

3D MT Inversion Dublin 2008 Workshop

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Inverse Problem Formulation



The goal of CSEM/MT inversion is to infer the electrical conductivity of earth's subsurface given CSEM/MT measurements.

The problem can be expressed as an inverse problem of the form

$$d_i = F_i(m) + e_i, \quad i = 1, 2, \dots, N, \quad m \in M.$$

where

m =model

F =forward modeling operator

d =observed data

e =observational error

Tikhonov Regularization



Tikhonov's method defines a regularized solution of the inverse problem to be a model m that minimizes the objective function

$$\phi(m) = (d - F(m))^T V^{-1} (d - F(m)) + \tau \|Lm\|^2$$

where

d =observed data vector

F =forward modeling operator

m =unknown model vector

V =error covariance matrix

L =a linear operator

τ =regularization parameter

Nonlinear Conjugate Gradients



- **Non-linear conjugate gradients** applies directly to the minimization of ϕ . The model sequence is given by

$$m_{j+1} = m_j + \alpha_{j+1}h_{j+1}$$

where h_{j+1} is a given search direction.

- Like linear CG, NLCG computes the search directions as

$$h_{j+1} = C_j g_j + \beta_j h_j$$

where C_j is the preconditioner.

- Unlike linear CG, α_{j+1} is computed to minimize ϕ along the search direction using an iterative line minimization procedure.

Nonlinear Conjugate Gradients



- In our NLCG algorithm, the line minimization automatically defaults to a one-step computation when the Forward problem can be well-approximated by its linear expansion around the previous model.
- Operations with A and A^T are computed efficiently using reciprocity relationships. They require solving two extra forward problems per frequency.

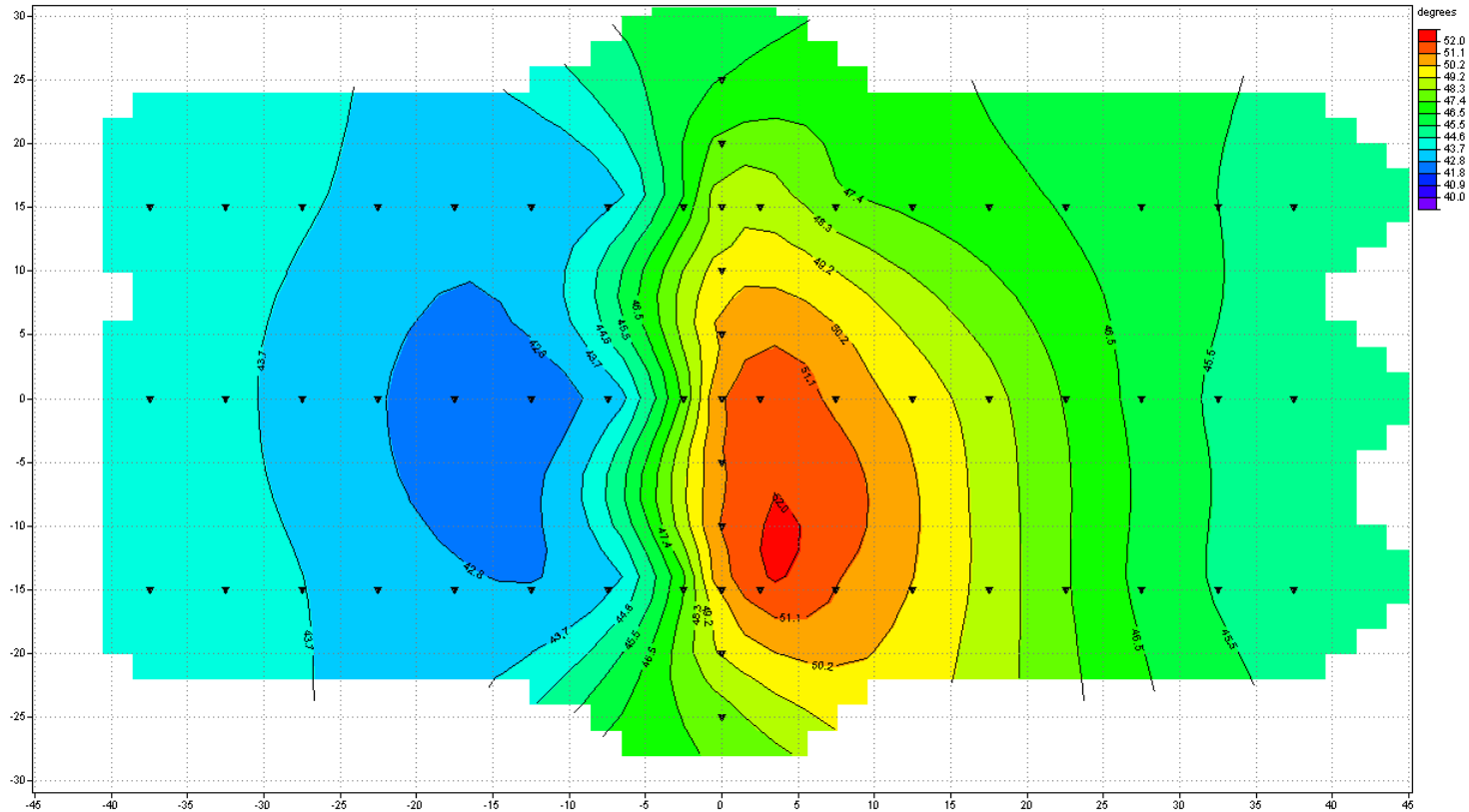


- The preconditioner is important for an efficient algorithm. The preconditioner should 'steer' the gradient in the direction of the true solution, but should not be computationally expensive.
- We take the preconditioner to be the inverse of an approximation to the Hessian:

