

Automated inversion for porosity in shallow reservoirs using a pseudo-transient adjoint solver for non-linear hydro-mechanical equations

Georg Reuber, Lukas Holbach, Ludovic Räss

Hydro-mechanics

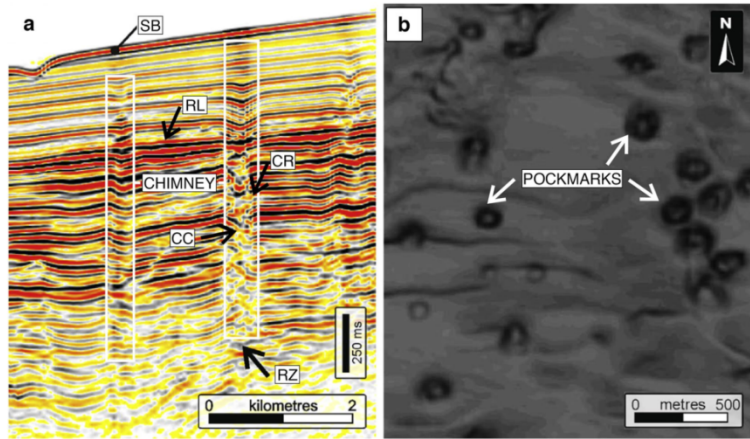
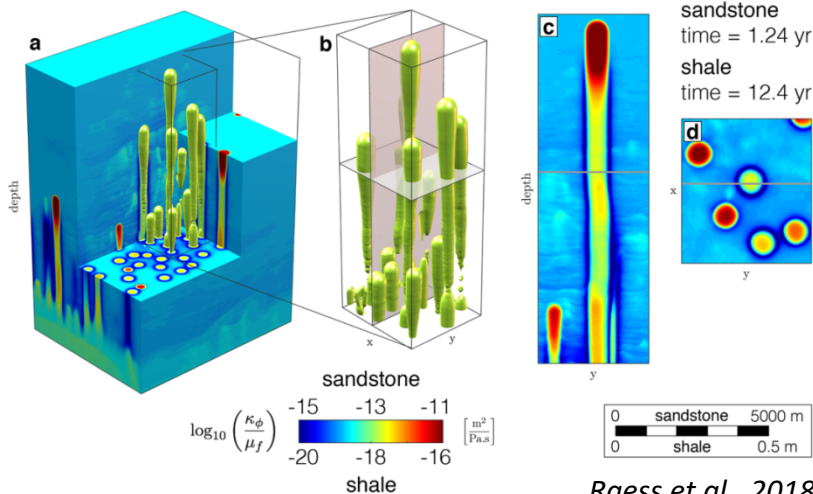


Figure 1. Seismic expression of chimneys and pockmarks. Figure modified from Cartwright and colleagues^{11,22}. (a) Vertical seismic profile through fluid migration pathways from offshore Namibia. SB = seabed, RL = reflective horizontal sedimentary layer, CR = chimney downward-bending compacted rim, CC = chimney core, RZ = root zone and diffuse base of the chimney. (b) Horizontal slice through a group of chimneys displaying the typical circular craters or pockmarks.

I. Can we invert for porosity in reservoir systems?

- Can we automatically invert for pointwise porosity and therefore the geometry of the chimney?

II. Solving adjoint systems with explicit pseudo-transient (PT) solvers (on GPU?)



Raess et al., 2018

Hydro-mechanics

2-phase flow in geology is the **combination** of **Stokes** (rock) and **Darcy** flow (fluid – water/melt), while **coupling** the two with **porosity** (void space filled with fluid).

$$\operatorname{div} \mathbf{u}_s = -\frac{p - p_f}{\eta(1 - \Phi)},$$

Stokes, mass conservation + coupling term

$$\operatorname{div} \mathbf{u}_f = \frac{p - p_f}{\eta(1 - \Phi)},$$

Darcy, mass conservation + coupling term

$$\operatorname{div} \boldsymbol{\sigma} = \rho g \mathbf{e}_d,$$

Stokes, momentum conservation

$$\mathbf{u}_f = -\kappa \left(\nabla p_f + \rho_f g \mathbf{e}_d \right),$$

Darcy, momentum conservation

$$\boldsymbol{\sigma} = \boldsymbol{\sigma}(\mathbf{u}_s, p) = 2\mu_s \dot{\boldsymbol{\epsilon}}(\mathbf{u}_s) - p \mathbf{I},$$

$$\rho = \rho_f \Phi + \rho_s(1 - \Phi),$$

$$\eta = \frac{\mu_s}{\Phi} C \left(1 + \frac{1}{2} \left(\frac{1}{R} - 1 \right) \left(1 + \tanh \left(-\frac{p - p_f}{\lambda} \right) \right) \right),$$

$$\kappa = \kappa_0 \mu_f^{-1} \left(\frac{\Phi}{\Phi_0} \right)^n,$$

Constitutive relations:

