## **Counteracting sensitivity accumulation near source and receiver locations** in 3-D inversion of controlled-source electromagnetic data

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## 1. Introduction & Objectives

- $\Rightarrow$  mitigate the influence of strong sensitivities • synthetic controlled-source electromagnetic (CSEM) inversion study with a single transmitter and a distant inclined conductive anomaly near source and receivers
- 3-D CSEM finite-element approximations [1,2] combined with non-linear con- $\Rightarrow$  guide the inversion to a reliable model via the jugate gradient (NLCG) inversion [3] implemented in inversion framework [4] model regularisation term

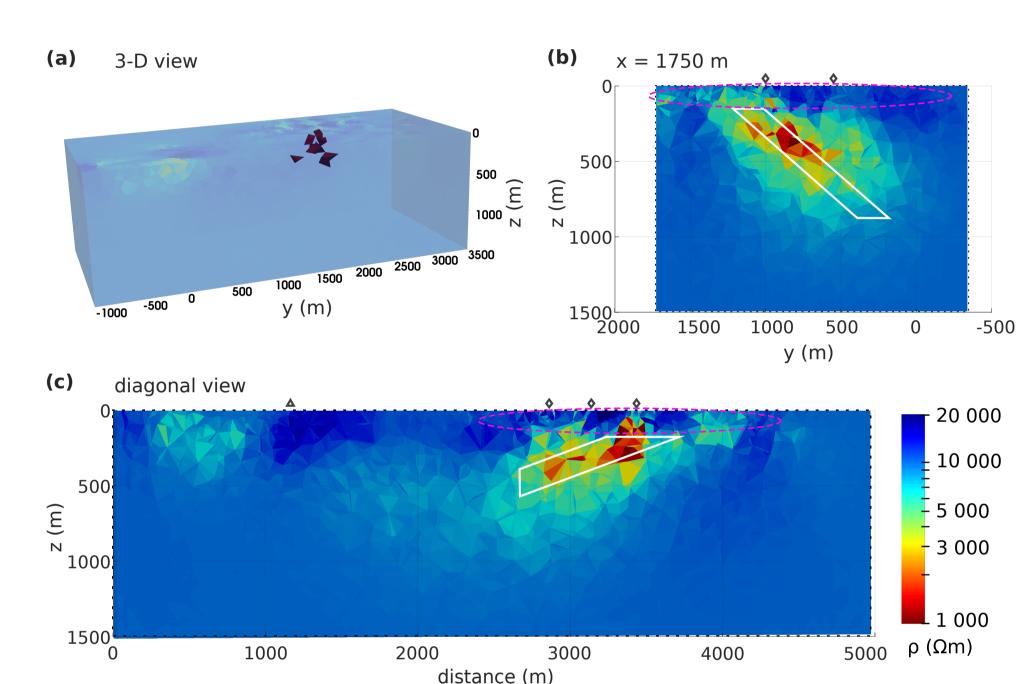
#### How can we prevent resistivity artefacts close to the source location?



# (a) 3-D view 1500 1000 500 0 -5

Zxy & Zyx input data

#### Full Z input data



**Figure 3:** (a) 3-D view and (b,c) vertical sections through the resistivity model obtained with preconditioned NLCG inversion and Zxy & Zyx input data.  $\rho$ -threshold in (a): 1 000  $\Omega$ m.

**Figure 4:** (a) 3-D view and (b,c) vertical sections through **Figure 5:** (a) 3-D view and (b,c) vertical sections through the resistivity model obtained with preconditioned NLCG inverthe resistivity model obtained with preconditioned NLCG inversion and all Z components as input data. sion and distance-to-source weights  $(r_d = 750 m)$ .

## Analysis

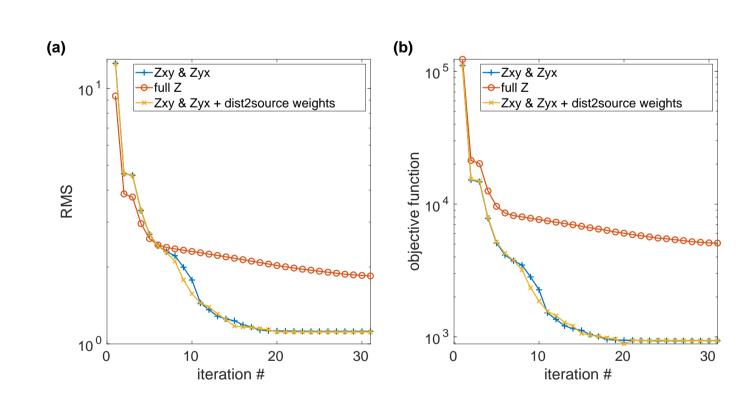


Figure 6: Convergence analysis: (a) RMS and (b) objective function versus iteration number.