$$C = (\Lambda + \lambda L^T L)^{-1}$$

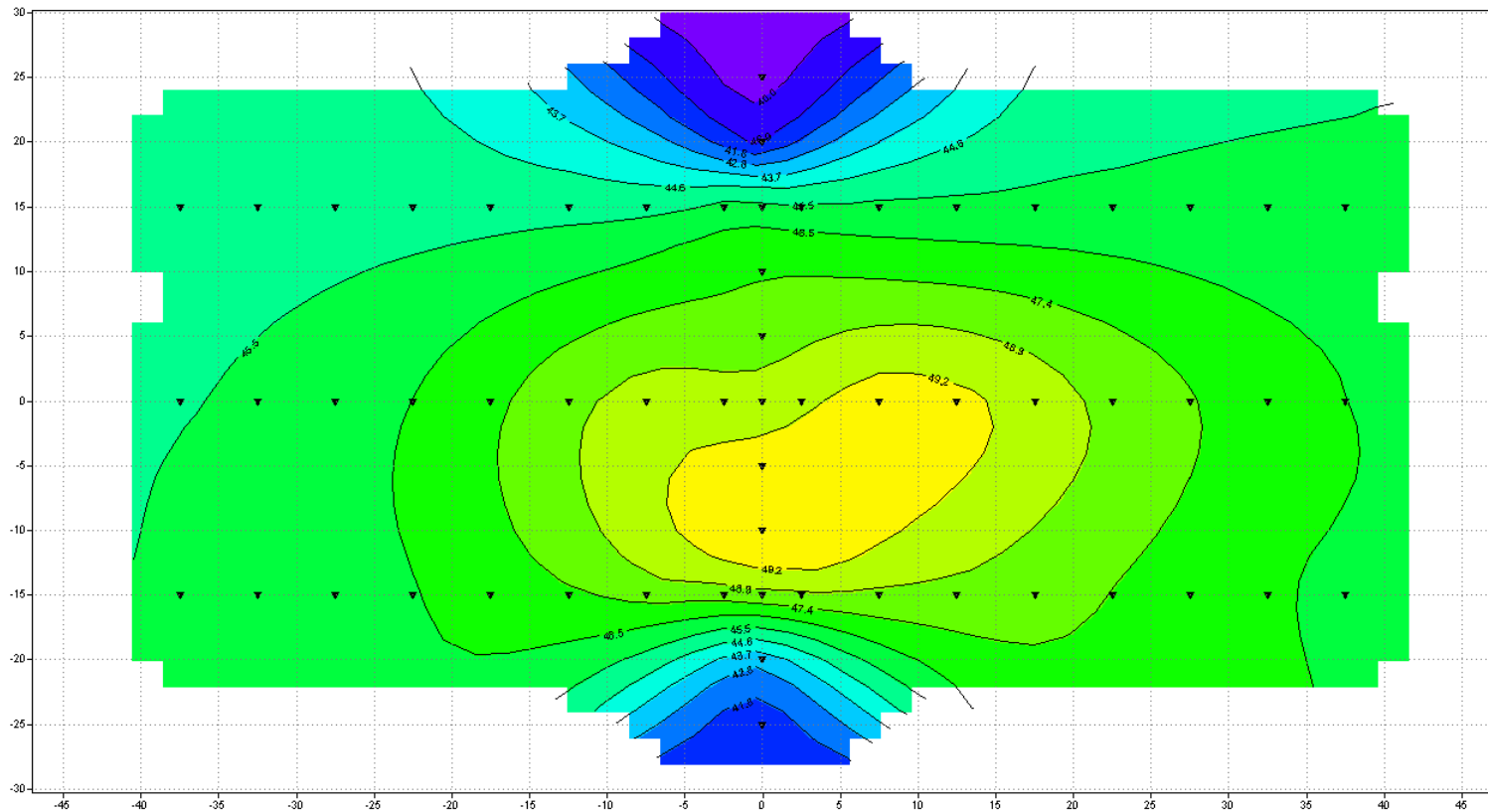
Where Λ approximates the diagonal part of $A^T V^{-1} A^T$

