

The UBC
Geophysical Inversion Facility



MT3D: 3D Inversion of MT Data

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Background for MT3D

- MT3D developed as part of a GIF industry consortium (2000-2003)
- Basic aspects of forward modelling and inversion parallel those for controlled sources
 - Finite volume discretization for forward modelling
 - Gauss-Newton methodology for inversion
- References:
 - Farquharson, Oldenburg, Haber, and Shekhtman, 2002, 3D Inversion of MT Data SEG Ext. Abstract
 - Haber, Ascher, and Oldenburg, 2004, 3D Inversion of frequency and time domain data using all-at-once. Geophysics 69, p 1214
 - Haber, Ascher, Aruliah and Oldenburg, 2000, Fast simulation of 3D EM problems using potentials. J. Comp. Phys. 163, p 150.
 - Haber and Ascher: 2001 Fast finite volume simulation of 3D EM problems with highly discontinuous coefficients. SIAM J. Sci. Comp. p 1943



Forward Modelling

- **FD Maxwell's equations ($e^{-i\omega t}$)**

$$\nabla \times \mathbf{E} - i\omega \mu \mathbf{H} = 0$$

$$\nabla \times \mathbf{H} - (\sigma + i\omega \epsilon) \mathbf{E} = \mathbf{J}^e$$

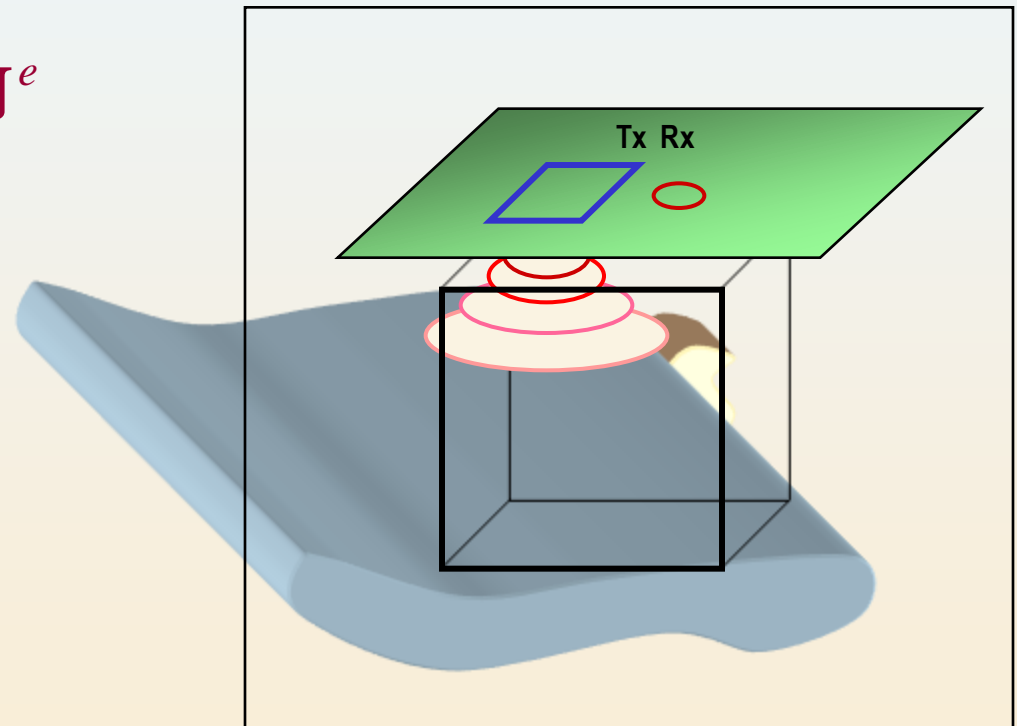
$$\nabla \cdot \epsilon \mathbf{E} = 0$$

$$\nabla \cdot \mu \mathbf{H} = 0$$

- **Boundary condition**

$$\mathbf{n} \times \mathbf{H} = 0$$

Electromagnetic induction



Forward Modelling

- A Helmholtz decomposition with Coulomb gauge

$$\mathbf{E} = -\mathbf{A} - \nabla \phi$$

$$\nabla \cdot \mathbf{A} = 0$$

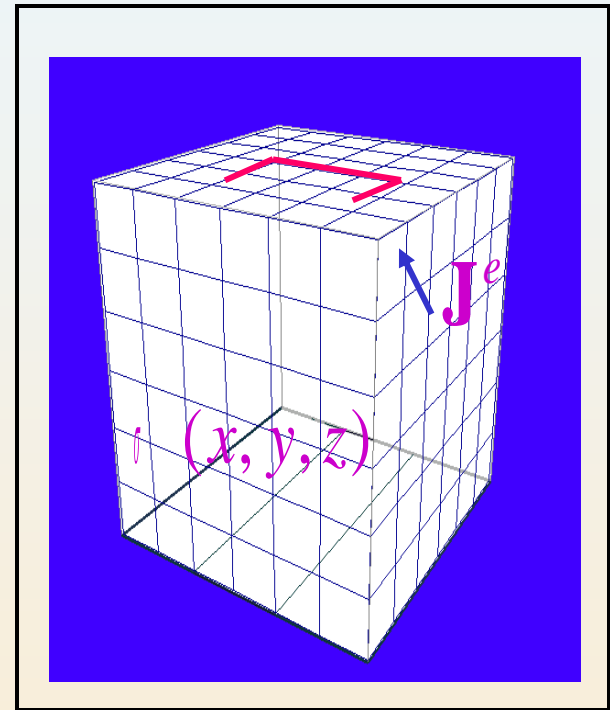
- System equations for \mathbf{A} and ϕ

$$\begin{pmatrix} \mathbf{L}_\mu - i\omega\hat{\sigma} & -i\omega\hat{\sigma}^\top \\ \nabla \cdot \hat{\sigma} & \nabla \cdot \hat{\sigma}^\top \end{pmatrix} \begin{pmatrix} \mathbf{A} \\ \phi \end{pmatrix} = \begin{pmatrix} -i\omega\mathbf{J}^e \\ -\nabla \cdot \mathbf{J}^e \end{pmatrix}$$

where

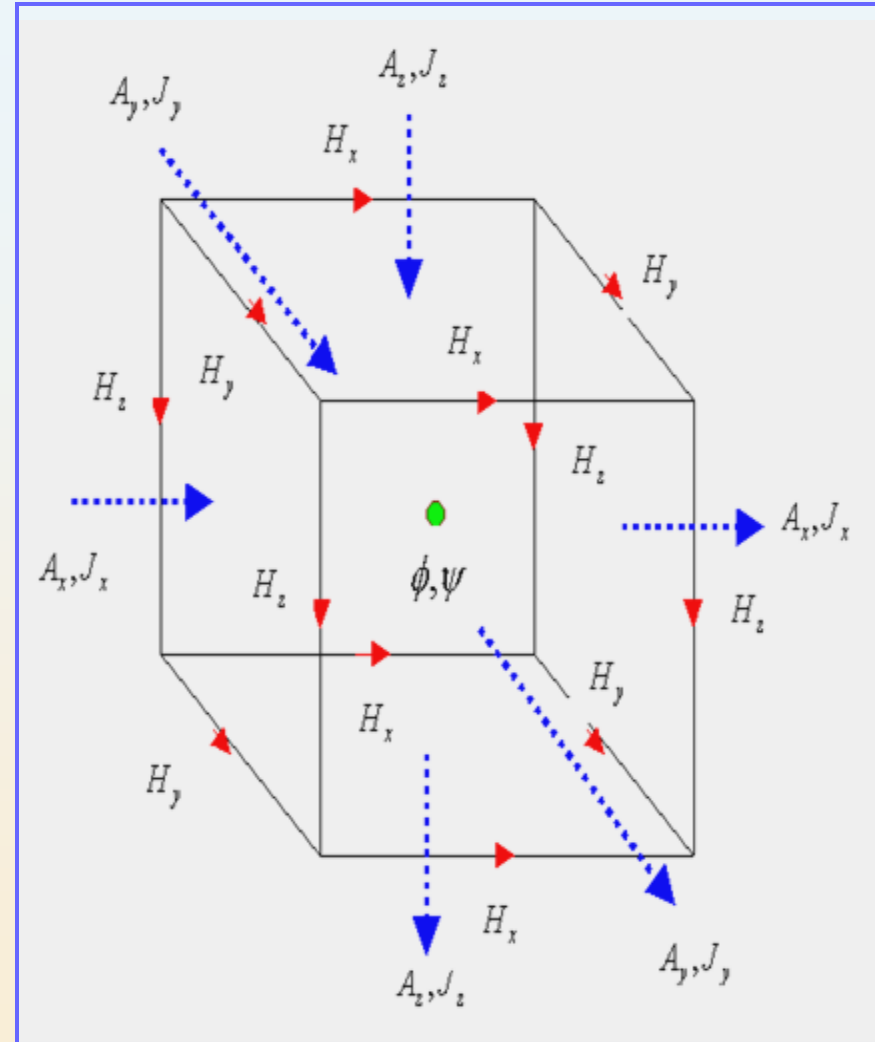
$$\mathbf{L}_\mu = \nabla \times \frac{1}{\mu} \nabla \times - \nabla \frac{1}{\mu} \nabla \cdot$$

