Inverse Probleme in der Geophysik Vorlesung (Vertretung K. Spitzer) TU Bergakademie Freiberg, SS 2020

Teil 12: Abschluss der Veranstaltung

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13. Juli 2020

### Heutige Veranstaltung

#### Abschluss nichtlineare Probleme

- Auflösungsmatritzen für nichtlineare Probleme (pdf)
- Inversion mit mehreren Datentypen (1D-MT JNB und Beleg Teil 2)
- Inversion mit mehreren Modellgrößen (NMR)
- 2D-Geoelektrik-Vorführung (JNB) mit struktureller A-priori-Information
- Parameter-Informationen und Regionskonzept (BERT)
- Joint inversion
  - klassische JI: gleicher Parameter (pg: DC+EM)
  - petrophysikalische JI: Überführung auf klassisch (pg Example)
  - strukturell gekoppelte Inversion (pg 1D MRS)
  - voll gekoppelte (hydrogeophysikalische) Inversion (pg)
- Alternative Inversionsmethoden (Genetische Algorithmen)

## Auflösungsmatritzen nichtlinearer Probleme

Iterativ ( $\mathbf{m}^{k+1} = \mathbf{m}^k + \Delta \mathbf{m}^k$ ) gelöstes inverses Subproblem:

$$\begin{split} (\mathbf{S}^T\mathbf{S} + \lambda^2\mathbf{C}^T\mathbf{C})\Delta\mathbf{m}^k &= \mathbf{S}^T(\mathbf{d} - \mathbf{f}(\mathbf{m})) - \lambda^2\mathbf{C}^T\mathbf{C}(\mathbf{m}^k - \mathbf{m}^0) \\ \mathbf{m}^{k+1} &= \mathbf{m}^k + \mathbf{S}^\dagger(\mathbf{d} - \mathbf{f}(\mathbf{m}^k)) + \mathbf{C}^\dagger\mathbf{C}(\mathbf{m}^k - \mathbf{m}^0) \end{split}$$

Mit

$$\mathbf{d} = \mathbf{f}(\mathbf{m}^{true}) + \mathbf{n} = \mathbf{f}(\mathbf{m}^k) - \mathbf{S}(\mathbf{m}^{true} - \mathbf{m}^k) + \mathbf{n}$$

ergibt sich

$$\mathbf{m}^{est} = \mathbf{m}^k + (\mathbf{S}^{\dagger}\mathbf{S} + \mathbf{C}^{\dagger}\mathbf{C})(\mathbf{m}^{true} - \mathbf{m}^k) + \mathbf{S}^{\dagger}\mathbf{n}$$
 (1)

$$= \mathbf{S}^{\dagger} \mathbf{S} \mathbf{m}^{true} - (\mathbf{S}^{\dagger} \mathbf{S} - \mathbf{C}^{\dagger} \mathbf{C}) \mathbf{m}^{k} + \mathbf{C}^{\dagger} \mathbf{C} \mathbf{m}^{0} + \mathbf{S}^{\dagger} \mathbf{n}$$
 (2)

$$= \mathbf{R}^{M} \mathbf{m}^{true} + (\mathbf{I} - \mathbf{R}^{M}) \mathbf{m}^{0} + \mathbf{S}^{\dagger} \mathbf{n}$$
 (3)

wie bei linearem Problem, hängt aber von der Sensitivität ab.  $\mathbf{R}^D$  analog. -