- Nonlinear in bulk viscosity (etas)

Reuber et al., 2020 (submitted)

Hydro-mechanics

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coupling term

$$\operatorname{div} \boldsymbol{\sigma} = \rho g \mathbf{e}_d,$$

ation

$$\mathbf{u}_f = -\kappa \left(\nabla p - \frac{1}{\eta} \nabla p_f \right)$$

tion

Assumptions in this work:

- Permeability scales with porosity
- We neglect advection of porosity (solid velocity) in the channel limit

$$\boldsymbol{\sigma} = \boldsymbol{\sigma}(\mathbf{u}_s, p) = 2\mu_s \dot{\boldsymbol{\epsilon}}(\mathbf{u}_s) - p\mathbf{I},$$

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Inversion

We seek to **minimize** the misfit between **observed** and **calculated vertical fluid velocity**:

$$\mathcal{J}(\Phi) := \frac{1}{2} \int_{\Omega_{\text{obs}}} \left((\mathbf{u}_f - \mathbf{u}_f^{\text{obs}}) \cdot \boldsymbol{\nu} \right)^2 d\mathbf{x},$$

By taking **variations** of the **Lagrangian function** (e.g. Tröltzsch 2010) one can calculate the **derivative of a cost function with respect to the design parameter**:

$$\mathcal{G}(\Phi) = \mathbf{v}_s \cdot (\rho_f - \rho_s) g \mathbf{e}_d - n \kappa_0 \mu_f^{-1} \frac{\Phi^{n-1}}{\Phi_0^n} \mathbf{v}_f \cdot (\rho_f g \mathbf{e}_d + \nabla p_f) - \frac{(q + q_f)(p - p_f)}{\eta \Phi (1 - \Phi)^2},$$

where, $\text{div } \boldsymbol{\sigma}(\mathbf{v}_s, q) = \mathbf{0}$,

$$-\text{div } \mathbf{v}_s = (q + q_f) \frac{\eta + p \psi(\Phi, p, p_f)}{\eta^2 (1 - \Phi)},$$

$$\mathbf{v}_f = \begin{cases} -\nabla q_f + \mathbf{e}_d (\mathbf{u}_f - \mathbf{u}_f^{\text{obs}}) \cdot \mathbf{e}_d & \text{on } \Omega_{\text{obs}} \\ -\nabla q_f & \text{on } \Omega \setminus \Omega_{\text{obs}} \end{cases},$$

$$-\text{div } (\kappa \mathbf{v}_f) = (q + q_f) \frac{\eta - p_f \psi(\Phi, p, p_f)}{\eta^2 (1 - \Phi)},$$

is the **adjoint system** of equations

Reuber et al., 2020 (submitted)

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By taking **variational derivative** of a cost function

$$\mathcal{G}(\Phi) = \mathbf{v}_s \cdot (\mathbf{u}_f - \mathbf{u}_f^{\text{obs}}),$$

where, $\text{div } \sigma = -\text{div } \mathbf{v}_s = (q + q_f) \frac{\eta - p_f \psi(\Phi, p, p_f)}{\eta^2(1 - \Phi)},$

Adjoint method:

- Pointwise gradient with only one additional linear solve

Gradient descent:

$$\Phi^{(i+1)} = \Phi^{(i)} - \alpha \mathcal{G}(\Phi^{(i)}),$$

can calculate the

$$\frac{(q + q_f)(p - p_f)}{\eta \Phi (1 - \Phi)^2},$$

$$\mathbf{v}_f = \begin{cases} -\nabla q_f + \mathbf{e}_d (\mathbf{u}_f - \mathbf{u}_f^{\text{obs}}) \cdot \mathbf{e}_d & \text{on } \Omega_{\text{obs}} \\ -\nabla q_f & \text{on } \Omega \setminus \Omega_{\text{obs}} \end{cases},$$

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Reuber et al., 2020 (submitted)

