

Characterisation of Structures and Transport Processes in Aquifers using Electrical Methods

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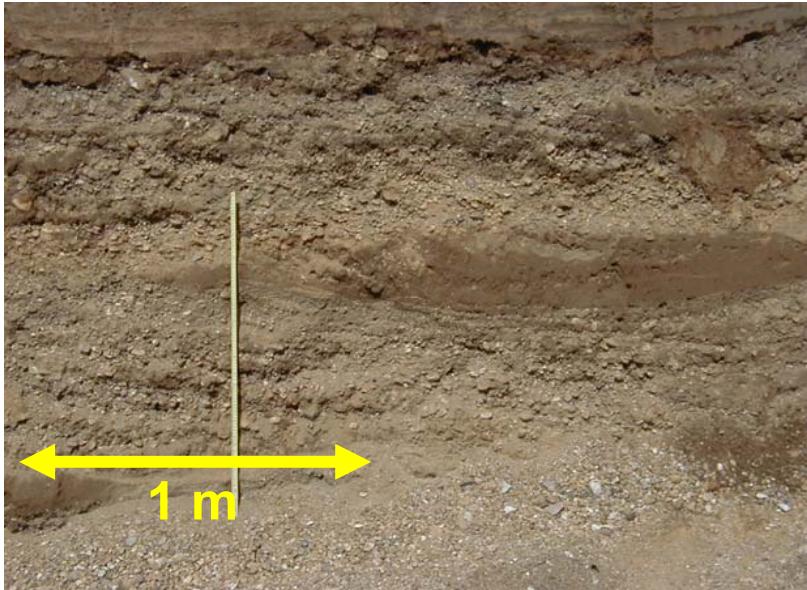
Outline

- Aquifer Structure and Processes
- Structural Characterisation
- Characterisation of Solute Transport
- Conclusions

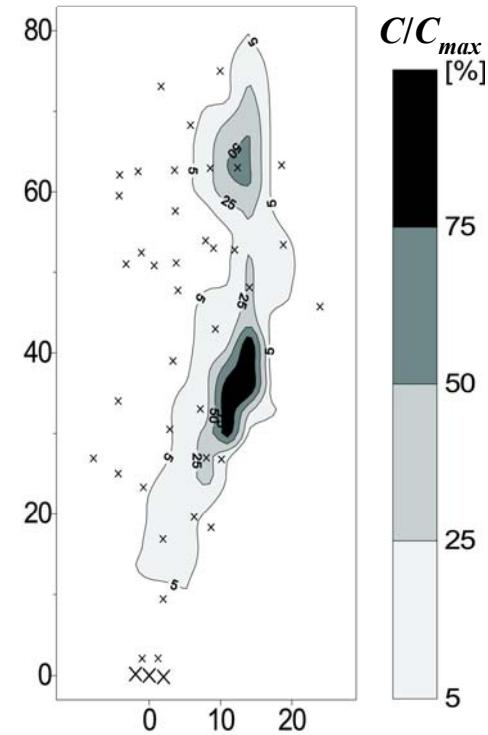


Flow and Transport in Aquifers

- Flow and transport processes in aquifers are strongly determined by the aquifer structure

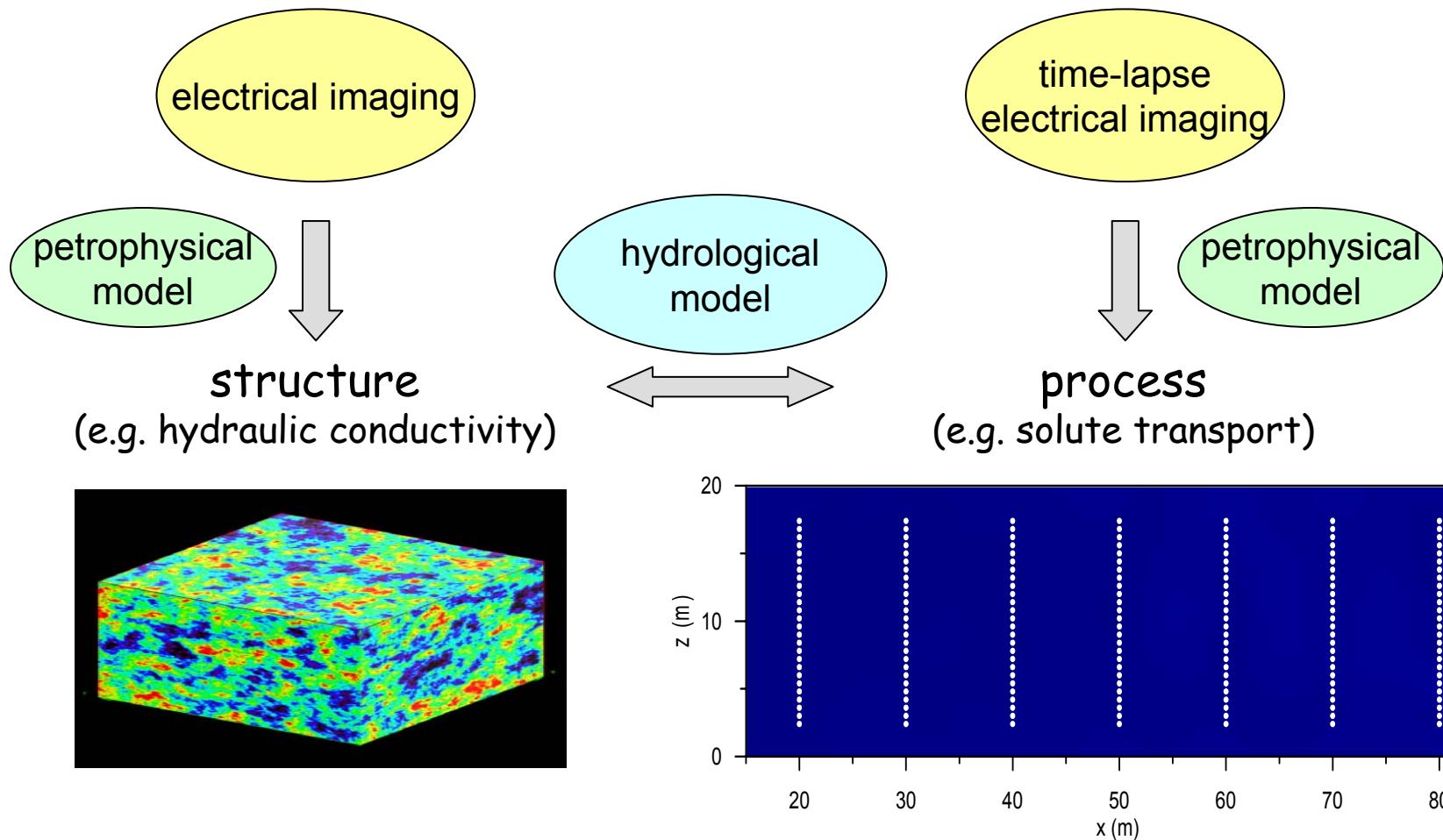


heterogeneous fluvial sediment
(variations in e.g. grain/pore size, clay content)



solute plume in a
heterogeneous aquifer
(variations in solute concentration)

Structure and Process Characterisation using Electrical Imaging



Outline

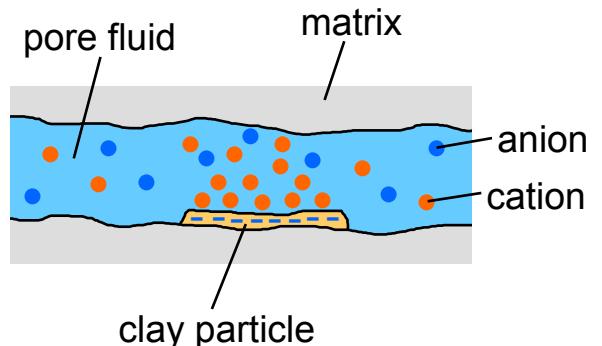
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Electrical Properties of Porous Rocks

- Electrolytic conductivity σ_{el}

$$\sigma_{\text{el}} = a \sigma_w \Phi^m S^n$$



- Surface (interface) conductivity σ_{surf}^*

→ complex, frequency-dependent electrical conductivity

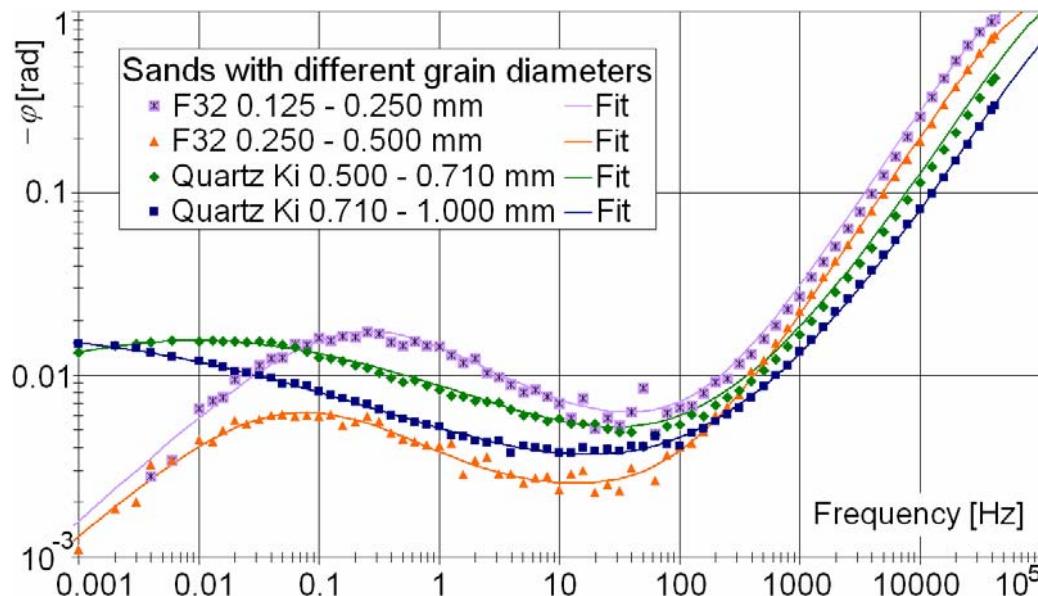
$$\sigma^* = |\sigma| e^{i\varphi} = [\sigma_{\text{el}} + \sigma'_{\text{surf}}(\omega)] + i \sigma''_{\text{surf}}(\omega)$$

contains information on:

e.g. salinity, water content, clay content, pore space topology, hydraulic conductivity

