

Efficient and robust open-source modelling platform for Energy Transition applications

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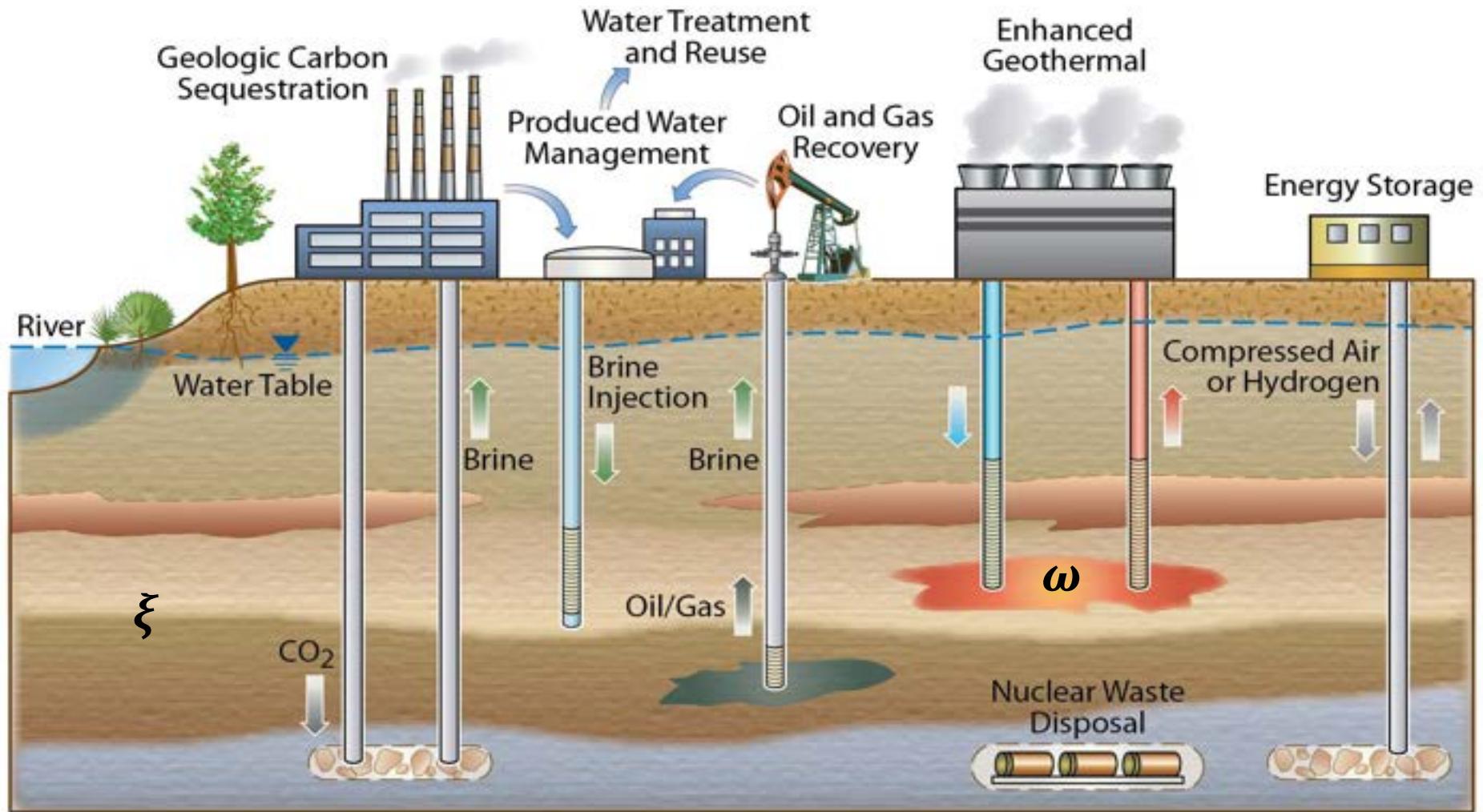


Modelling of energy transition problems

- Reservoir simulation is an important tool for an integration of different information from various sources and scales
- To describe industrial applications relevant to energy transition, we need to employ complex formulation of **multiphase** flow and **thermal-compositional reactive** transport in porous media
- Strong heterogeneities in subsurface properties demand uncertainty analysis with **ensembles** of simulation models

The robust and efficient reservoir simulation capabilities are wanted!

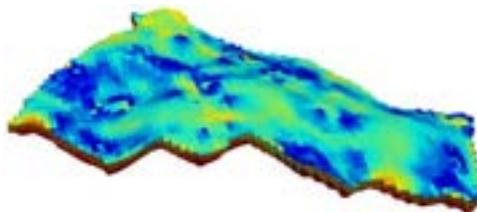
Multiphase thermal-compositional formulation



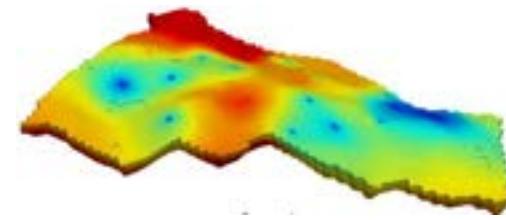
PDE for Energy Transition applications

$$g(\omega) = a_t(\omega, \xi) + \nabla \cdot b(\omega, \xi) + \Delta c(\omega, \xi) + d(\omega, \xi) = 0$$

$$\xi = \{G, \phi, K\}$$



$$\omega = \{p, T, z\}$$



Compressibility,
phase change
and convection

$$g(\omega) = \frac{\phi_0 V}{\Delta t} [\alpha(\omega) - \alpha(\omega_n)] + \sum_l v_t^l \beta(\omega) = \mathbf{0}$$

$$\alpha_c(\omega) = c(p) \sum_{j=1}^{n_p} x_{cj} \rho_j s_j,$$

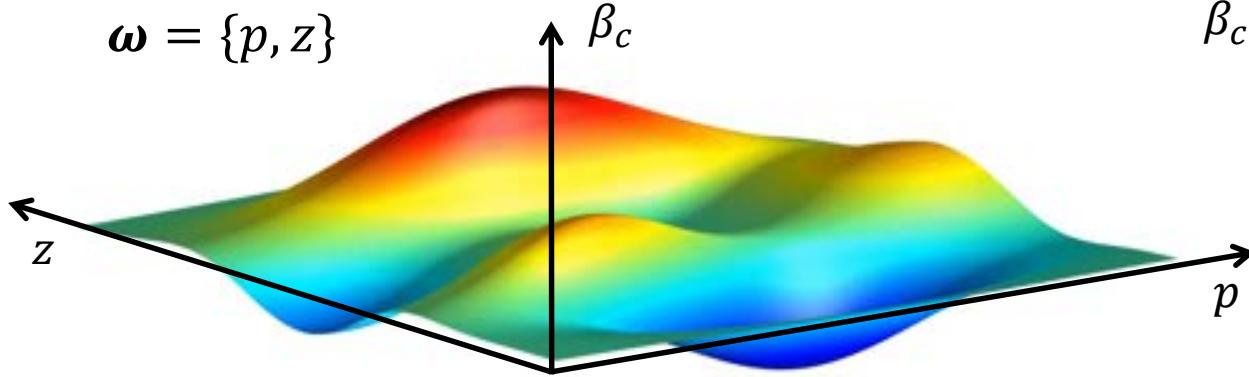
$$\beta_c(\omega) = \frac{1}{\Lambda} \sum_{j=1}^{n_p} x_{cj}^l \rho_j^l \frac{k_{rj}^l}{\mu_j^l}$$

$$g(\omega) = \frac{\phi_0 V}{\Delta t} [\alpha(\omega) - \alpha(\omega_n)] + \sum_l v_t^l \beta(\omega) + \sum_l \mathbf{D}^l (\chi^l - \chi) \gamma(\omega) + V \delta(\omega) = \mathbf{0}$$

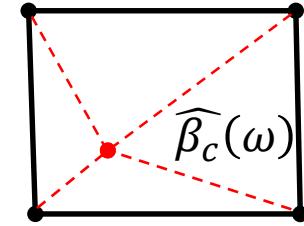
+ diffusion and
reactions

$$\gamma_c(\omega) = c(p) \sum_{j=1}^{n_p} x_{cj} \rho_j s_j d_{cj}, \quad \delta_c(\omega) = \sum_{k=1}^{n_k} v_{ck} r_k$$

Operator-Based Linearization



$$\beta_c(\omega^1) = \frac{1}{\Lambda} \sum_{j=1}^{n_p} x_{cj}^l \rho_j^l \frac{k_{rj}^l}{\mu_j^l} \quad \beta_c(\omega^2) = \frac{1}{\Lambda} \sum_{j=1}^{n_p} x_{cj}^l \rho_j^l \frac{k_{rj}^l}{\mu_j^l}$$

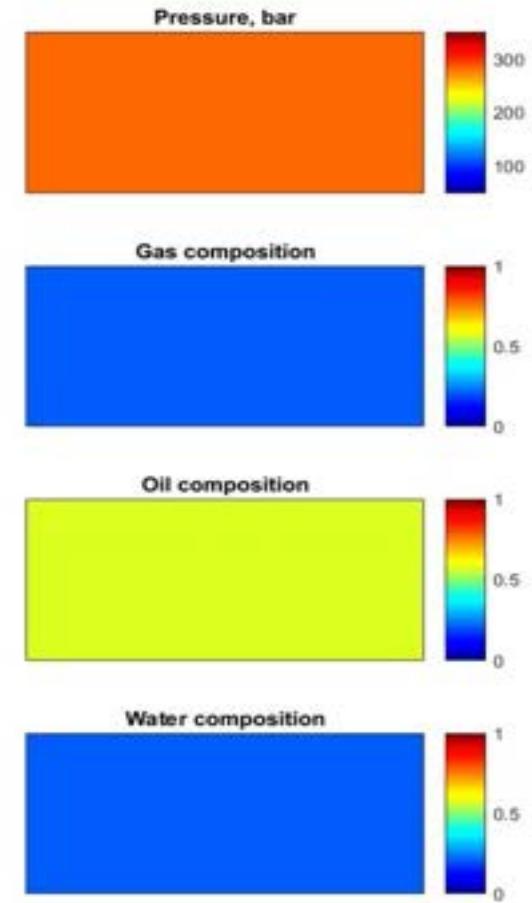
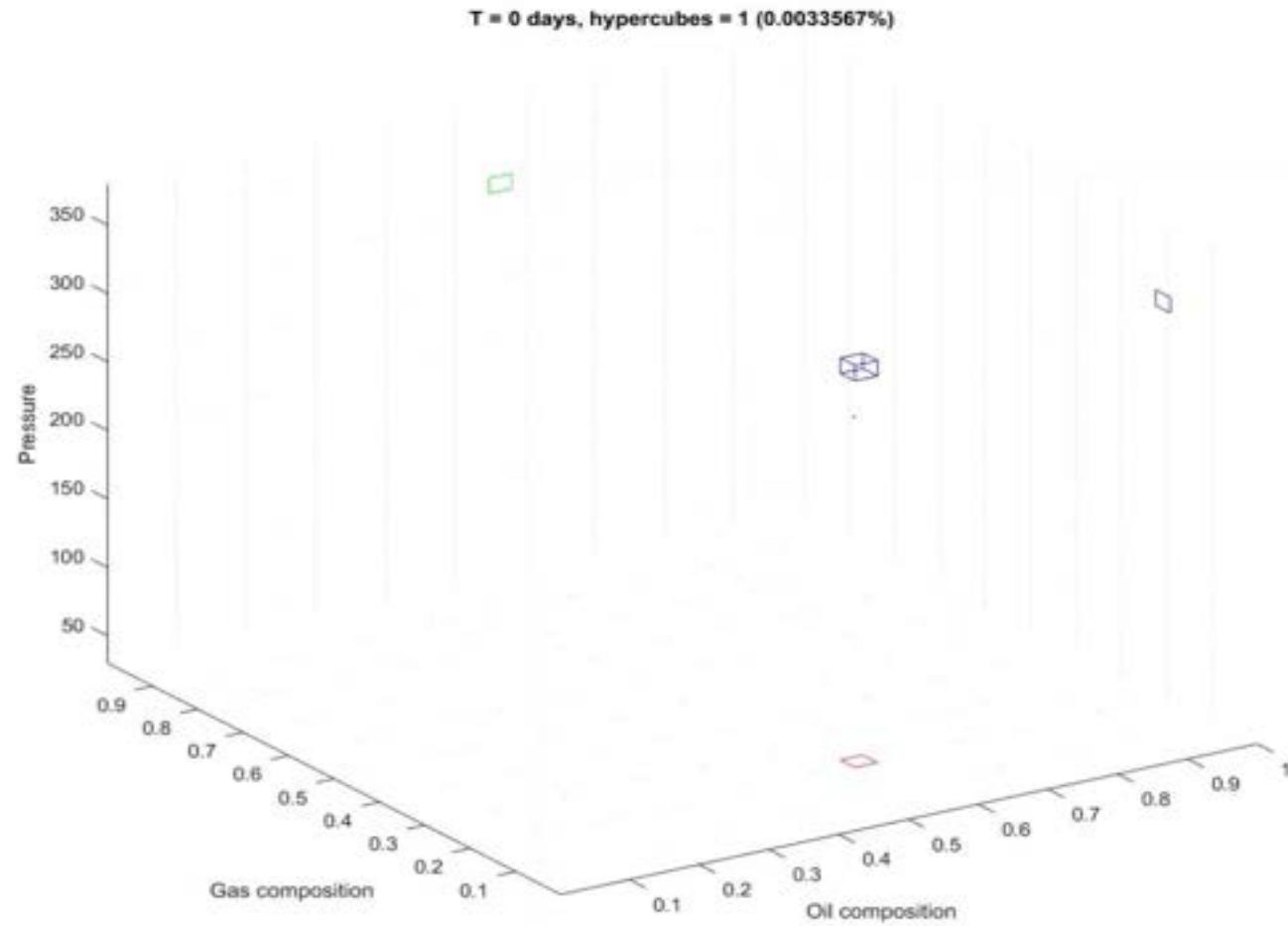


$$\beta_c(\omega^3) = \frac{1}{\Lambda} \sum_{j=1}^{n_p} x_{cj}^l \rho_j^l \frac{k_{rj}^l}{\mu_j^l} \quad \beta_c(\omega^4) = \frac{1}{\Lambda} \sum_{j=1}^{n_p} x_{cj}^l \rho_j^l \frac{k_{rj}^l}{\mu_j^l}$$

$$\frac{\partial g}{\partial \omega} = \frac{\partial \alpha}{\partial \omega} \bar{a}(\omega, \xi) + \frac{\partial \beta}{\partial \omega} \bar{b}(\omega, \xi) + \frac{\partial \gamma}{\partial \omega} \bar{c}(\omega, \xi) + \frac{\partial \delta}{\partial \omega} \bar{d}(\omega, \xi) + \bar{f}(\omega, \xi)$$

$$|\widehat{\beta}_c - \beta_c| \leq cA^2 \sup_{\omega} |\nabla^2 \beta_c|$$

Adaptive parametrization



Generic thermal-compositional formulation

$$\frac{\partial}{\partial t} \int_{\Omega} M^c d\Omega + \int_{\Gamma} \mathbf{F}^c \cdot \mathbf{n} d\Gamma = \int_{\Omega} Q^c d\Omega$$