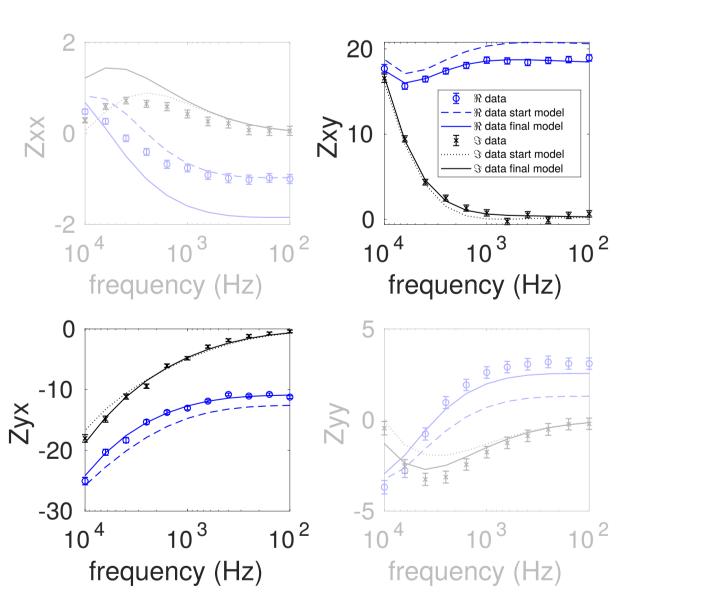


Figure 7: Data-fit S1 with distance-to-source weights.

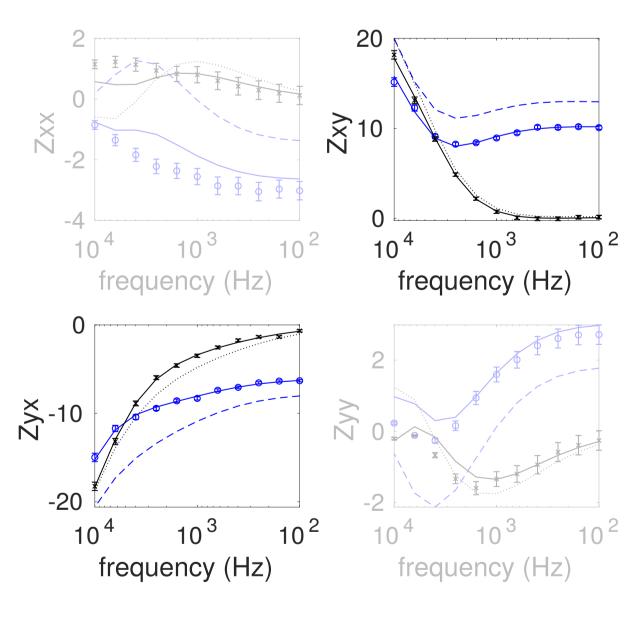
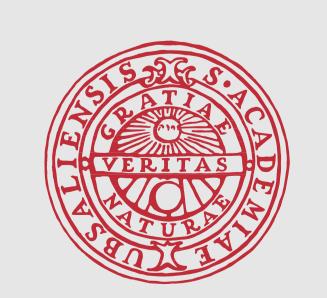


Figure 8: Data-fit S14 with distance-to-source weights.