Spectral Induced Polarisation (SIP)

- Phase spectra on sands with different grain size



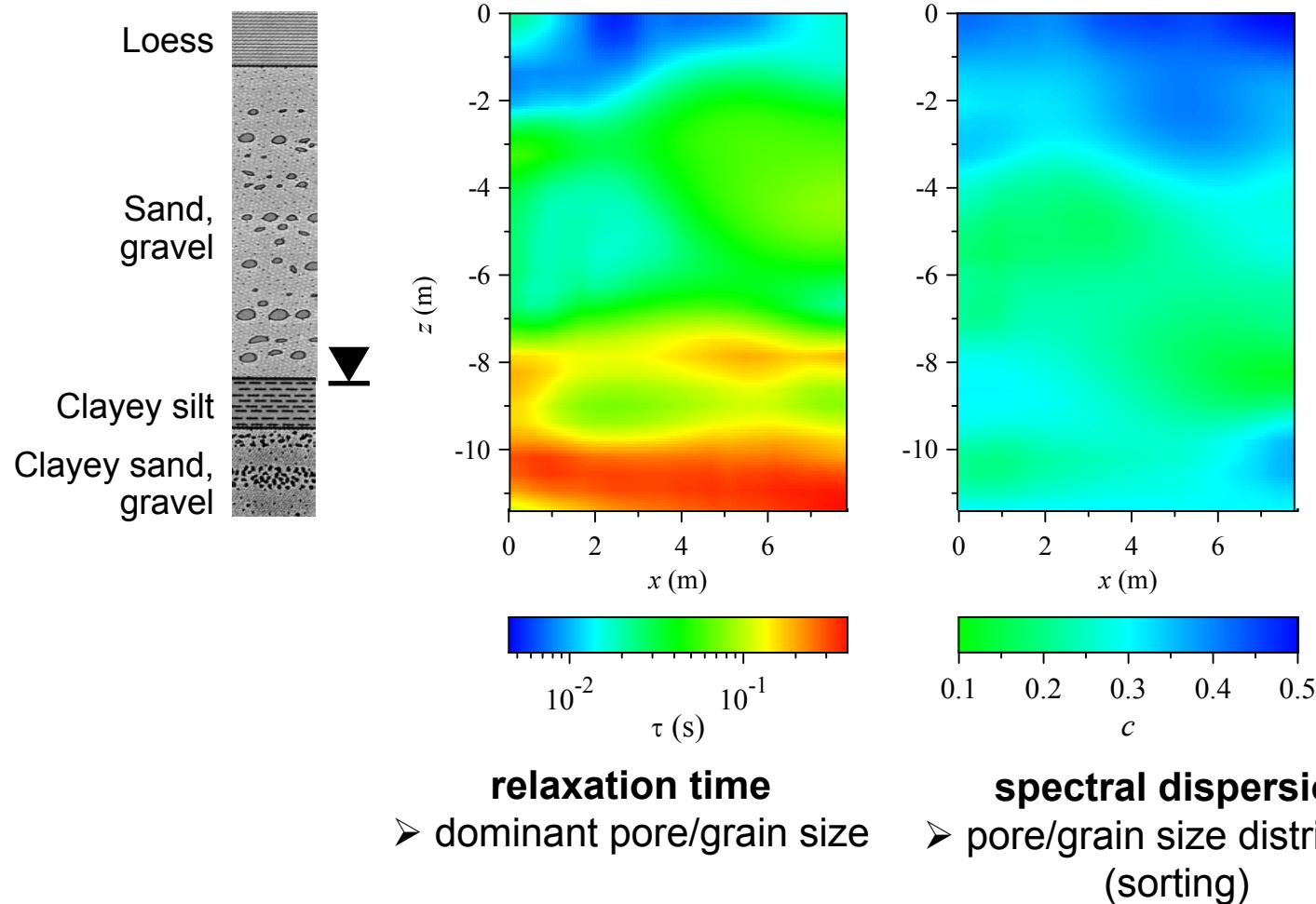
$$Z(\omega) = R_0 - R_0 m + R_0 m \int_0^{\infty} \frac{g(\tau')}{1 + (i\omega\tau')^c} d\tau'$$

→ inverse modelling of relaxation time distribution

→ pore size distribution (relaxation diffusion length $L \sim \sqrt{\tau}$)

SIP Imaging

- Improved textural characterisation by analysing the spectral response



Hydraulic Permeability Estimation

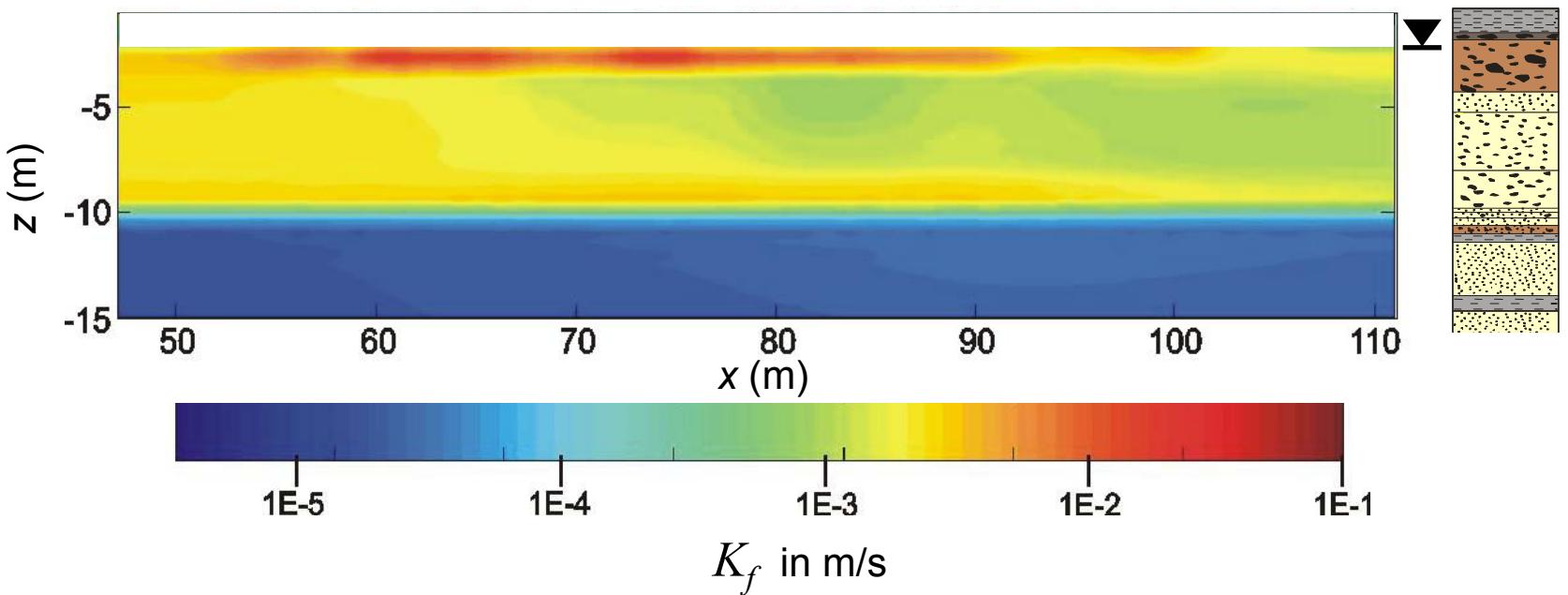
- Surface imaging result

$$S_{\text{por}} = a \sigma'', \quad K \approx \frac{b}{F S_{\text{por}}^c}$$

Börner et al. (1996)