Mass conservation

$$M^c = \phi \sum_{j=1}^{n_p} x_{cj} \rho_j s_j \quad c = 1, \dots, n_c,$$

$$\mathbf{F}^c = \sum_{j=1}^{n_p} x_{cj} \rho_j \mathbf{u}_j + s_j \rho_j \mathbf{J}_{cj}, \quad c = 1, \dots, n_c$$

$$Q^c = \sum_{k=1}^{n_k} v_{ck} r_k, \quad c = 1, \dots, n_c$$

Energy conservation

$$M^{n_c+1} = \phi \sum_{j=1}^{n_p} \rho_j s_j U_j + (1 - \phi) U_r$$

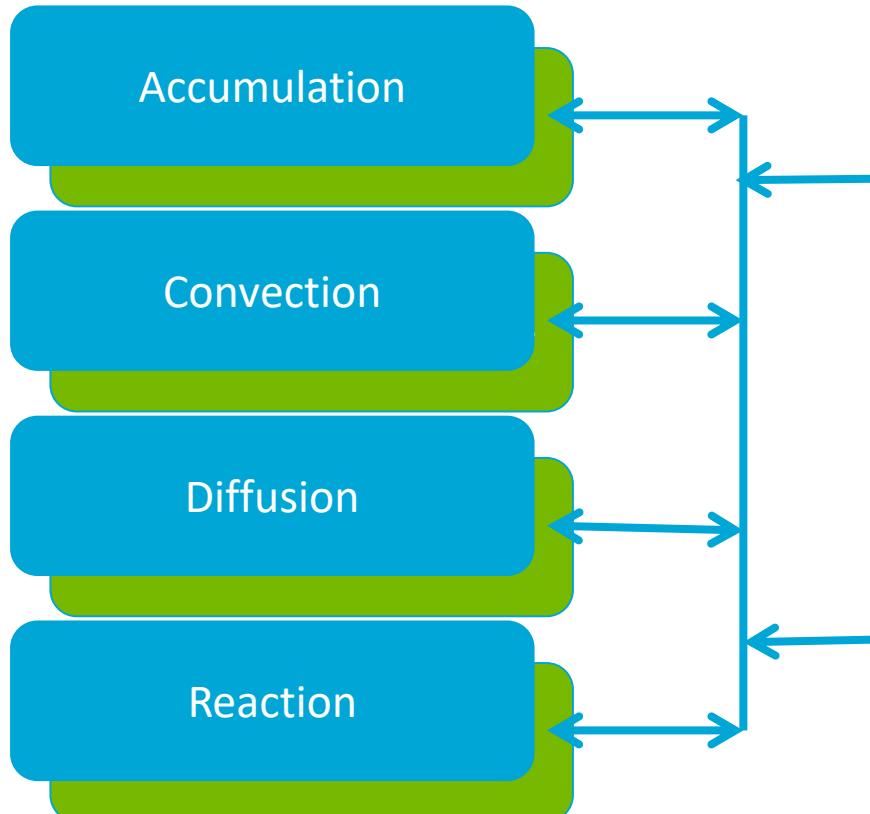
$$\mathbf{F}^{n_c+1} = \sum_{j=1}^{n_p} h_j \rho_j \mathbf{u}_j + \kappa \nabla T$$

$$Q^{n_c+1} = \sum_{k=1}^{n_k} v_{ek} r_{ek}$$

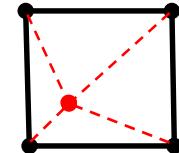
Delft Advanced Research Terra Simulator



DARTS-engine: C++ & CUDA

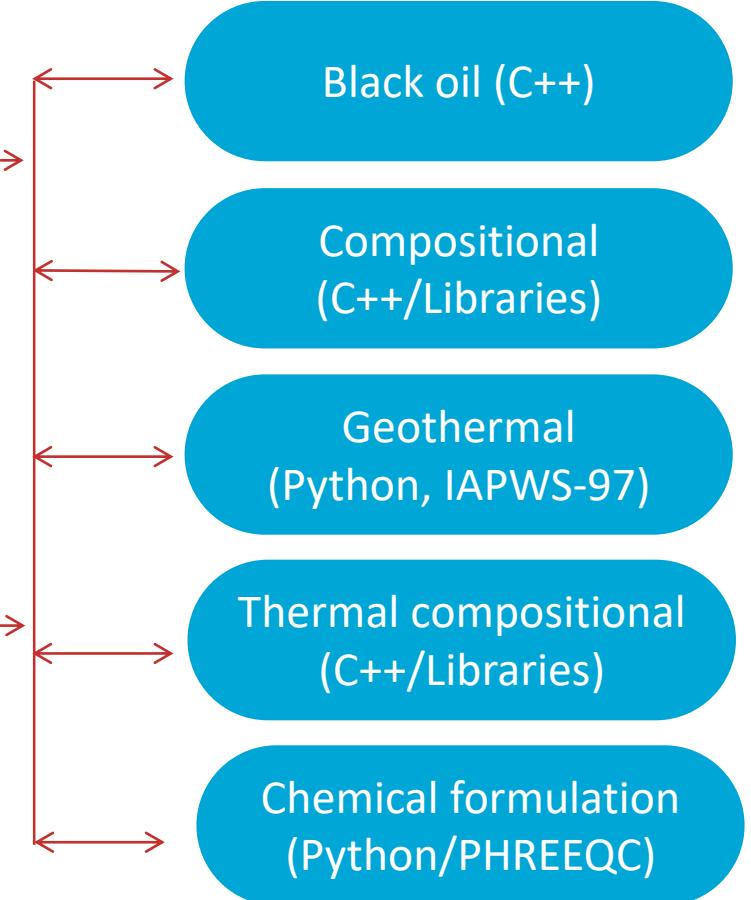


static_itor
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adaptive_itor
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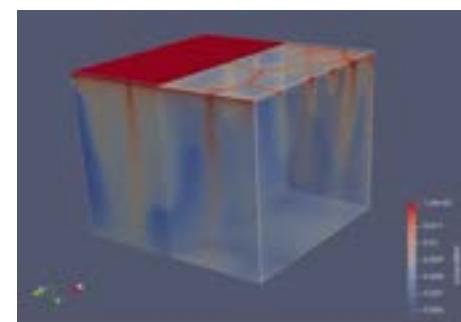
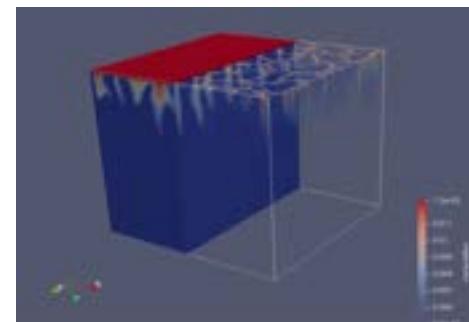
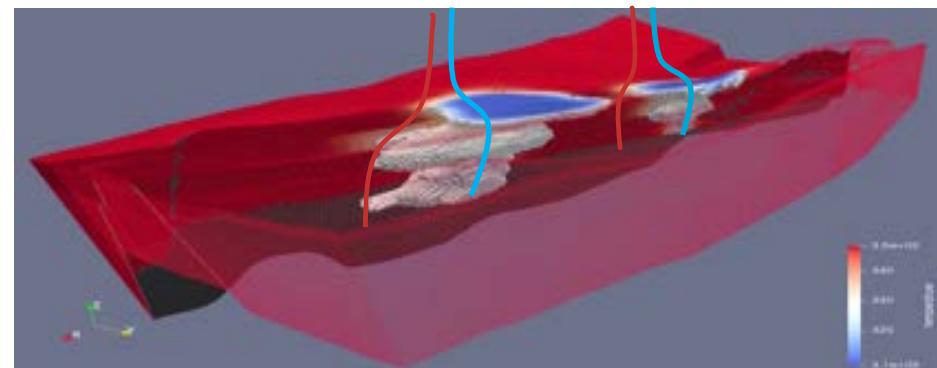
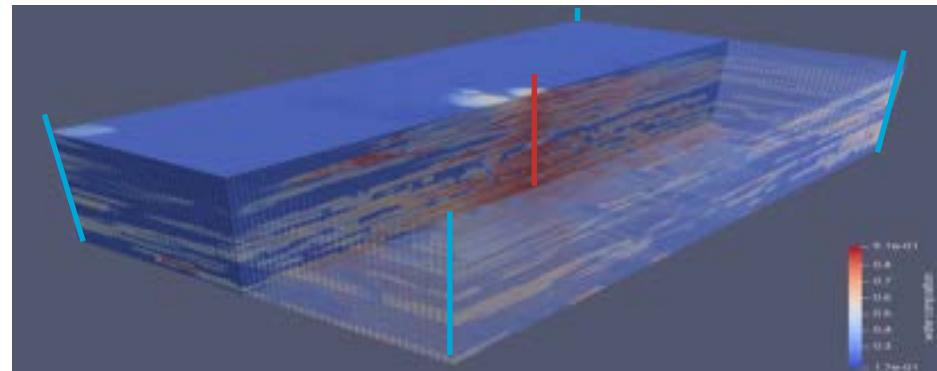
DARTS-physics: hybrid





Delft Advanced Research Terra Simulator

- CO₂ injection for EOR
 - 1.1M active blocks, 5.5 years
 - 4 unknowns per block
 - CPU*: **20 min**, GPU: **3.5 min**
- Geothermal model
 - 3.2M active blocks, 100 years
 - 2 unknowns per block
 - CPU*: **49 min**, GPU: **8 min**
- CO₂ sequestration
 - 1.0M active blocks, 3000 years
 - 2 unknowns per block
 - CPU*: **3.8 hours**, GPU: **55 min**



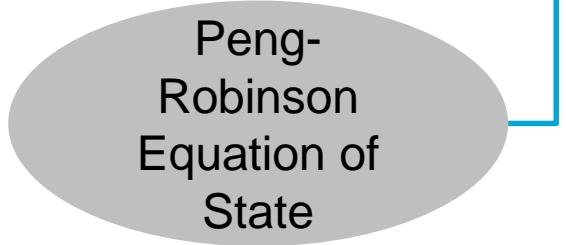
Thermodynamic model for CO₂-gas-brine-oil systems

- Gas component model: ignore effects of water component presence in gas phase

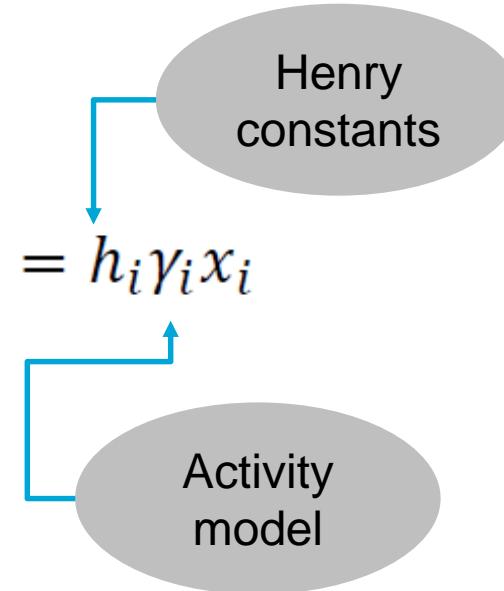
$$f_i^g = p\Phi_i y_i$$

$$f_i^g = f_i^l$$

$$f_i^l = h_i \gamma_i x_i$$



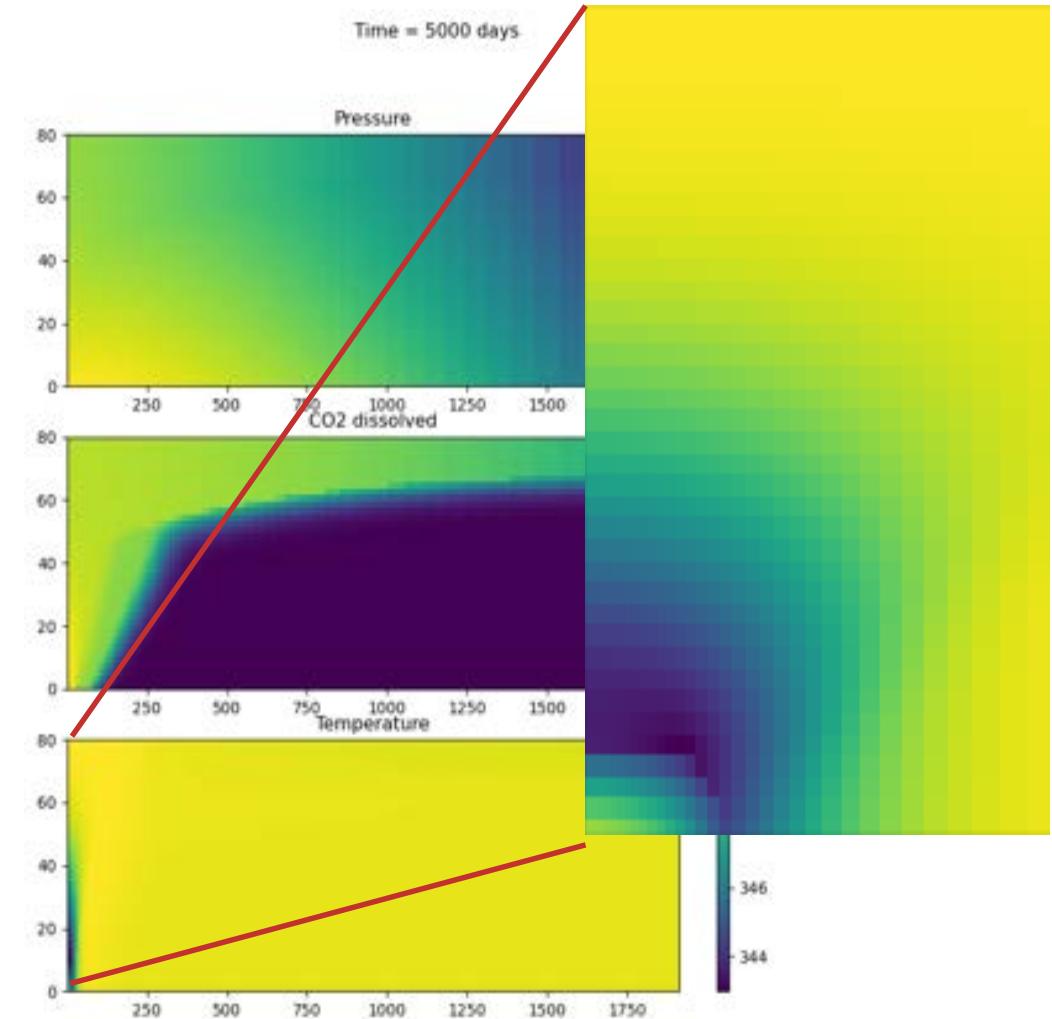
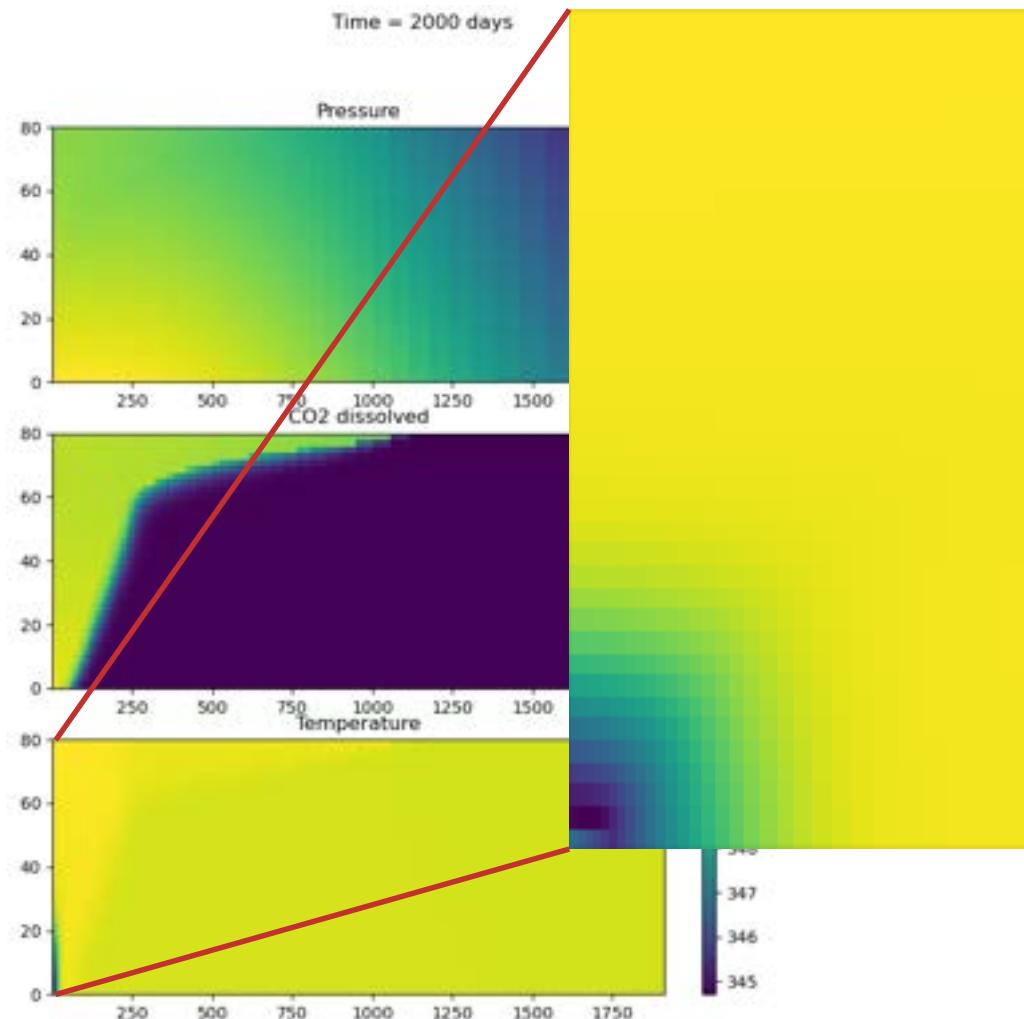
$$K_i = \frac{y_i}{x_i} = \frac{h_i \gamma_i}{p\Phi_i}$$