PT solver

Frankel, 1950, Otter, 1966, [...], Räss et al., 2018

On the example of nonlinear diffusion:

$$\frac{\partial H}{\partial t} = -\frac{\partial}{\partial x} \left(H^n \frac{\partial H}{\partial x} \right)$$

$$\frac{\partial H}{\partial \tau} = -\frac{\partial}{\partial x} \left(H^n \frac{\partial H}{\partial x} \right)$$

→ pseudo timestepping to converge

$$\partial \tau = \frac{1}{\frac{1}{dx H^{n2.1}} + \frac{1}{dt}}$$

Matlab main function example:

```
1 function [H,evol,res] = ForwardSolve(H,n,dx,dt,niter,epsi)
2     evol=[];
3     i = 1;
4     for iter=1:niter
5         Dx      = H.^n;
6         Dxa     = (Dx(1:end-1)+Dx(2:end))/2;
7         dtau    = 1/(1/(min(dx*dx)/max(Dx)/2.1) + 1/dt);
8         qx      = -Dxa.*diff(H)/dx;
9         dHdt    = -diff(qx)/dx;
10        H(2:end-1) = H(2:end-1) + dtau*dHdt;
11        H([1,end]) = [1,0.5];
12    end
13 end
14
```

PT solver

Frankel, 1950, Otter, 1966, [...], Räss et al., 2018

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$$\frac{\partial H}{\partial \tau} = -\frac{\partial}{\partial x} \left(H^n \frac{\partial H}{\partial x} \right)$$

$$\partial \tau = \frac{\partial t}{dx^n}$$

- + Matrix free method (explicit FD)
→ perfect for accelerator architecture (GPUs)
- + Solver for any (non) linear system
- + Easy to develop
- Many more iterations (versus many more FLOPS)
- How to find the pseudo-timestep?

`dx,dt,niter,epsi)`

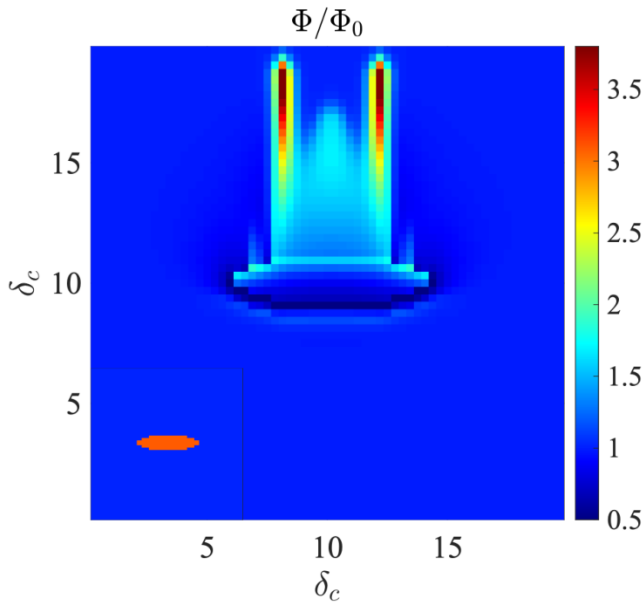
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/2.1) + 1/dt);`