Krauthausen test site with monitoring wells



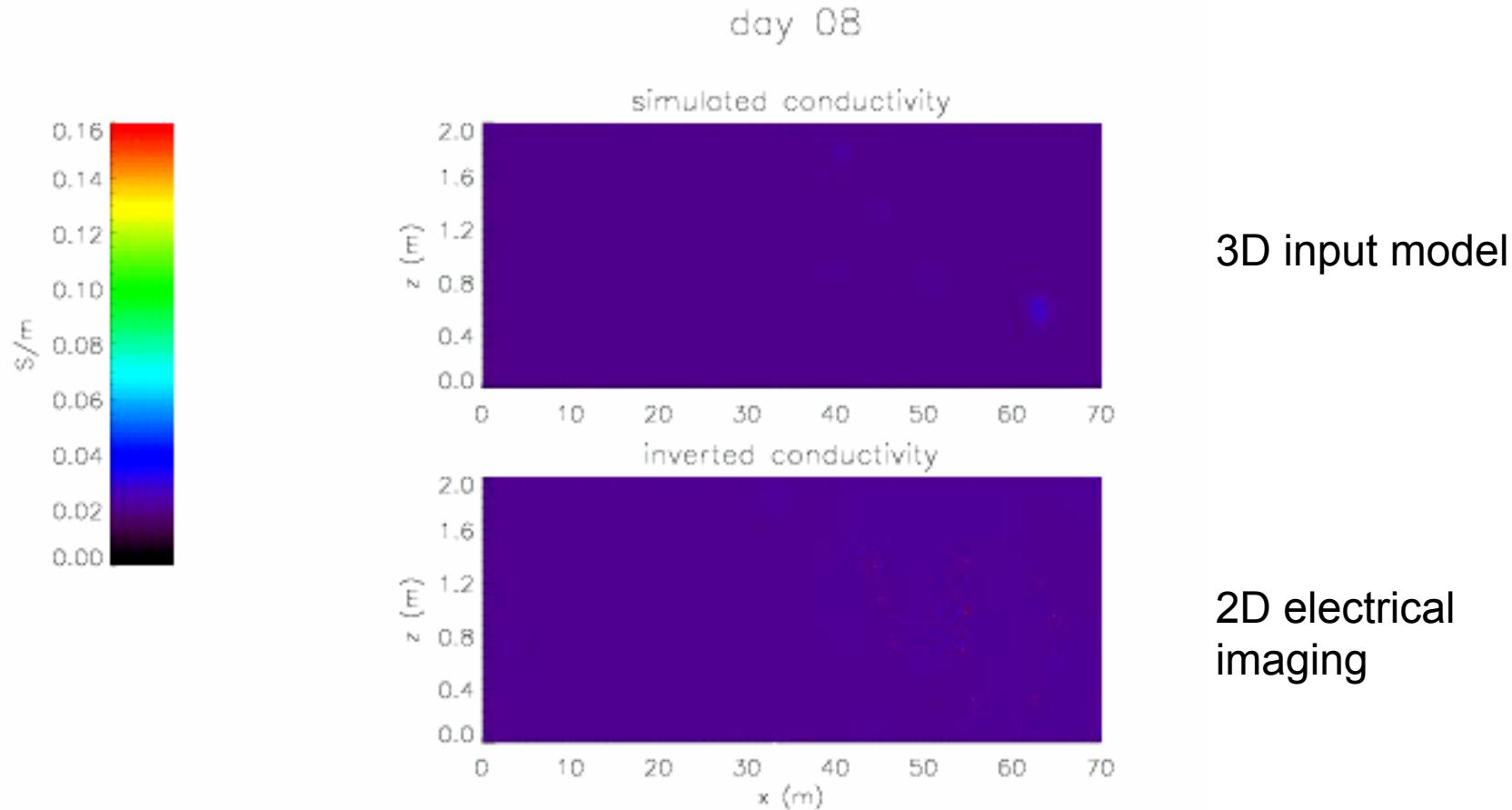
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Electrical Imaging of Solute Transport

- Synthetic experiment



Solute Transport in Heterogeneous Porous Media

- Flow equation $\nabla \cdot [K(\mathbf{x}) \nabla \psi(\mathbf{x})] = 0$ $\mathbf{q}(\mathbf{x}) = -K(\mathbf{x}) \nabla \psi(\mathbf{x})$
(sink/source free region)

- Convection dispersion equation (CDE) $\theta(\mathbf{x}) \frac{\partial C(\mathbf{x}, t)}{\partial t} = -\mathbf{q}(\mathbf{x}) \nabla C(\mathbf{x}, t) + \nabla \cdot [\theta(\mathbf{x}) \mathbf{D}(\mathbf{x}) \nabla C(\mathbf{x}, t)]$
(non-reactive solute)



heterogeneous flow field
and local dispersion



homogeneous flow field
and local dispersion

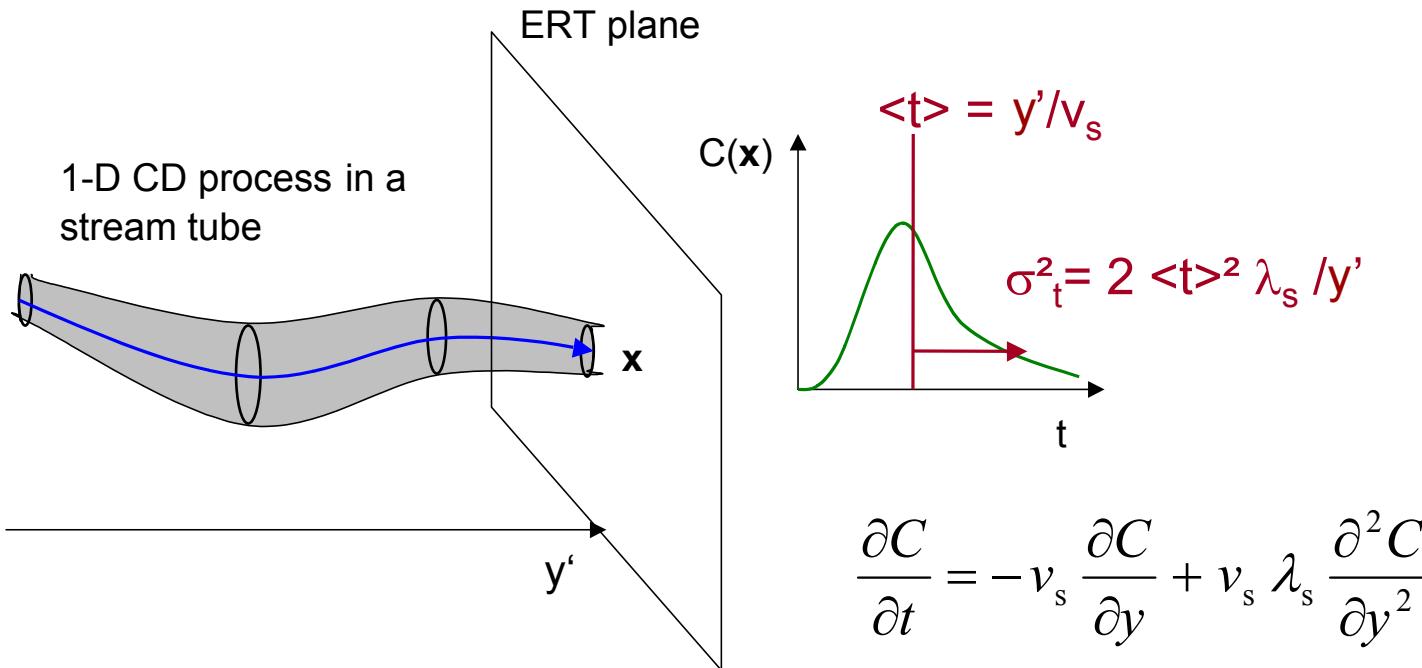


heterogeneous flow field
without local dispersion

Englert, 2004

Quantification of Transport Properties

- Conversion of electrical conductivity changes in concentrations
- Fitting an equivalent 1D (“stream tube”) convection dispersion model to the local breakthrough curves (BTCs)