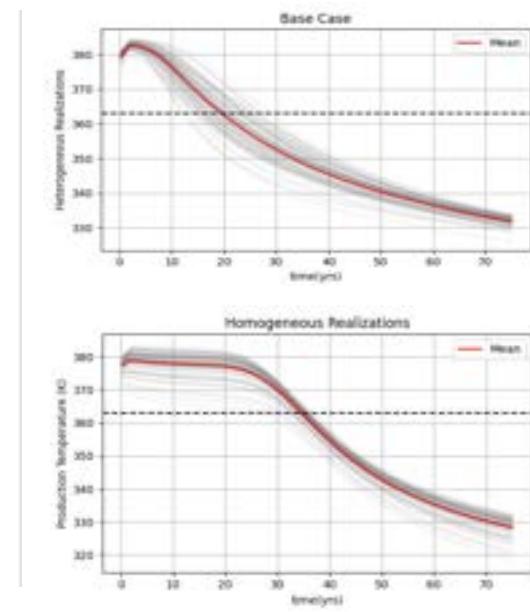
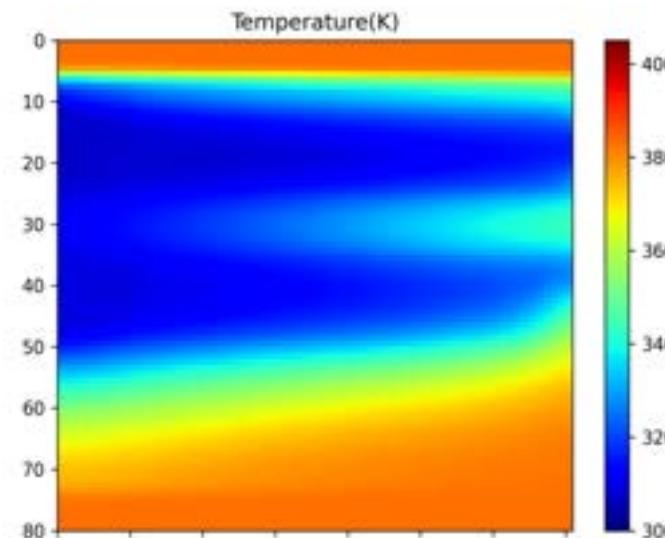
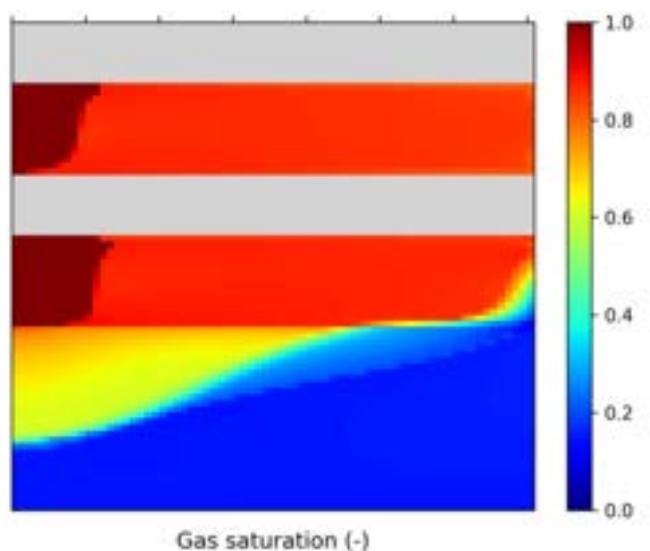
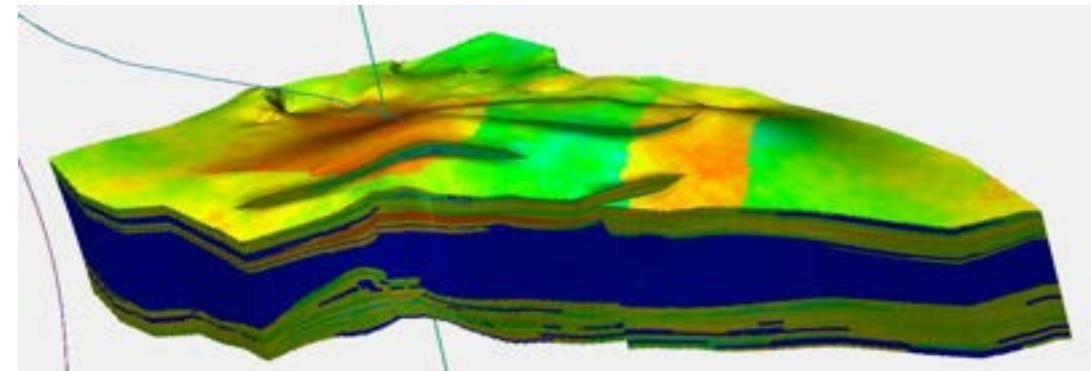
- Water component model: activity of water can be approximated by its mole fraction in liquid phase

$$K_{H_2O} = \frac{y_{H_2O}}{x_{H_2O}} = \frac{K_{H_2O}^0}{\Phi_{H_2O} p} \exp \left[\frac{(p - 1)V_{H_2O}}{RT} \right]$$

Joule-Thomson cooling for mixture CO₂-C₁-brine



CO₂ Plum Geothermal in depleted gas field



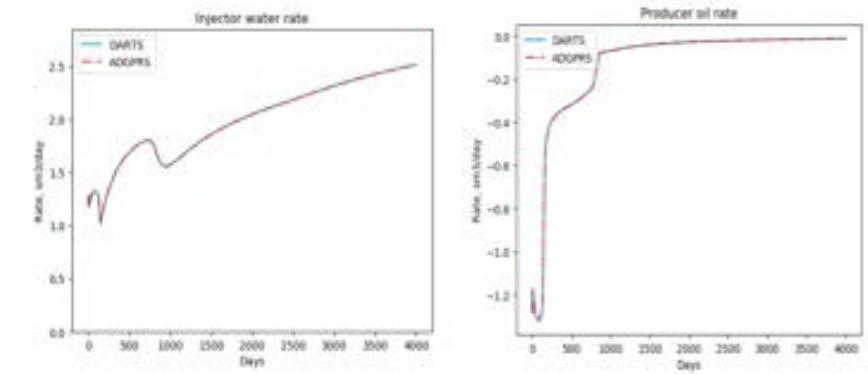
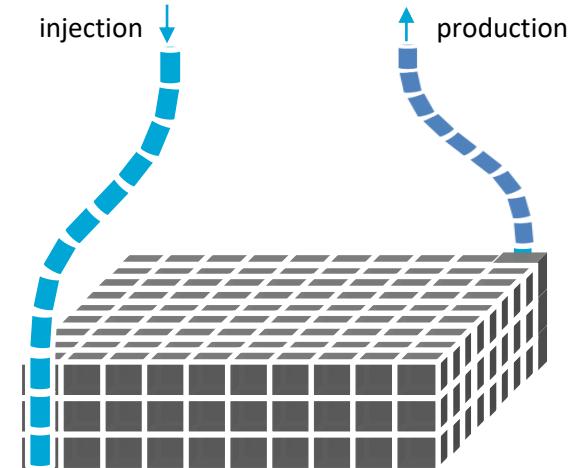
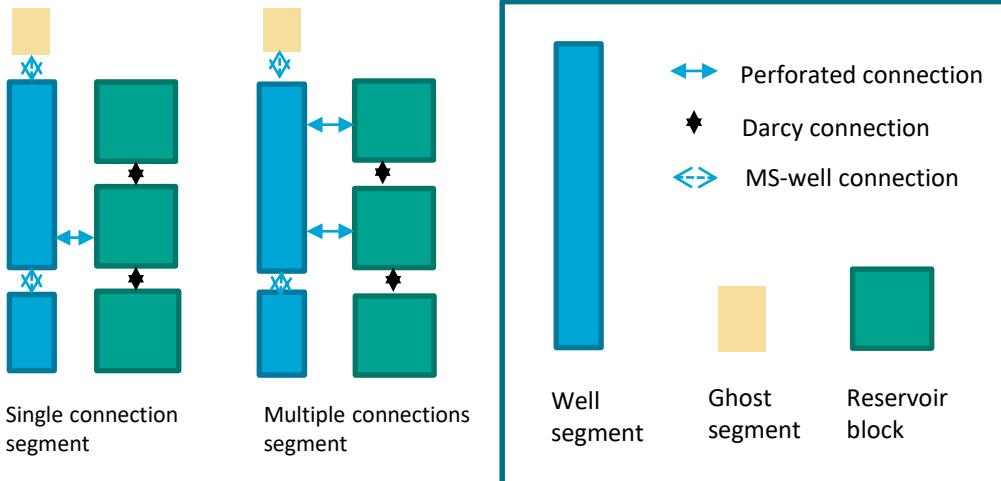
Multi-segmented well model

- Multi-physics in wellbore (thermal multiphase multi-component reactive flow and transport with heat losses)
- Complex well network (deviated, multilateral, annulus, etc.)

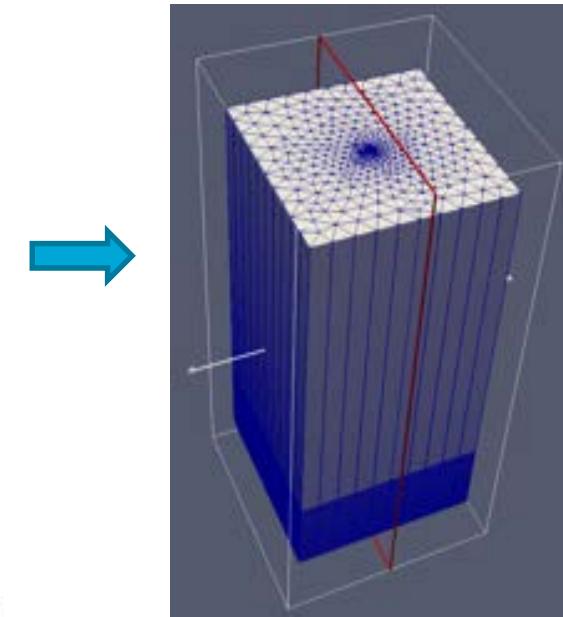
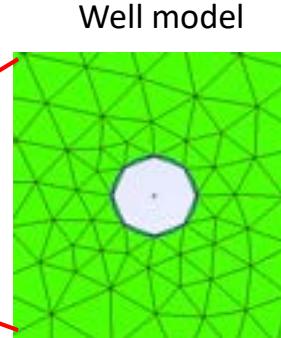
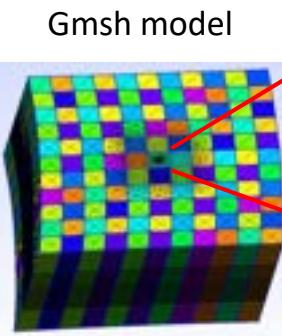
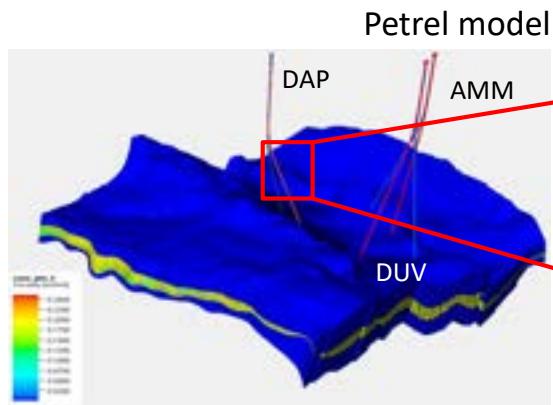
$$\frac{\phi_0 V}{\Delta t} [\alpha(\omega) - \alpha(\omega_n)] + \sum_l v_t^l(\xi, \omega) \beta(\omega) = g(\omega)$$

$$\Delta p - \theta_h(\omega, \xi) - \theta_f(\omega, \xi, v_t) - \theta_a(\omega, \xi, v_t) = g^w(\omega)$$

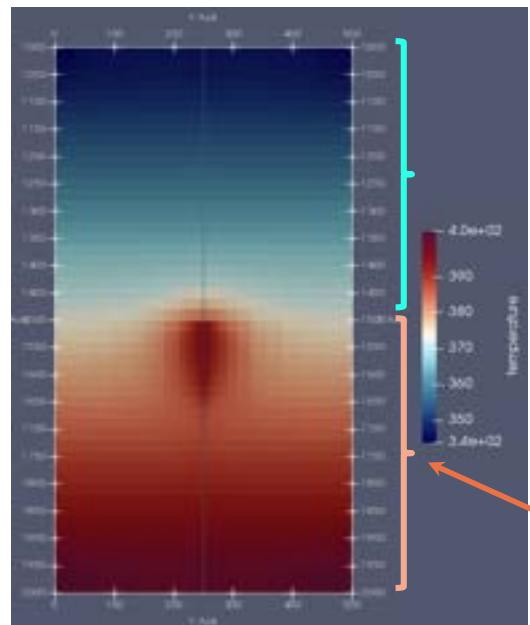
Hydrostatic losses Friction losses Acceleration losses



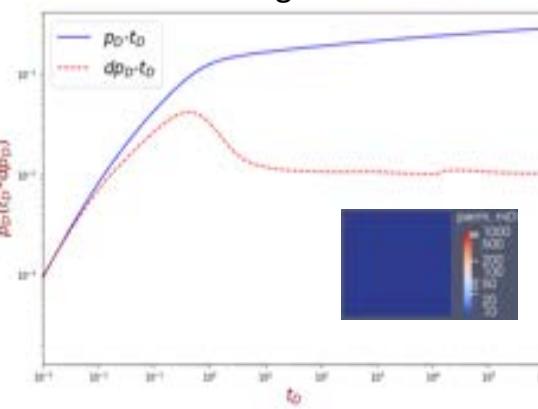
Near-well modeling



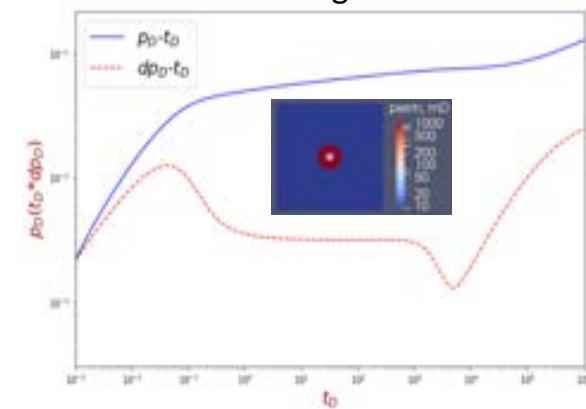
Effect of insulation



Homogeneous

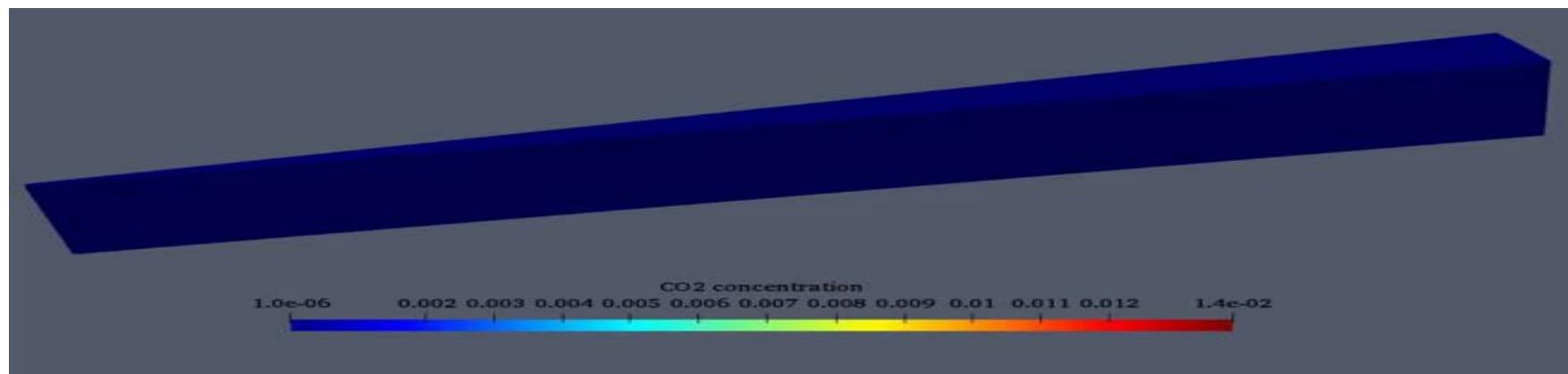
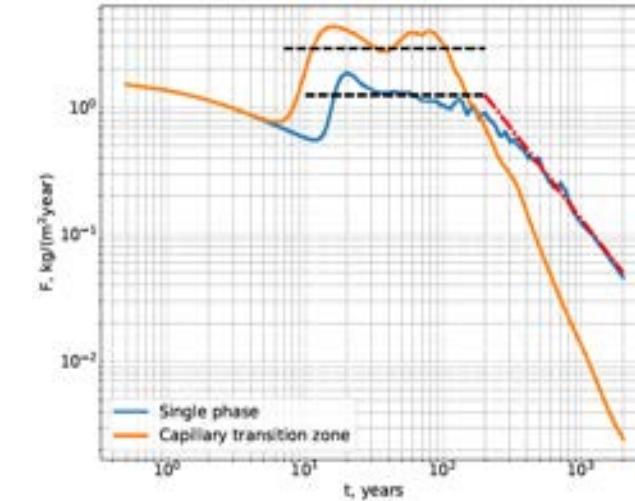
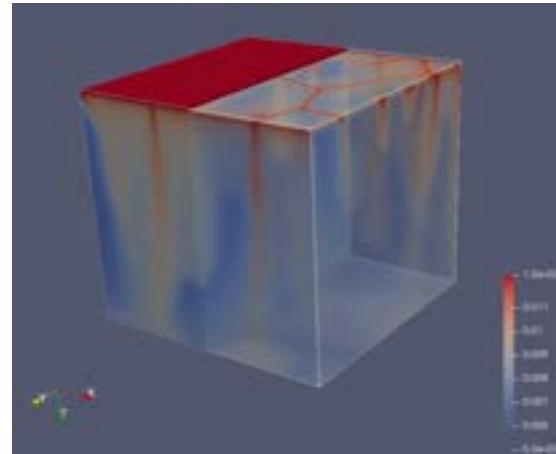
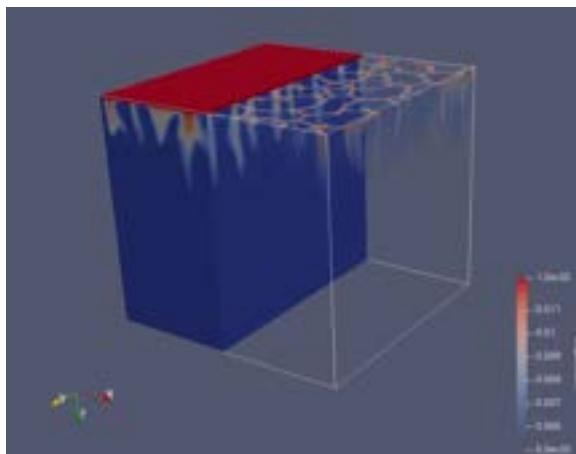