3D Forward Model Data



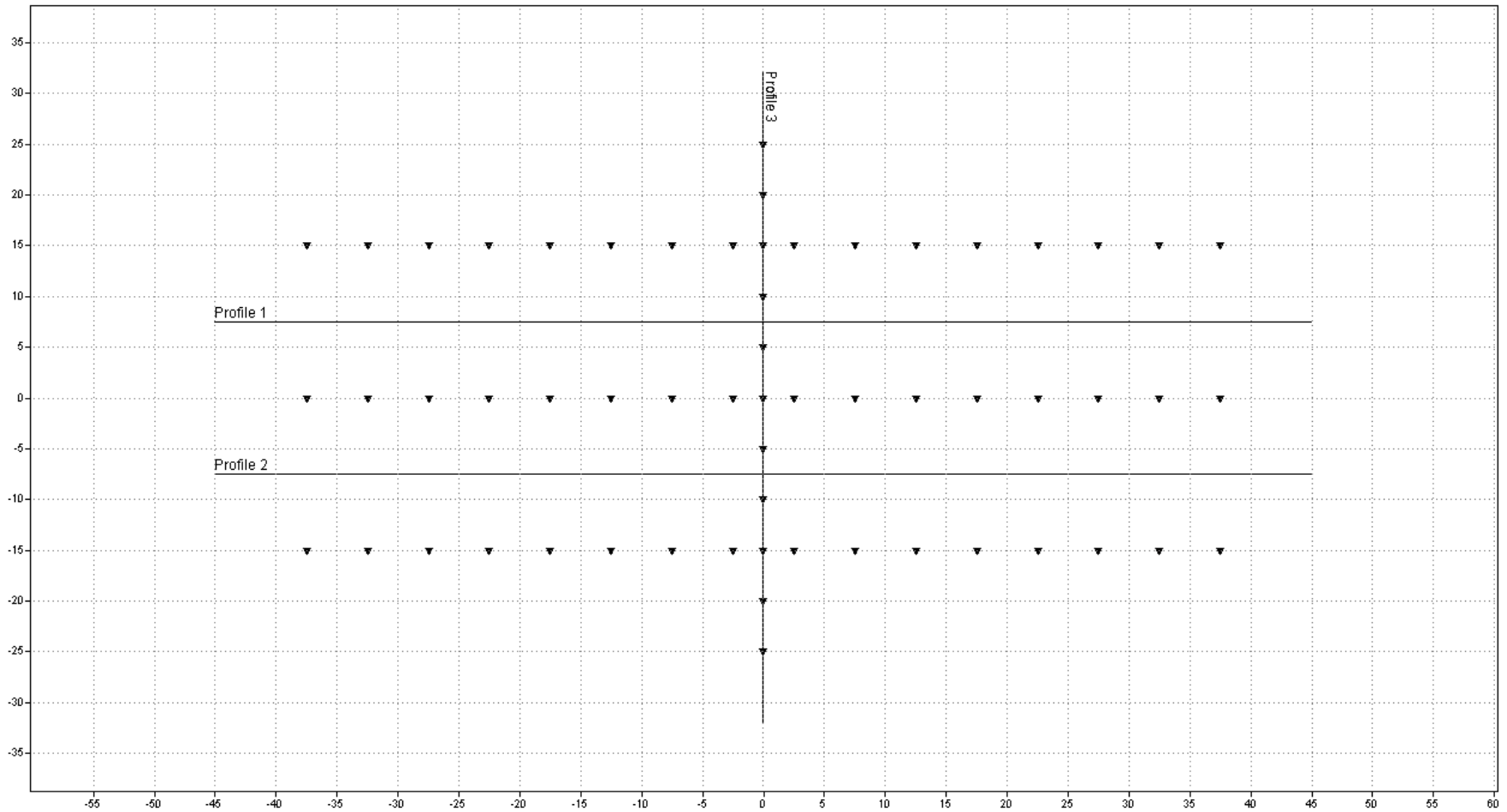
3D Phase XY Map at 30s

3D Forward Model Data

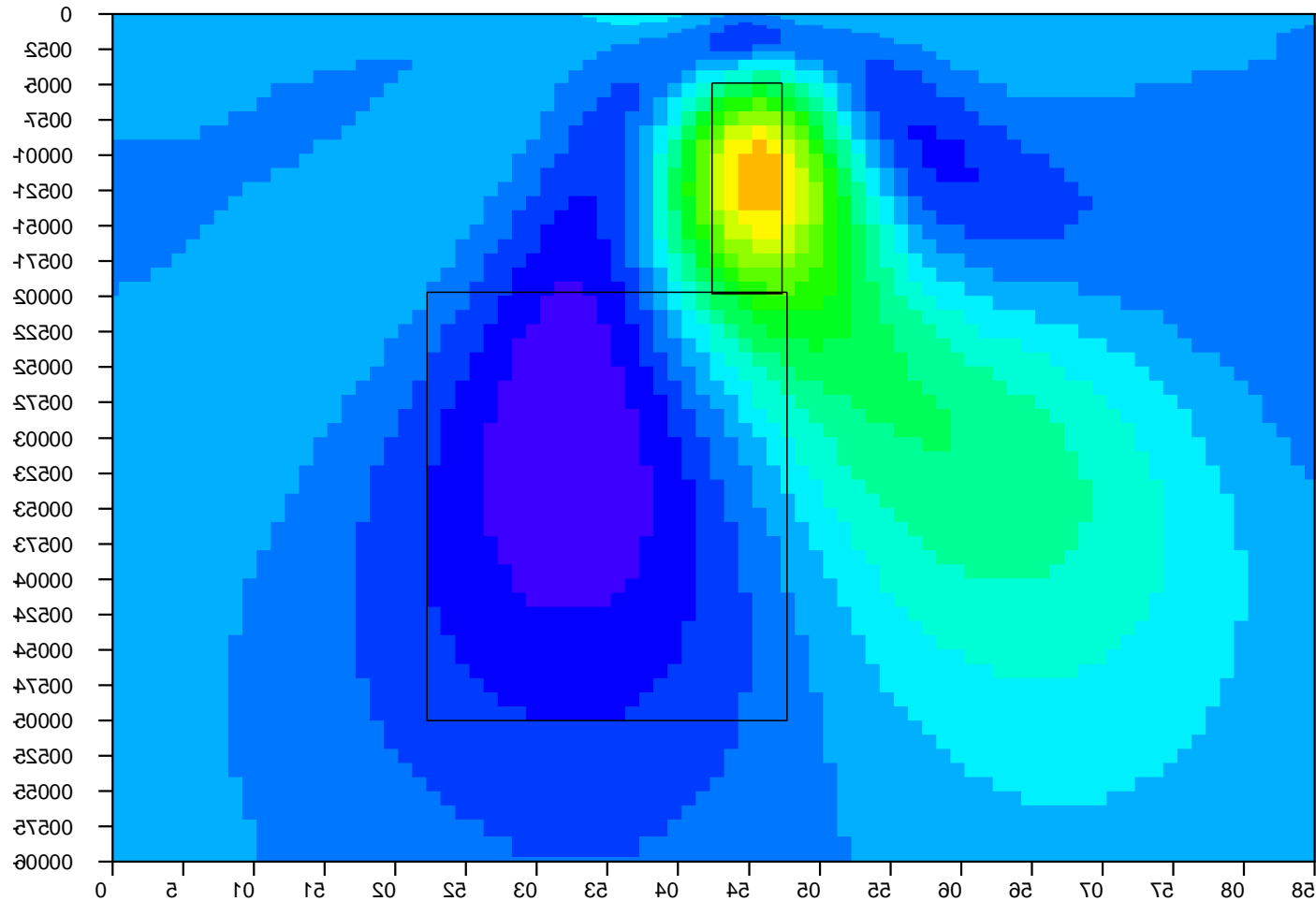