Matlab main fu

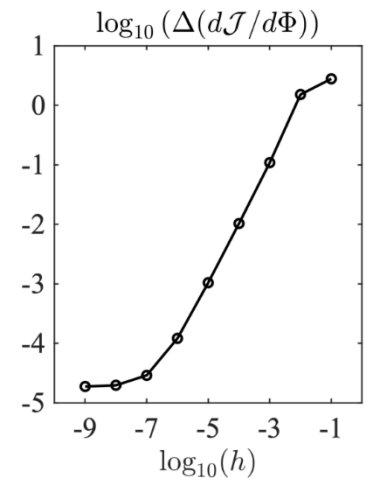
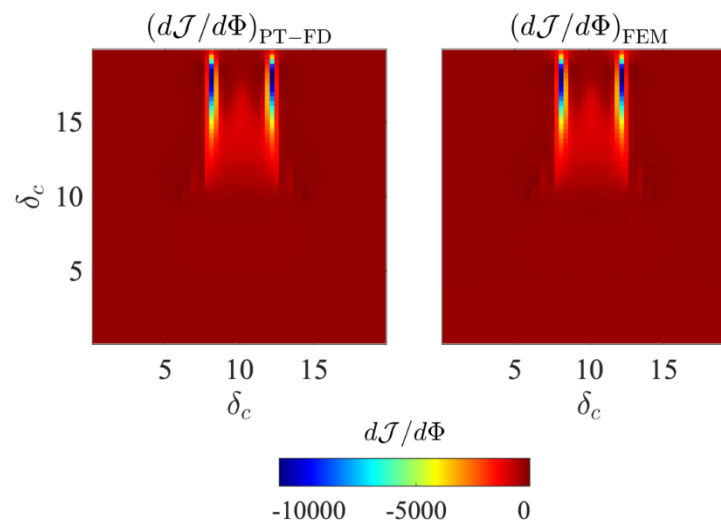
```
9      dHdt      = -diff(qx)/dx;  
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```


Benchmark

Comparison of the gradient with **direct finite difference**, a FEniCS **FEM** python code and the **pseudo-transient** matlab code (64 x 64 elements/cells/parameters):



The synthetic porosity field is created by solving the transient forward problem for some time

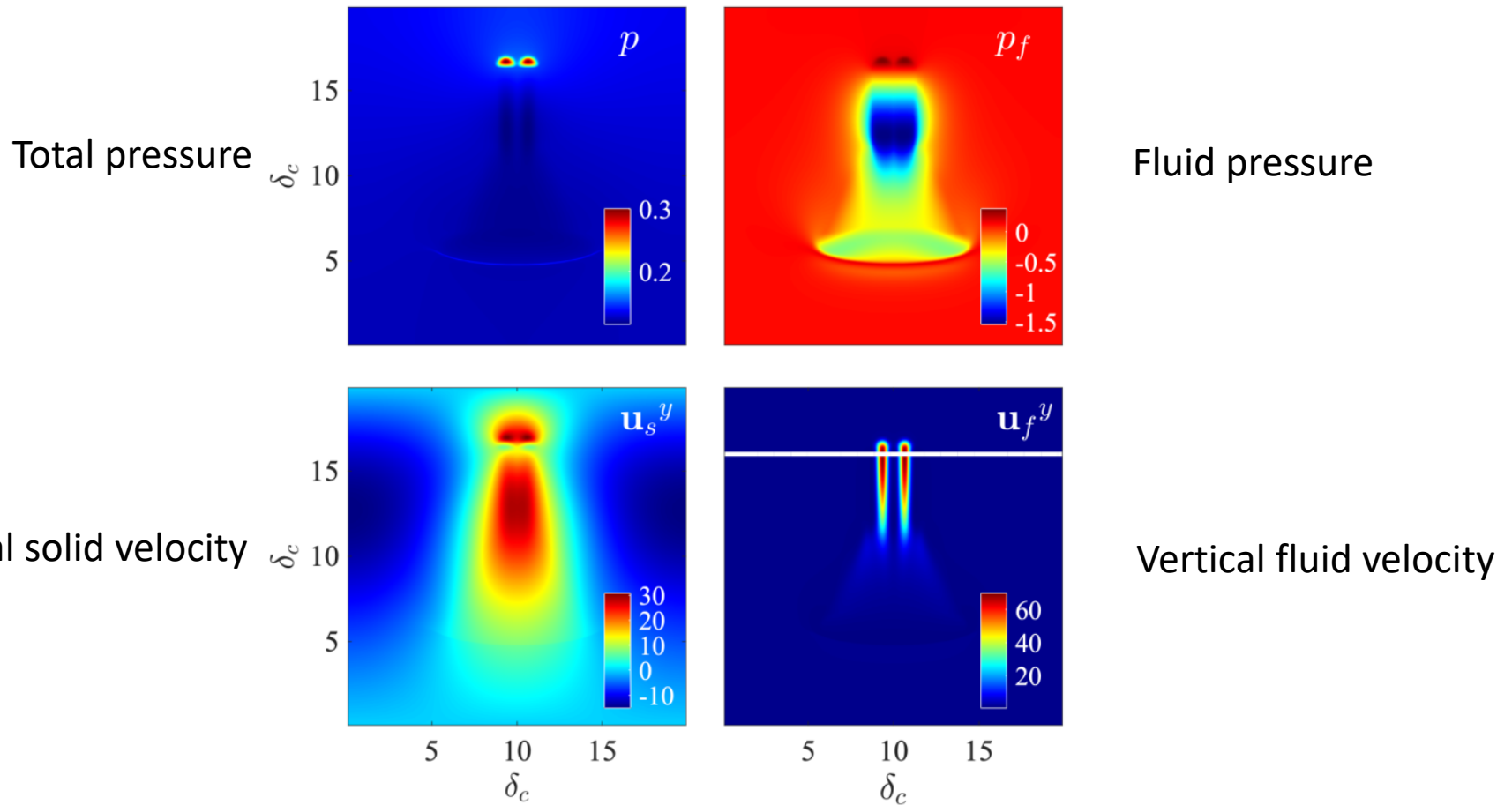