$$\hat{\sigma} = \sigma - i\omega\epsilon$$



• Discretization with staggered grid

- \mathbf{A}, \mathbf{J} are defined on the faces
- ϕ is in the centers
- \mathbf{H} is on the edges



• **Discretized matrix system**

$$\begin{pmatrix} \mathbf{L}_h \cdot i\omega \mathbf{M}_{\hat{\sigma}} & -i\omega \mathbf{M}_{\hat{\sigma}} \nabla_h \\ \nabla_h \cdot \mathbf{M}_{\hat{\sigma}} & \nabla_h \cdot \mathbf{M}_{\hat{\sigma}} \nabla_h \end{pmatrix} \begin{pmatrix} \mathbf{A} \\ \mathbf{u} \end{pmatrix} = \begin{pmatrix} -i\omega \mathbf{J}^e \\ \nabla_h \cdot \mathbf{J}^e \end{pmatrix}$$

where $\mathbf{L}_\mu = \nabla_h^{(e)}$, $\mathbf{M}_e^{-1} \nabla_h^{(f)}$, $\nabla_h \mathbf{M}_c^{-1}$

$\nabla_h^{(e)}$, $\nabla_h^{(f)}$: Discretization of **Curl**

$\nabla_h \cdot$: **Div**

∇_h : **Grad**

$\mathbf{M}_{\hat{\sigma}}$: Conductivity

$\mathbf{M}_e, \mathbf{M}_c$: Magnetic permeability

or simplified as

$$\mathcal{A}(\mathbf{m})\mathbf{u} = \mathbf{q} \quad \longrightarrow \quad \mathbf{u} = \mathcal{A}(\mathbf{m})^{-1}\mathbf{q}$$

- Solving the matrix system

$$\mathcal{A}(\mathbf{m})\mathbf{u} = \mathbf{q}$$

- Iterative solver BiCGSTAB
- Preconditioner

Find $\mathbf{M} \cong \mathcal{A}(\mathbf{m})$
 where \mathbf{M}^{-1} is readily computed

Solve $\mathbf{M}^{-1}\mathcal{A}(\mathbf{m})\mathbf{u} = \mathbf{M}^{-1}\mathbf{q}$

$$\mathbf{M} = \begin{pmatrix} \mathbf{L}_h & i\omega\mathbf{M}_{\hat{\sigma}} & i\omega\mathbf{M}_{\hat{\sigma}}^{\top} & h \\ 0 & & \mathbf{M}_{\hat{\sigma}}^{\top} & h \end{pmatrix}$$



Inverse Problem

- **Minimize** $\phi = \phi_d + \beta \phi_m$

where

$$\phi_d = \left\| \mathbf{W}_d (F[\mathbf{m}] \cdot \mathbf{d}^{obs}) \right\|^2, \quad F[\mathbf{m}] = f(\mathbf{Q}\mathbf{u})$$

$$\phi_m = \left\| \mathbf{W} (\mathbf{m} \cdot \mathbf{m}_{ref}) \right\|^2$$

β : Regularization parameter

\mathbf{Q} : Projection matrix

\mathbf{u} : Potentials

\mathbf{d}^{obs} : Observed data

$\mathbf{m}, \mathbf{m}_{ref}$: Model and Reference model

\mathbf{W}_d, \mathbf{W} : Data error, model weighting

• Solving the inverse problem

-- Differentiating the objective function ϕ with model \mathbf{m}

$$\frac{\partial \phi}{\partial \mathbf{m}} : \mathbf{g}(\mathbf{m}) : \mathbf{J}^T \mathbf{W}_d^T [F[\mathbf{m}] \cdot \mathbf{d}^{obs}] + \mathbf{W}^T \mathbf{W}(\mathbf{m} \cdot \mathbf{m}_{ref})$$

where sensitivity matrix

$$\mathbf{J} = \frac{\partial F[\mathbf{m}]}{\partial \mathbf{m}} = \frac{\partial f[\mathbf{Q}\mathbf{u}]}{\partial \mathbf{m}} = \mathbf{S}\mathbf{Q} \frac{\partial \mathbf{u}}{\partial \mathbf{m}} = -\mathbf{S}\mathbf{Q}\mathcal{A}(\mathbf{m})^{-1}\mathbf{G}(\mathbf{m}, \mathbf{u})$$

and

$$\mathbf{G}(\mathbf{m}, \mathbf{u}) = \partial[\mathcal{A}(\mathbf{m})\mathbf{u}]/\partial \mathbf{m} \quad \mathbf{S} = \frac{\partial f}{\partial (\mathbf{E}, \mathbf{H})}$$

f : Function that converts E, H fields to other data types

• Gauss-Newton method

-- Solve $\mathbf{g}(\mathbf{m}) = 0$, and let $\mathbf{F}[\mathbf{m} + \delta\mathbf{m}] = \mathbf{F}[\mathbf{m}] + \mathbf{J} \delta\mathbf{m}$

$$(\mathbf{J}^T \mathbf{J} + \beta \mathbf{W}^T \mathbf{W}) \delta\mathbf{m} = -\mathbf{g}(\mathbf{m})$$

The sensitivity matrix \mathbf{J} has been normalized by \mathbf{W}_d

$$\mathbf{J} = -\mathbf{W}_d \mathbf{S} \mathbf{Q} \mathcal{A}(\mathbf{m})^{-1} \mathbf{G}(\mathbf{m}, \mathbf{u}).$$

and the gradient is

$$\mathbf{g}(\mathbf{m}) = -\mathbf{G}(\mathbf{m}, \mathbf{u})^T \mathcal{A}(\mathbf{m})^{-T} \mathbf{Q}^T \mathbf{S}^T \mathbf{W}_d^T \mathbf{W}_d [\mathbf{F}[\mathbf{m}] - \mathbf{b}] + \beta \mathbf{W}^T \mathbf{W} (\mathbf{m} - \mathbf{m}_{ref}) .$$

Matrices \mathbf{W}_d , \mathbf{W} , \mathbf{S} , \mathbf{Q} , $\mathcal{A}(\mathbf{m})$, $\mathbf{G}(\mathbf{m}, \mathbf{u})$ are **SPARSE!**