3D Phase YX Map at 30s

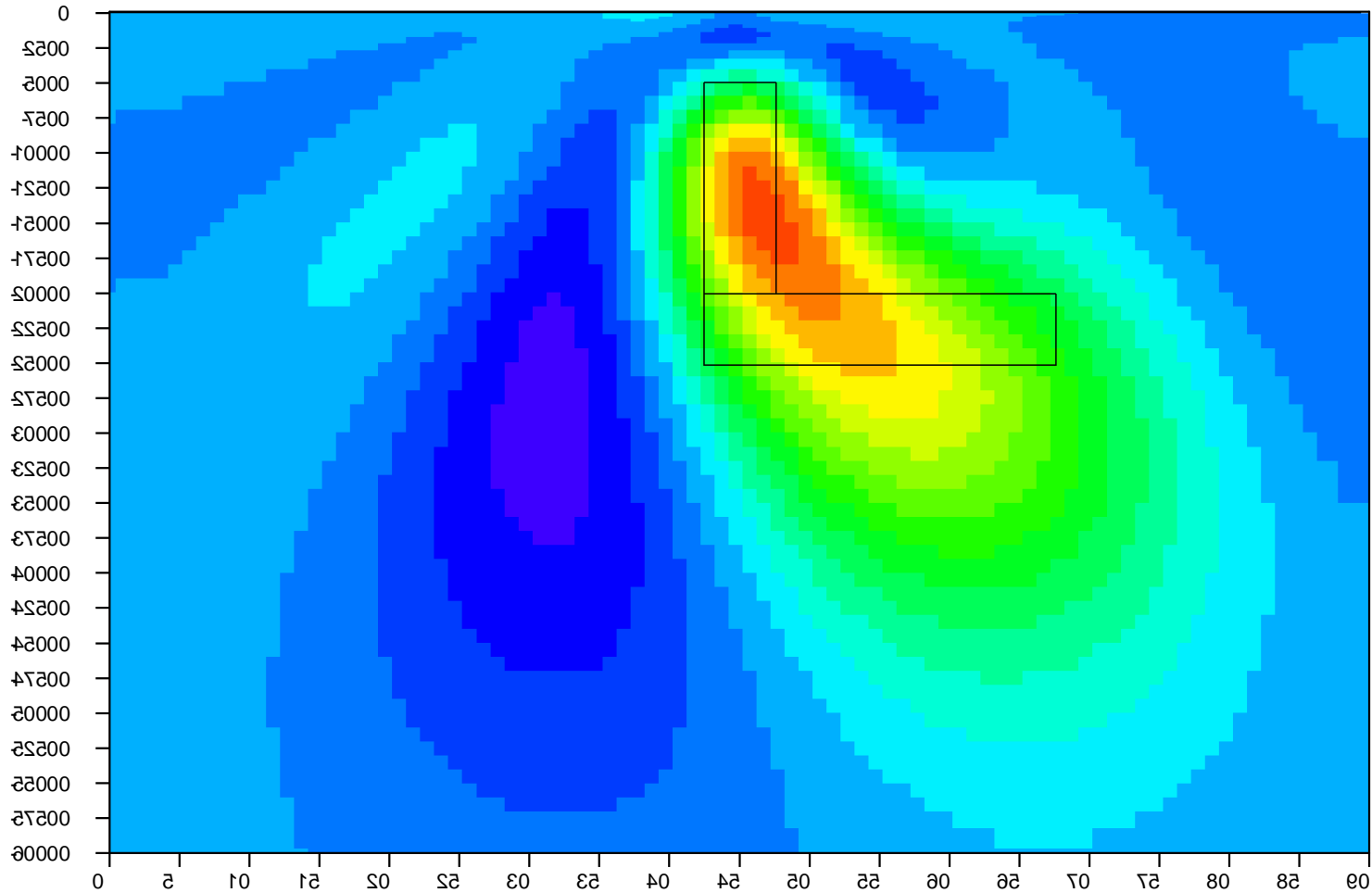
3D Forward Model Data



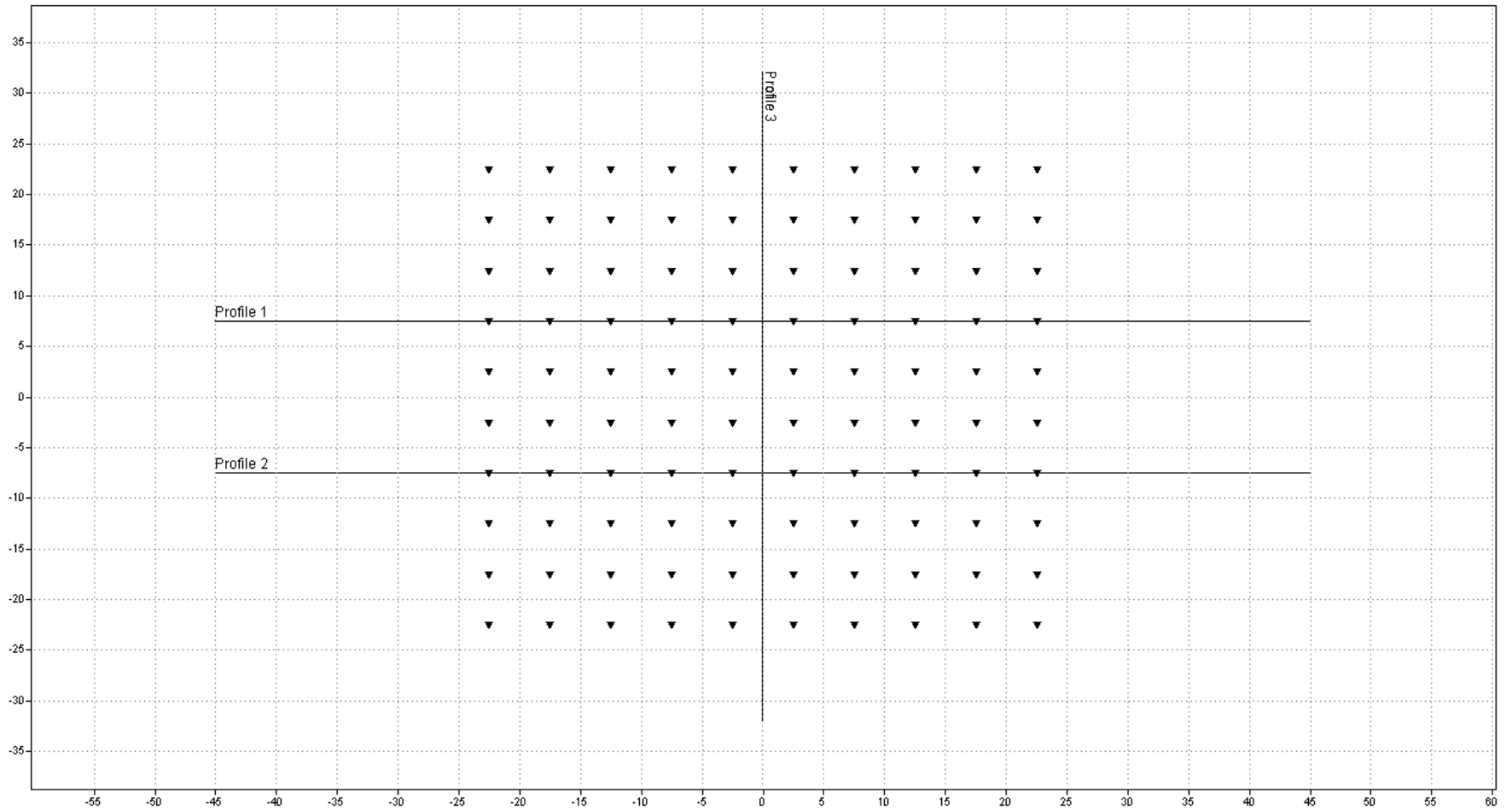
Profile 1: Inversion Result from FWD data