Heterogeneous

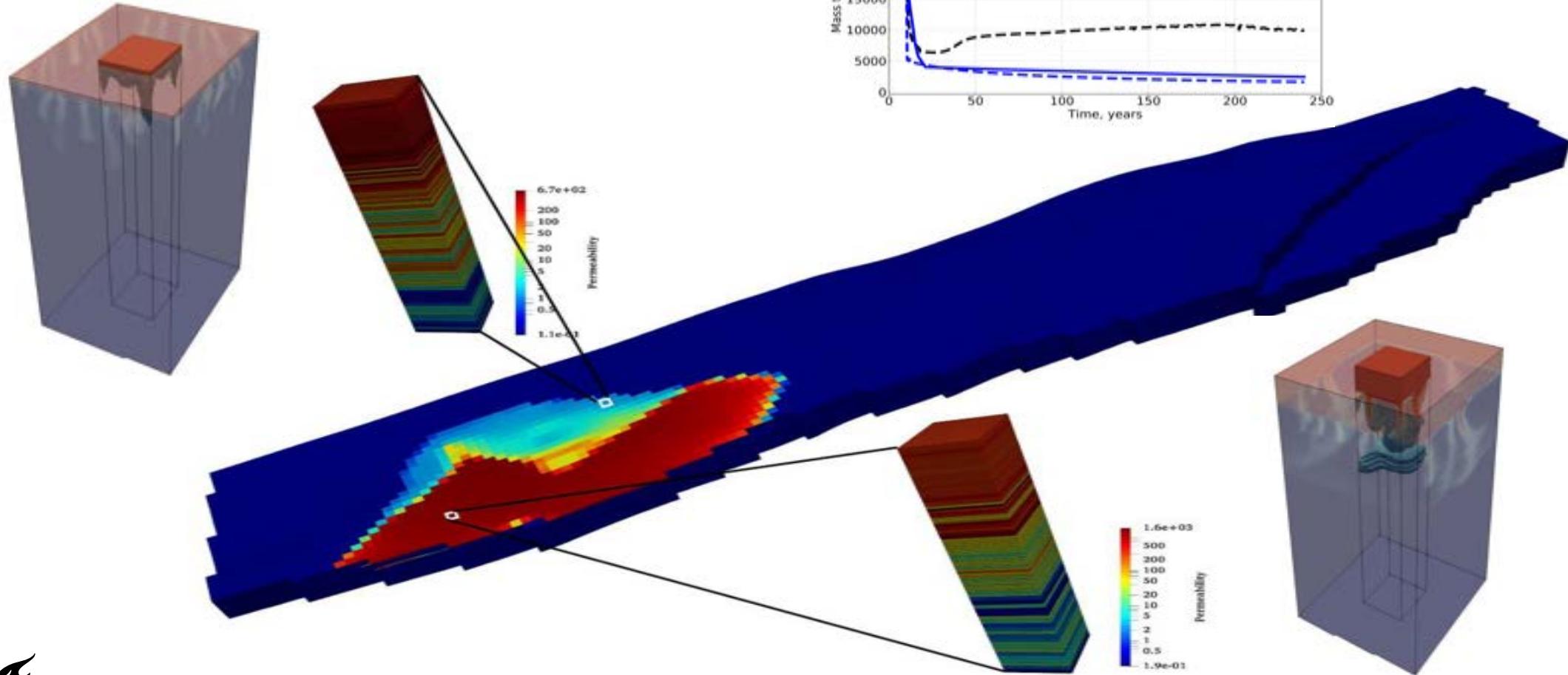


Non-insulated interval

Enhanced CO₂ dissolution



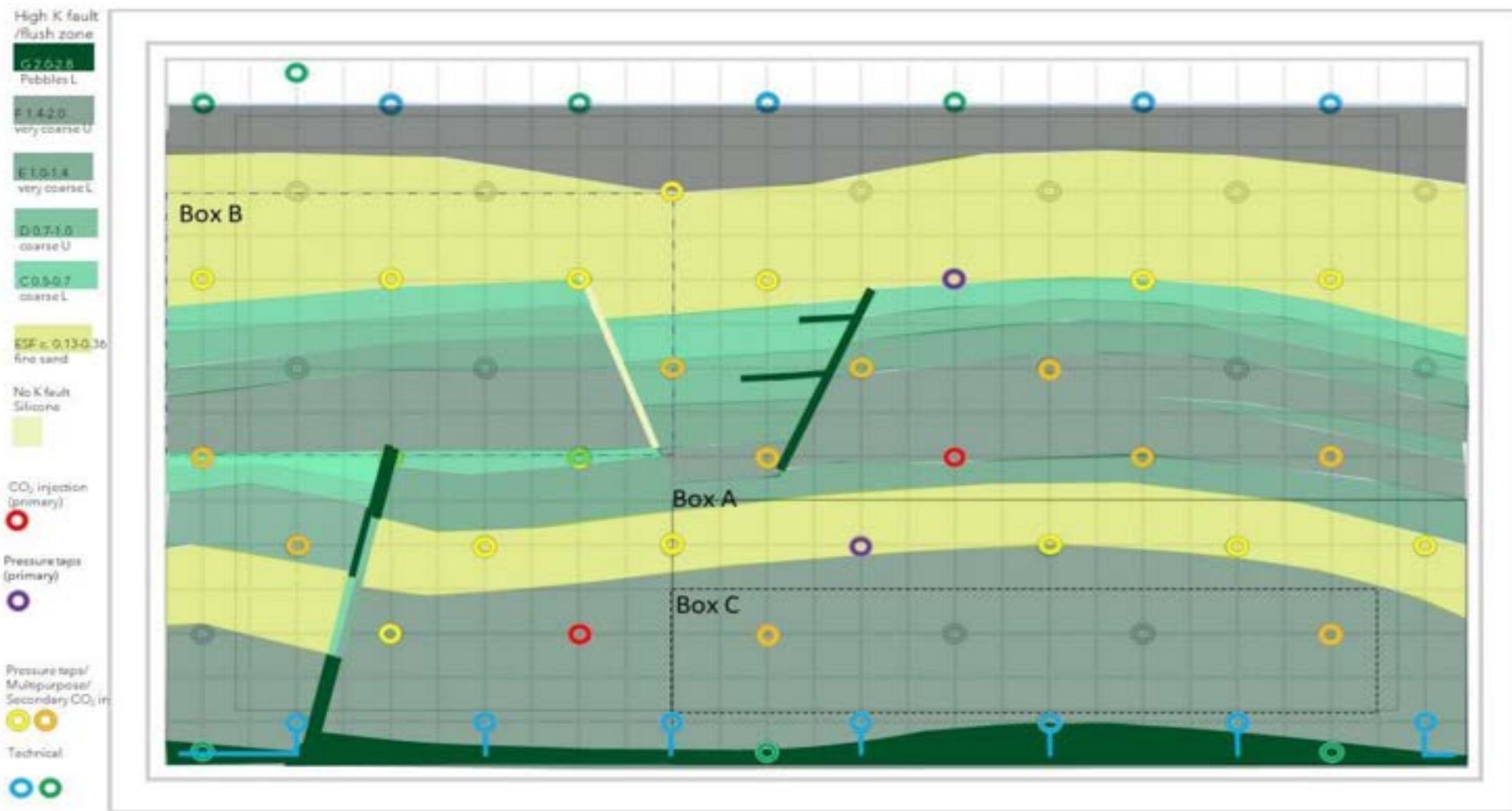
Heterogeneous reservoir



FluidFlower benchmark (University of Bergen)

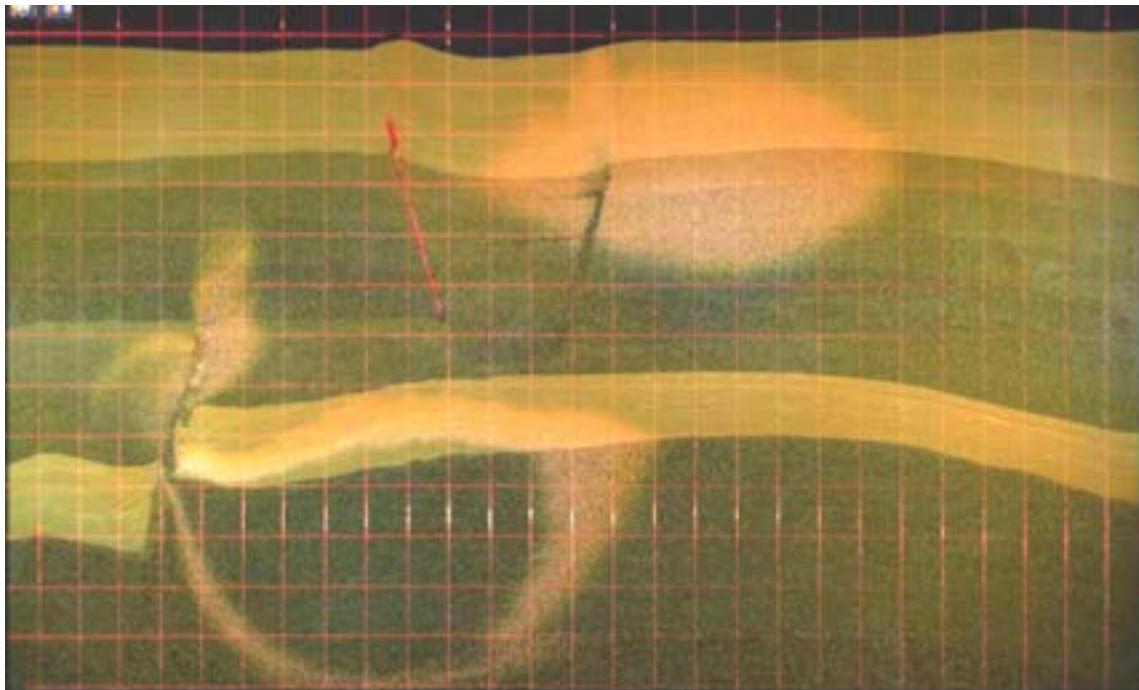


Model description

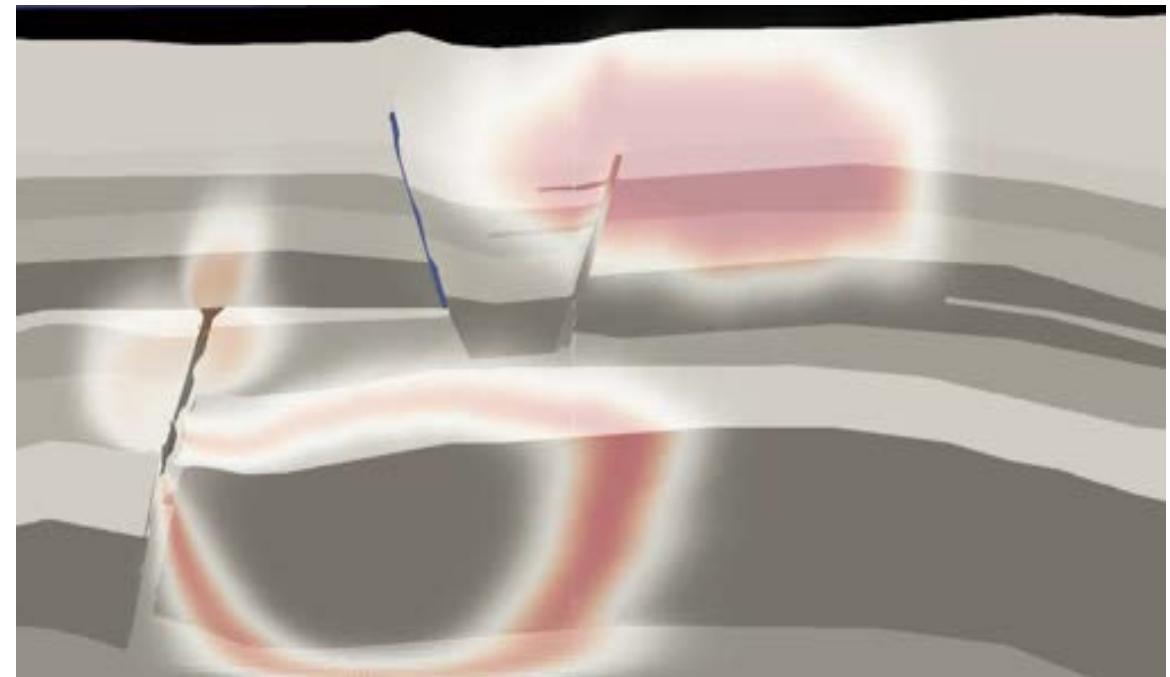


History matching using RML (single realization)

Tracer observations (high resolution images)

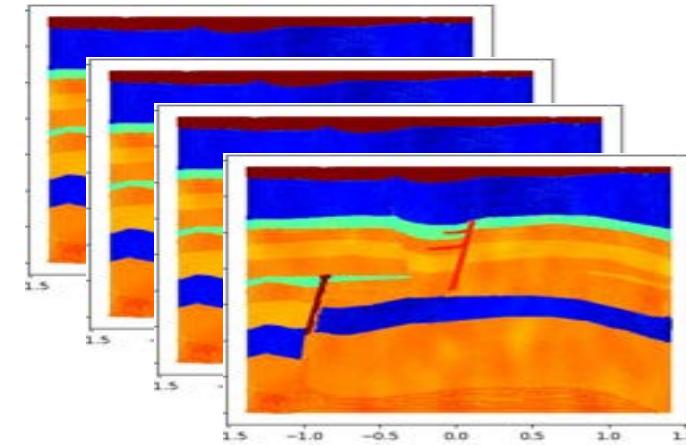
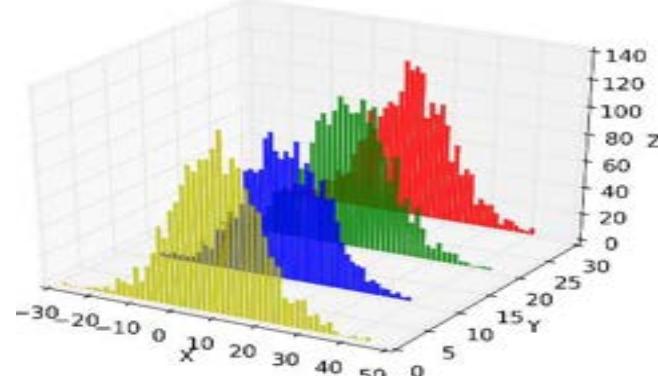


Inversed model (RML, 18278 forward runs for 100 priors)