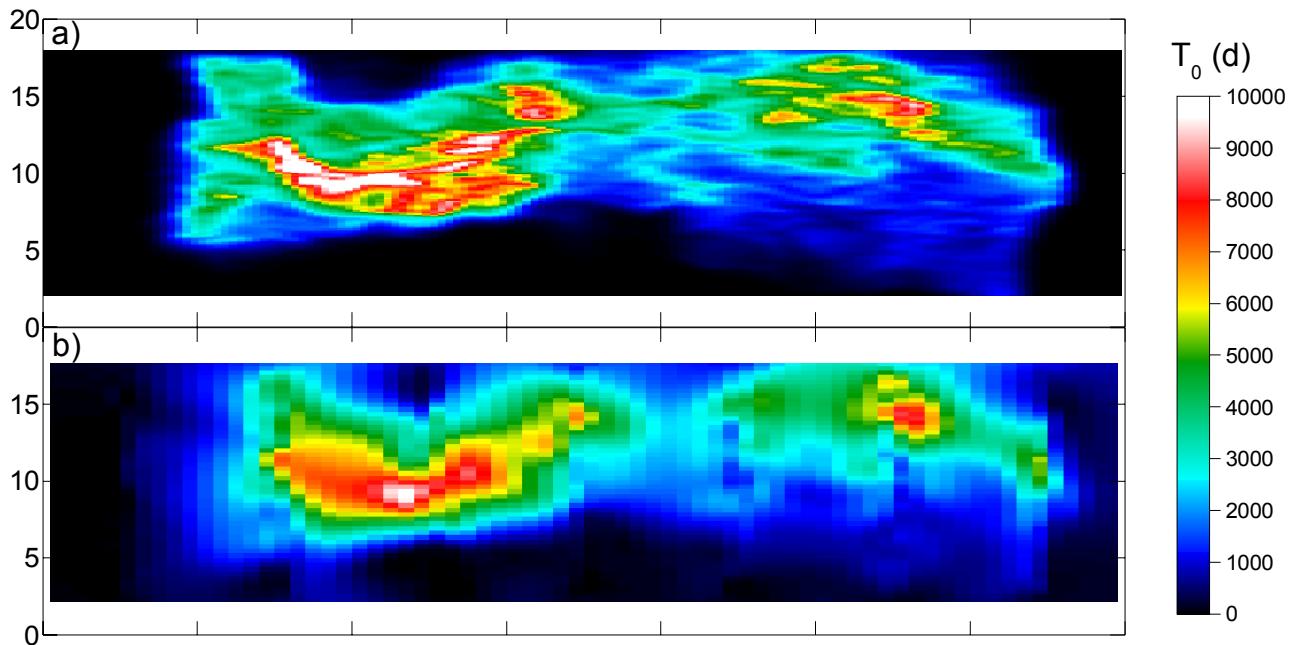


→ Spatial variability of flow velocity and (longitudinal) dispersivity

Synthetic Experiment

- Spatial variability of 0th moment of local BTCs

3D input model



2D imaging

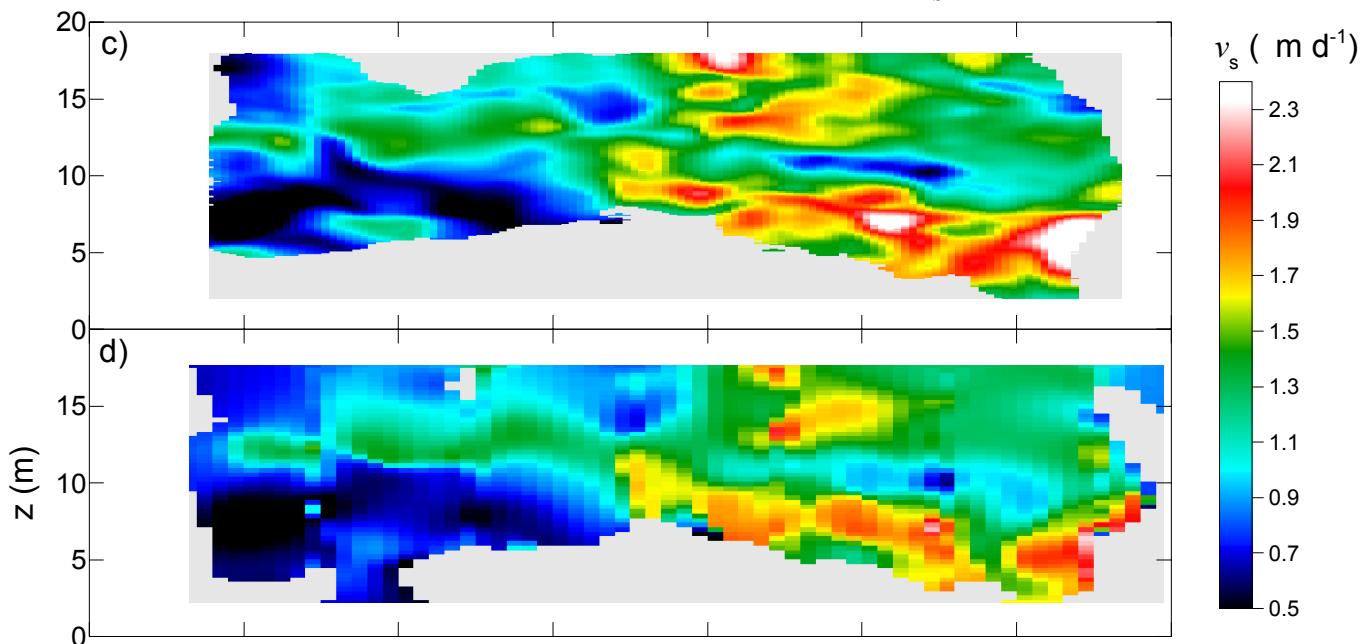
Synthetic Experiment

- Spatial variability of equivalent velocity

3D input model

$$\langle v_s \rangle = 1.15 \text{ m/d}$$

$$\sigma^2_{v_s} = 0.13 \text{ (m/d)}^2$$



2D imaging

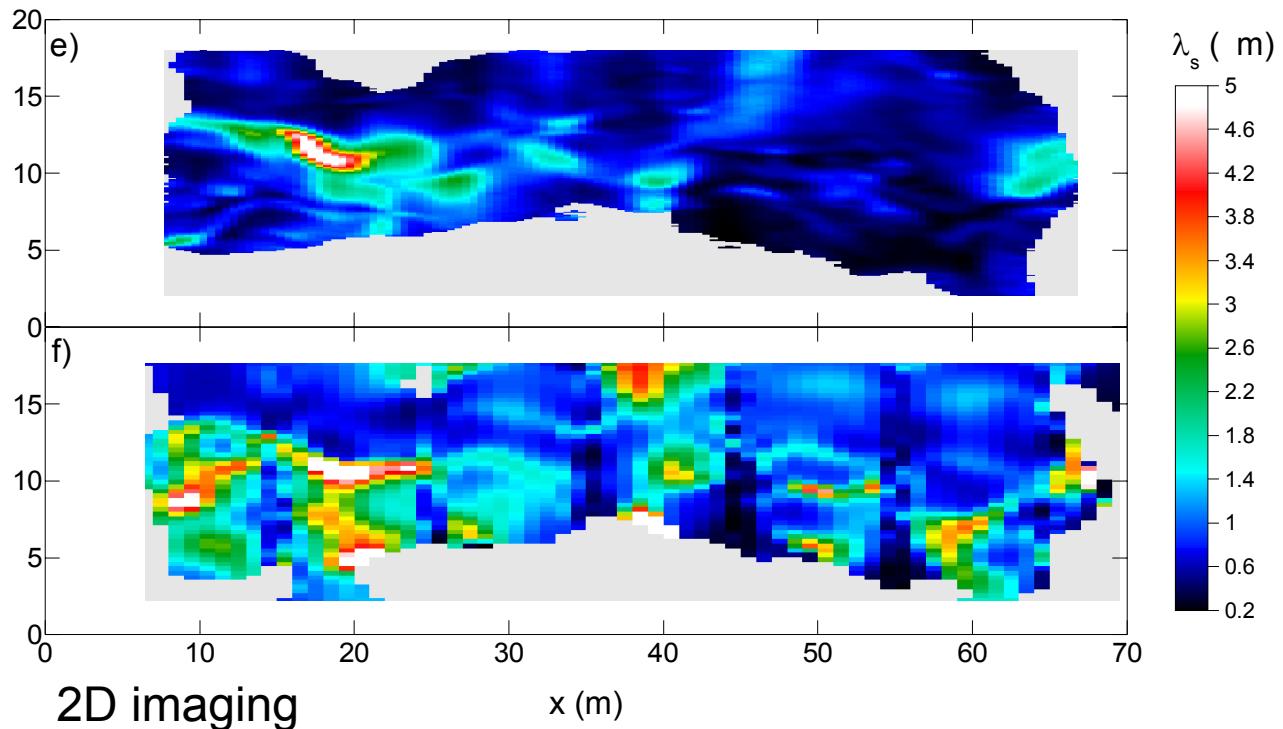
$$\langle v_s \rangle = 1.13 \text{ m/d}$$

$$\sigma^2_{v_s} = 0.10 \text{ (m/d)}^2$$

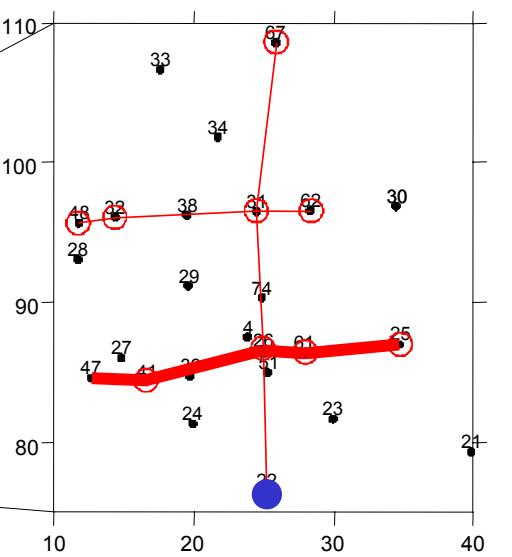
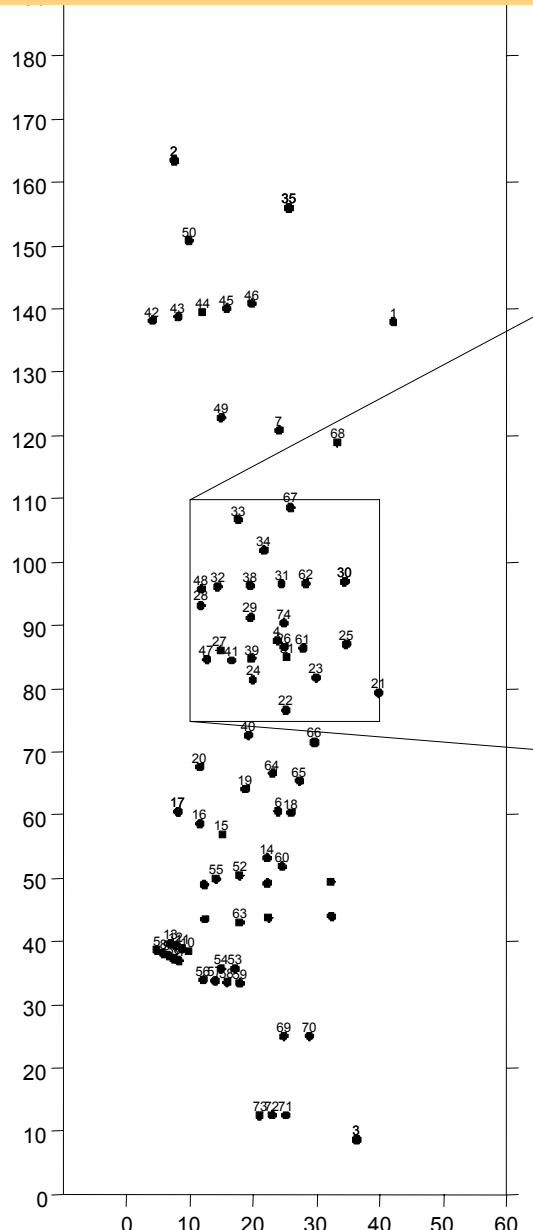
Synthetic Experiment