• Solution of the matrix system

$$(\mathbf{J}^T \mathbf{J} + \mathbf{W}^T \mathbf{W}) \delta \mathbf{m} = -\mathbf{g}(\mathbf{m})$$

IPCG solver with preconditioner

$$\mathbf{M}^{-1}(\mathbf{J}^T \mathbf{J} + \mathbf{W}^T \mathbf{W})$$

Main computations:

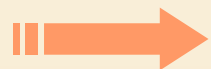
$$(1) \mathbf{J} \mathbf{v} = -\mathbf{W}_d \mathbf{S} \mathbf{Q} \underbrace{\mathcal{A}(\mathbf{m})^{-1} \mathbf{G} \mathbf{v}}_{\mathbf{w}}$$



Solve $\mathcal{A}(\mathbf{m}) \mathbf{f} = \mathbf{w}$

: *Forward modelling*

$$(2) \mathbf{J}^T \mathbf{v} = -\mathbf{G}^T \underbrace{\mathcal{A}(\mathbf{m})^{-T} \mathbf{Q}^T \mathbf{S}^T \mathbf{W}_d^T \mathbf{v}}_{\mathbf{w}}$$



Solve $\mathcal{A}^T(\mathbf{m}) \mathbf{f} = \mathbf{w}$

: *Adjoint modelling*

$$\text{UPDATE } \mathbf{m}_{k+1} = \mathbf{m}_k + \delta \mathbf{m}$$

Flow chart

Recall we are solving ...

$$d = d + \beta m$$

$$\phi(m) = \left\| \mathbf{W}_d (F[m] \cdot \mathbf{d}^{obs}) \right\|^2 + \beta \left\| \mathbf{W} (m - m_{ref}) \right\|^2$$

Choose β_0, m_{ref}

Evaluate $\phi(m_{ref}), g(m_{ref}),$ matrices $\mathbf{W}_d, \mathbf{W}...$

For β cooling loop

For $k = 1 \rightarrow$ max iterations

- IPCG to solve $\mathbf{W}_d \mathbf{W}_d^{-1} \mathbf{W}_d \mathbf{m}_k = \mathbf{W}_d \mathbf{W}_d^{-1} \mathbf{W}_d \mathbf{m}_k + \mathbf{W}_d \mathbf{W}_d^{-1} \mathbf{W}_d \mathbf{g}_k$
- Line search for step length α
- Update model $\mathbf{m}_{k+1} = \mathbf{m}_k + \alpha \mathbf{d}_k$
- Exit if $\phi(m_{k+1}) < \phi(m_k)$

