# Joint inversion of GPR and ER Data using the Adjoint Method

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#### INTRODUCTION

Ground penetrating radar (GPR) and electrical resistivity (ER) data are sensitive in complementary ways to electrical conductivity: GPR is sensitive to conductivity through reflection and attenuation in high frequencies while ER is directly sensitive to conductivity in the DC frequency. We have developed a joint inversion algorithm that takes into account the GPR and ER sensitivities of the conductivity at each step of a non-linear adjoint method inversion. We formulate our imaging algorithm by finding parameters  $\varepsilon_*$ ,  $\sigma_*$  that satisfy,

$$\{\varepsilon_*, \sigma_*\} = \arg\min E(\varepsilon, \sigma)$$

$$E = E_w(\varepsilon, \sigma; d_w^o) + E_{dc}(\sigma; d_{dc}^o).$$
(1)

Both  $E_w$  and  $E_{dc}$  are the sum squared errors of the synthetic and observed data. We optimize (1) using gradient descent.

#### **GPR INVERSION**

The physics of the GPR experiment are given by the time dependent Maxwell's equations,

$$\begin{pmatrix}
\mu_{o} & 0 & 0 \\
0 & \mu_{o} & 0 \\
0 & 0 & \varepsilon
\end{pmatrix}
\begin{pmatrix}
H_{z} \\
-\dot{H}_{x} \\
\dot{E}_{y}
\end{pmatrix} = \begin{pmatrix}
0 & 0 & \partial_{x} \\
0 & 0 & \partial_{z} \\
\partial_{x} & \partial_{z} & 0
\end{pmatrix}
\begin{pmatrix}
H_{z} \\
-H_{x} \\
E_{y}
\end{pmatrix} -$$

$$\sigma \begin{pmatrix}
0 \\
0 \\
E_{y}
\end{pmatrix} + \begin{pmatrix}
0 \\
0 \\
-J_{y}
\end{pmatrix}.$$
(2)

The discretized version of equation (2),

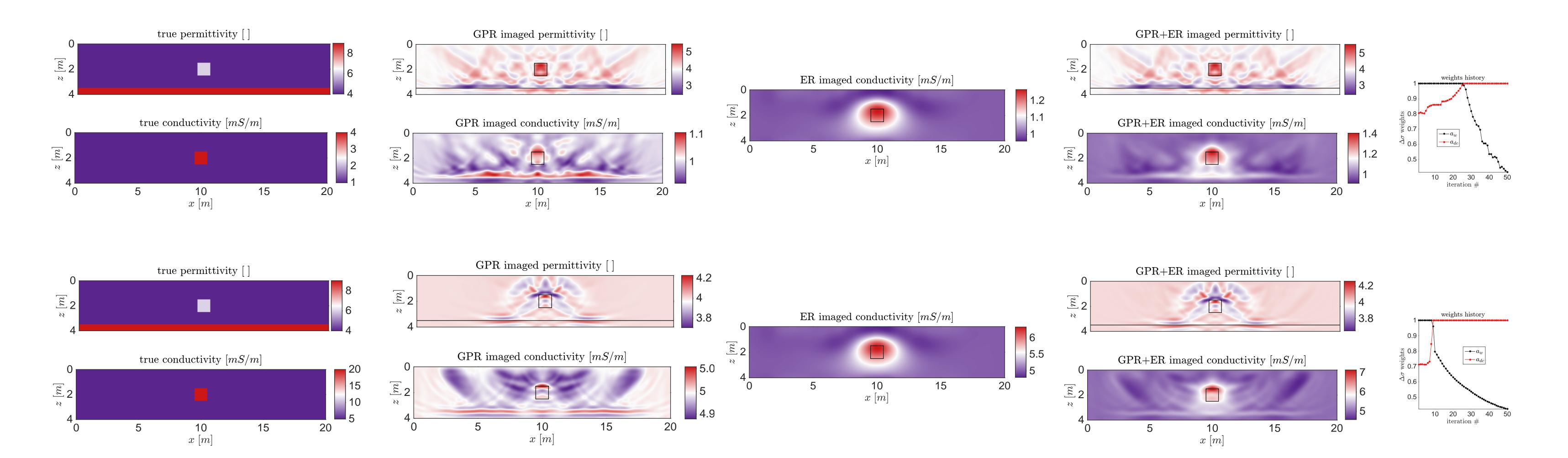
$$u = L_w s_w, d_w = M_w u$$
 (3)

where  $L_w$  is the discretized differential operator of (2), u is the electric field y component,  $s_w$  is the source term,  $M_w$  is the measuring operator and  $d_w = M_w u$  is the data. We compute the gradients of  $E_w$  with respect to  $\sigma$  and  $\varepsilon$  by using the adjoint wave field  $a_w$ ,

$$a_w = L_w e_w(-t),$$
  $g_{w,\sigma} = -\sum_t u(-t) \odot a_w(t) \cdot \Delta t,$   $g_{w,\varepsilon} = -\sum_t \dot{u}(-t) \odot a_w(t) \cdot \Delta t,$  (4)

where  $e_w=d_w-d_w^o$  is the error of the synthetic vs the observed data.