### Inversion verschiedener Datentypen

### Beispiel Magnetotellurik

Modell: spezifischer Widerstand als Funktion der Tiefe (logarithmisch)

$$\mathbf{m} = [\log \rho_1, \log \rho_2, \dots, \log \rho_M]^T$$

Daten: scheinbarer spezifischer Widerstand Amplitude (log) und Phase

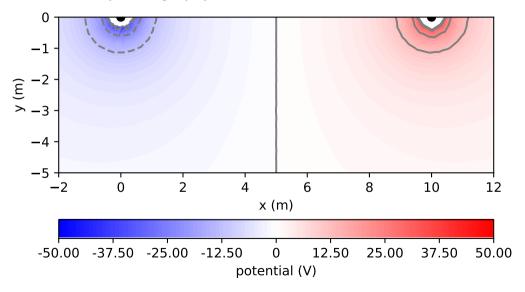
$$\mathbf{d} = [\log \rho_1^a, \dots, \log \rho_N^a, \phi_1, \dots, \phi_N]^T$$

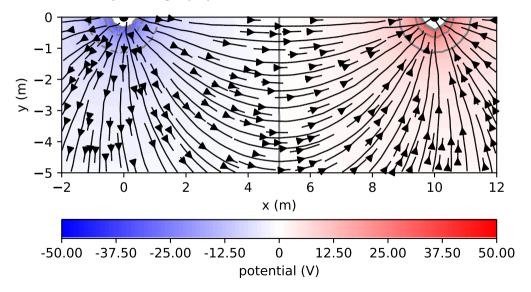
Einheitenproblem gelöst durch Fehlermodell  $\delta \rho^a$  und  $\delta \phi$ 

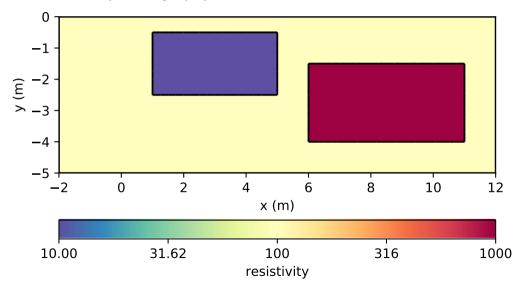
$$\boldsymbol{\varepsilon} = [\delta \rho^a / \rho_1^a, \dots, \delta \rho^a / \rho_N^a, \delta \phi, \dots, \delta \phi]^T$$

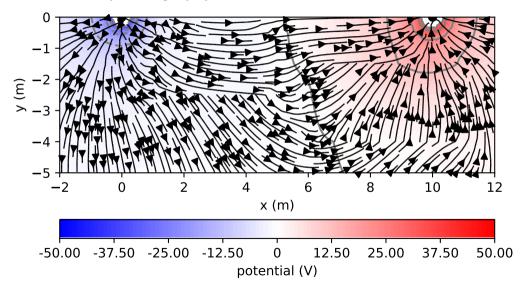
Jacobi-Matrix durch vertikales Aneinander-hängen von Matritzen

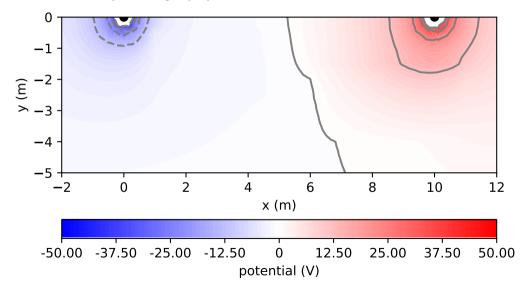
$$S_i = \left[\partial \log \rho_1^a / \partial \rho, \dots, \partial \log \rho_N^a / \partial \log \rho, \partial \phi_1 / \partial \log \rho, \dots, \partial \phi_N / \partial \log \rho\right]^T$$