- Spatial variability of equivalent dispersivity

3D input model



Recent Tracer Experiments at Krauthausen



ERT
(Electrical Resistivity Tomography)

injection well

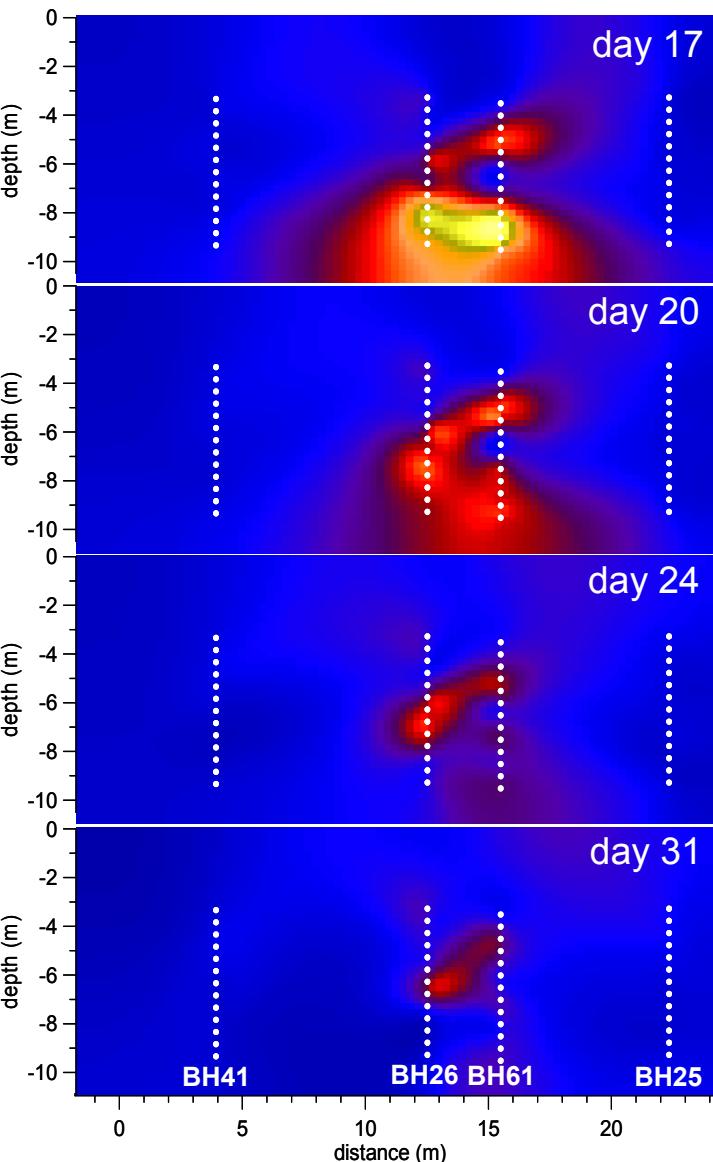
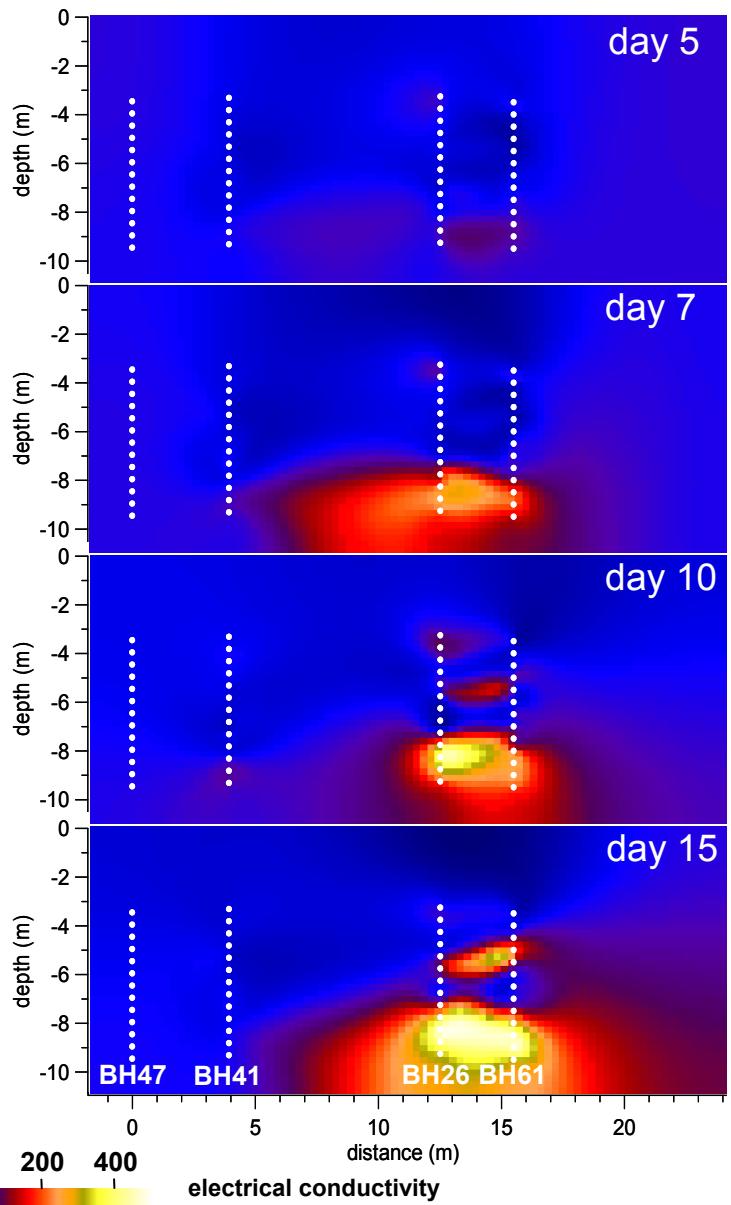
Sep - Nov 2002

400 kg CaCl₂ / 140 000 l
injected over 7 d
between -3 and -10 m
(~ 2.86 g/l @ 835 l/h)
electrical contrast ~7:1

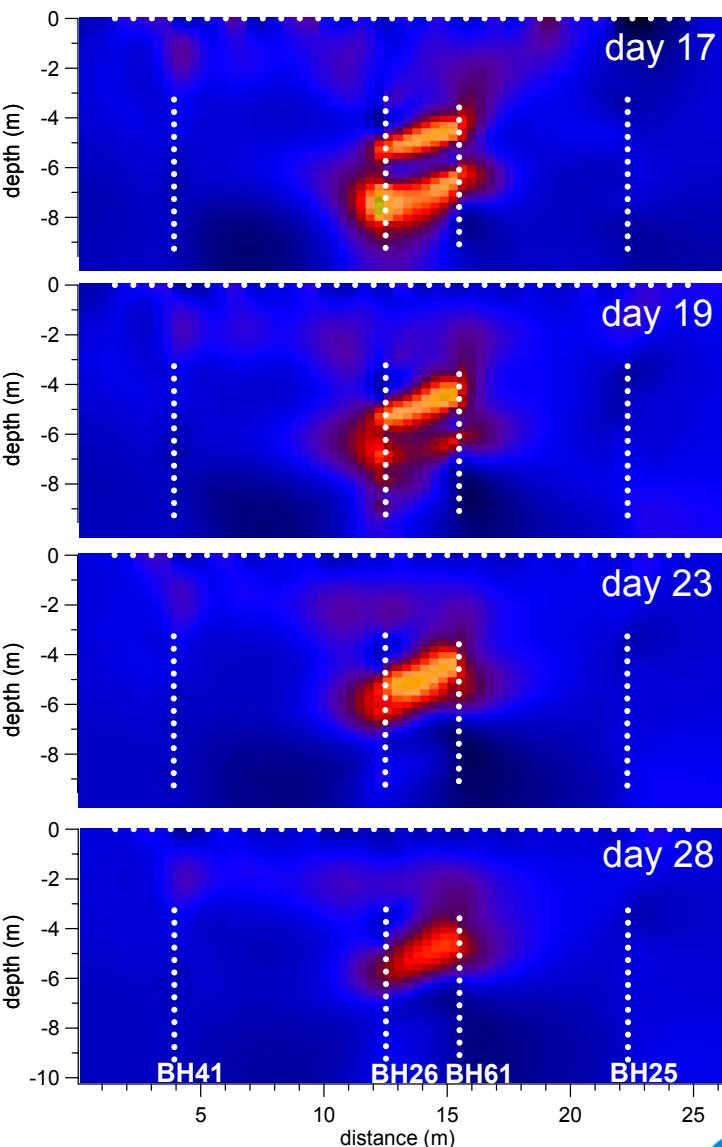
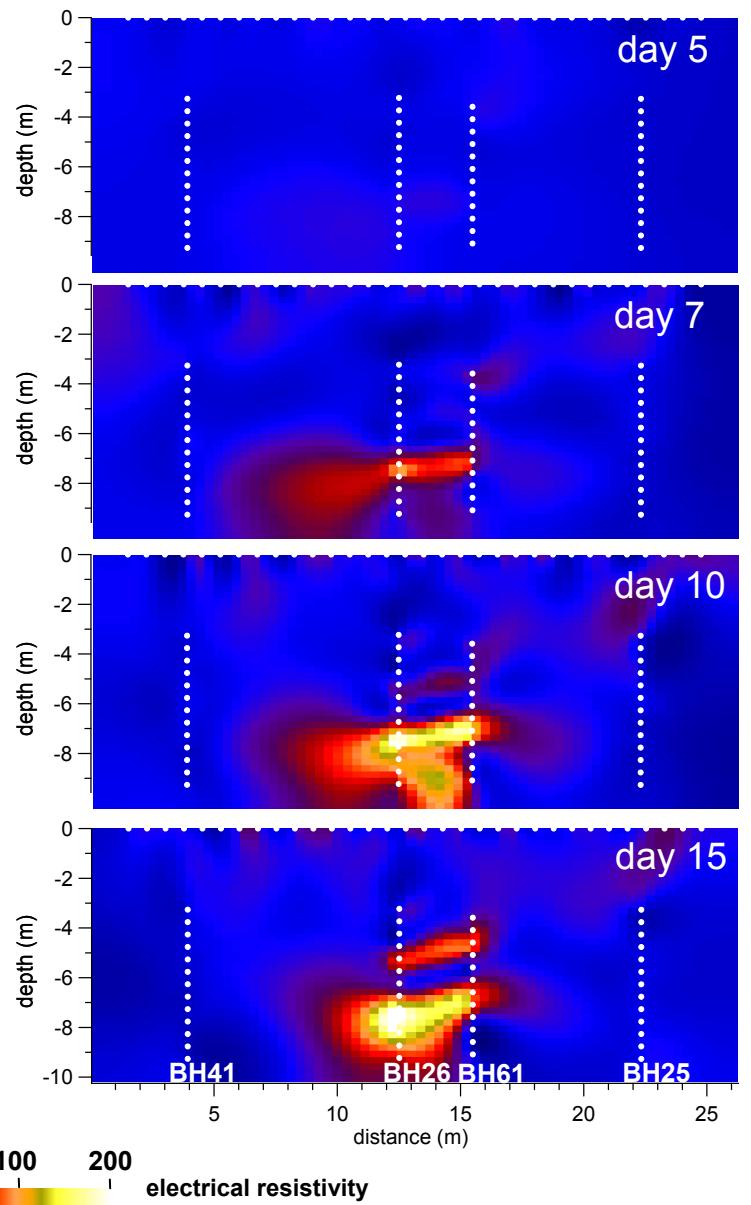
Sep - Nov 2003