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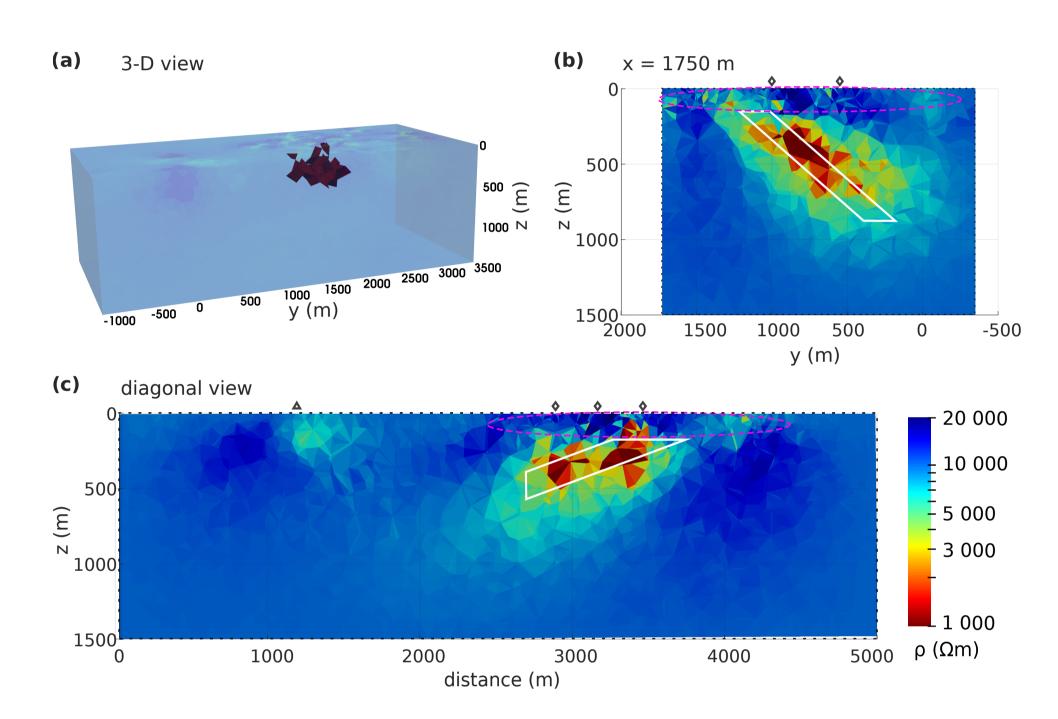
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known

(subscripts 1 & 2)

#### **3. Results**

#### Zxy & Zyx input data with distance-to-source weights



#### Take-aways

- $\Rightarrow$  It is difficult to remove the source signature in the resistivity model, but possible to reduce artefacts by increasing the smoothing in the source region.
- $\Rightarrow$  Diagonal components of the impedance tensor add information, but are more difficult to fit than off-diagonal components.
- $\Rightarrow$  Do not place the transmitter above known resistivity anomalies!