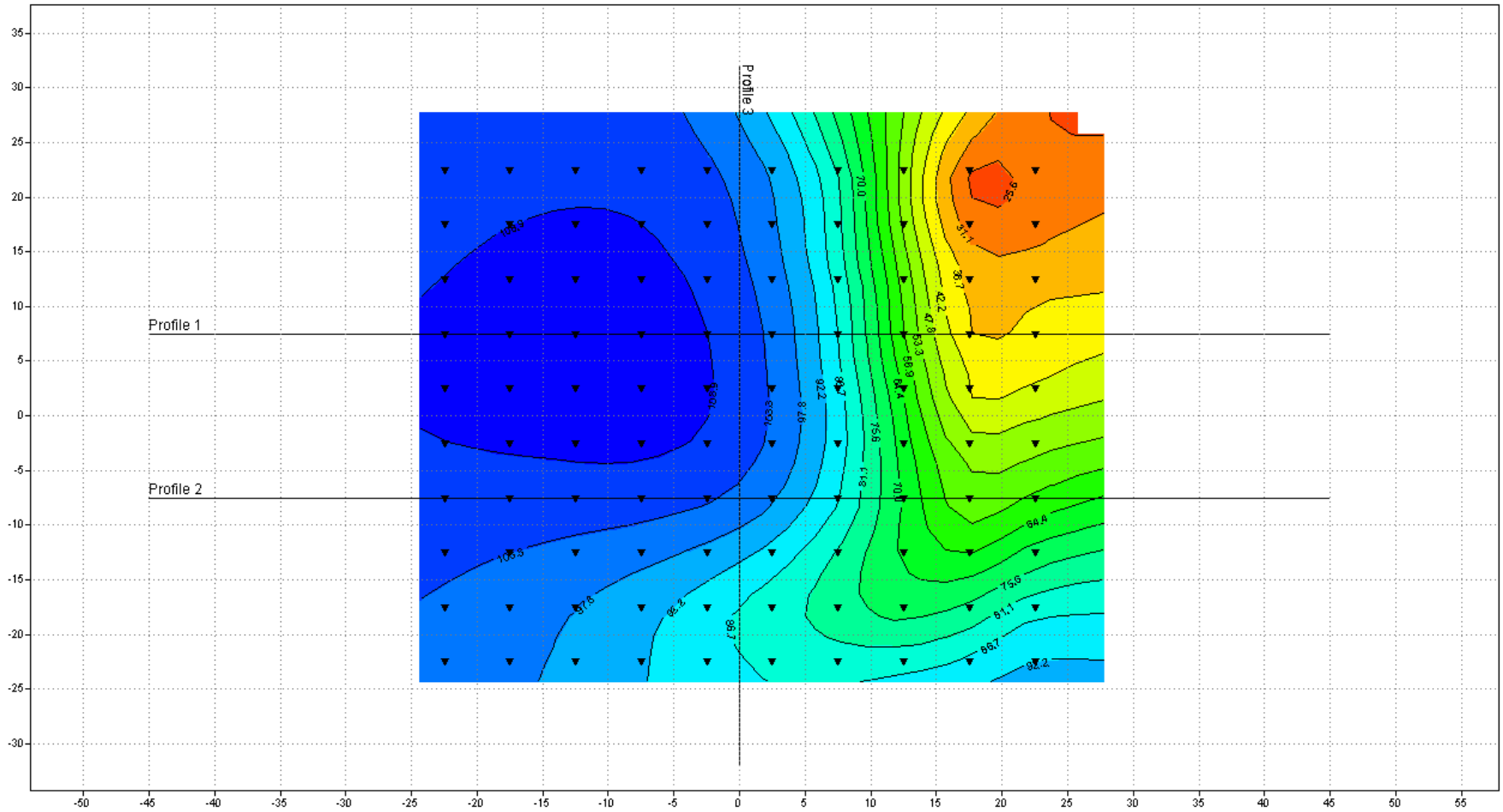
Profile 2: Inversion Result from FWD data



