. Uchida, Toshihiro. 2022. "Consideration on Three-Dimensional Inversion of Magnetotelluric Data by Finite-Difference Modeling Including Topography: Data Interpretation at Northern Hakkoda Area." *Journal of the Geothermal Research Society of Japan* 44 (1): 7–21.
23. Uchida Toshihiro, and Yusuke Yamaya. 2023. "Three-Dimensional Finite-Element and Finite Difference Inversions of Magnetotelluric Data: Application at Northern Hakkoda Geothermal Area." *Journal of the Geothermal Research Society of Japan* 45 (3): 175–94. <https://doi.org/10.11367/grsj.45.175>.
24. Usui, Yoshiya. 2015. "3-D Inversion of Magnetotelluric Data Using Unstructured Tetrahedral Elements: Applicability to Data Affected by Topography." *Geophysical Journal International* 202 (2): 828–49. <https://doi.org/10.1093/gji/ggv186>.
25. Usui, Yoshiya, Makoto Uyeshima, Hideaki Hase, Hiroshi Ichihara, Koki Aizawa, Takao Koyama, Shin'ya Sakanaka, et al. 2024. "Three-Dimensional Electrical Resistivity Structure Beneath a Strain Concentration Area in the Back-Arc Side of the Northeastern Japan Arc." *Journal of Geophysical Research: Solid Earth* 129 (5). <https://doi.org/10.1029/2023JB028522>.
26. Zhu, Xiaoxiong, Jie Liu, Yian Cui, and Chunye Gong. 2022. "A Scalable Parallel Algorithm for 3-D Magnetotelluric Finite Element Modeling in Anisotropic Media." *IEEE Transactions on Geoscience and Remote Sensing* 60. <https://doi.org/10.1109/TGRS.2021.3078735>.

Disclaimer/Publisher's Note: The statements, opinions and data contained in all publications are solely those of the individual author(s) and contributor(s) and not of MDPI and/or the editor(s). MDPI and/or the editor(s) disclaim responsibility for any injury to people or property resulting from any ideas, methods, instructions or products referred to in the content.