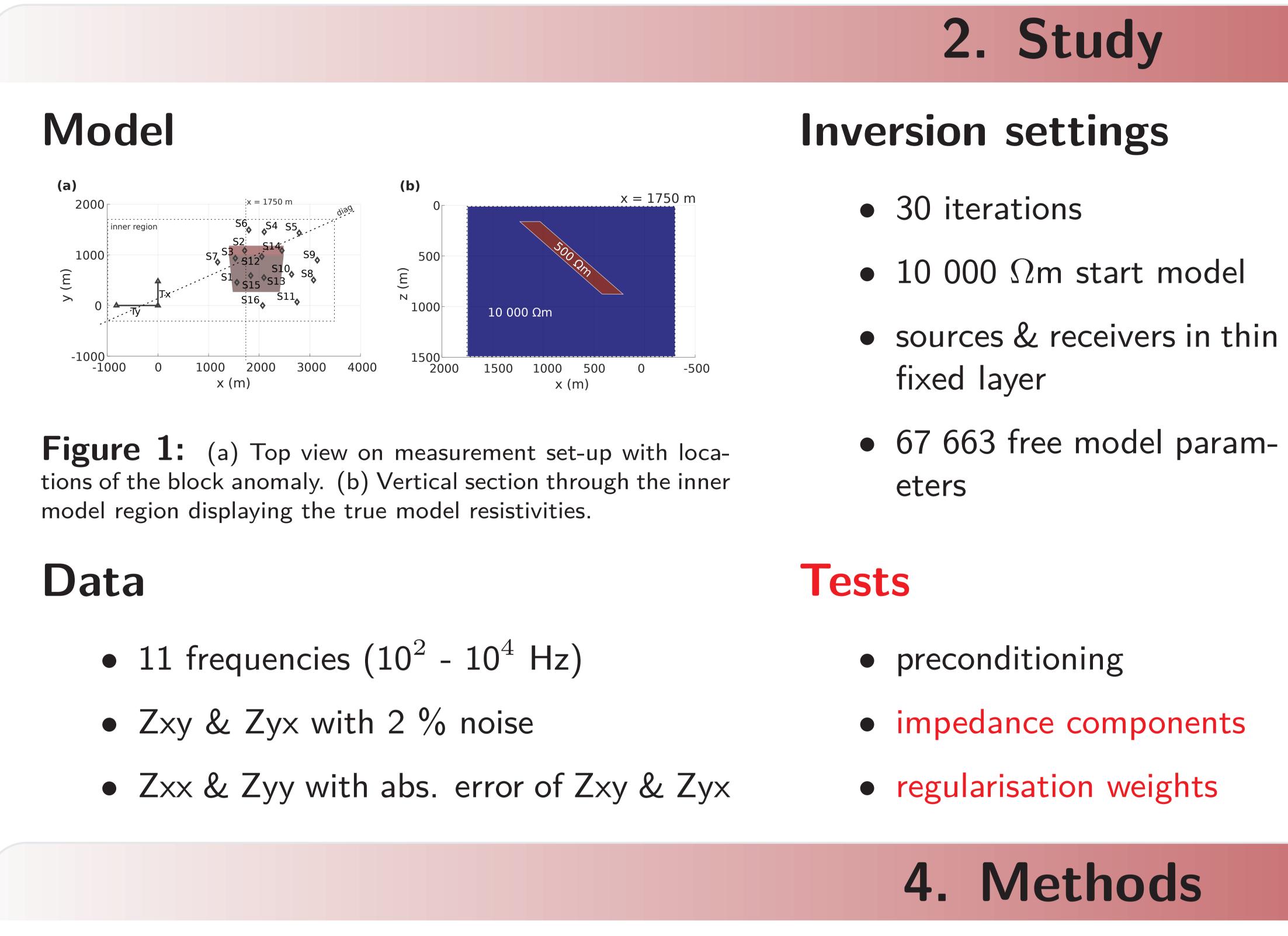
#### **CSEM Impedance Tensor Z**

•	"Controlled-source tensor magnetotel- lurics" [9,10]
•	impedance tensor <b>Z</b> defined via hori- zontal electric and magnetic fields
•	two independent source polarisations (subscripts $1 \ \ell \ 2$ )

• source current does not need to be

$Z_{xx}$	_	$\frac{E_{x1}H_{y2} - E_{x2}H_{y1}}{H_{x1}H_{y2} - H_{x2}H_{y1}},$
$Z_{yx}$	—	$\frac{E_{y1}H_{y2} - E_{y2}H_{y1}}{H_{x1}H_{y2} - H_{x2}H_{y1}},$

7. —	$E_{x2}H_{x1} - E_{x1}H_{x2}$
$Z_{xy} =$	$\frac{E_{x2}H_{x1} - E_{x1}H_{x2}}{H_{x1}H_{y2} - H_{x2}H_{y1}},$
Z —	$E_{y2}H_{x1} - E_{y1}H_{x2}$
$Z_{yy} =$	$\overline{H_{x1}H_{y2} - H_{x2}H_{y1}}.$