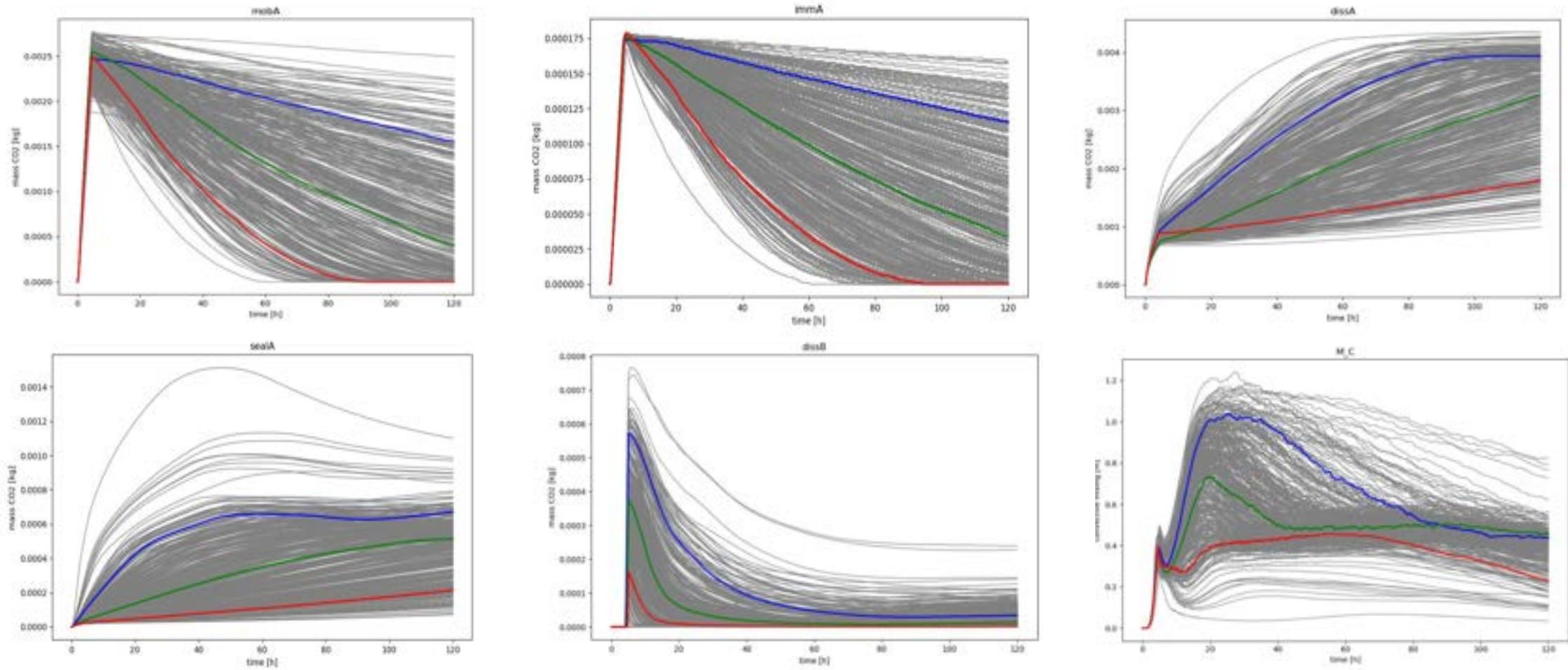


Uncertainty Quantification

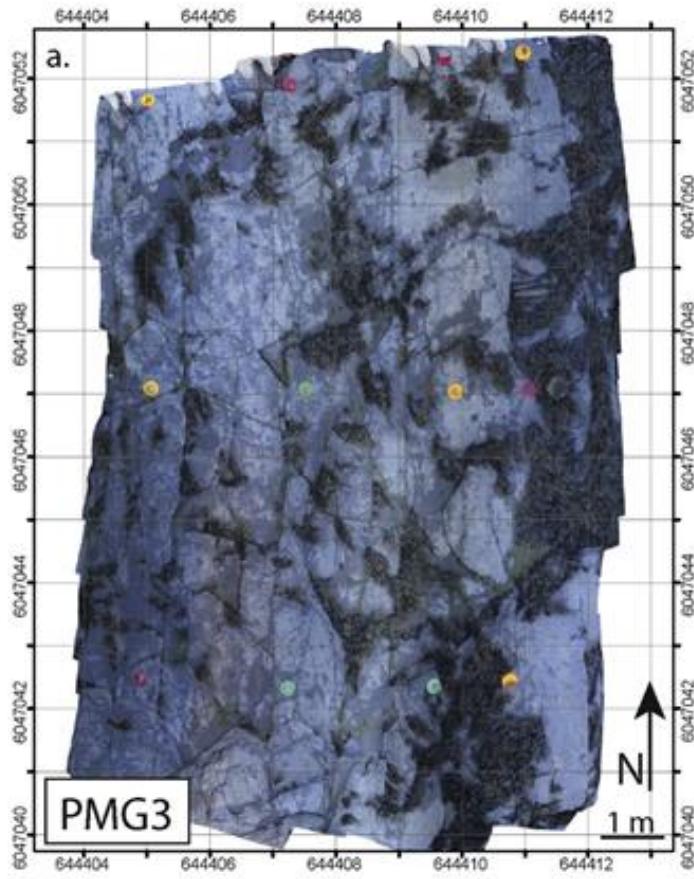
- Temperature: $23^{\circ}\text{C} \pm 2$, normal distribution
- Diffusion: $2 \cdot 10^{-10} - 2 \cdot 10^{-9}$, log-normal distribution
- Corey: base values with standard deviation from 5 to 50% for different parameters, normal distribution
- Models: 100 history matched permeability maps



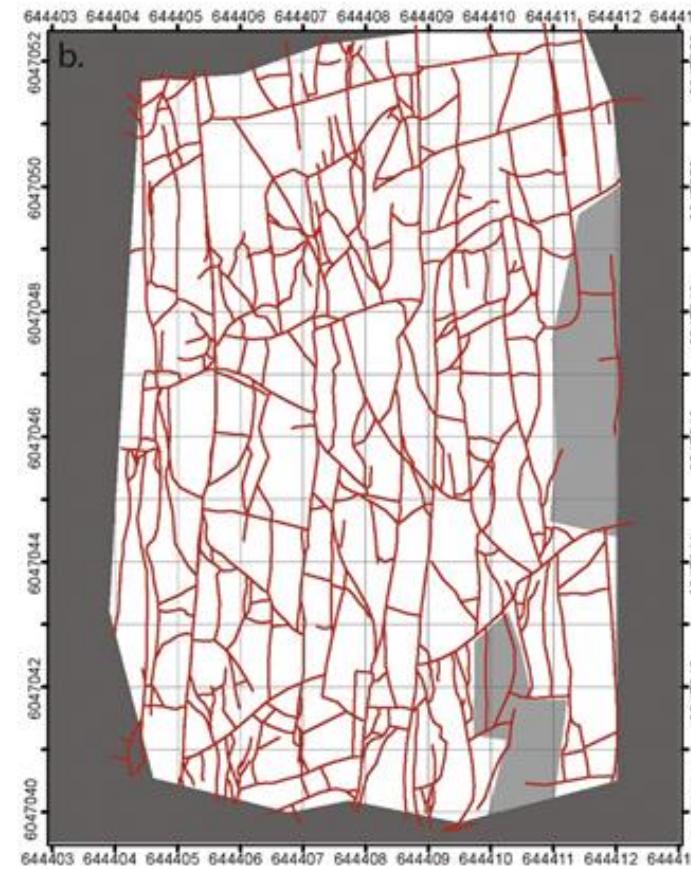
UQ investigation (400 runs)



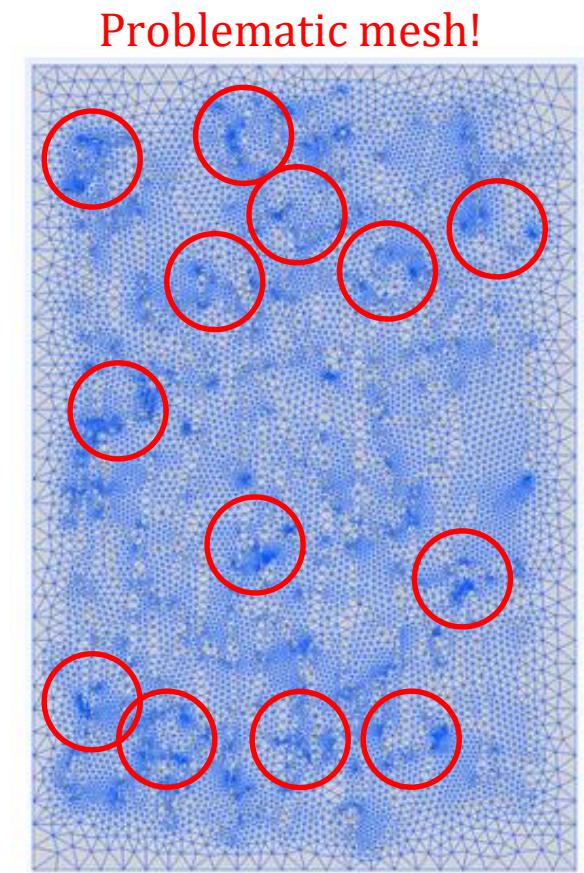
Modeling fractures



Acquire data



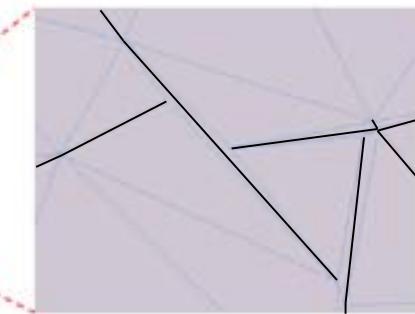
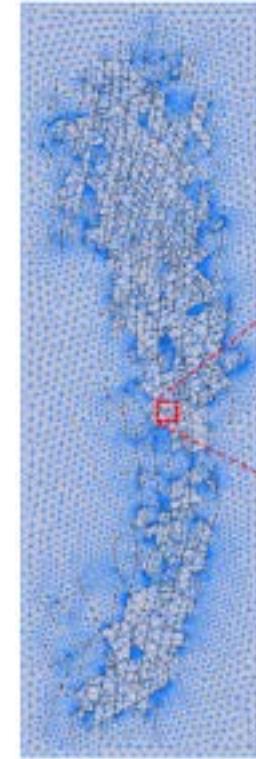
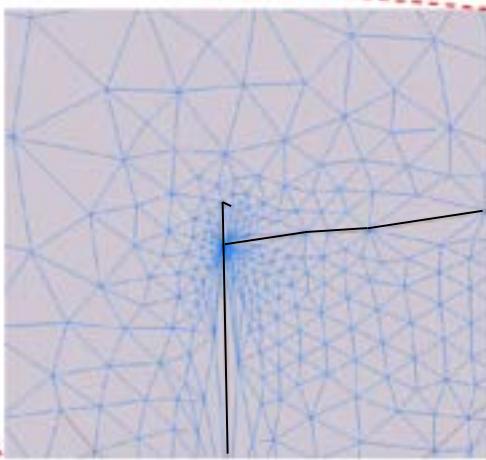
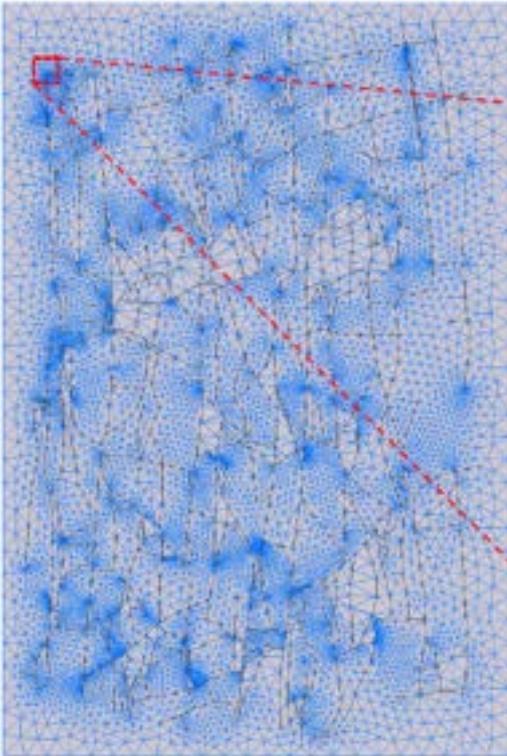
Interpret data



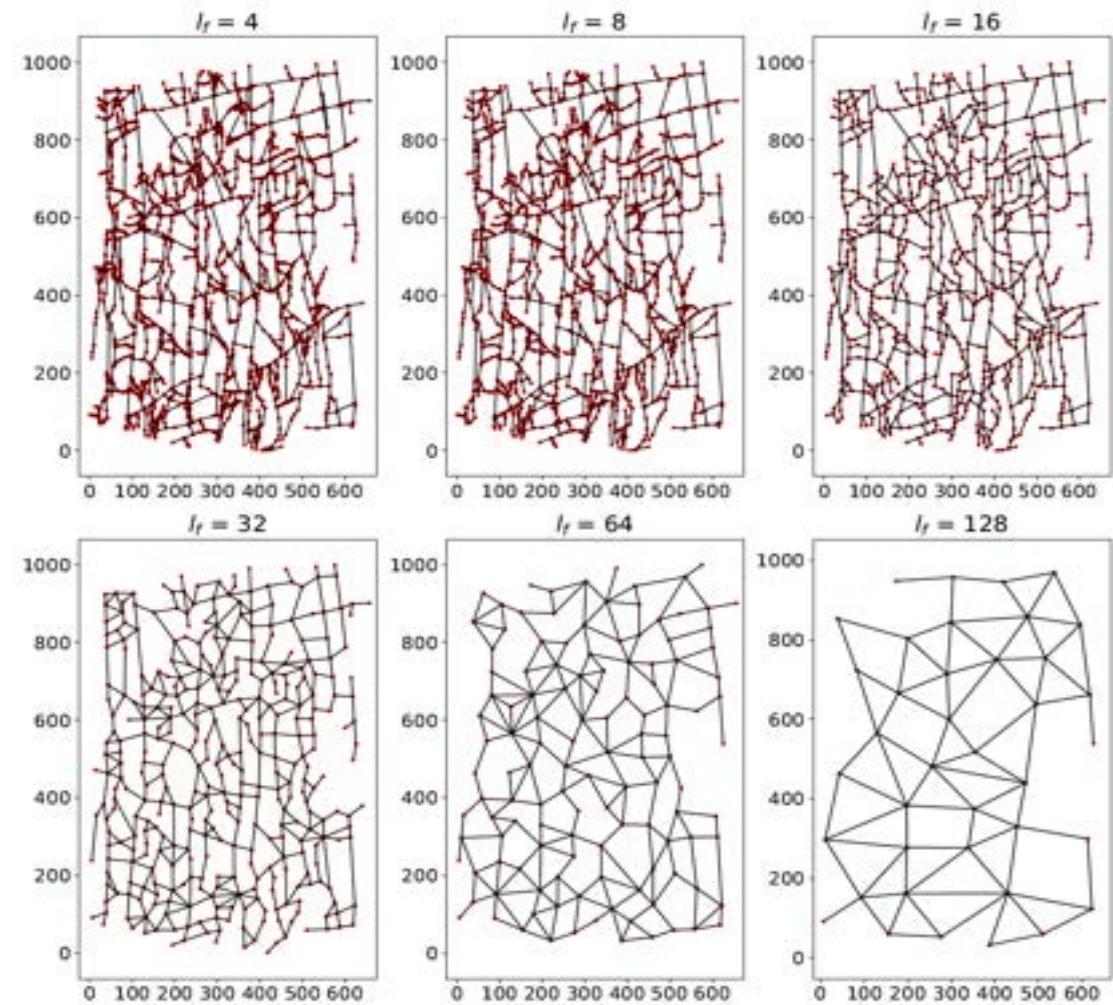
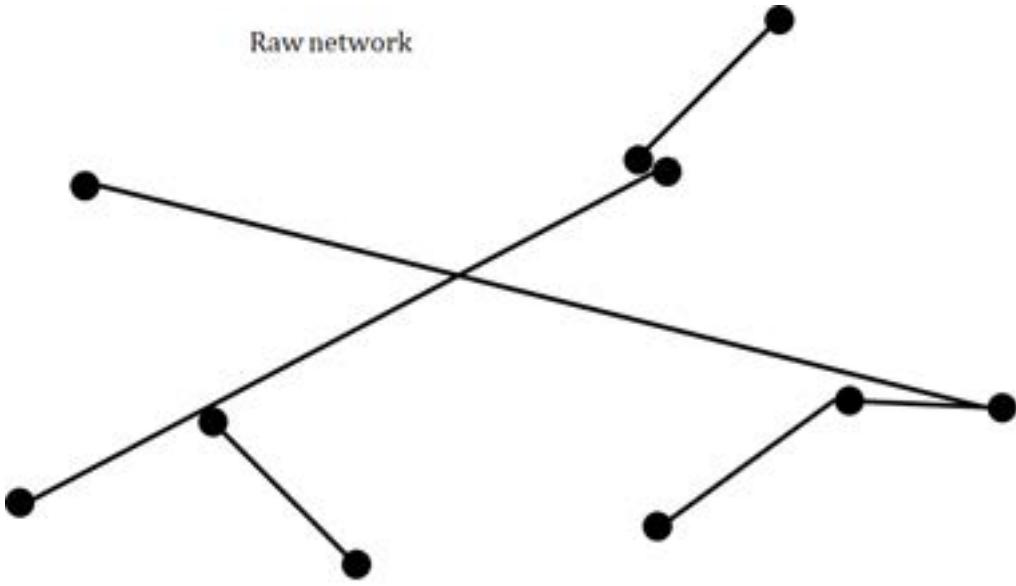
Create numerical model (mesh)