## EXAMPLES



**GPR experiments**: 5 sources equally spaced with receivers all along the transect spaced  $\lambda_o/4$  between each other and  $\lambda_o$  away from the source. The source was a Ricker wavelet with  $f_o$  =250MHz. **ER experiments**: electrodes placed 1m apart with all possible dipole-dipole and Wenner arrays. **Inversion**: All initial models were taken with a homogeneous background. Aside from removing high amplitudes near the sources and smoothing the gradients, no further regularization was made.

#### **ER INVERSION**

The physics of the ER experiment are given by the steady state Maxwell's equations where Ohm's law holds,

$$-\nabla \cdot \sigma \nabla \varphi = \mathbf{i}(\delta(x - s_+) - \delta(x - s_-)), \qquad (5)$$

where  $\varphi$  is the electric potential,  $\mathbf{i}$  is the current intensity and  $s_{\pm}$  is the source-sink position. We write the discretized version of (5) as,

$$L_{dc}\varphi = s_{dc},$$

$$d_{dc} = M_{dc}\varphi,$$
(6)

where  $L_{dc}$  is the discretized differential operator of (5),  $\varphi$  is the electric potential,  $s_{dc}$  is the source term,  $M_{dc}$  is the measuring operator that computes observed voltages and  $d_{dc}$  is the data. We compute the gradient of  $E_{dc}$  with respect to  $\sigma$  by using the adjoint potential field  $a_{dc}$ ,

$$L_{dc}^{\top} a_{dc} = M_{dc}^{\top} e_{dc},$$

$$g_{dc} = S_{dc} a_{dc},$$

$$(7)$$

where  $S_{dc} = -((\nabla_{\sigma} L_{dc})\varphi)^{\top}$  and  $e_{dc} = d_{dc} - d_{dc}^{o}$  is the error of the synthetic vs the observed data.

## JOINT INVERSION

Let the updates for  $\sigma$  be,

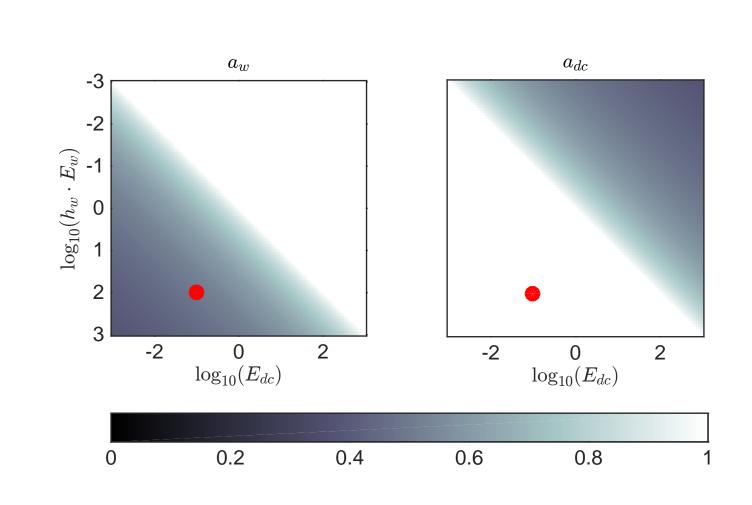
$$\Delta \sigma_{w} = -\frac{1}{n_{w}} \sum_{i=1}^{n_{w}} \alpha_{\sigma} g_{w,\sigma},$$

$$\Delta \sigma_{dc} = -\frac{1}{n_{dc}} \sum_{i=1}^{n_{dc}} \alpha_{dc} g_{dc},$$
(8)

where  $n_w$  and  $n_{dc}$  denote the number of GPR and ER experiments respectively. We define the joint update,

$$\Delta \sigma = a_w \, \Delta \sigma_w + a_{dc} \, \Delta \sigma_{dc}, \tag{9}$$

where  $a_w$  and  $a_{dc}$  are scalar weights. The choice for weights  $a_w$  and  $a_{dc}$  is made with the paradigm of letting both updates  $\Delta\sigma_w$  and  $\Delta\sigma_{dc}$  always contribute to  $\Delta\sigma$  in proportion to their objective function value at a given iteration.



#### CONCLUSIONS

- Our joint inversion algorithm of GPR and ER data allows for both data sets to work cooperatively and regularize each other while honoring the physics.
- Our algorithm performs well on one sided acquisition and makes no assumption of the subsurface geometry.
- Both when the media is strongly and weakly conductive, our algorithm enhances the spatial and magnitude resolution of  $\sigma$ .
- Possible enhancements to the algorithm are a stepped-frequency GPR inversion as in Adam Mangel's poster NS31B-0753.

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### References

- [1] D. Domenzain, J. Bradford & J. Mead, *Joint inversion* of *GPR and ER data*, SEG Technical Program Expanded Abstracts 2018, 4763-4767.
- [2] https://github.com/diegozain/gerjoii