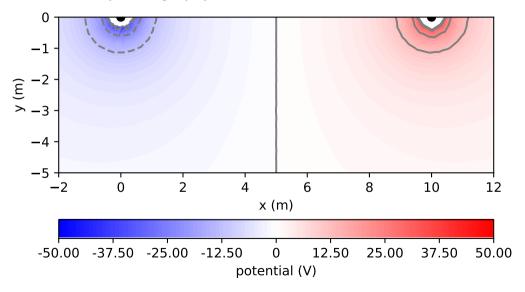


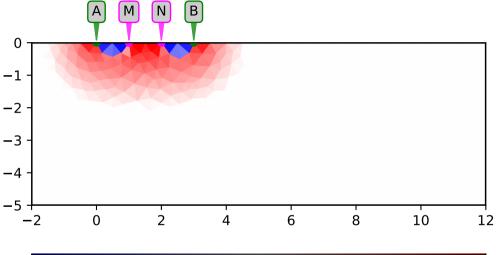


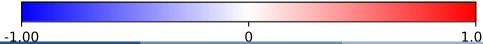


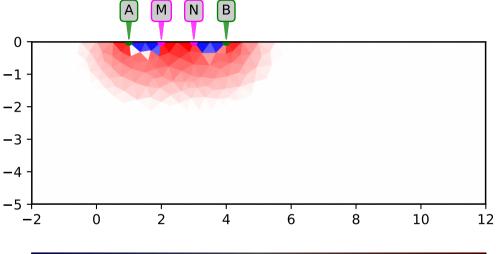


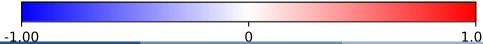


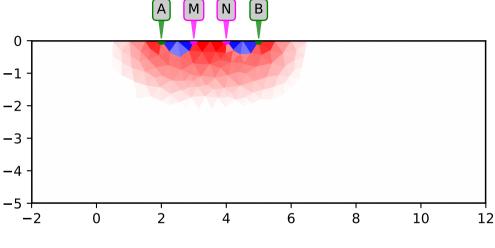


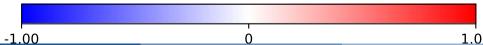


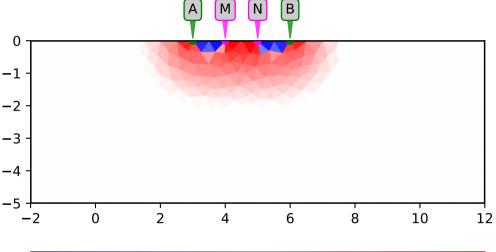


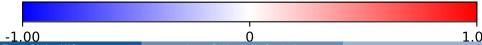




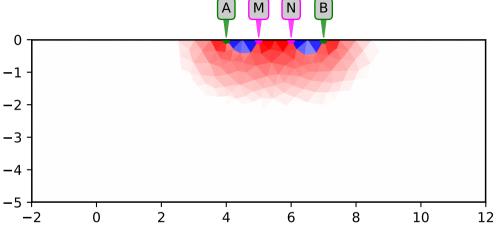


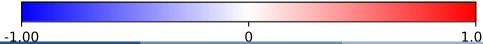


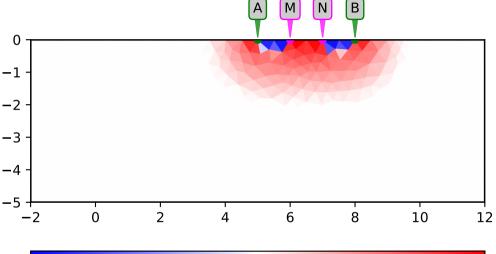


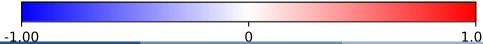


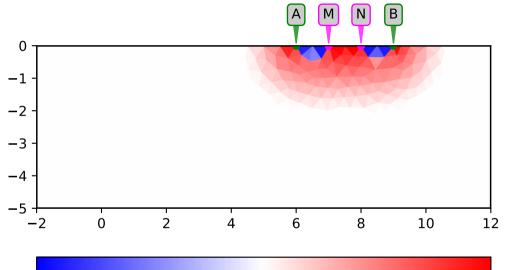
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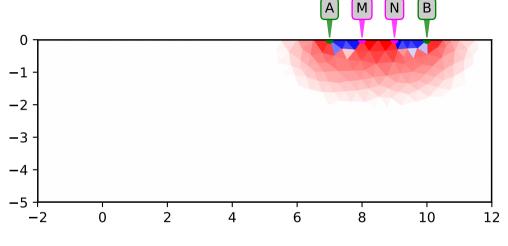


