Comparison of different regularization schemes for the 1D laterally constrained inversion of seismic surface wave data.

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Context

At the University of Potsdam, we recently have developed 1D (Guillemoteau *et al*, 2016) and 3D (Guillemoteau *et al*, 2017, 2019) inversion procedures for loop-loop electromagnetic induction (EMI) data.

Our 1D non-linear inversion algorithm is a modified version of the OCCAM procedure described in Aster, *et al*, (2005), which incorporates prior information on data uncertainties and inverts the model in the log space to impose positivity constraints:

$$log m^{k+1} = \left[\mathbf{G}^T (log m^k) \mathbf{W}_{\mathbf{d}} \mathbf{G} (log m^k) + \lambda \mathbf{W}_{\mathbf{m}} \right]^{-1} \mathbf{G}^T (log m^k) \mathbf{W}_{\mathbf{d}} d^*$$
$$d^* = d - d(log m^k) + \mathbf{G} (log m^k) log m^k$$

 W_m now (Klose *et al*, in prep) includes smooth (*sm*) (Auken *et al*, 2005, Socco *et al*, 2009) and sharp (*sh*) (Vignoli *et al*, 2014) constraints, which are applied on both x and z directions (LCI):

$$\Phi_{m}^{sm/sm} = \sum_{i=1}^{N_{s}} \sum_{j=1}^{N_{l}} \left[\nabla_{z} m_{ij} \right]^{2} + w_{LCI} \cdot \left[\nabla_{x} m_{ij} \right]^{2}$$

$$\Phi_{m}^{sm/sh} = \sum_{i=1}^{N_{s}} \sum_{j=1}^{N_{l}} \frac{\left[\nabla_{z} m_{ij} \right]^{2}}{\sqrt{\left[\nabla_{z} m_{ij} \right]^{2} + \epsilon^{2}}} + w_{LCI} \cdot \left[\nabla_{x} m_{ij} \right]^{2}$$

$$\Phi_{m}^{sh/sh} = \sum_{i=1}^{N_{s}} \sum_{j=1}^{N_{l}} \frac{\left[\nabla_{z} m_{ij} \right]^{2}}{\sqrt{\left[\nabla_{z} m_{ij} \right]^{2} + \epsilon^{2}}} + w_{LCI} \cdot \frac{\left[\nabla_{x} m_{ij} \right]^{2}}{\sqrt{\left[\nabla_{x} m_{ij} \right]^{2} + \epsilon^{2}}}$$



Context

Example of LCI inversions of DUALEM-21 data collected above a conductive peat layer:



You can see the most recent application of this algorithm to EMI CMD explorer data in this EGU 2020 presentation!

EGU2020-19011 | Displays | <u>HS8.1.5/SSS6.12</u> Exploration of electromagnetic induction potential to understand groundwater infiltration within the Chalk <u>critical zone</u> Marc Dumont *et al*, Thu, 07 May, 16:15–18:00 | D370

Goal

Our aim is to develop a common inversion platform for different kinds of geophysical data set, for example, EMI and Surface Wave (SW):



In this respect, the purpose of this study is to evaluate how our inversion strategy for EMI data can be applied to the 1D inversion of SW dispersion curves.

The present study shows preliminary results only.

Forward modelling and inversion

The 1D forward modeling code is based on the finite elements approach described in Haney and Tsai, (2020).

Our novel inversion algorithm relies on two decoupled grids: one discretizing the model parameters, the other (denser) for the FE computation.



Forward modelling of SW data

First example of 1D synthetic data set



Smooth and sharp VCI inversions

First tests with noise free data: smooth(z) vs sharp(z)



Hybrid smooth and sharp LCI inversions

LCI: smooth(x)sharp(z) vs sharp(x)sharp(z)



Forward modelling of SW data

Second example of 1D synthetic data set: adding some geological complexity



VCI sharp vs LCI sharp

VCI sharp(z) vs LCI sharp(x) sharp(z)



VCI smooth(z)



LCI smooth(x) smooth(z)



LCI sharp(x) sharp(z)



Synthetic data + model noise+ 2% data random noise



Synthetic data + model noise+ 2% data random noise



Synthetic data + model noise+ 5% data random noise



Synthetic data + model noise+ 5% data random noise



Summary: results obtained with the same inversion setup



Conclusion

- We applied our EMI inversion tools to the 1D inversion of seismic surface wave data
- We implemented both VCI and LCI approaches incorporating smooth and sharp constraints in both directions (x,z)
- From this preliminary synthetic study, the LCI with sharpness constraints in both directions (x,z) shows the best imaging capabilities
- Next step is to invert a real data set!

Thank you for your comments and feedback!

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Data misfit



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