End

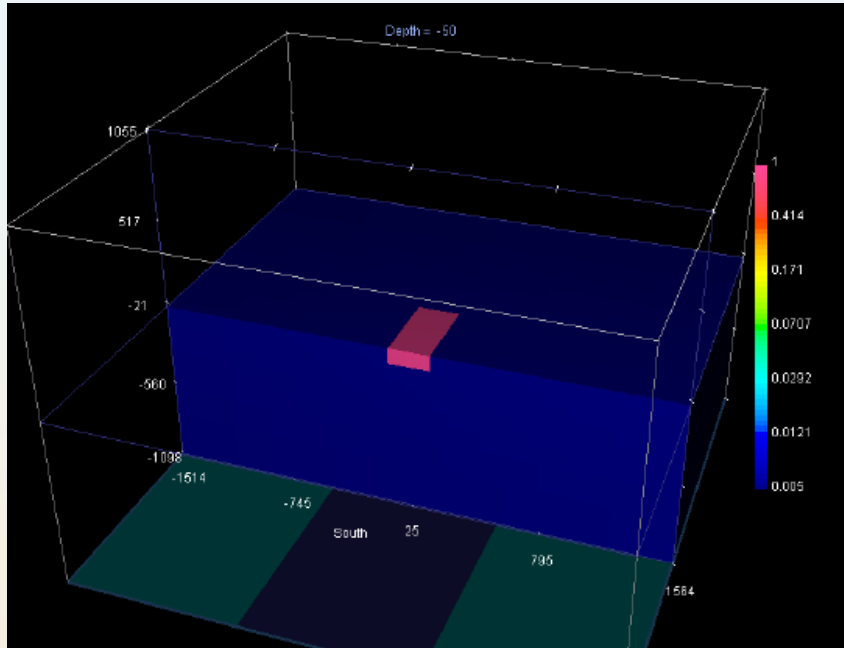
Reduce β

End



Synthetic Model

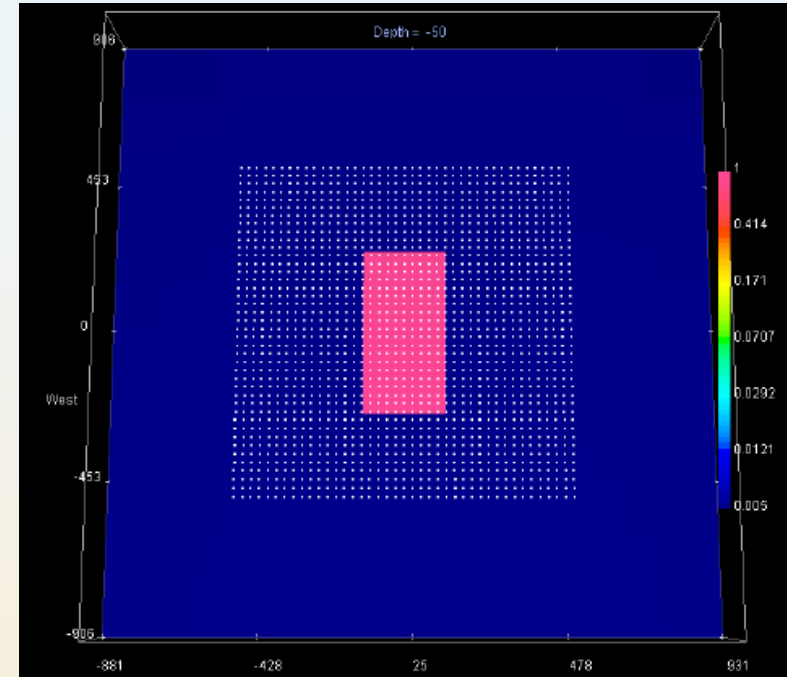
Cross section



Block: 250 X 500 X 100m
Buried 50 m

Block: 1 Ω -m
Background 200 Ω -m

Plan view showing model and station coverage

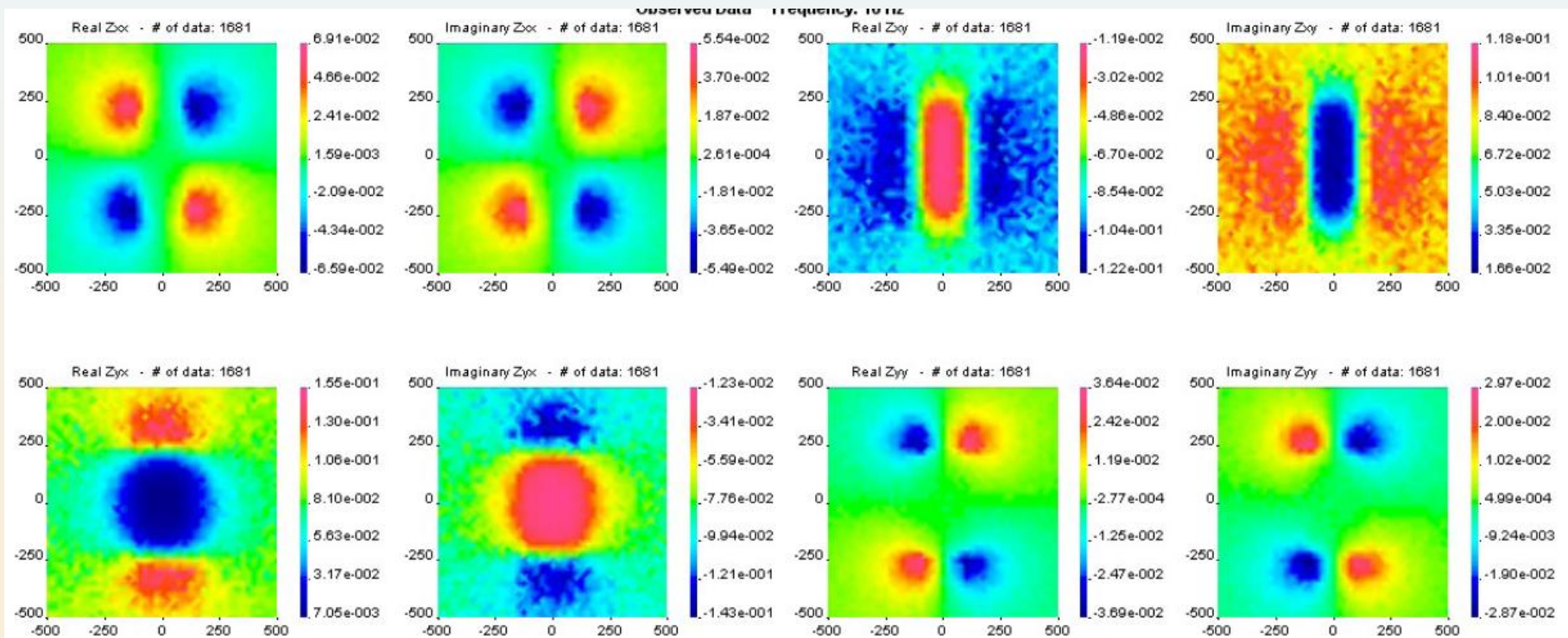