repeated with
“negative tracer”
electrical contrast ~4:1

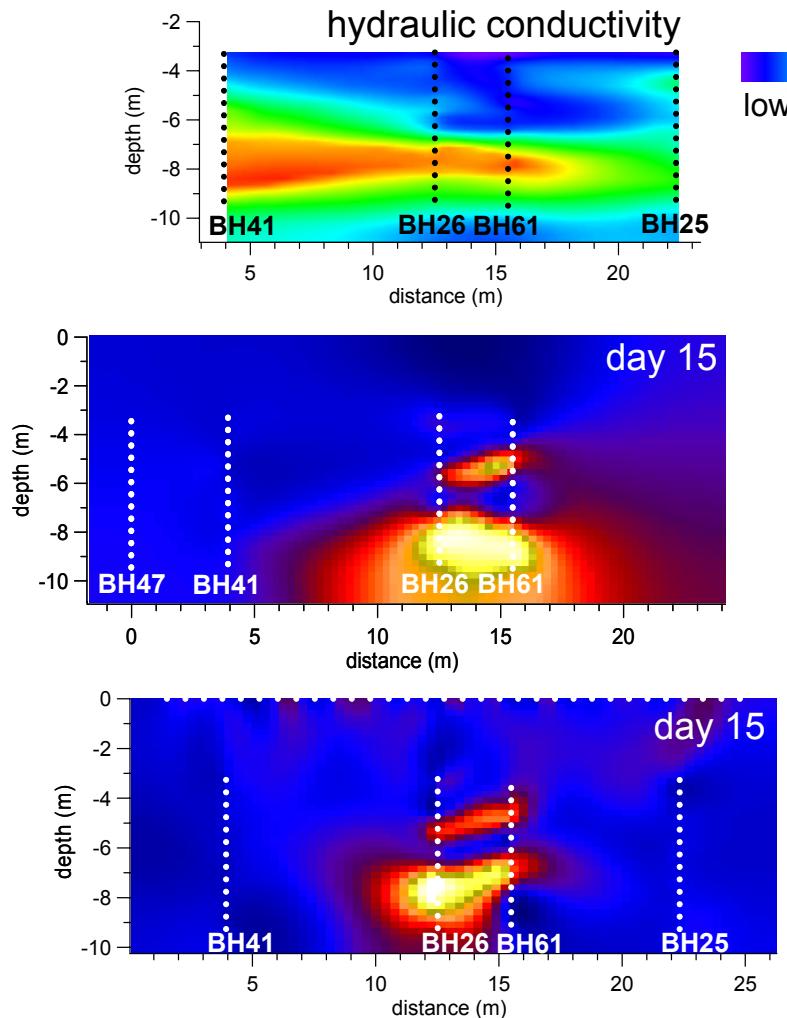
ERT Results from Conductive Tracer Experiment



ERT Results from Resistive Tracer Experiment



Structure vs. Process Characterisation



conductive tracer

resistive tracer

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Conclusions

- In combination with petrophysical and hydrological models, electrical imaging can provide quantitative information about structures (e.g. hydraulic conductivity) and processes (e.g. solute transport) in aquifers with relatively high spatial resolution.

However ...

- Improved hydrologic-geophysical (e.g. hydraulic-electrical) parameter relationships are needed for the linkage of measured electrical properties with structural characteristics and/or state variables of hydrologic systems.
- Improved data fusion methodologies are needed to optimally integrate hydrological and geophysical information – either in terms of data, models or general concepts.

- improved understanding of structures and processes in hydrologic systems
- improved hydrological models for a more reliable scenario prediction
- improved (sustainable) management of water resources



Acknowledgements

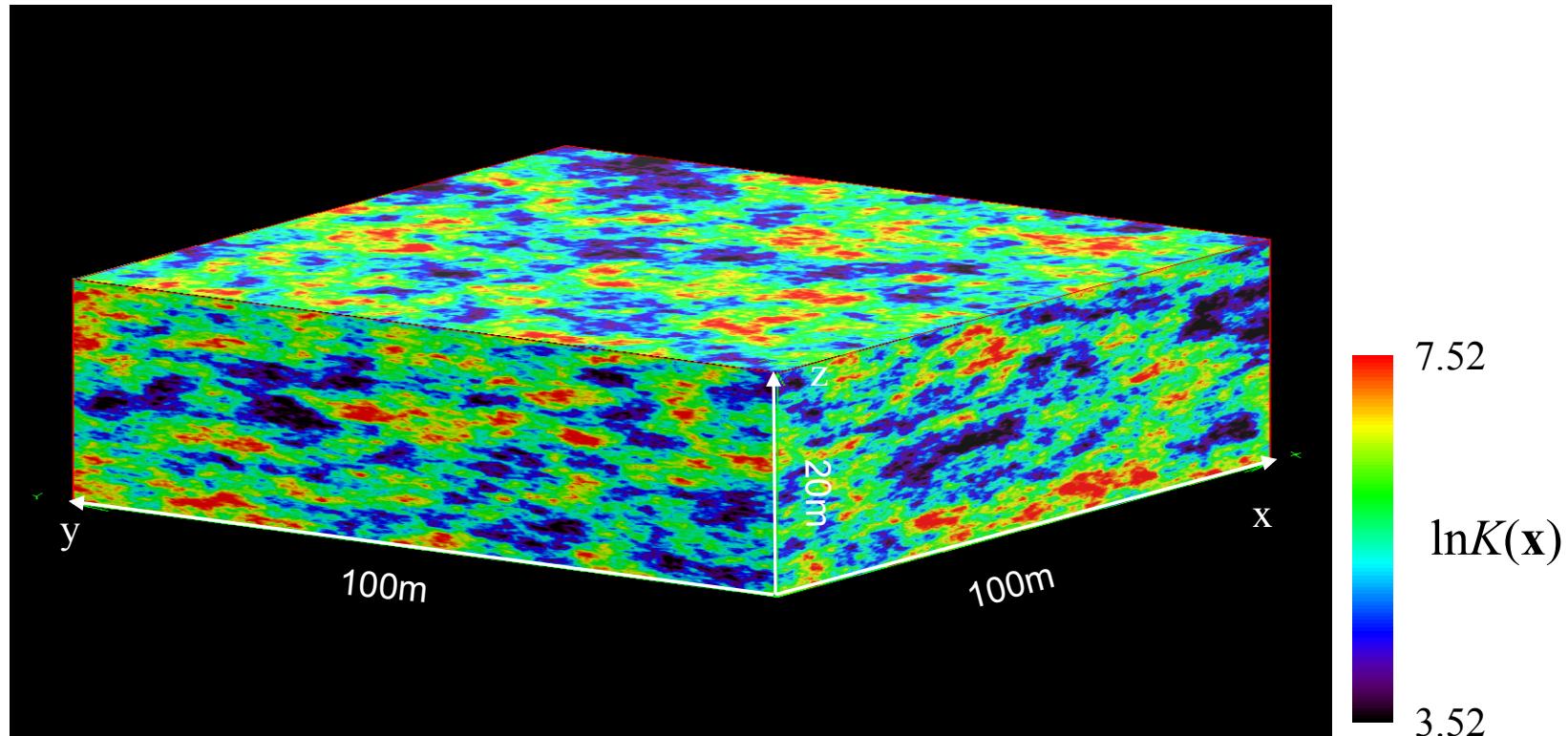
Thanks to my colleagues at the Agrosphere Institute, in particular:

- Andreas Englert
- Kerstin Müller
- Martin Münch
- Jan Vanderborght



3D Hydraulic Conductivity Field

- $\ln K(\mathbf{x})$ represented as spatial stochastic process with mean value $K_0 = 250 \text{ m/d}$, variance $\sigma_{\ln K} = 1$, and exponential covariance structure (5 m horizontal and 0.5 m vertical correlation lengths)