Data from: Houben et al., 2017; Boersma et al., 2019

Meshering artifacts



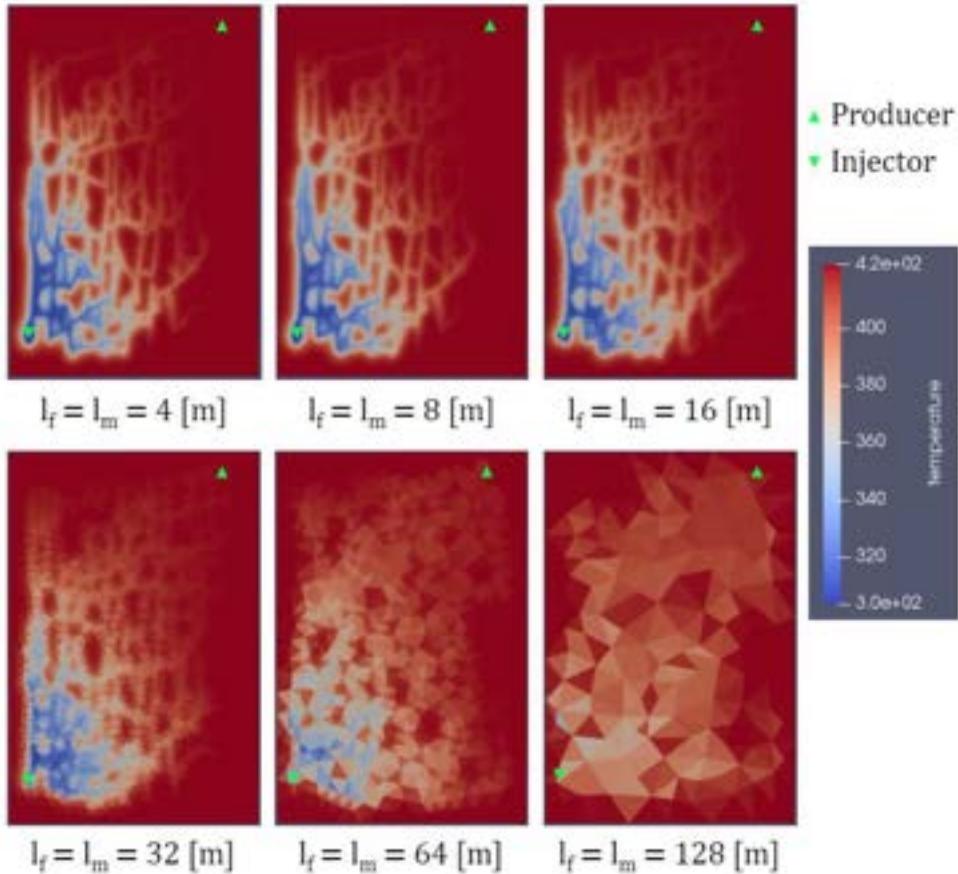
Efficient fracture modeling



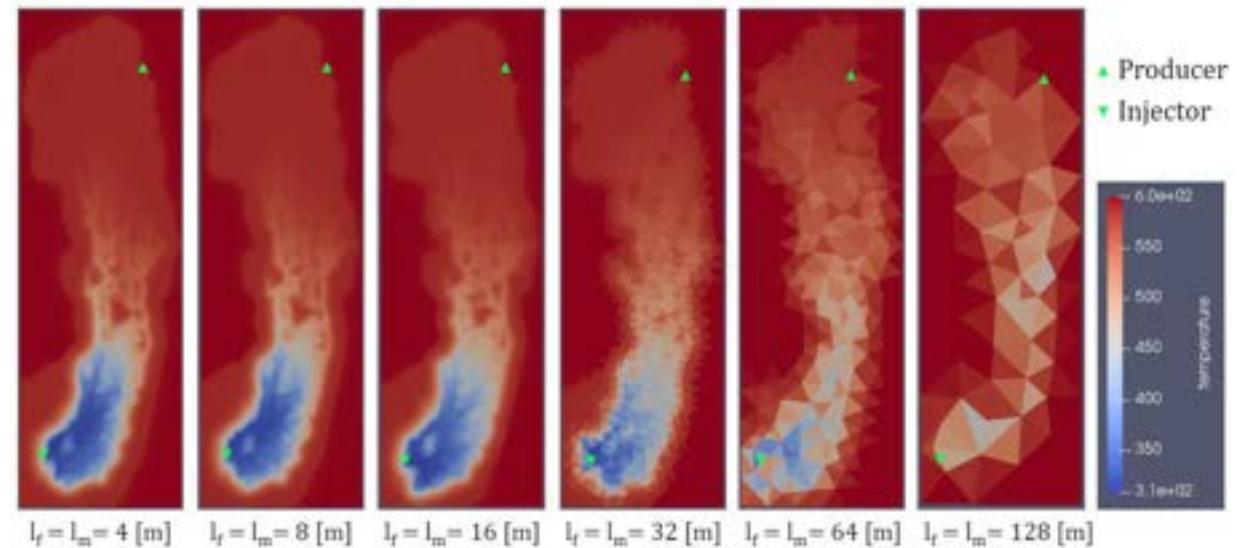
de Hoop et al. (2022)

Sensitivity of geothermal systems

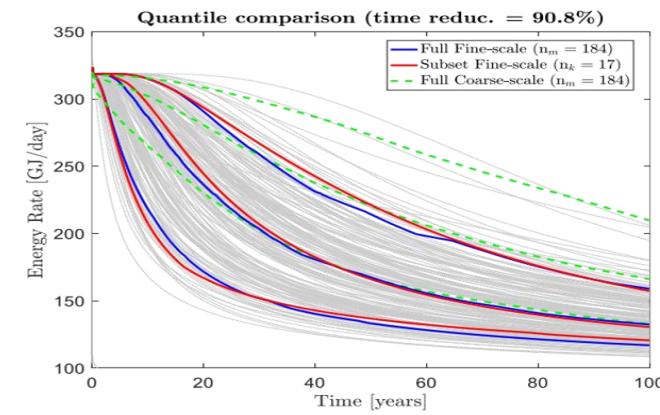
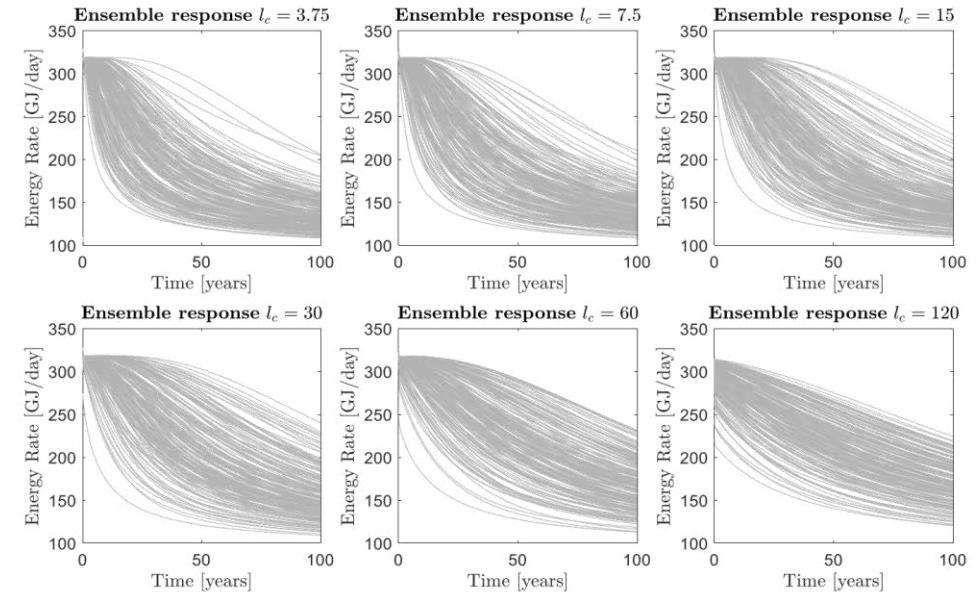
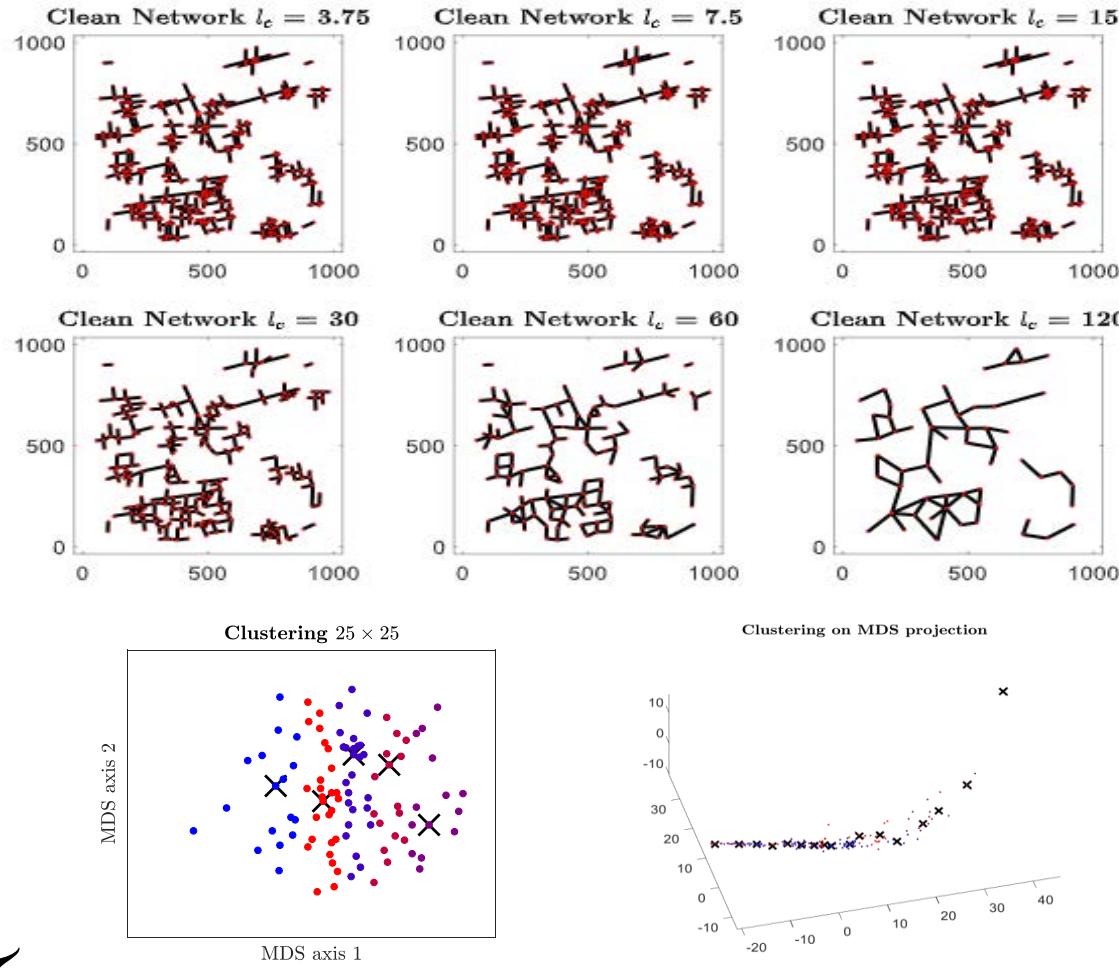
High-enthalpy super critical water



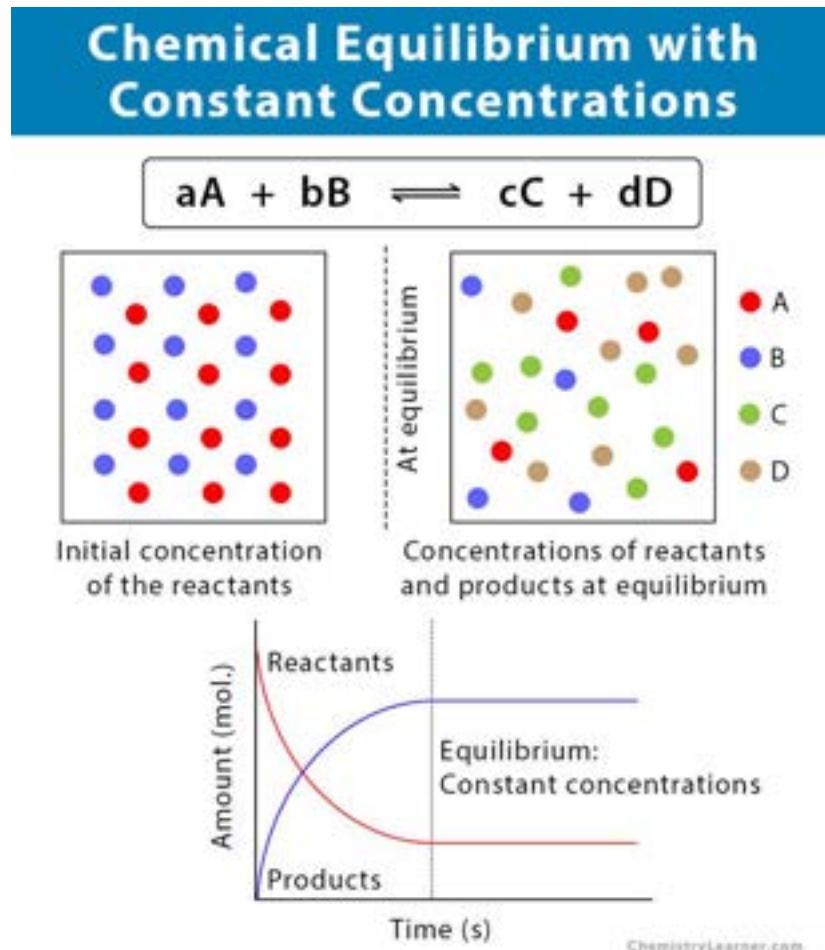
High-enthalpy steam-water system



Uncertainty quantification in fractured reservoirs



Equilibrium chemical reactions



Component mass balance:

$$\frac{\partial}{\partial t} \left(\phi \sum_p \rho_p S_p x_{cp} \right) + \nabla \cdot \sum_p \rho_p x_{cp} u_p = \sum_{r=1}^{n_r} v_{c,r} r_r$$

\downarrow \downarrow \downarrow

$$a_c + l_c = Vr$$

$$a_c + l_c = Vr \times E \Rightarrow a_e + l_e = 0$$

$$K_{sp} - Q_{sp} = 0 \quad \text{equilibrium}$$

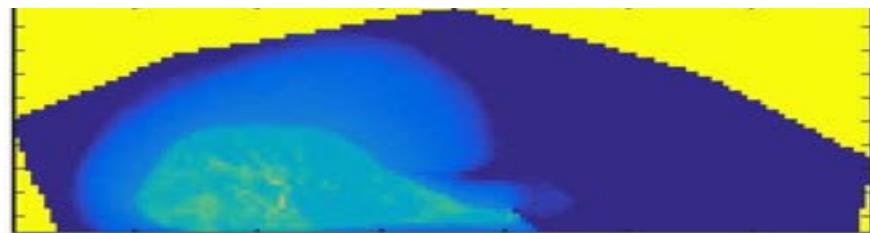
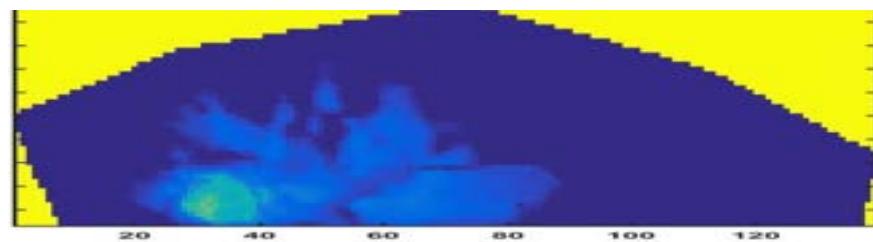
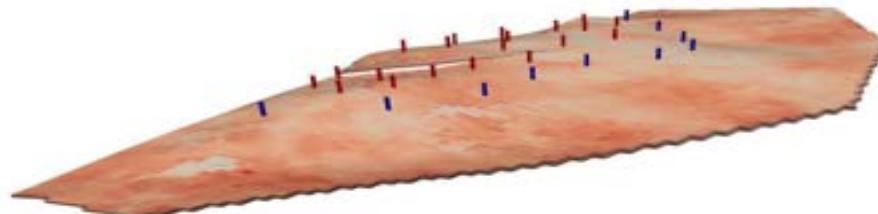
Chemical reactions:

$$a_c^k + l_c^k = vr^k \quad \text{kinetic}$$

Equilibrium reactions in brine- CO_2

$$\mathbf{E} \times \frac{\partial}{\partial t} (\phi \rho_t z_c) + \operatorname{div}(l_c) = \sum_{q=1}^{n_q} v_{cq} r_q$$