of stations: 1661
of data: 13,448

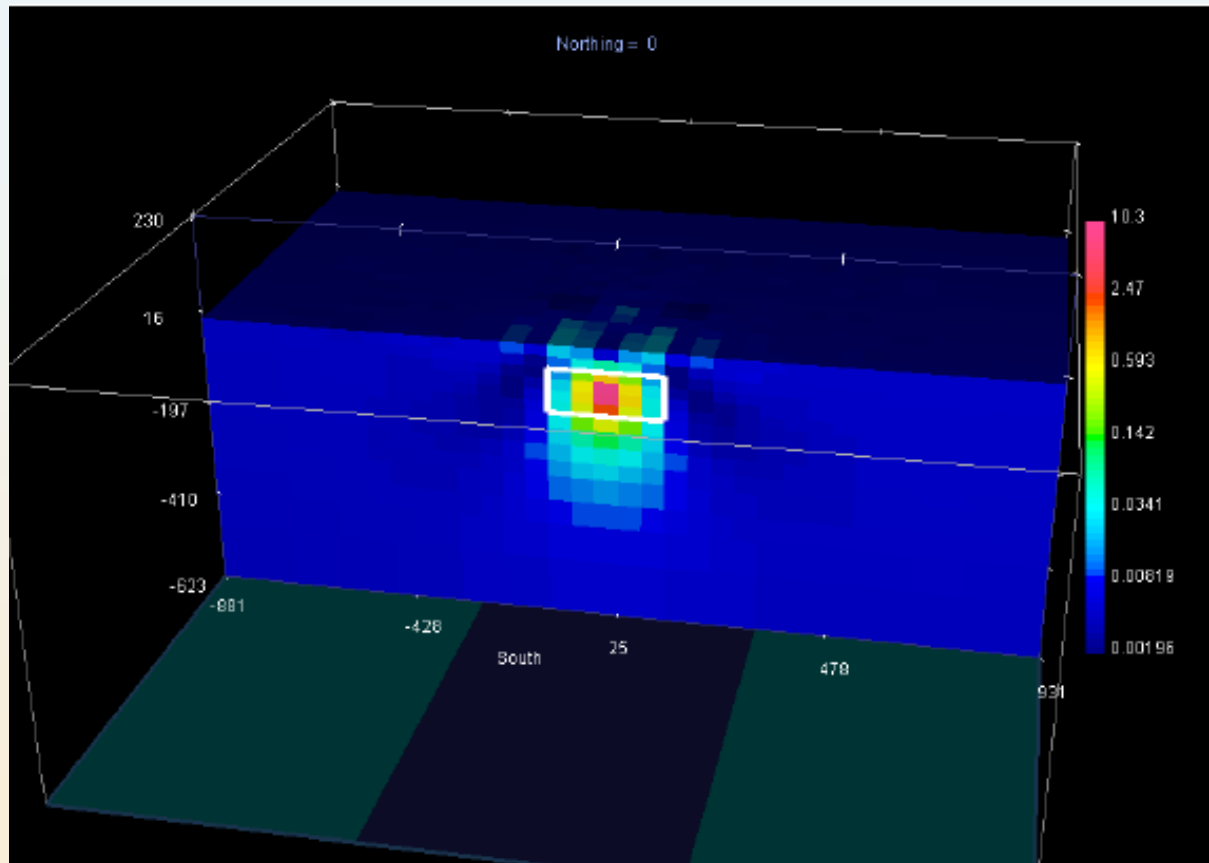
Mesh: 32 32 28 cells

Data for Inversion: Impedances 10Hz

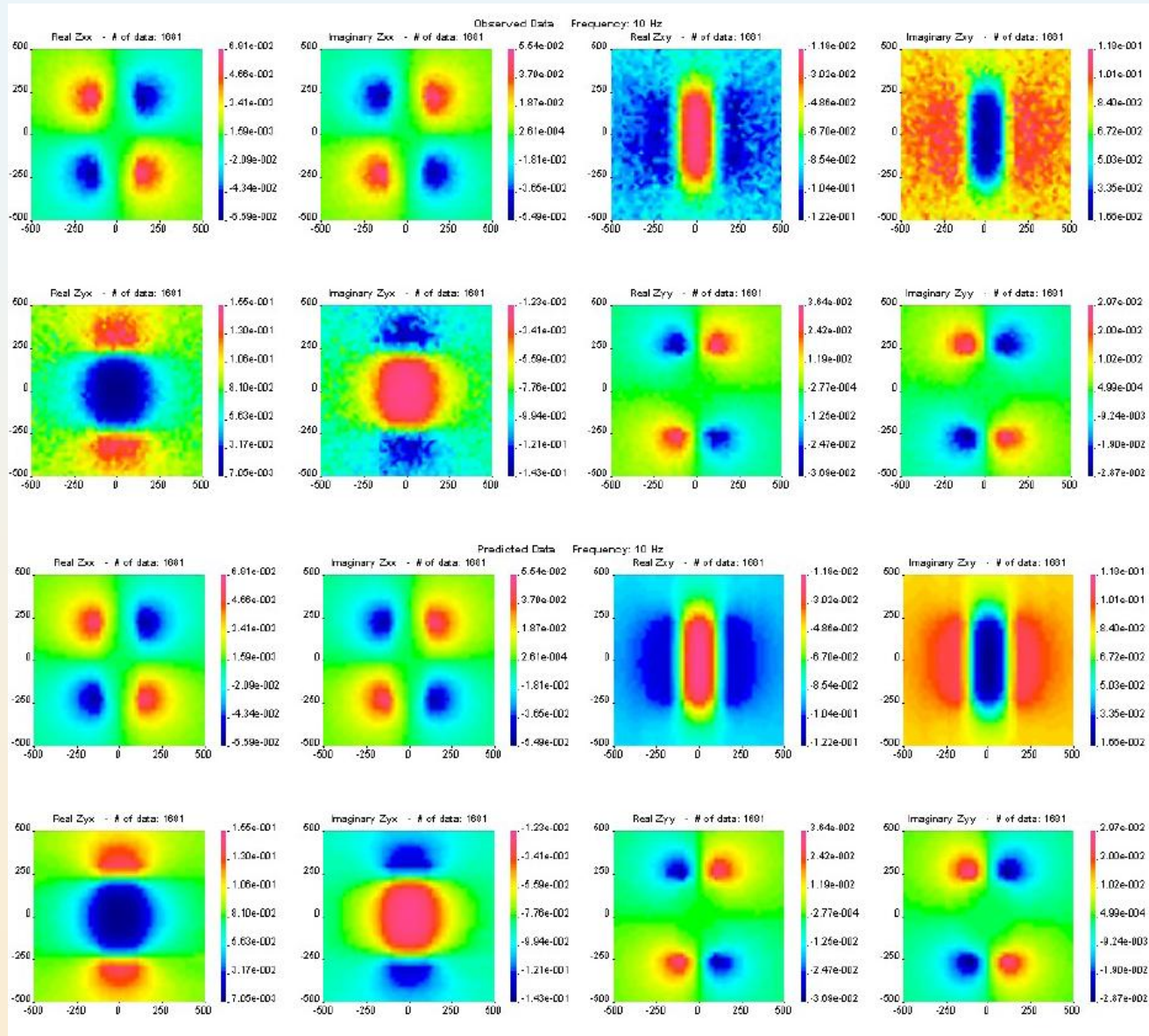
Noise is Gaussian and equivalent to:
5% of apparent resistivity
2 degrees in phase