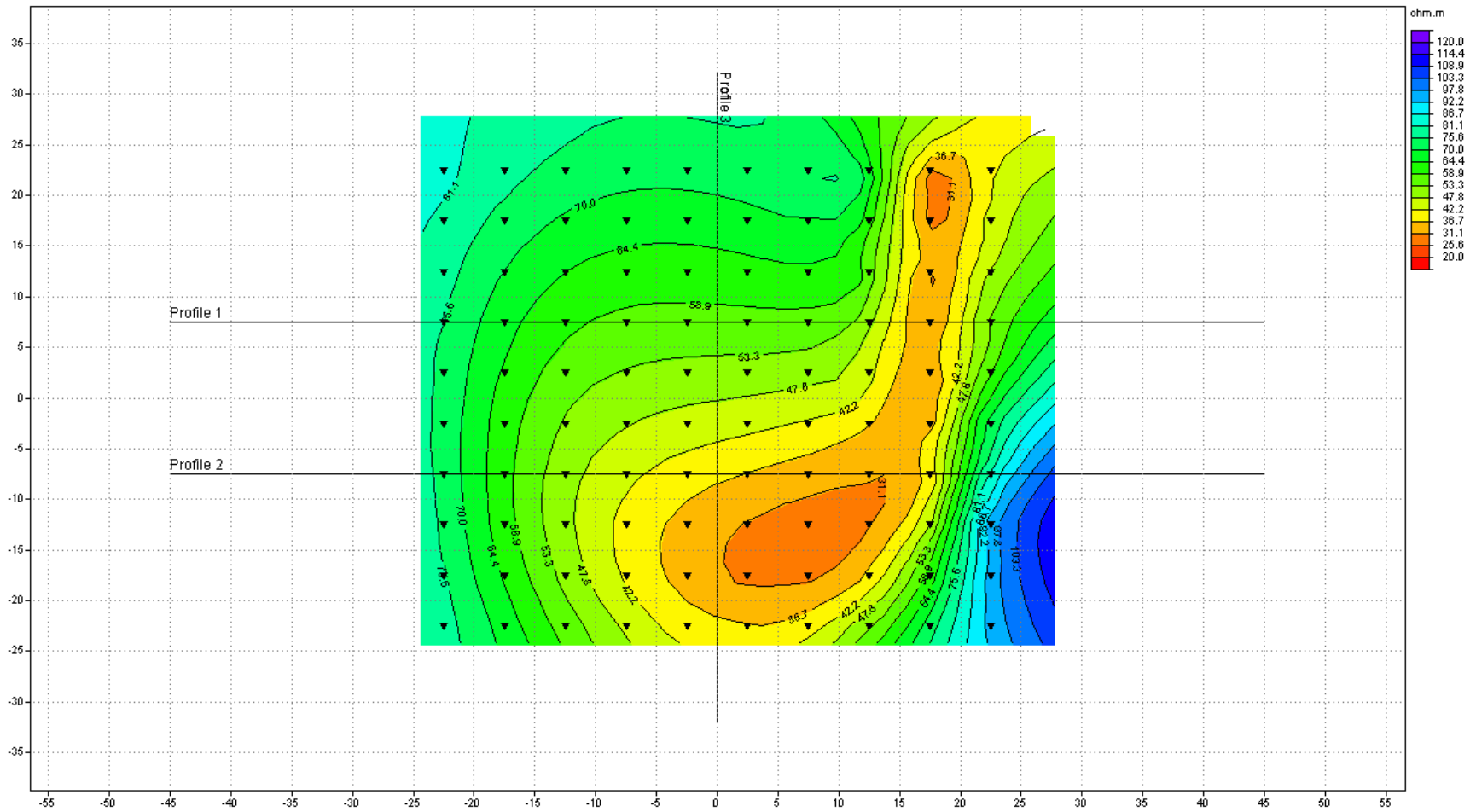
Secret Data Site Plan



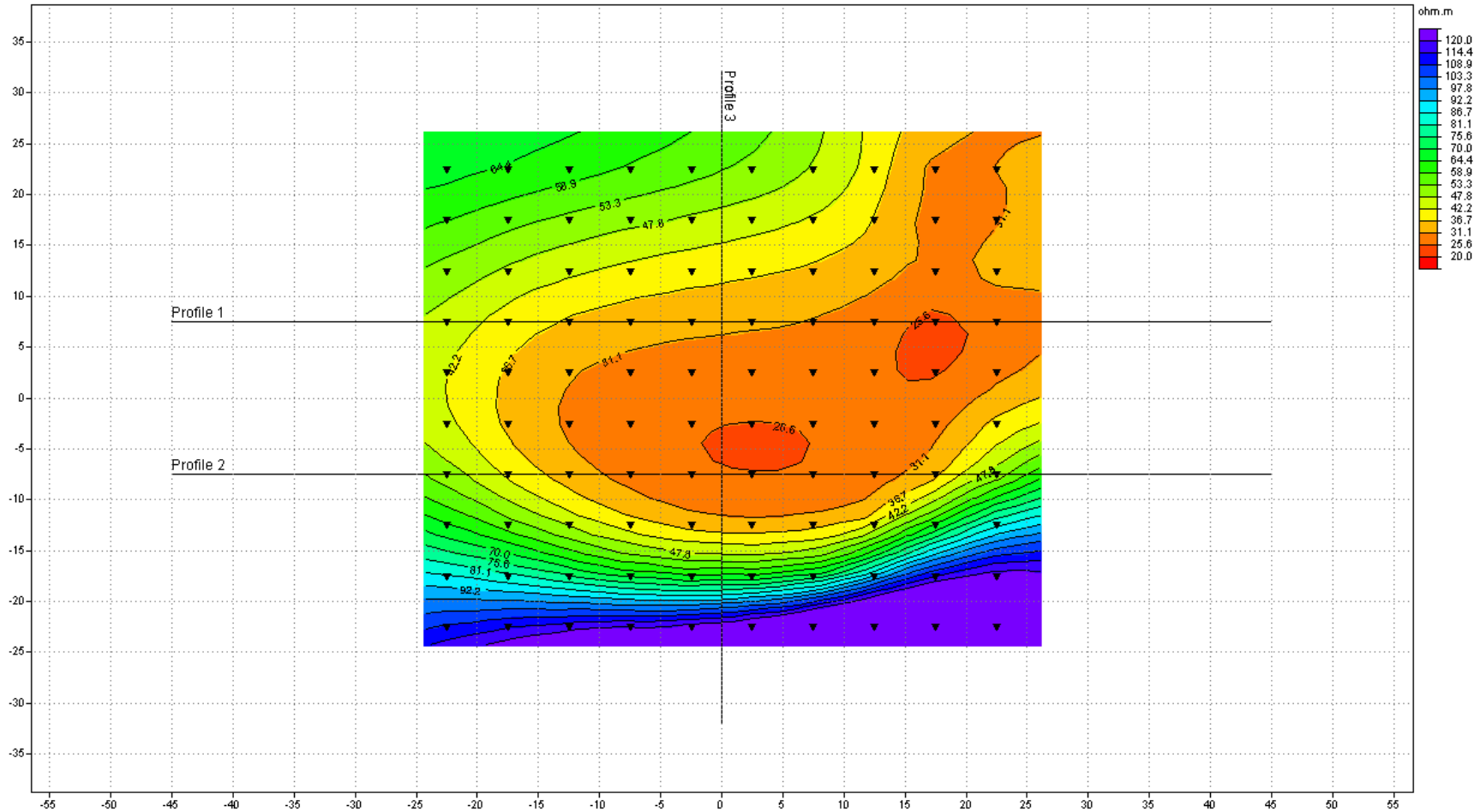
App Res XY at 10 sec



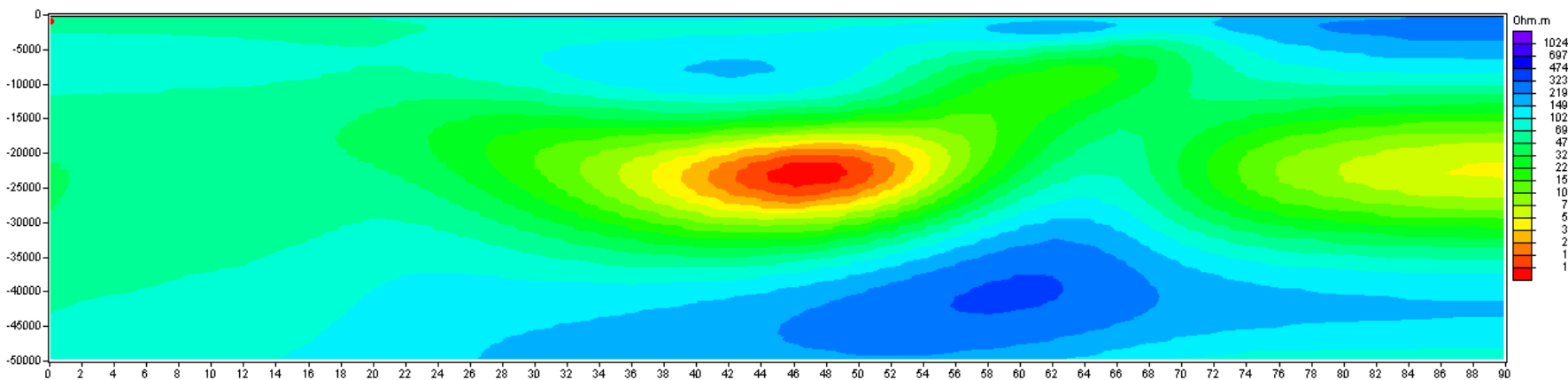
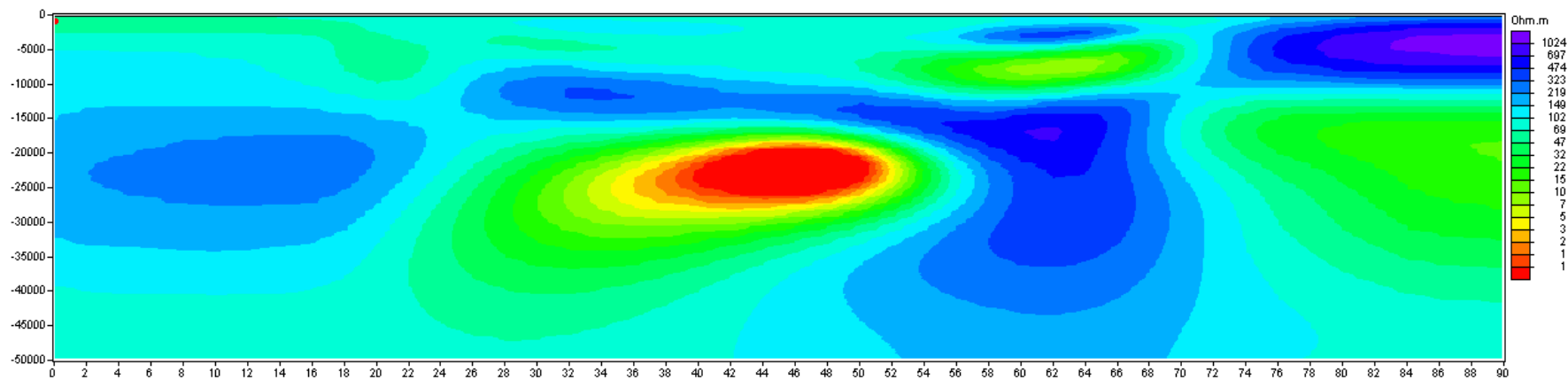
App Res XY at 100 sec