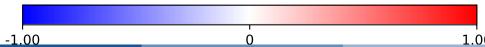


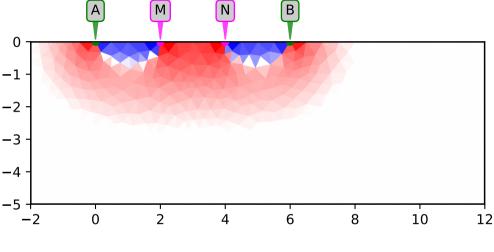
Thomas Günther (LIAG)

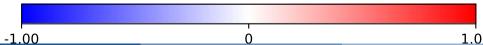
Inverse Probleme in der Geophysik

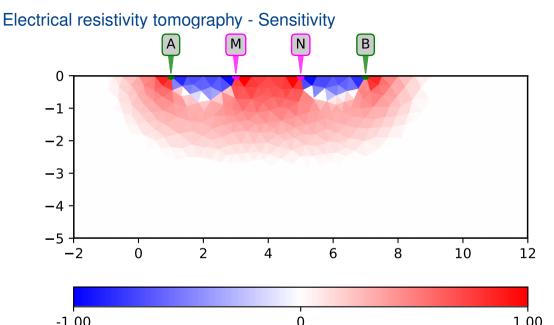
13. Juli 2020

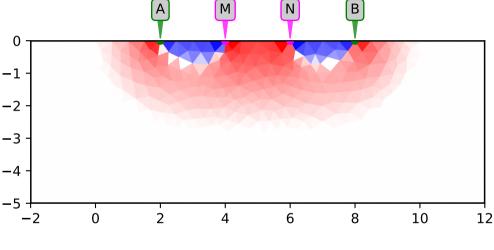


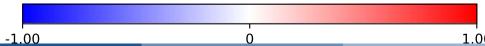


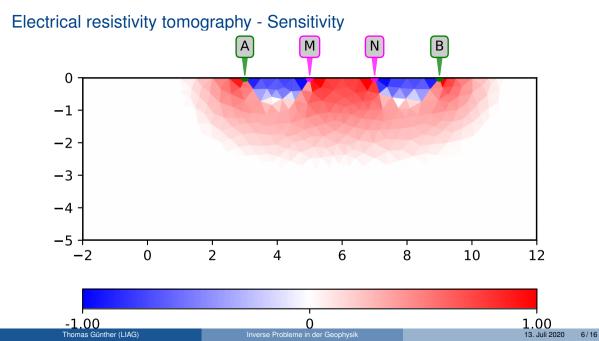


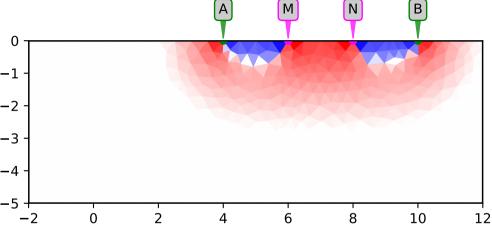


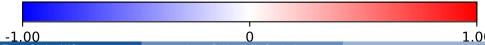


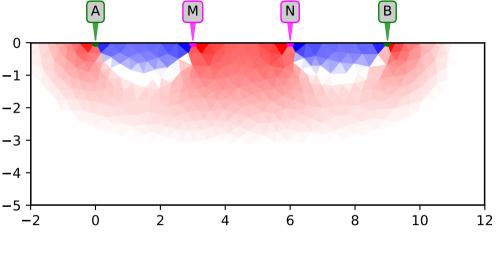


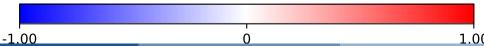


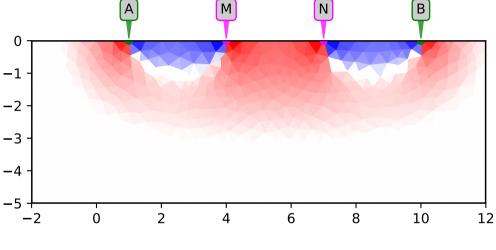


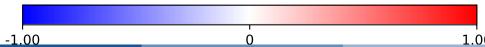












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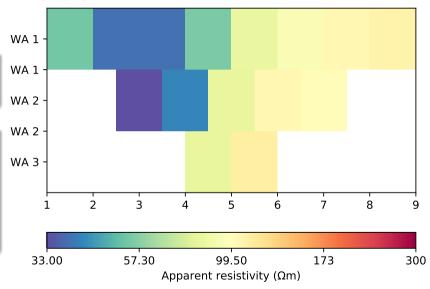
### Electrical resistivity tomography - data

#### Pseudosection

coloured data table overlain by data errors

#### Inversion

reconstruct resistivity image ambiguous ⇒ smoothest distribution improve by additional data



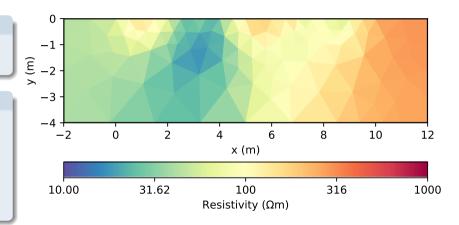
### Electrical resistivity tomography - data

#### **Pseudosection**

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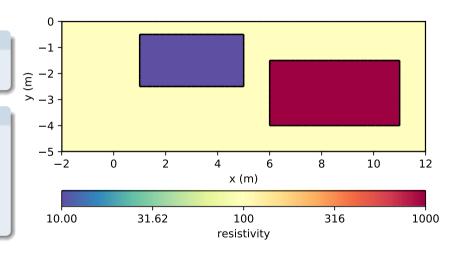