Recovered model



Observed and predicted data



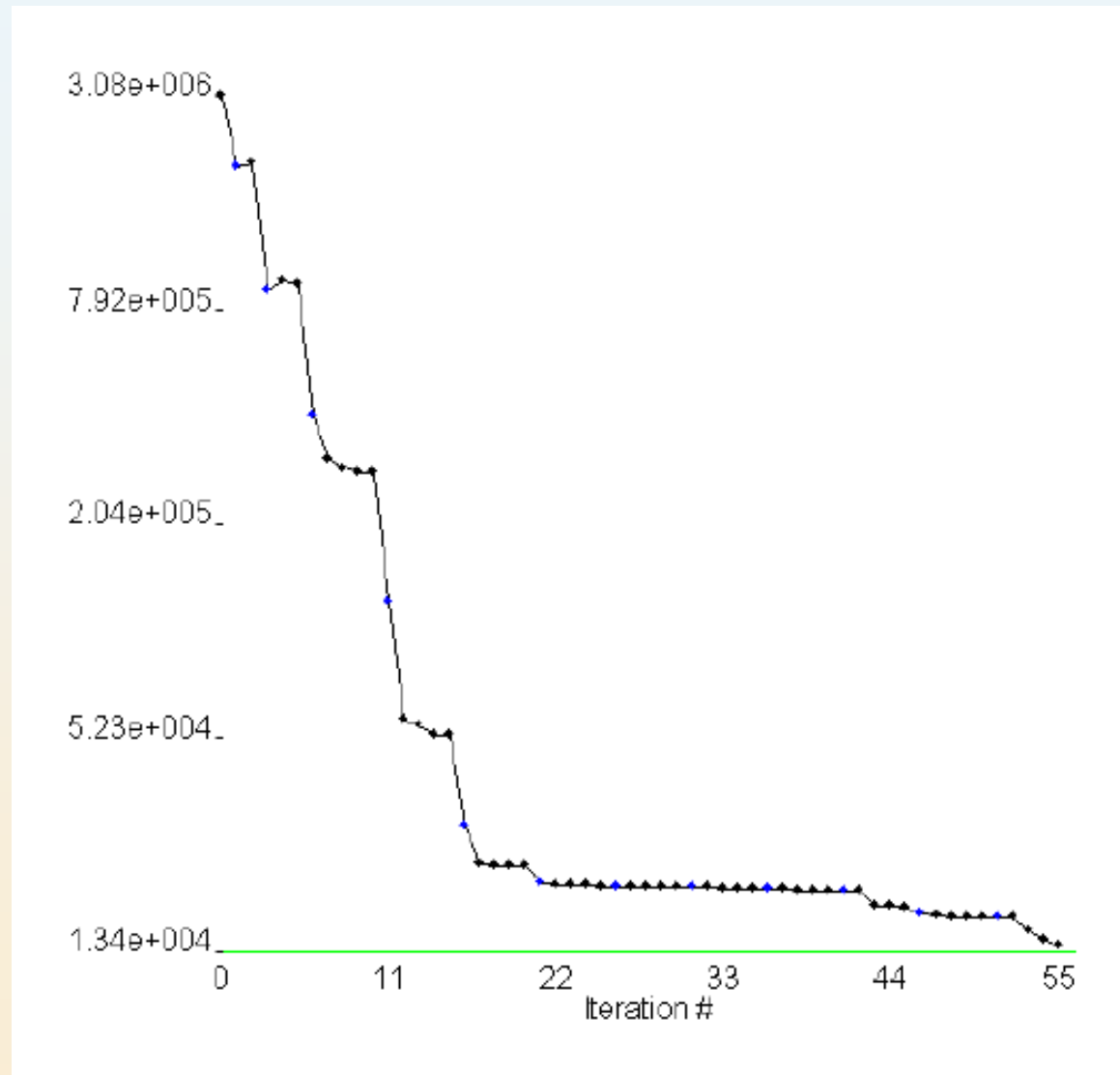
Observed

Predicted



Convergence curve

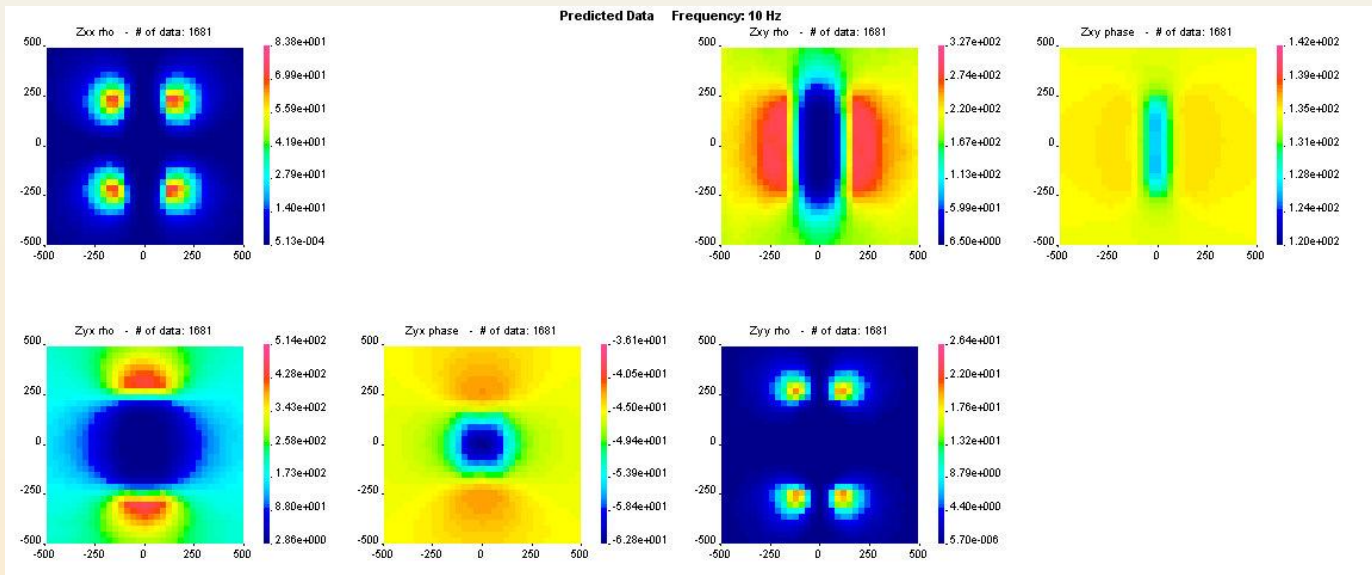
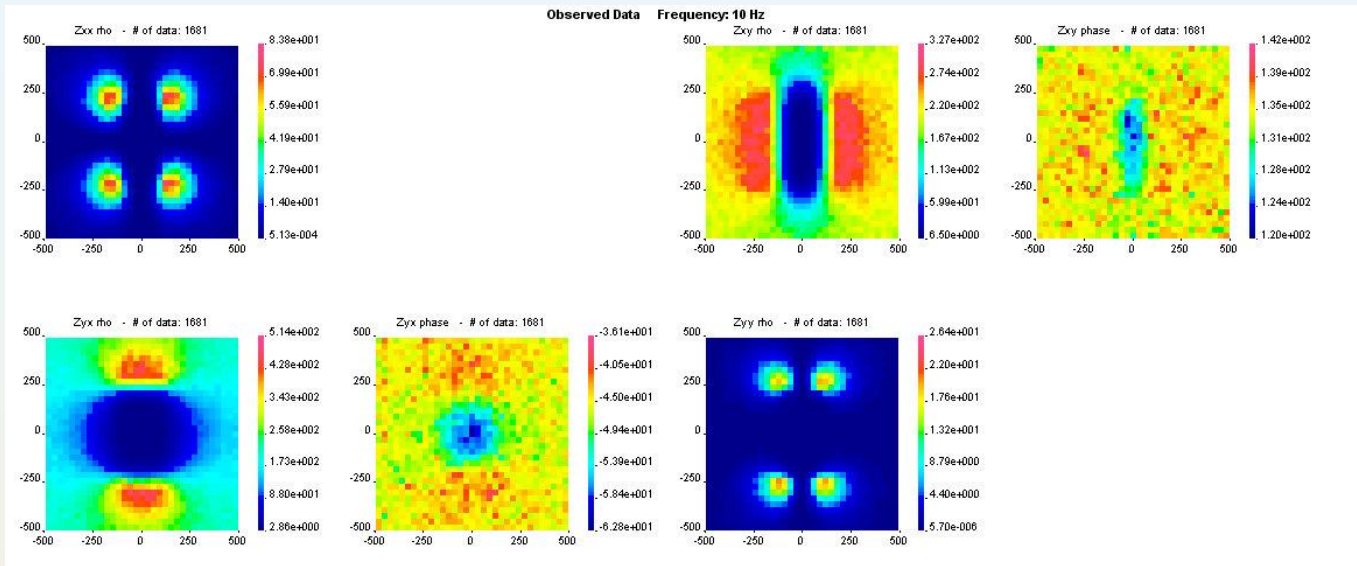
Misfit