Very good agreement among the methods

Reuber et al., 2020 (submitted)

Results

Forward problem result (1024 x 1024 cells) used as **synthetic measurements**:

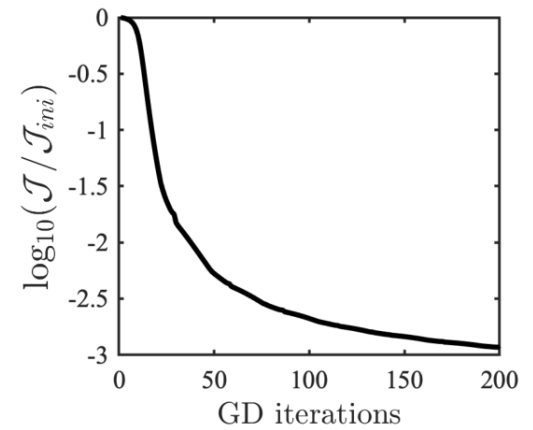
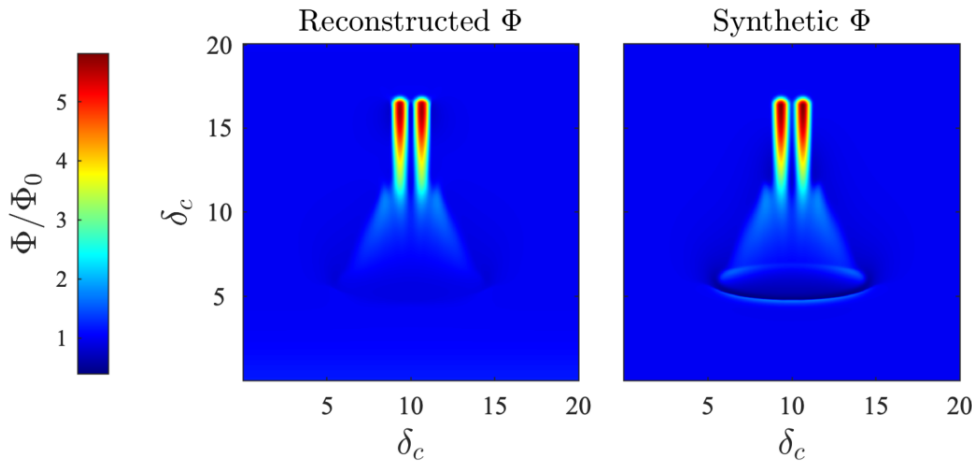


Reuber et al., 2020 (submitted)

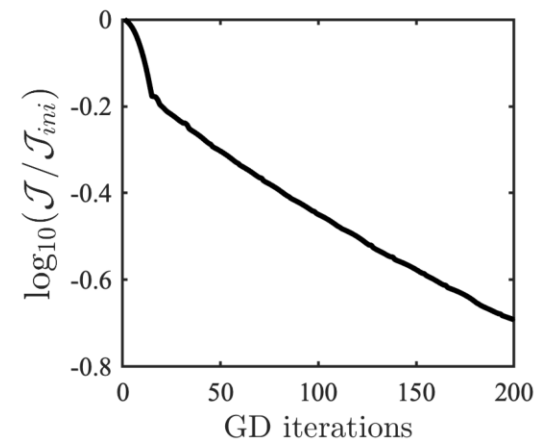
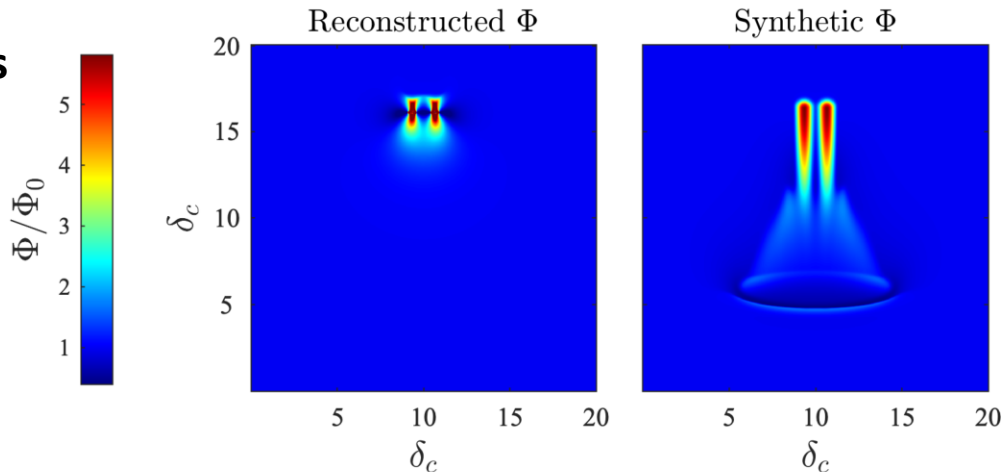
Results

We try to **recover** the synthetic porosity field from an initial **homogeneous porosity field** (1024 x 1024 cells/parameters):