### Electrical resistivity tomography - data

#### **Pseudosection**

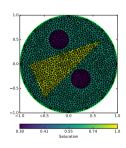
coloured data table overlain by data errors

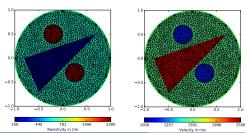
### Inversion

reconstruct resistivity image ambiguous ⇒ smoothest distribution improve by additional data



## Petrophysical Joint Inversion ERT/Ultrasonic

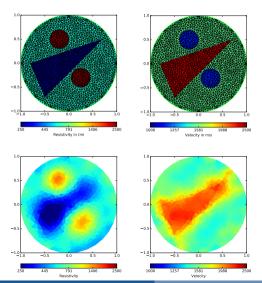




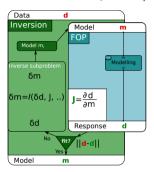
### Synthetic model

- Circular heterogeneous domain
- ERT + Ultrasonic with 16 sensors
- Target: saturation  $S_w$
- Assumption: known porosity Φ=40%
- Archie-Equation
- Wyllie-Equation
- Data  $\rho_a$ , t with noise (2%, 10  $\mu$ s)

### Inversion ERT/Ultrasonic



### Separate inversion for $\rho$ and $v_p$

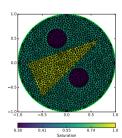


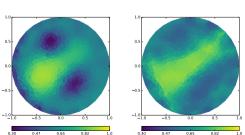
fERT = ERTManager.createFop()
iERT = Inversion(fERT, rhoa)
resistivity = iERT.run()

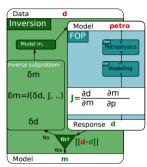
fSRT = Traveltime.createFop()