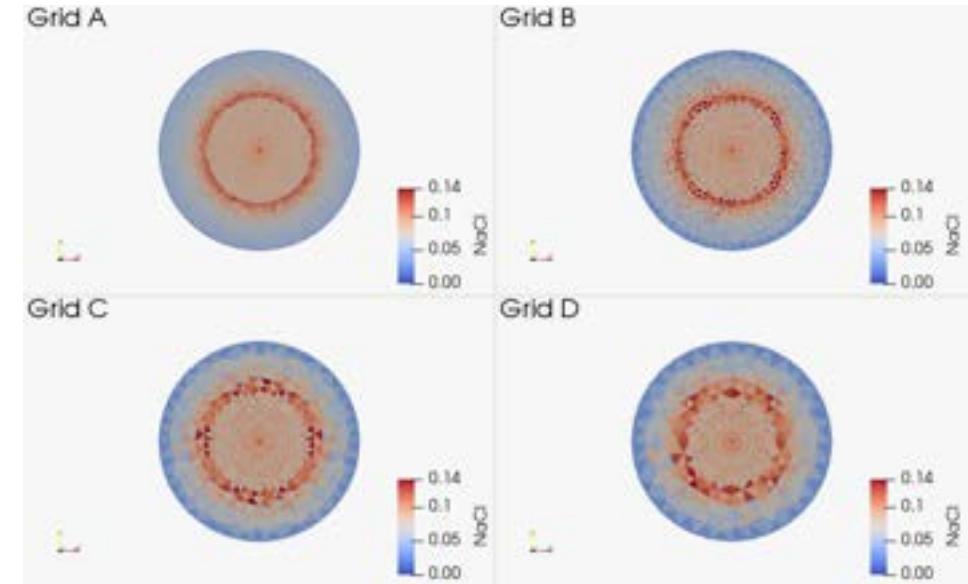
$$f_i^g = f_i^l$$



$$\frac{\partial}{\partial t} (\phi^T \rho_t^E z_i^E) + \operatorname{div}(\mathbf{e}_i \mathbf{l}) = 0$$

$$f_i^g = f_i^l \quad \rho_t^E = \rho_t \sum_{i=1}^{n_e} \mathbf{e}_i \mathbf{z}$$

$$\prod_{c=1}^{n_c} a_c^{v_{cq}} - K_q = 0 \quad \mathbf{z}^E \sum_{i=1}^{n_e} \mathbf{e}_i \mathbf{z} - \mathbf{E} \mathbf{z} = 0$$

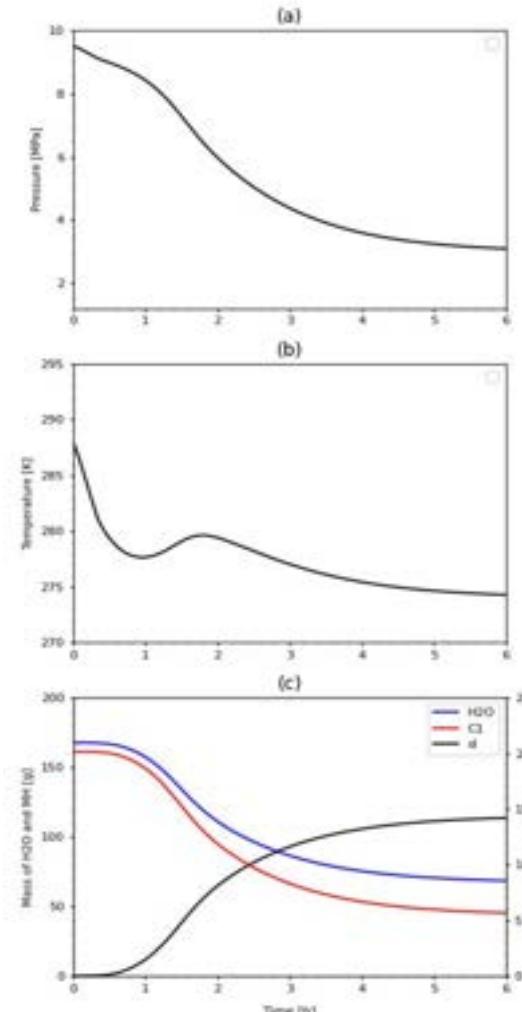


Kala and Voskov, Comput. Geosci., 2020

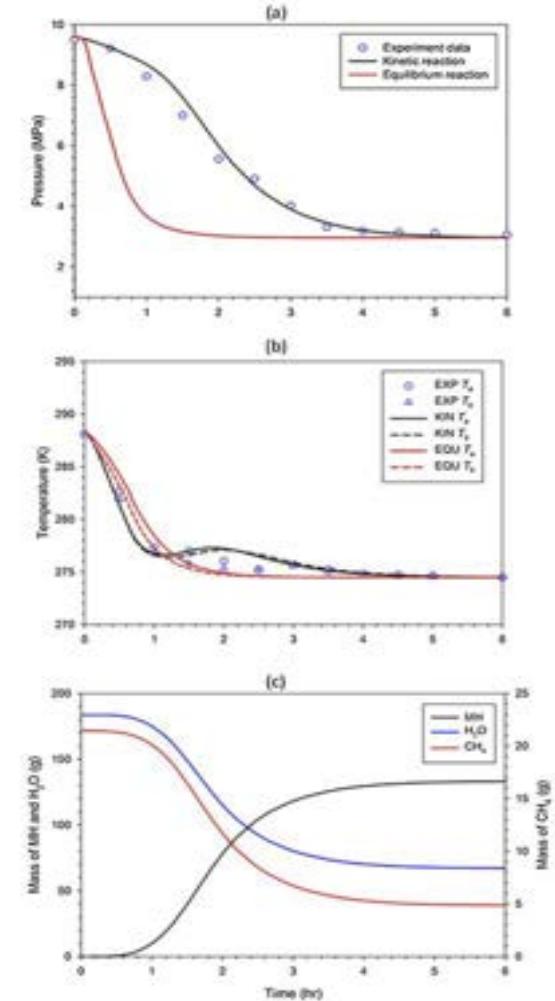
Modeling of hydrate formation

- Numerical simulation of hydrate formation experiment
 - Core filled with brine and free gas CH_4
 - Initially above hydrate formation pressure
 - Cooled down to hydrate formation conditions
- Pressure, temperature and mass recorded over time

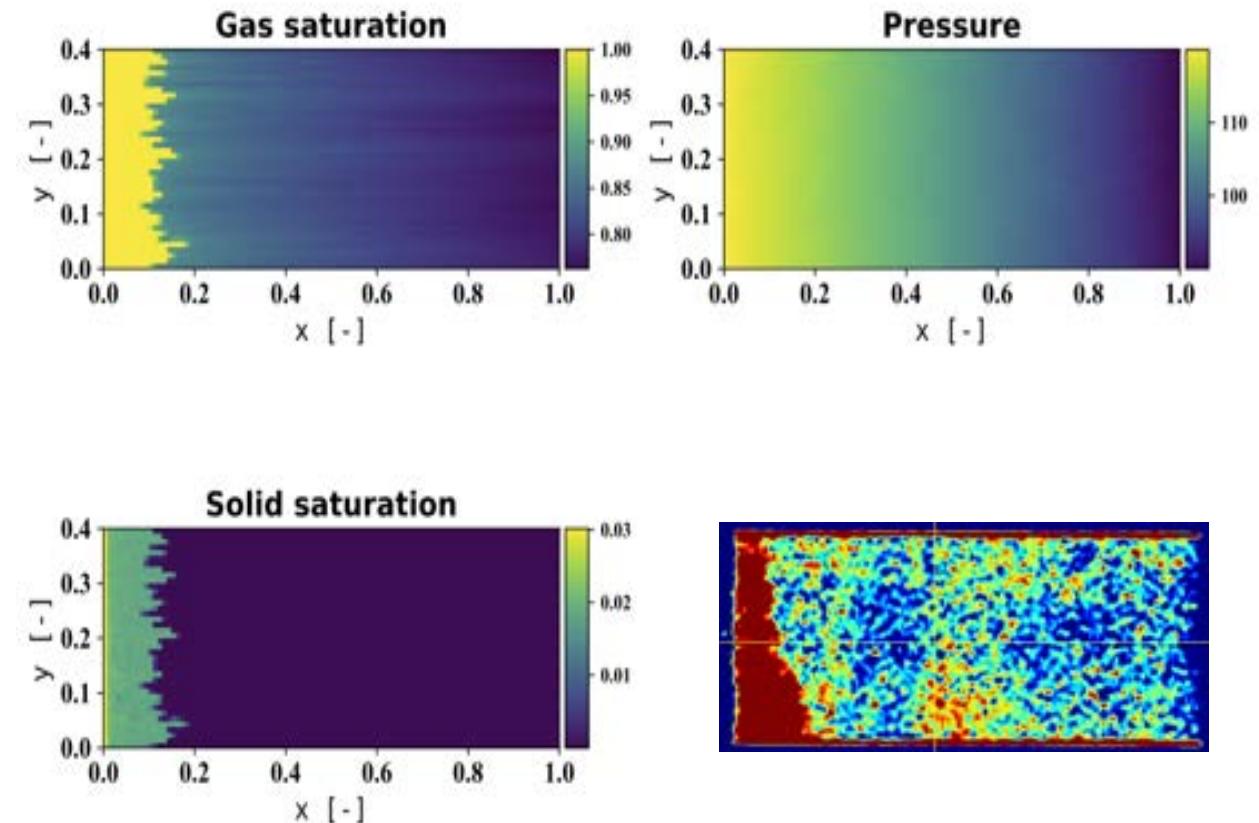
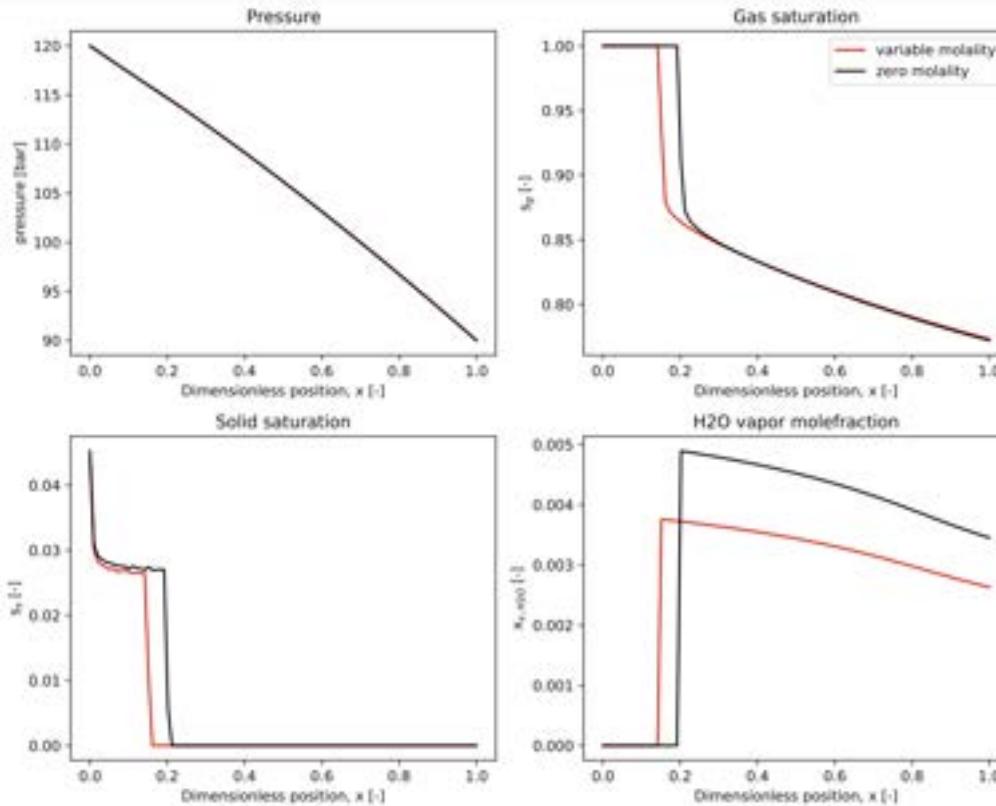
DARTS



TOUGH+Hydrate

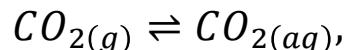


Modeling of salt formation due to dry-out

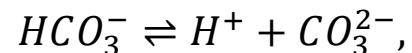
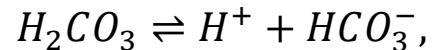
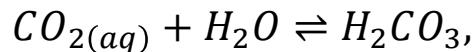


CO₂ injection into calcite core

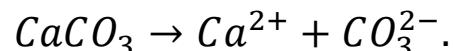
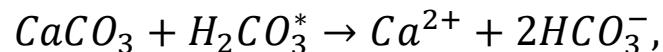
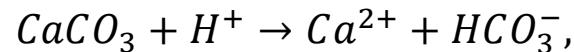
- Carbon dioxide dissolution:



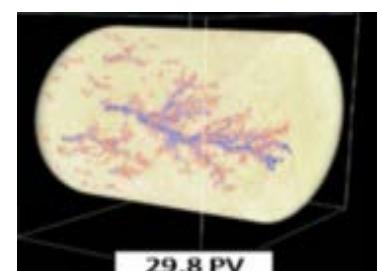
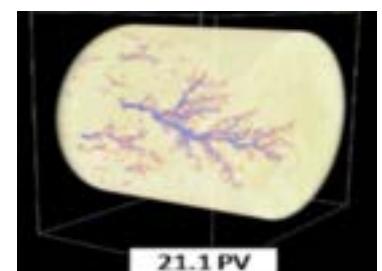
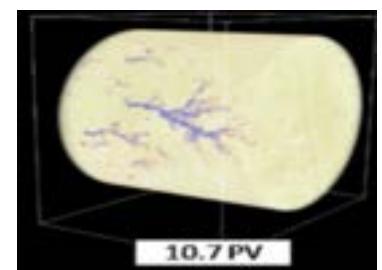
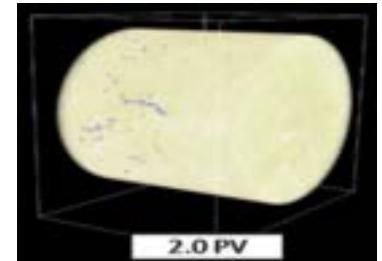
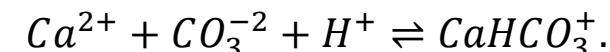
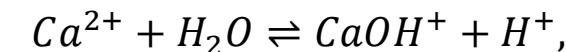
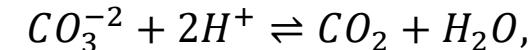
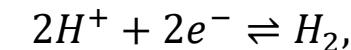
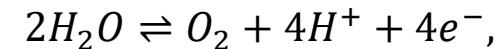
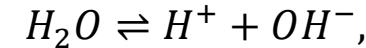
- Acid formation:



- Calcite dissolution:



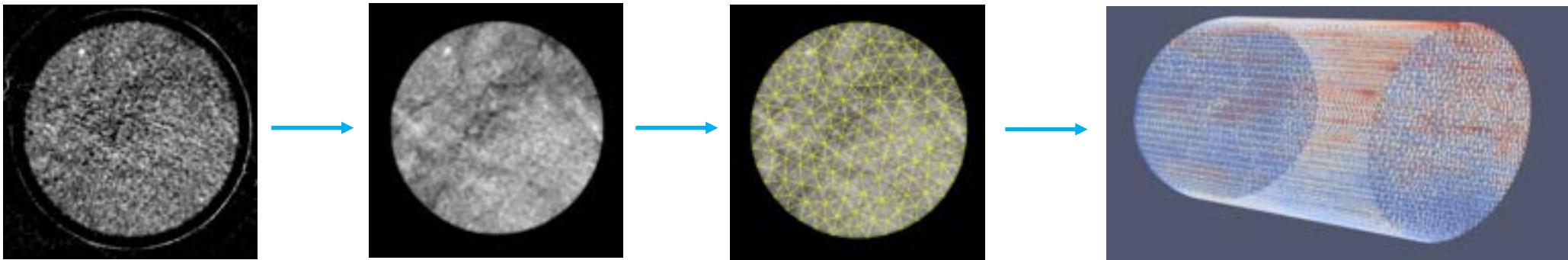
- Other aqueous reactions considered:



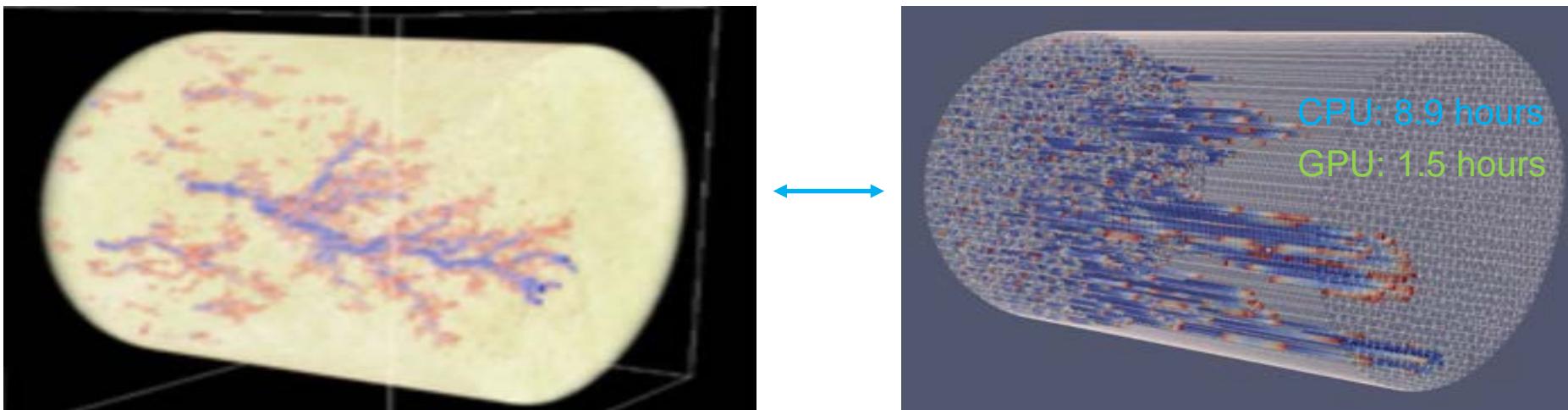
Use PHREEQC for equilibrium chemistry calculation.