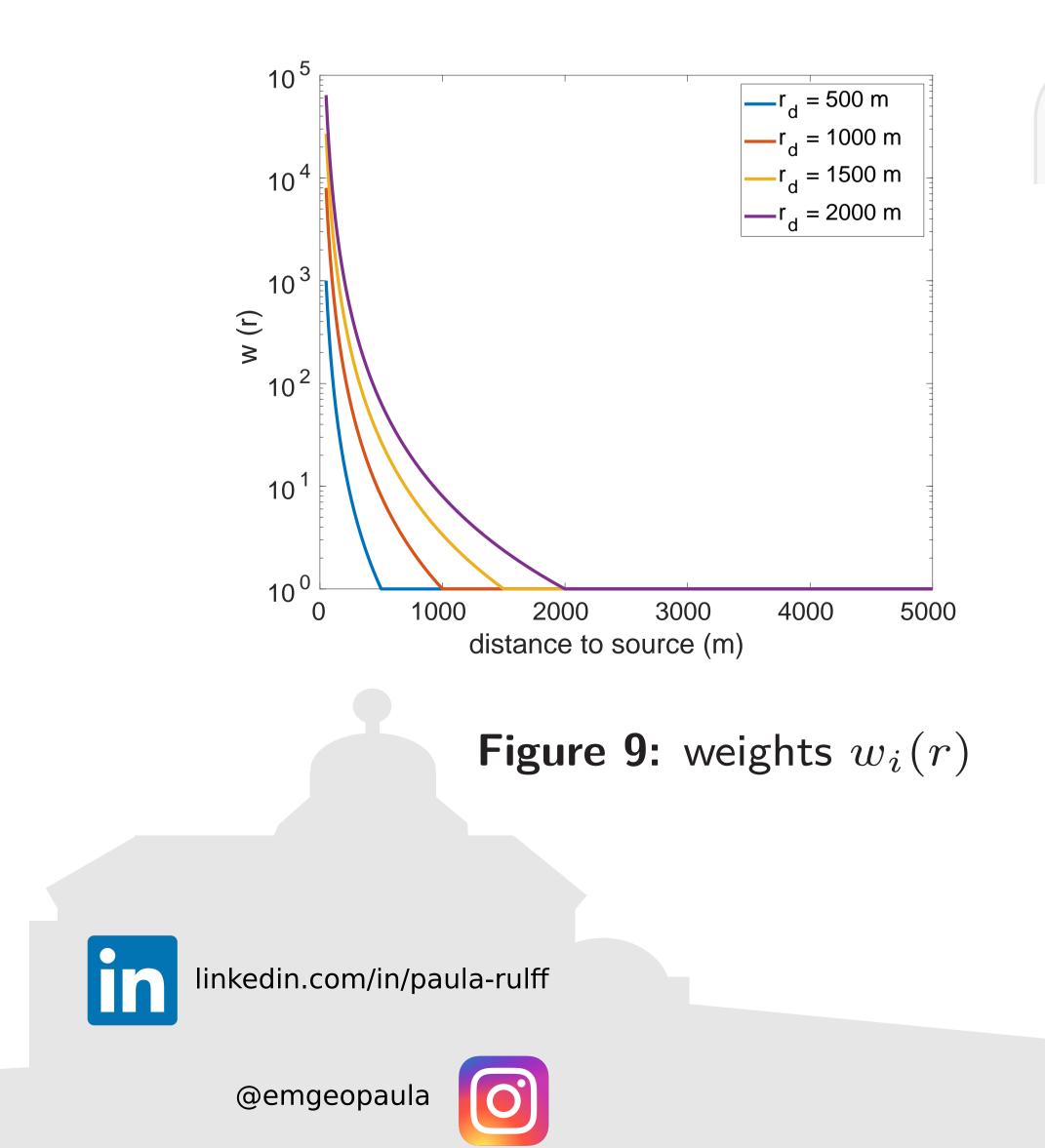


### Distance-to-source (r) weights

- weights  $w_i(r)$  in the model regularisation term
- increase the smoothing in model regions, where sensitivities are high, but no variations in subsurface resistivity are expected (at source location)

$$w_i(r) = \begin{cases} \left(\frac{r_i}{r_d}\right)^{-3} & \text{for} \quad r_i \leq r_d; \\ 1 & \text{for} \quad r_i > r_d. \end{cases}$$

• for first-order difference regularisation [7] and gradient approximation via the weighted sum of first differences [8]