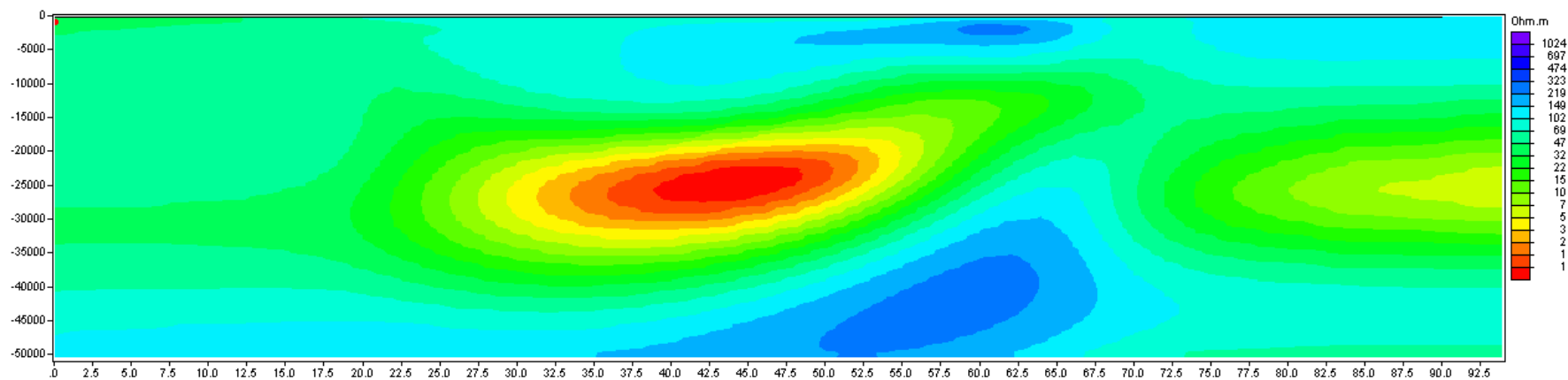
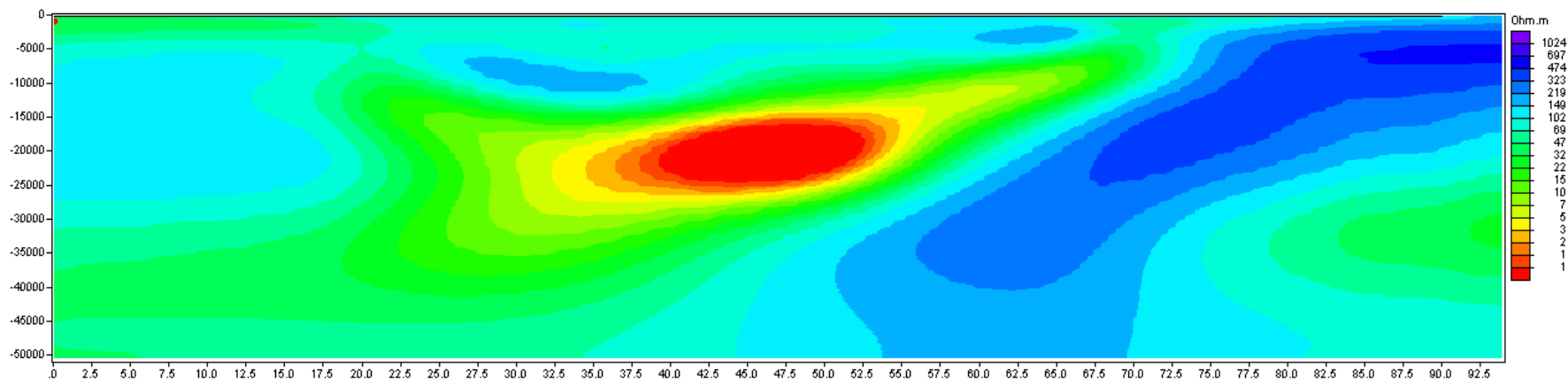
App Res XY at 1000 sec



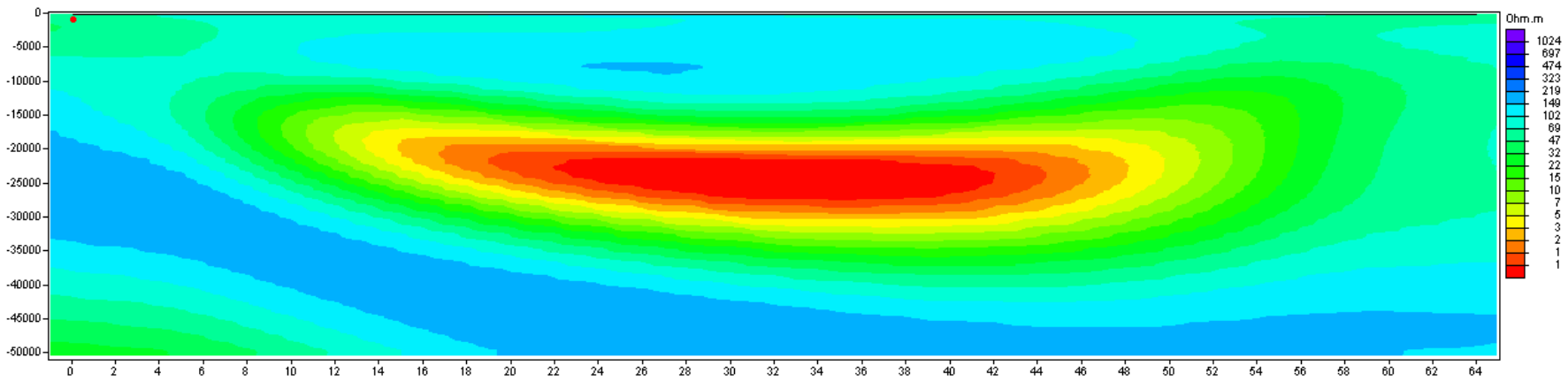
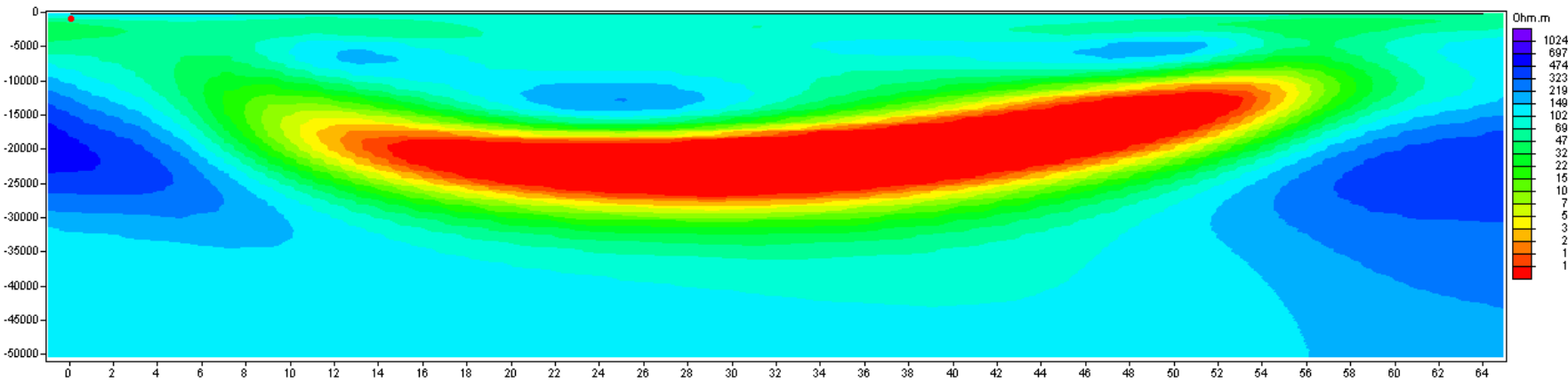
Secret Data: Profile 1