Extra slides: apparent resistivity inversion

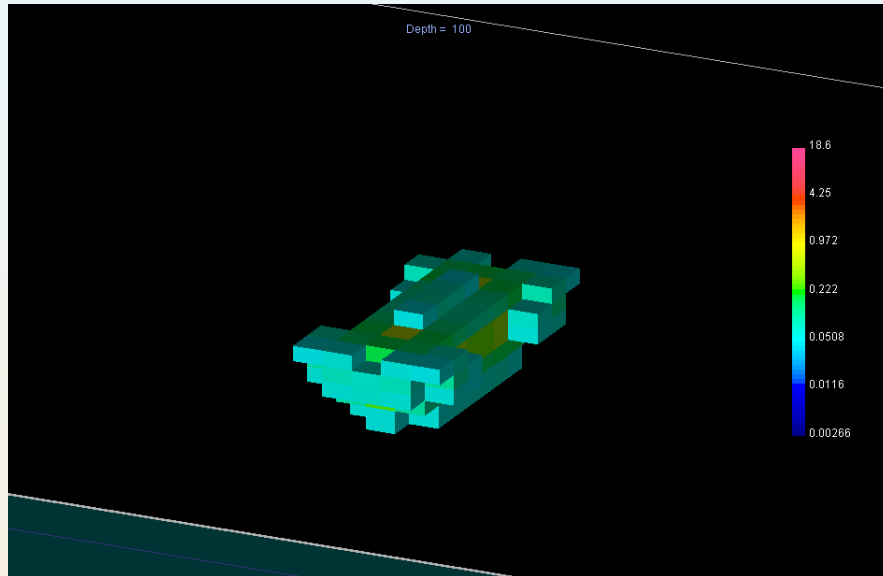


Invert apparent resistivities and phases



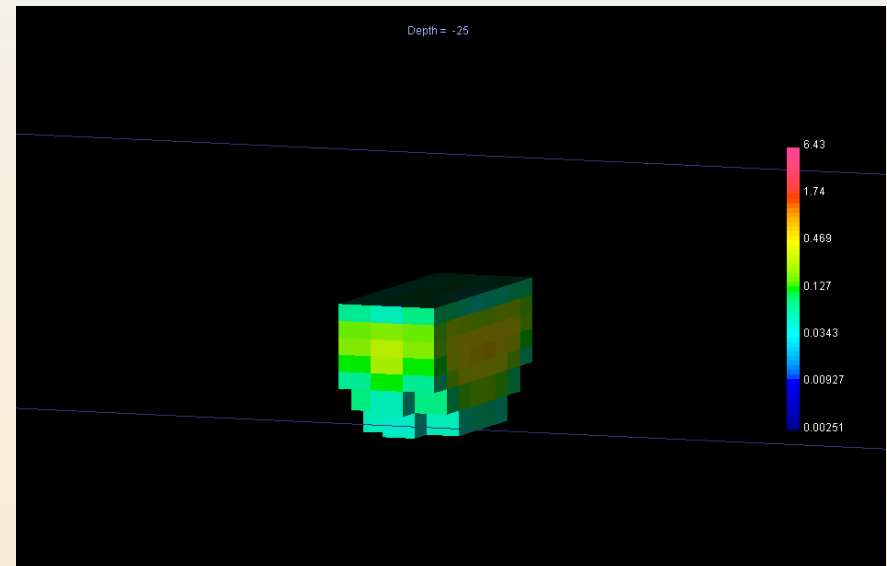
Inversion Comparisons

Apparent resistivity and phase



Cutoff 0.05

Real Imaginary Impedances

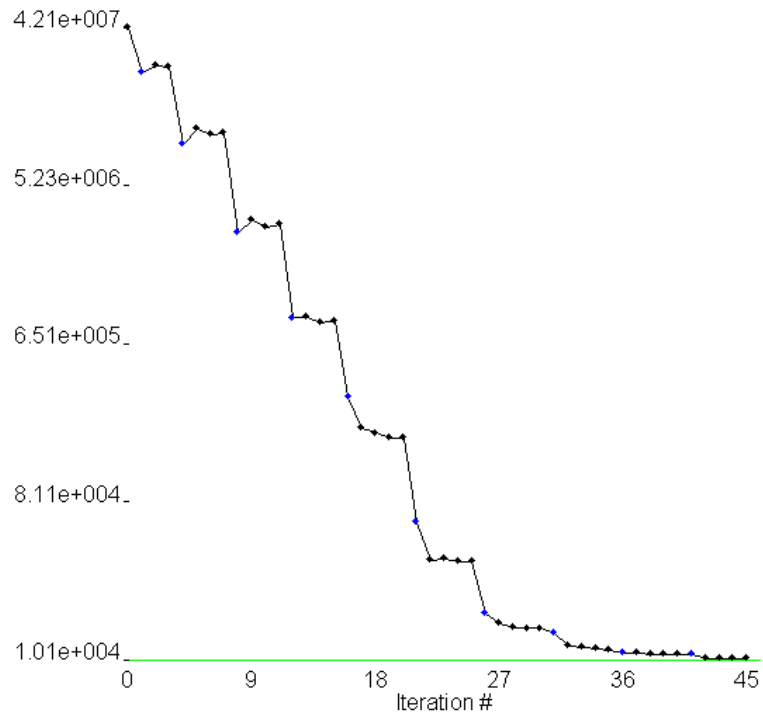


Convergence

Resistivity and phase

target misfit: 1.00860E+04
final misfit: 1.01970E+04

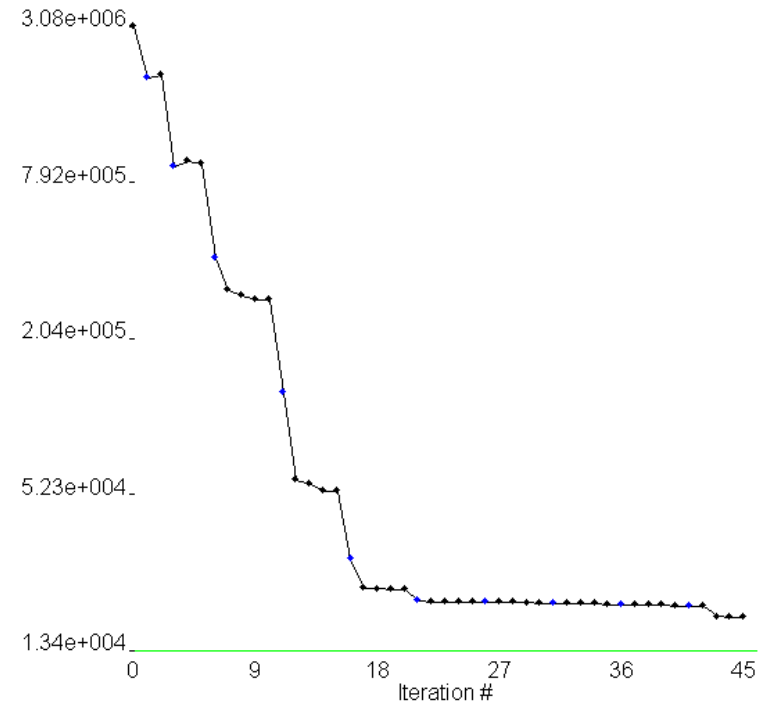
TOTAL cpu time: 12:57:01.78



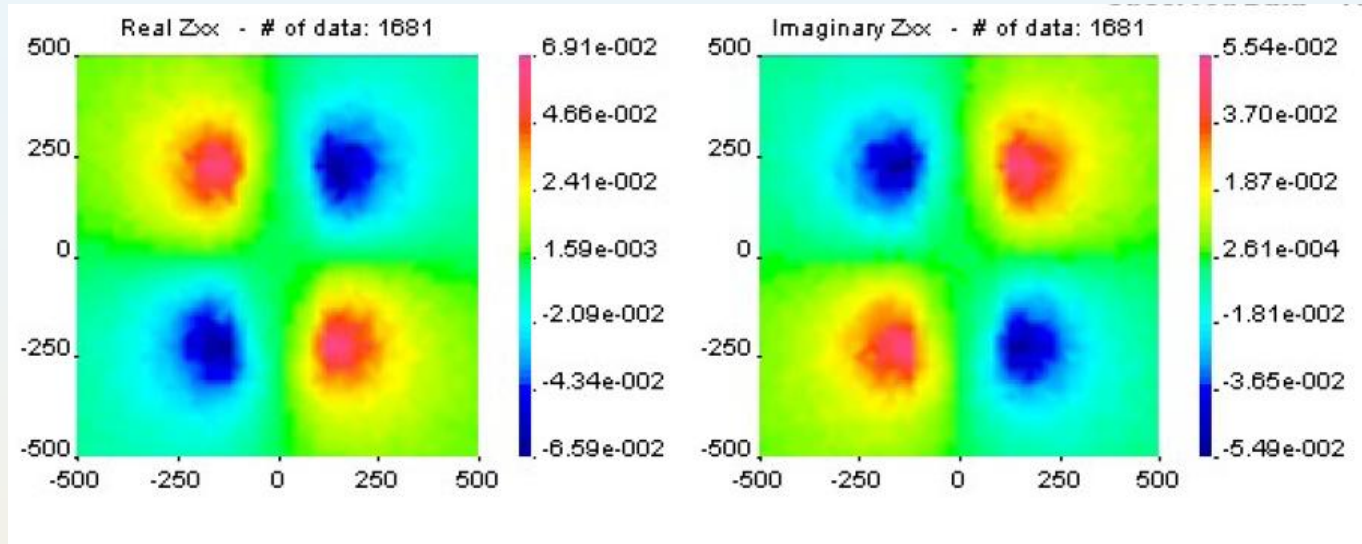
Real and Imaginary

target misfit: 1.34480E+04
final misfit: 1.78897E+04

TOTAL cpu time: 14:13:20.26

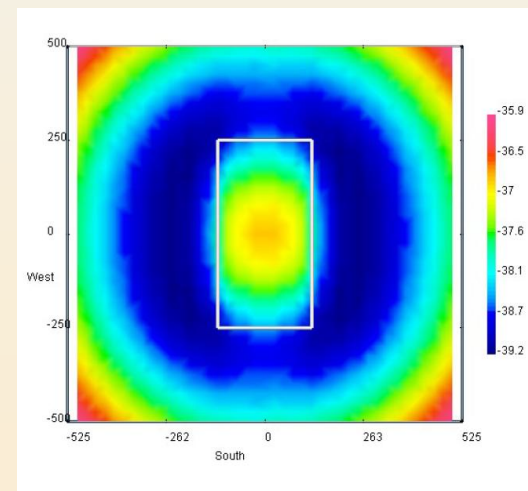
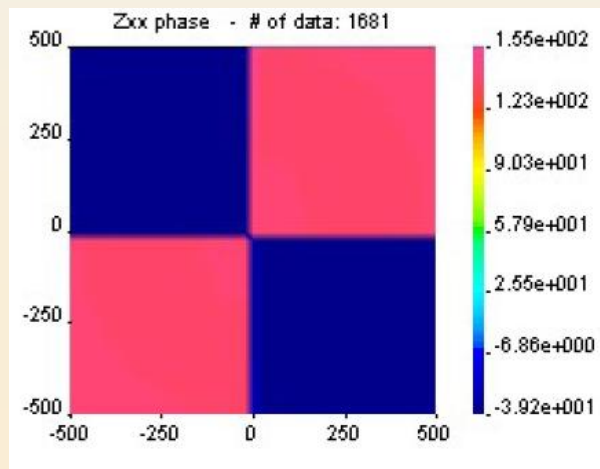