### Petrophysical Inversion ERT/Ultrasonic

#### Separate petrophysical inversion for $S_w$





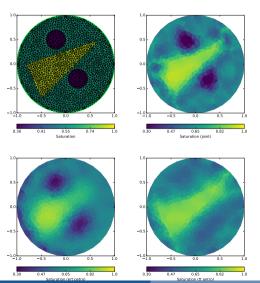


```
fERTpetro = PetroModelling(
fERT, Archie(phi=0.4, rhof=20))
fSRTpetro = PetroModelling(
fSRT, Wyllie(phi=0.4))
```

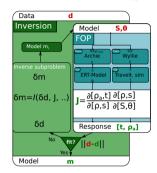
Saturation (ert petro)

Saturation (tt petro)

## Petrophysical Joint-Inversion ERT/Ultrasonic

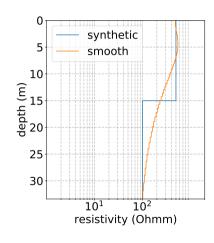


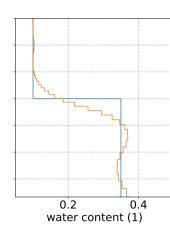
#### Common inversion for $S_w$

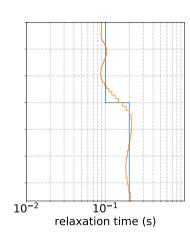


fArchie = PetroModelling(
fERT, Archie(phi=0.4, rhof=20))
fWyllie = PetroModelling(
fTT, Wyllie(phi=0.4))
fop = JointModelling([fArchie,

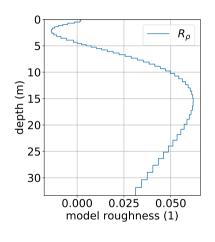
#### Smooth inversion

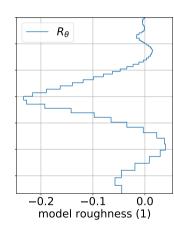


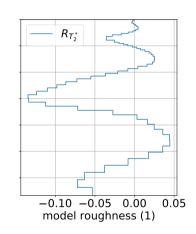




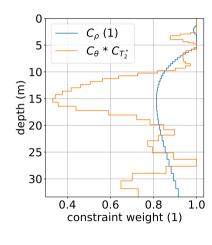
#### Roughness

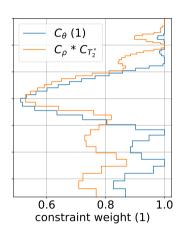


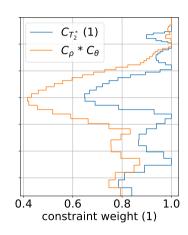




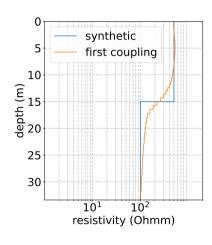
#### Constraint weights

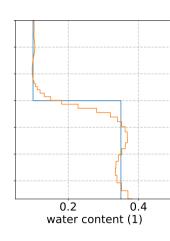


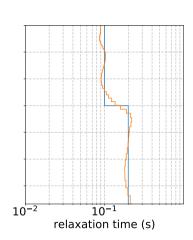




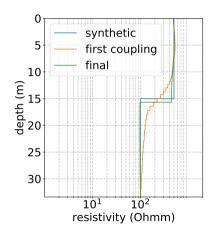
#### First coupling

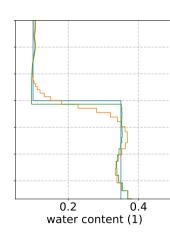


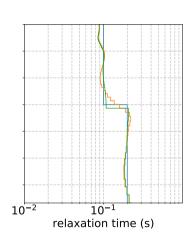




#### Final result









## Problemangepasste Regularisierung

Inhalt...

### Globale Inversionsstrategien

### Bio-inspirierte Inversionsalgorithmen

- Genetische Algorithmen (GA)
- Simulated Annealing (SA)
- Particle Swarm Optimization (PSO)
- Ant Colony System (ACS)
- Artificial Bee colony Algorithm