Simulation of Solute Transport

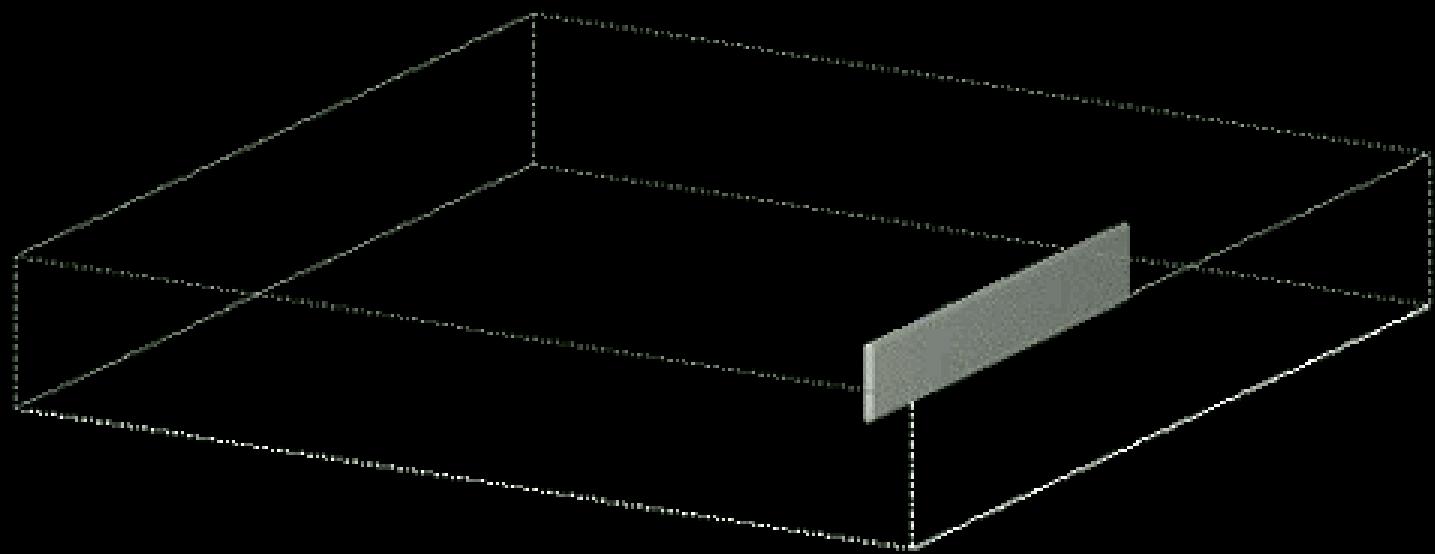
- $\frac{\partial C}{\partial t} = -\mathbf{v}(\mathbf{x}) \nabla C + \nabla (\mathbf{D}_d \nabla C); D_{dL} = |\mathbf{v}(\mathbf{x})| \lambda_{dL}; D_{dT} = |\mathbf{v}(\mathbf{x})| \lambda_{dT}$

mean hydraulic gradient: $J_y = 0.001, J_x = J_z = 0$

porosity: $\phi_0 = 0.25$

mean pore water velocity: $v = 1 \text{ m/d}$

longitudinal and transverse dispersivities: $\lambda_{dL} = 0.1 \text{ m}, \lambda_{dT} = 0.01 \text{ m}$

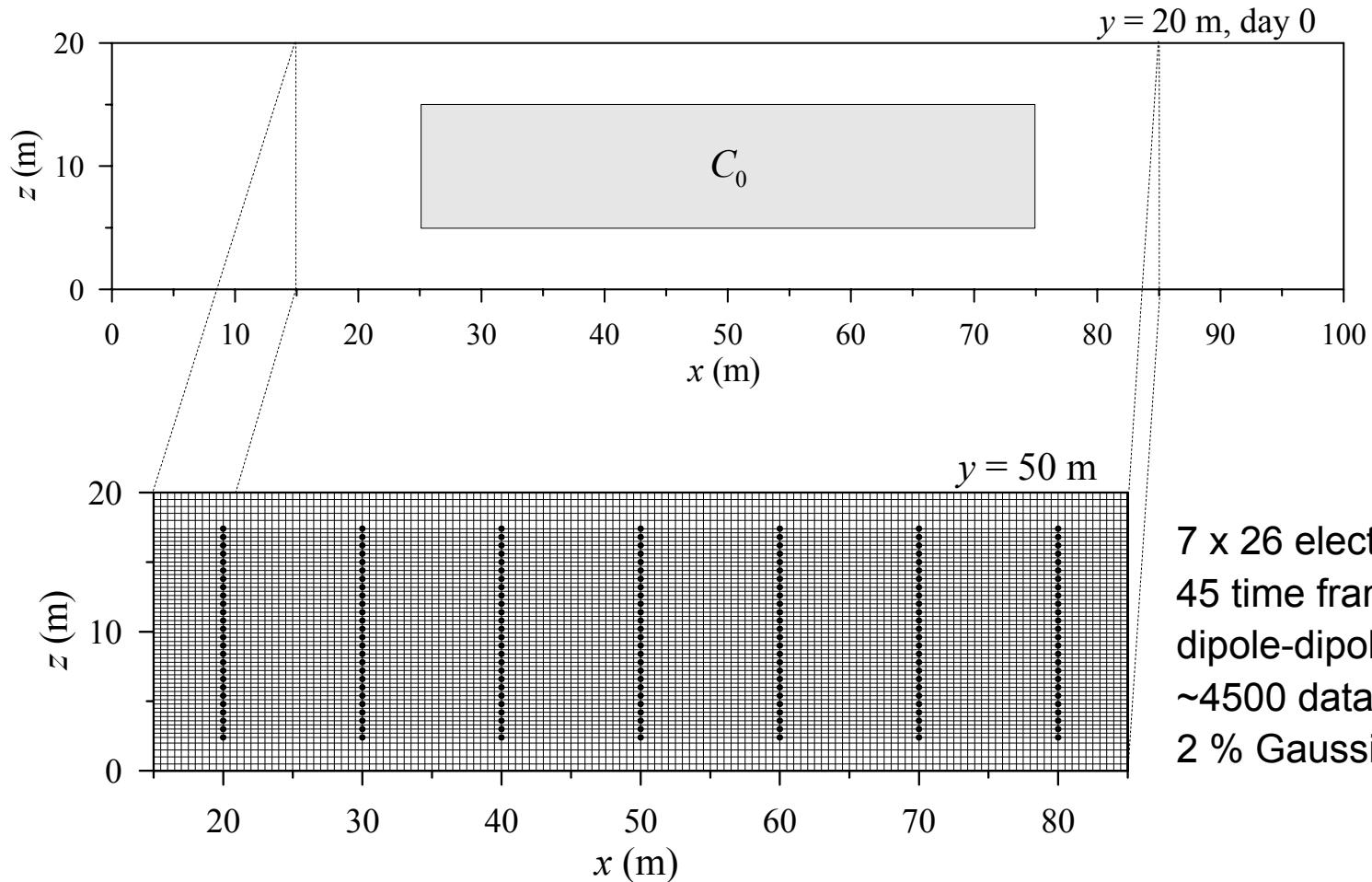


evolution of solute plume (day 0-100)



Simulation of Time-lapse ERT Data Acquisition

- Assumption of linear calibration relation: $\sigma(\mathbf{x}, t) = \sigma_{\text{in}} + \beta C(\mathbf{x}, t)$



ERT 2D Imaging Approach

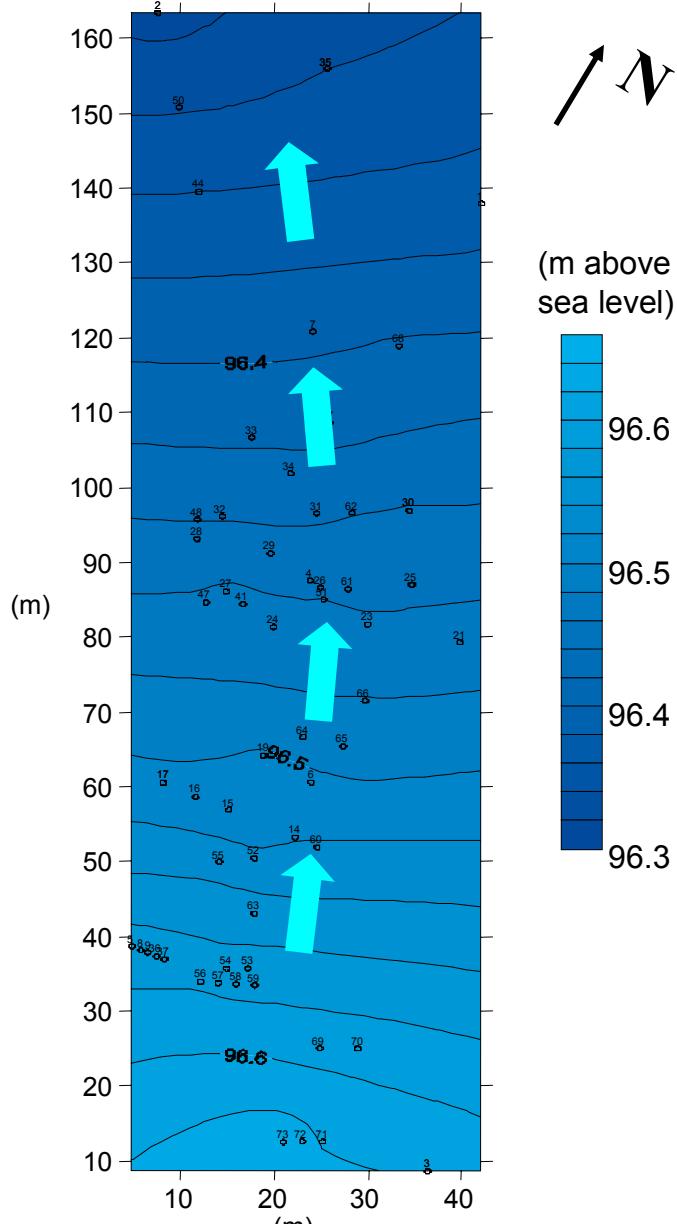
- 2.5D (finite-element) modeling
- Smoothness-constraint difference inversion (*LaBrecque & Yang, 2000*):

$$\Psi_{\text{diff}}(\mathbf{m}_t) = \|\mathbf{W}[\mathbf{d}_t - \mathbf{d}_0 + \mathbf{f}(\mathbf{m}_0) - \mathbf{f}(\mathbf{m}_t)]\|^2 + \alpha \|\mathbf{R}(\mathbf{m}_t - \mathbf{m}_0)\|^2$$

- “robust” iterative data reweighting (*LaBrecque & Ward, 1990*)