Phases for Z_{xx} , Z_{yy}

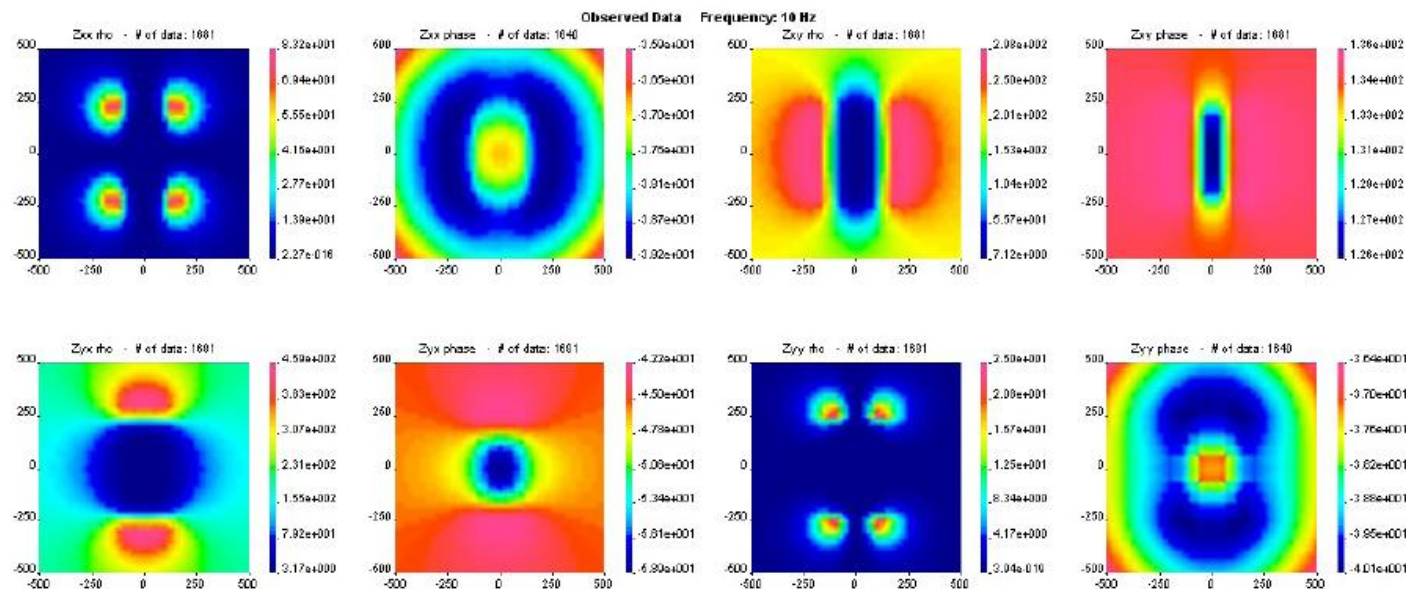
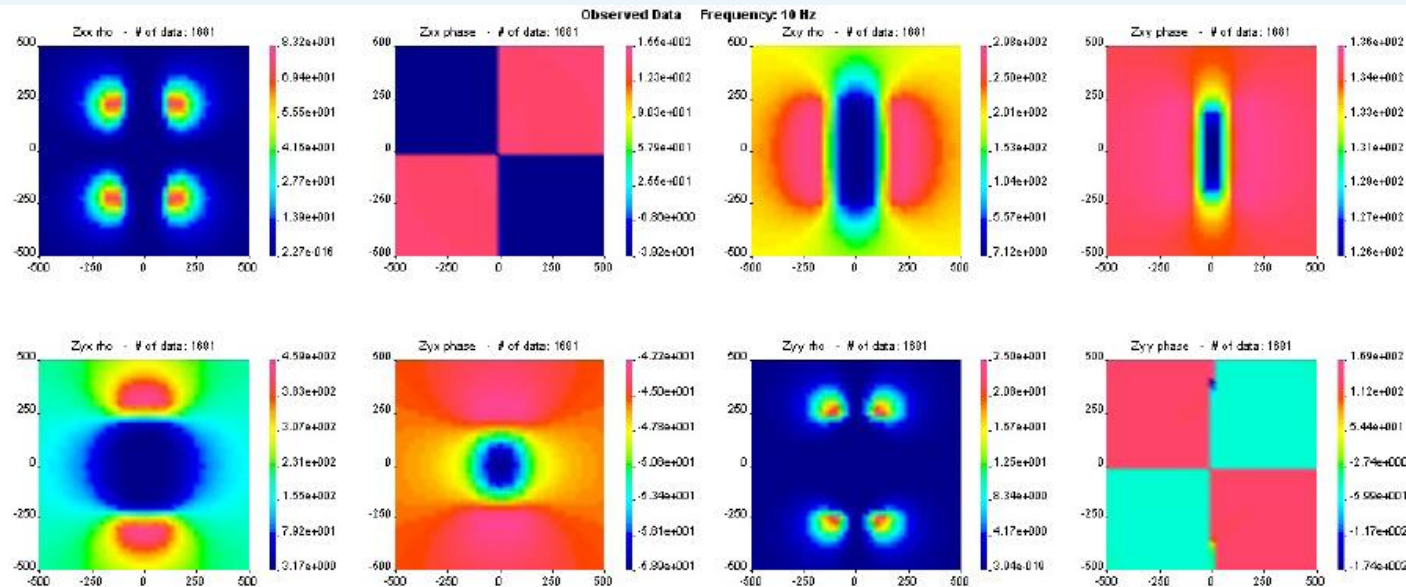


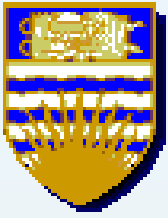
Phase evaluated with atan2
Lies in $[-40, 155]$ degrees

Phase evaluated with atan
Lies in $[-40, -35]$



Block data with altered phases





The UBC Geophysical Inversion Facility



Summary Discussion

Doug Oldenburg and Greg Newman

**March 2008
Dublin**



MT3D Workshop: Wednesday : Forward Modelling

Summary Comments:

- Finite difference, finite element, IE modelling
- Synthetic model generated by Marion and Pilar:
- Range of times for running
- Major difficulties sorting out the sign conventions of the impedance were of major difficulties (conjugate gradients, rotations) should be able to make comparisons.
- Results were quite comparable (major discrepancies were for diagonal components when impedance values were low).



MT3D Workshop: Wednesday : Forward Modelling

- Discussion points:
 - Need compatibility for comparison of results. (EDI format?)
 - What accuracy is required by the modelling?
 - How should workshop forward modelling results be archived?
 - What other models should be considered? (topography, complicated structure, near surface heterogeneity, multi-scale)
 - Modelling large scale feature (galvanic distortions).
 - How do we handle unknown boundary conditions?



MT3D Workshop: Thursday Inversion

Summary comments:

- Most inversion codes gave similar results on Dublin test

Discussion points:

- Archive results for Dublin test?
- Assignment of errors to impedance. Workflows and methodologies.
- Incorporation of tipper.
- Grid design (forward modelling and inversion meshes; frequency dependent grids; validation of grid, unstructured meshes)



MT3D Workshop: Thursday Inversion: More Discussion

- Regularization functionals: reference models, inclusion of a priori information.
- Including other information (bound constraints, tears, other geophysical measurement, geologic information, petrophysical constraints)
- Ways to ameliorate high sensitivity due to receiver locations. (filtering, weighting)
- Static Shift:
 - don't worry—just model them out;
 - distortion process then invert;
 - invert for conductivity and distortion



MT3D Workshop: Thursday Afternoon

Secret Model

- Most inversion results provided similar “shoe” structure
- Discussion points:
 - Why didn’t inversions detect the inconsistent data?
 - Other approaches for delineating inconsistencies. (e.g. coordinate systems)
 - Anything more to be done with “old secret model”?
- Another Secret Model?
 - Same model but true station locations
 - Data set contaminated with noise.
 - A more complicated model?



Model assessment or Appraisal (next workshop?)

- What are the approaches for ground-truth
- Constructing other inverse solutions
- Resolution estimates

MT3D Workshop: Friday

- Single RMS misfit value is not enough!
- Can we map fractures (eg for geothermal energy)
- Further ways for exploiting marine MT for hydrocarbons
- ?????



MT3D Workshop: Dublin

Thanks to Alan Jones

Thanks to Marion Miensopust