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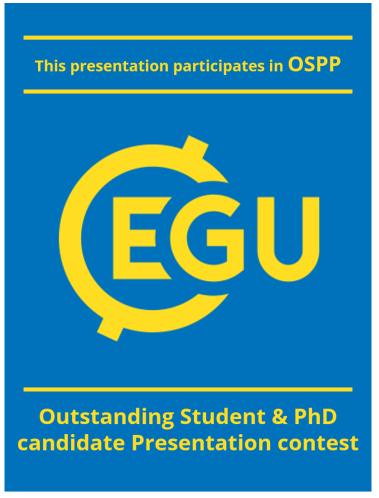
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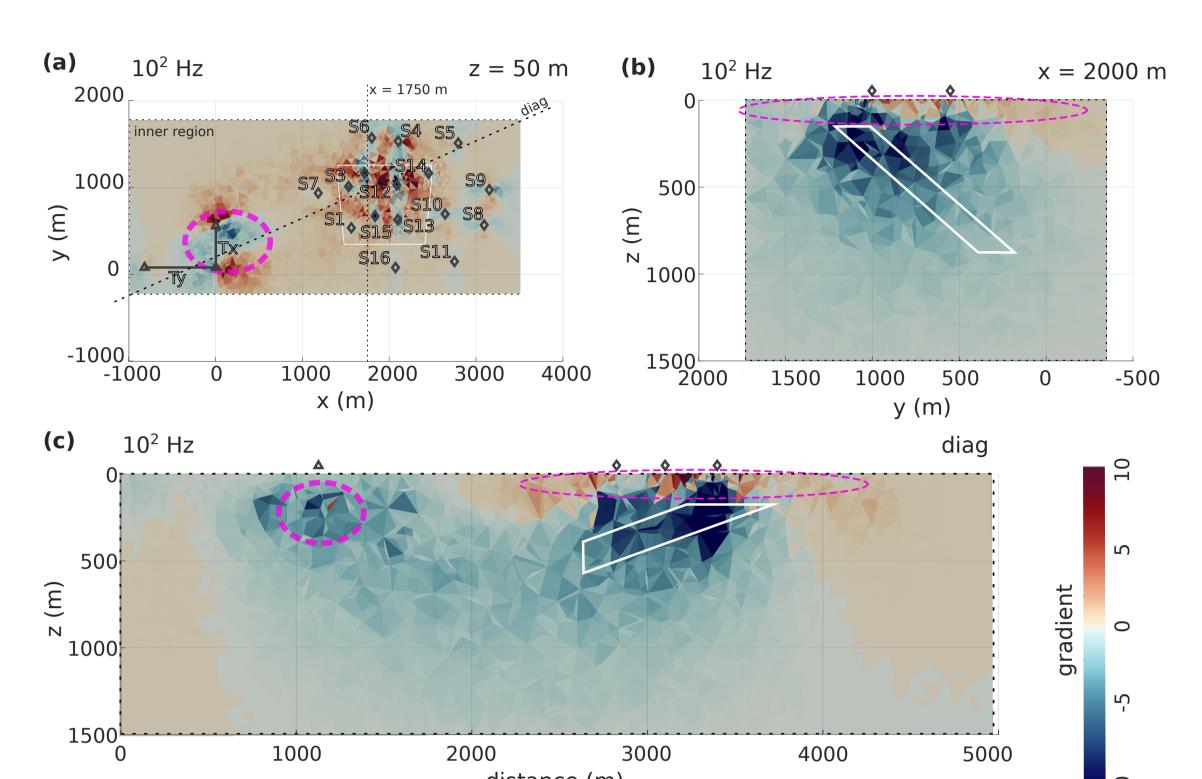
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#### Data gradients



**Figure 2:** Sections through the inner model region displaying gradients at  $10^2$  Hz of Zxy & Zyx data. The location of the conductive anomaly is outlined in white. Areas of high sensitivities near source and receivers are outlined in pink. Mesh: tetgen [5]

#### Preconditioning

- approximates the Hessian of the objective function  $\Phi$
- distributes the model update  $\mathbf{m}_{k+1}$  more uniformly in the domain due to better search directions  $\mathbf{u}_k$
- Preconditioner suggested in [6] with step size  $\alpha_k$ :

$$\mathbf{M}_{k+1} = \mathbf{M}_k + \frac{\nabla \Phi(\mathbf{m}_k) \nabla \Phi(\mathbf{m}_k)^T}{\nabla \Phi(\mathbf{m}_k)^T \mathbf{u}_k} + \frac{\mathbf{v}_k \mathbf{v}_k^T}{\alpha_k \mathbf{v}_k^T \mathbf{u}_k}$$

- diagonal preconditioner updated at each iteration k
- $\mathbf{M}_0$  is set to the identity matrix,  $\mathbf{v}_k = \nabla \Phi(\mathbf{m}_{k+1}) \nabla \Phi(\mathbf{m}_k)$

#### References

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(cc)