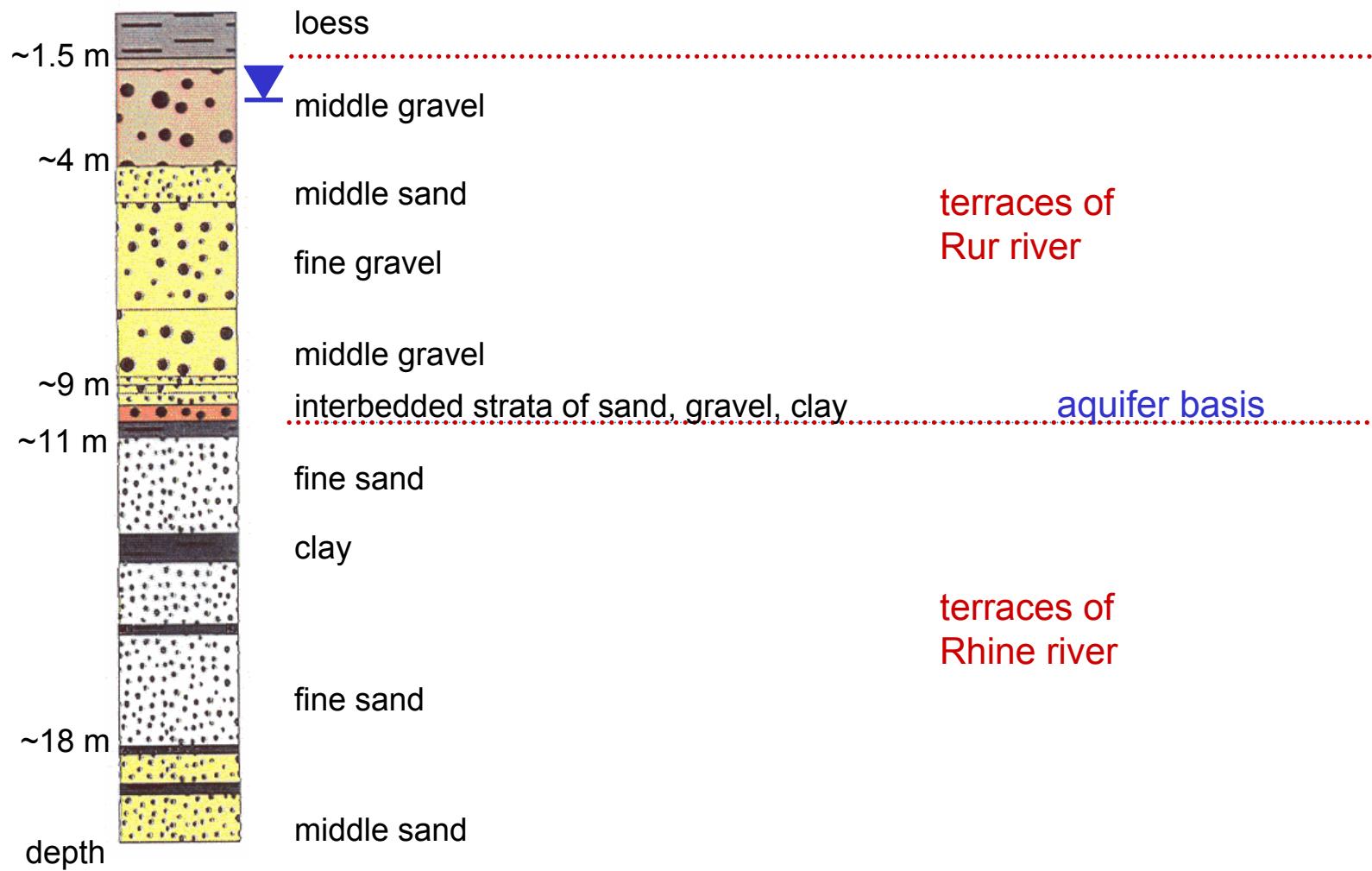


The Krauthausen Test Site: Hydrogeology



- heterogeneous aquifer
- mean porosity 26 %
- mean hydraulic conductivity $3.8 * 10^{-3}$ m/s
- hydraulic gradient 0.2 %
- mean flow velocity 0.65 m/d
- ground water table between 2.5 m (summer) and 1 m (winter)

The Krauthausen Test Site: Lithology

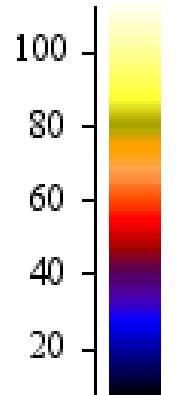


Time-lapse Electrical Imaging (Monitoring)

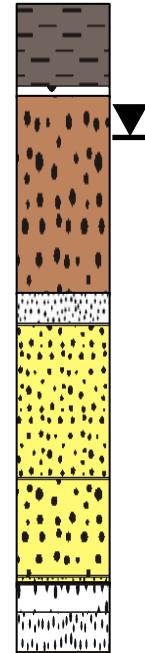
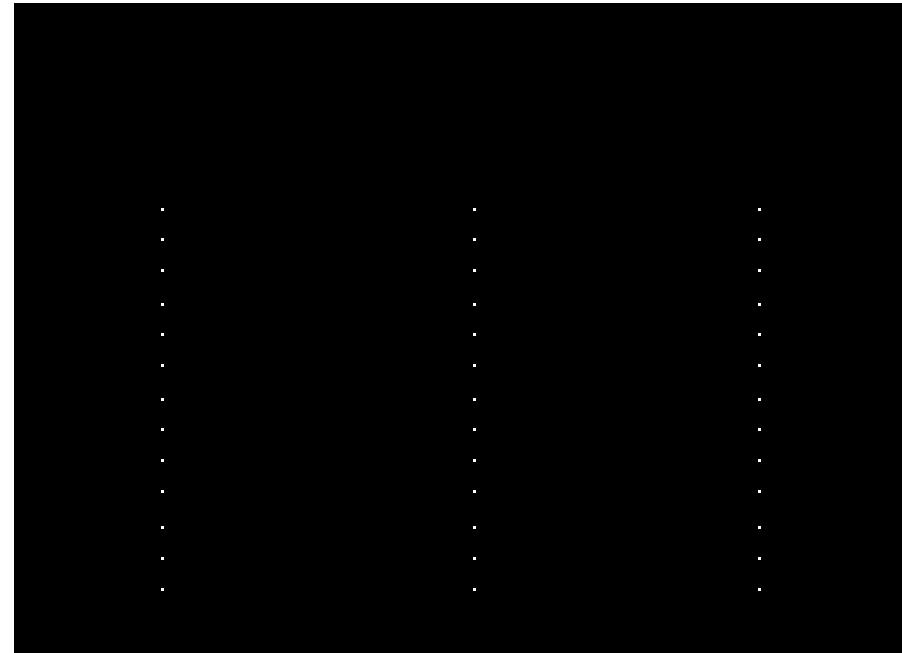
- Solute tracer experiment in heterogeneous aquifer (Krauthausen)



day 1



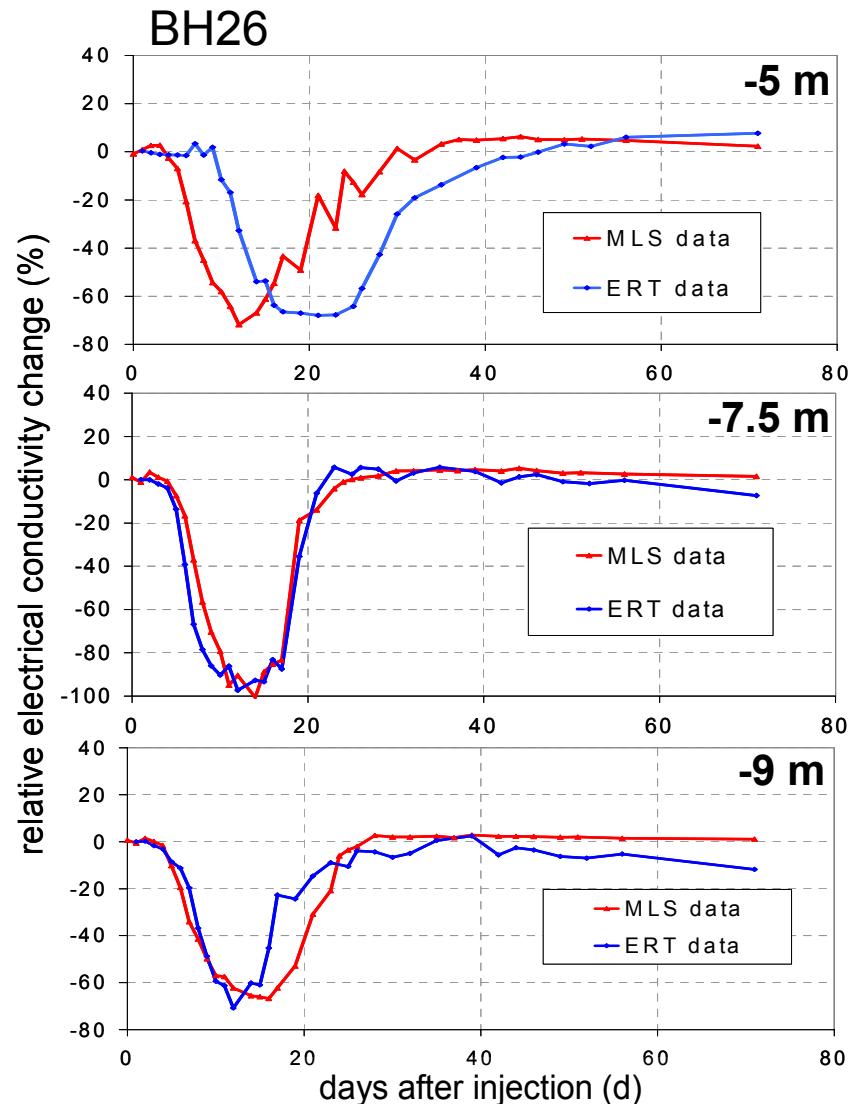
electrical conductivity changes (%)
due to injected NaBr solution



lithology

ERT vs. Multi-Level Sampling (MLS)

- Resistive tracer experiment



EIT Results at Krauthausen

- Background images in 2003