Measurement is available everywhere:
Very good reconstruction

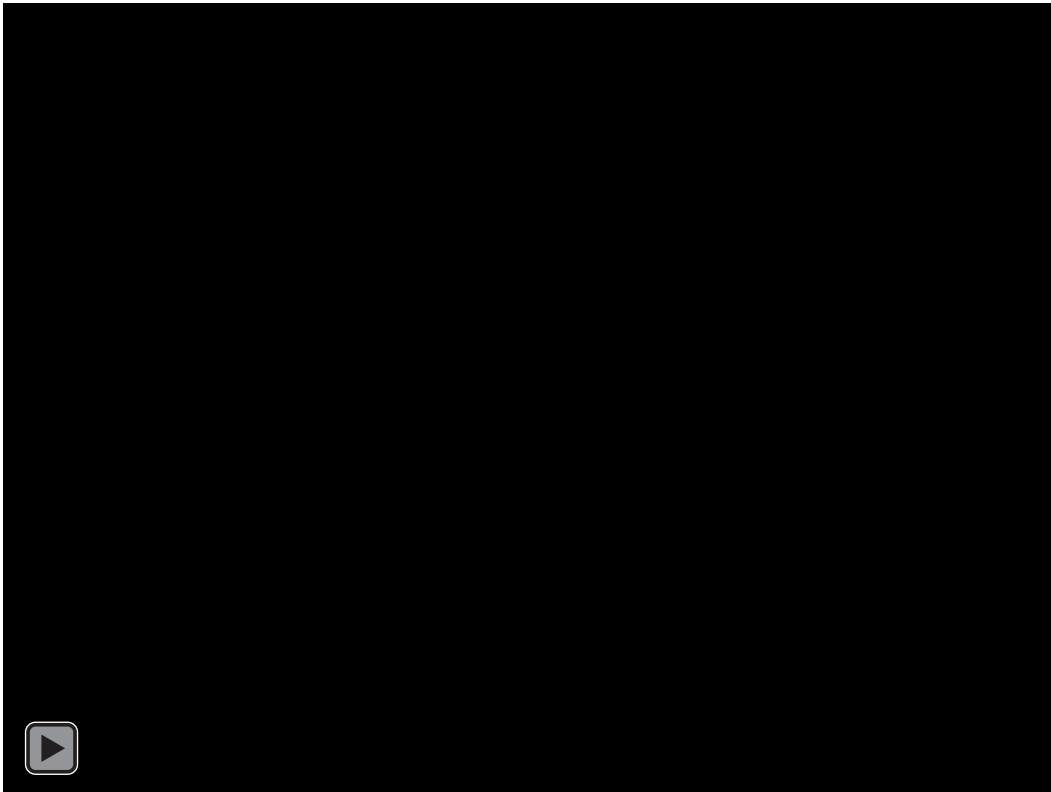


Measurement is available along a surface:
Very localized sensitivity/reconstruction



Outlook

1. Julia/CUDA MPI 3D pseudo-transient solver (ParallelStencil and ImplicitGlobalGrid Julia package by Sam Omlin) - **HPC**
2. Transient inverse problem - **improved reconstruction**
3. Regularization (H^1 Tikhonov) – **improved convergence**



2D slice through a 3D
porosity inversion
(256x256x256 grid
points) in Julia on GPU