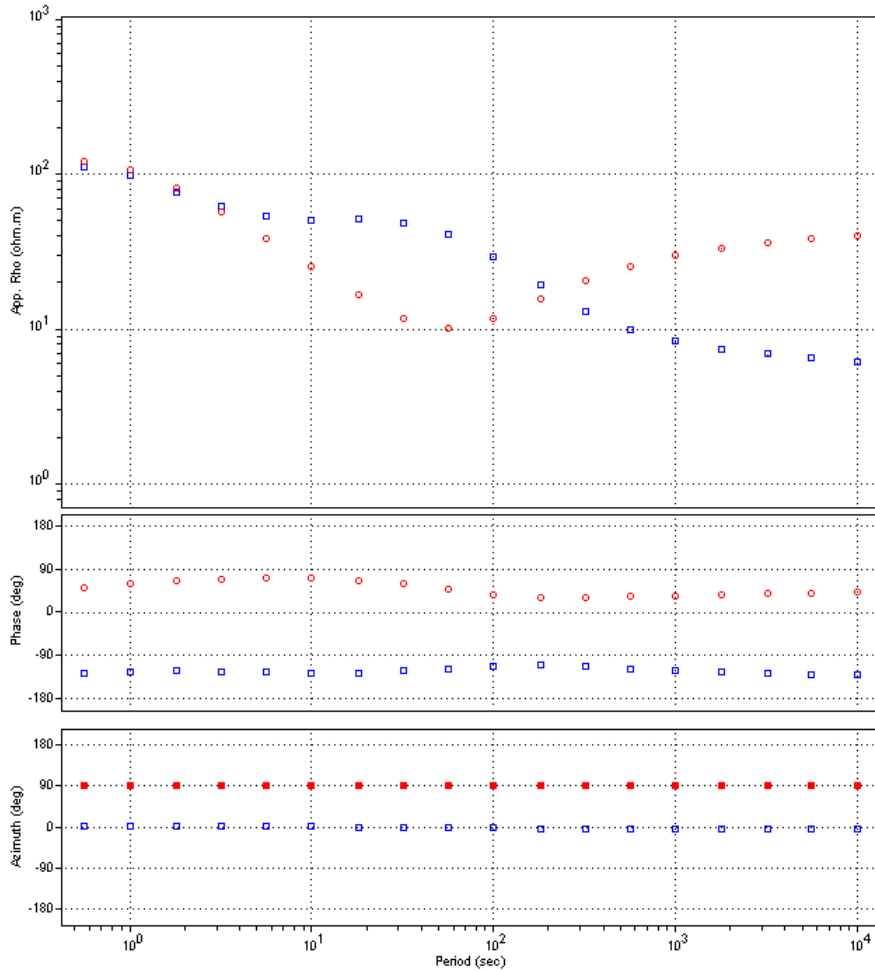
Secret Data: Profile 2



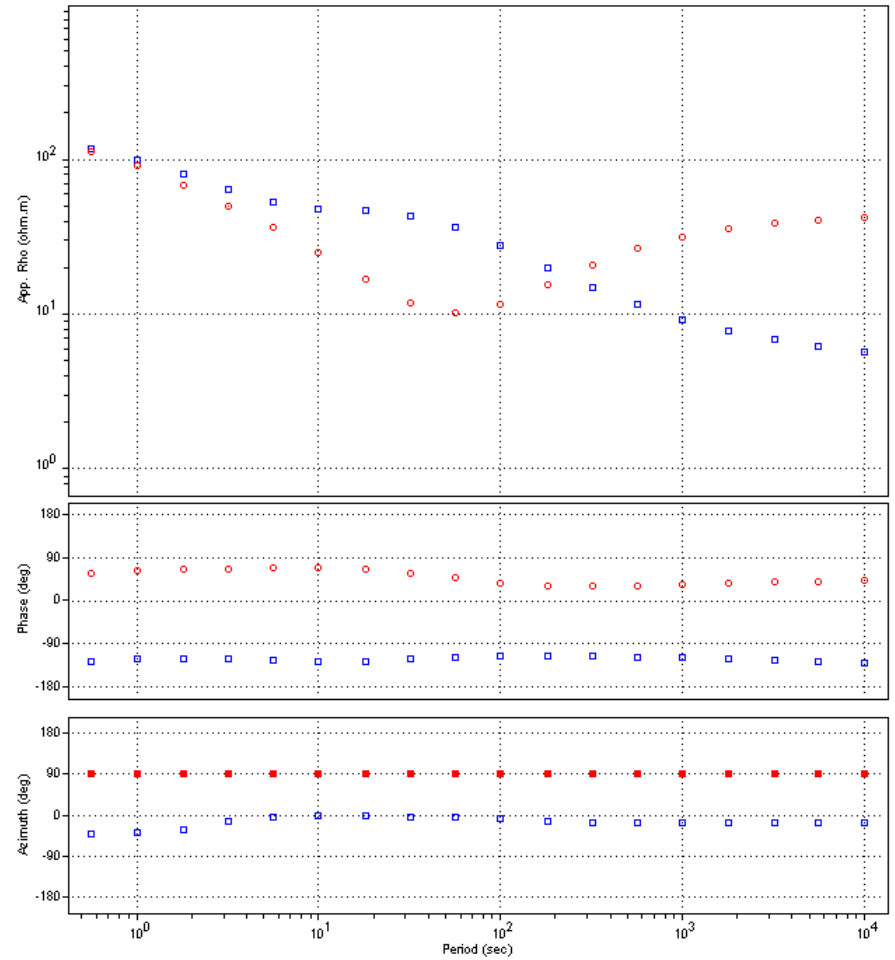
Secret Data: Profile 3



Data Fits: A09



Observed



Predicted

Data Fits: Site J07