Modeling of dissolution at core scale

Step 1: porosity interpretation (image subtraction, filtering, gridding)



Step 2: modeling of dissolution (combination of DARTS + PHREEQC)



Run time:
8.9 hours (CPU engine)
PHREEQC call: 12 min

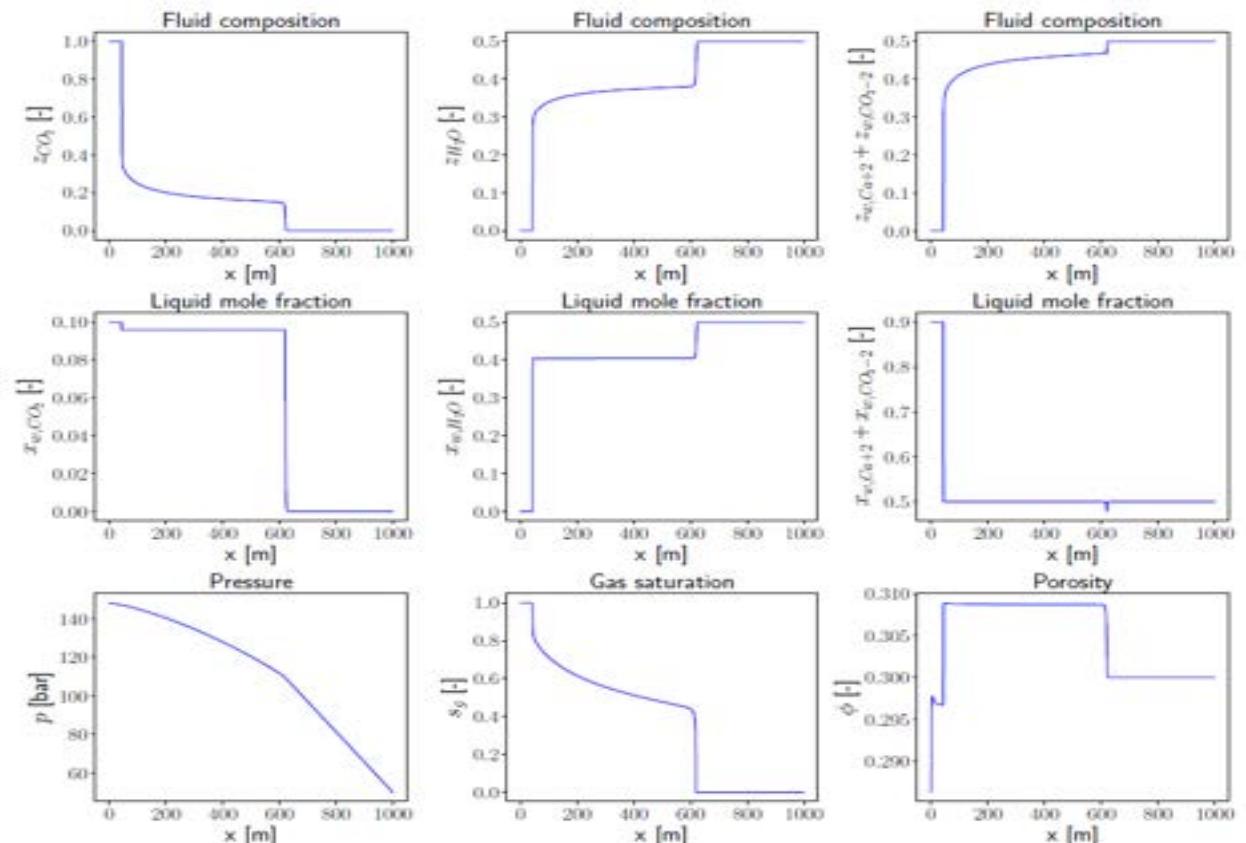
Multiphase flow with reactions (1D benchmark)

$$\frac{\partial n_c}{\partial t} + l_c + q_c = \sum_{k=1}^K v_{ck} r_k^K + \sum_{q=1}^Q v_{cq} r_q^Q, \quad c = 1, \dots, C + M,$$

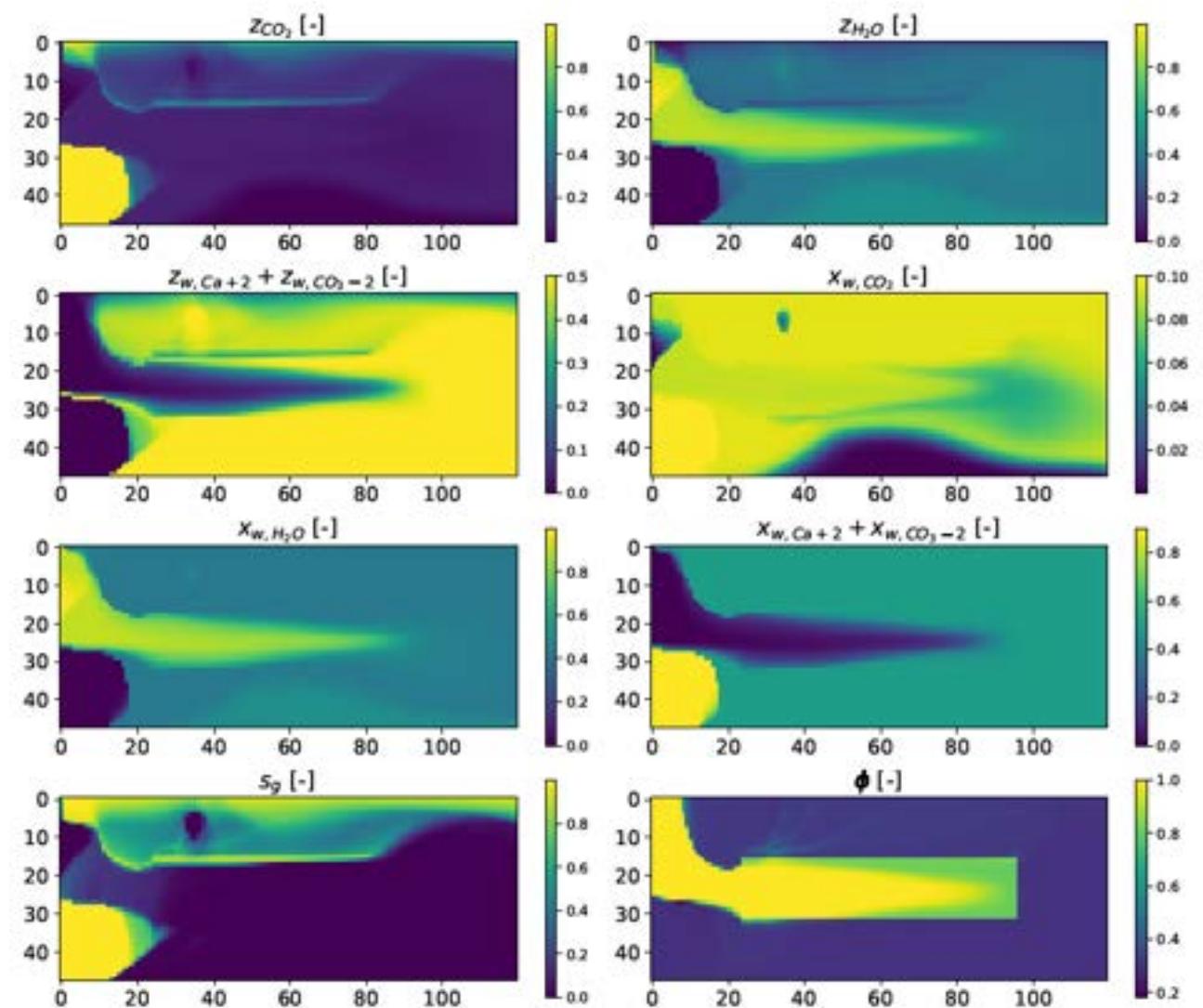
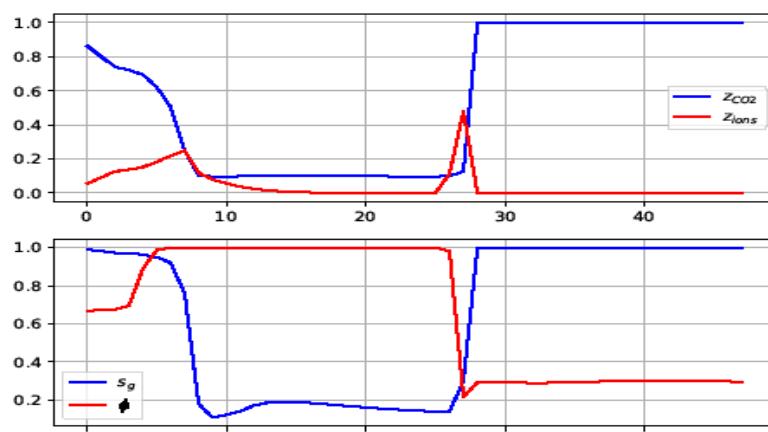
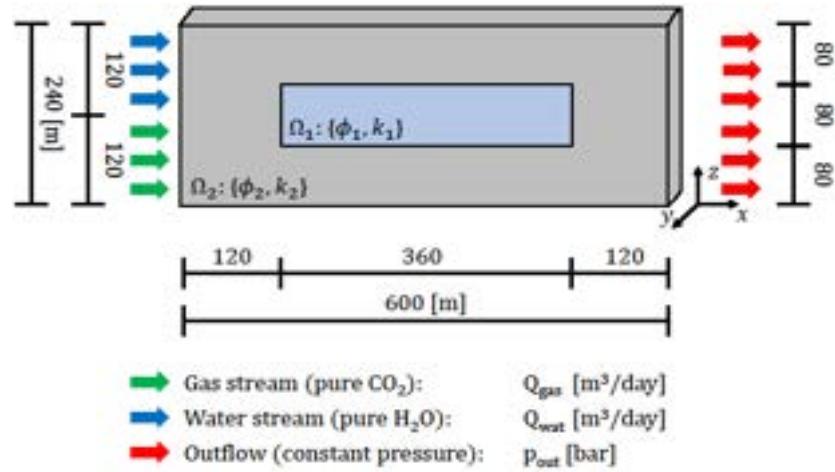
$$n_c = \phi^T \sum_{j=1}^P (\rho_j s_j x_{cj}), \quad c = 1, \dots, C.$$

$$n_m = \phi^T \rho_m z_m, \quad m = C + 1, \dots, M.$$

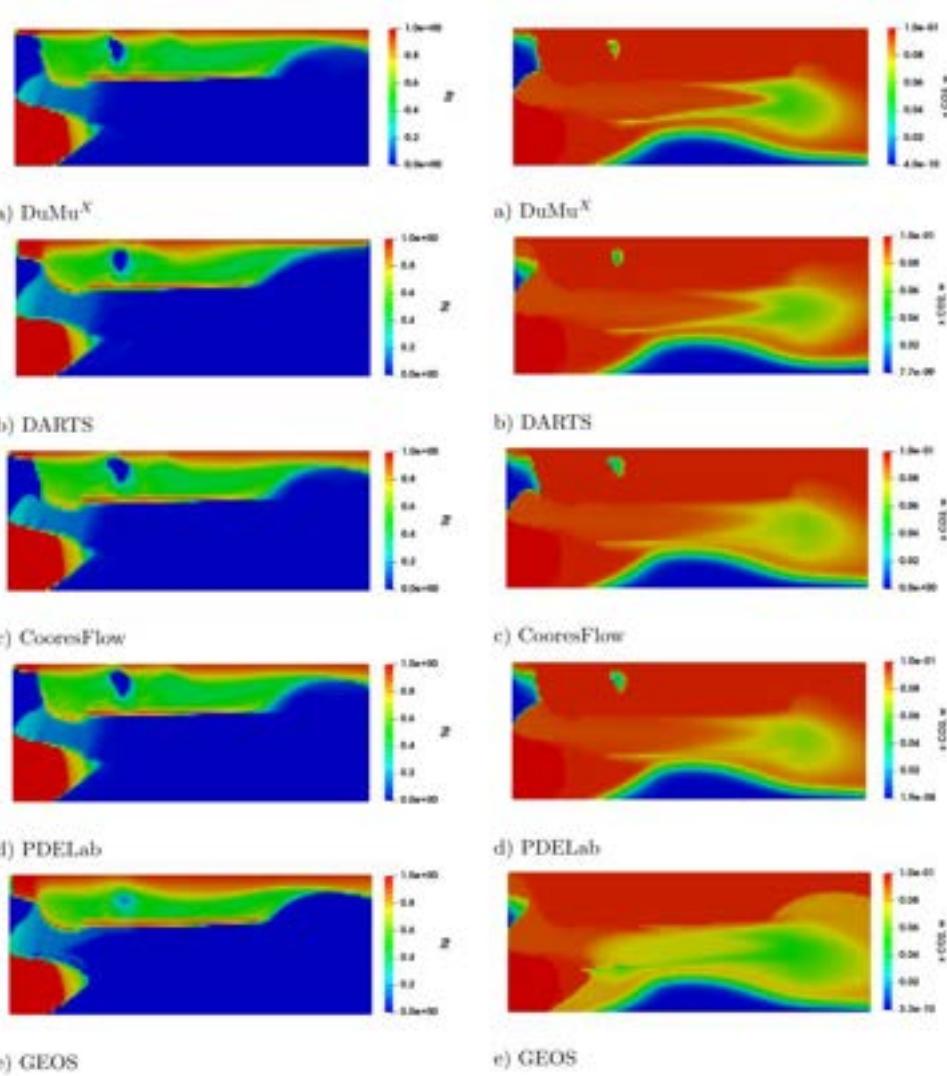
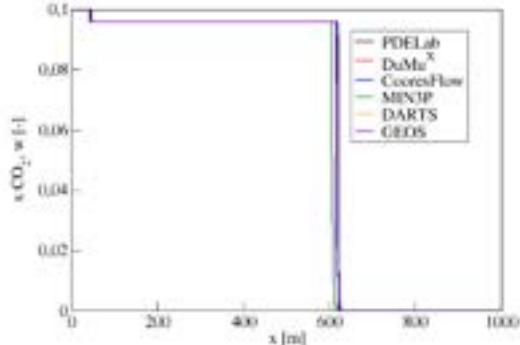
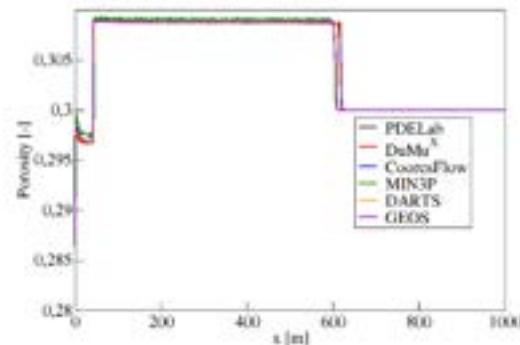
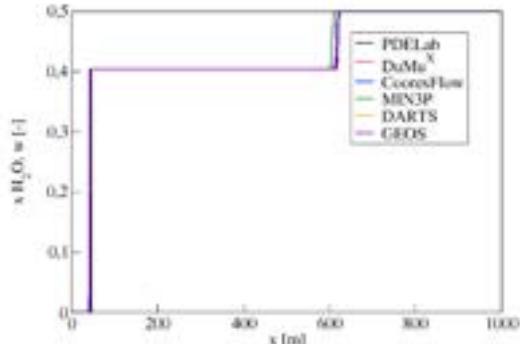
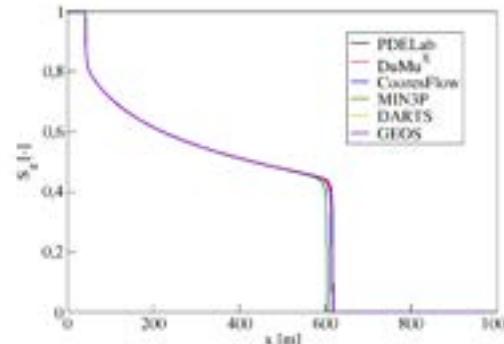
$$\phi = \phi^T \left(1 - \sum_{m=1}^M \hat{s}_m \right) \quad k = k_0 \left(\frac{\phi}{\phi_0} \right)^A$$



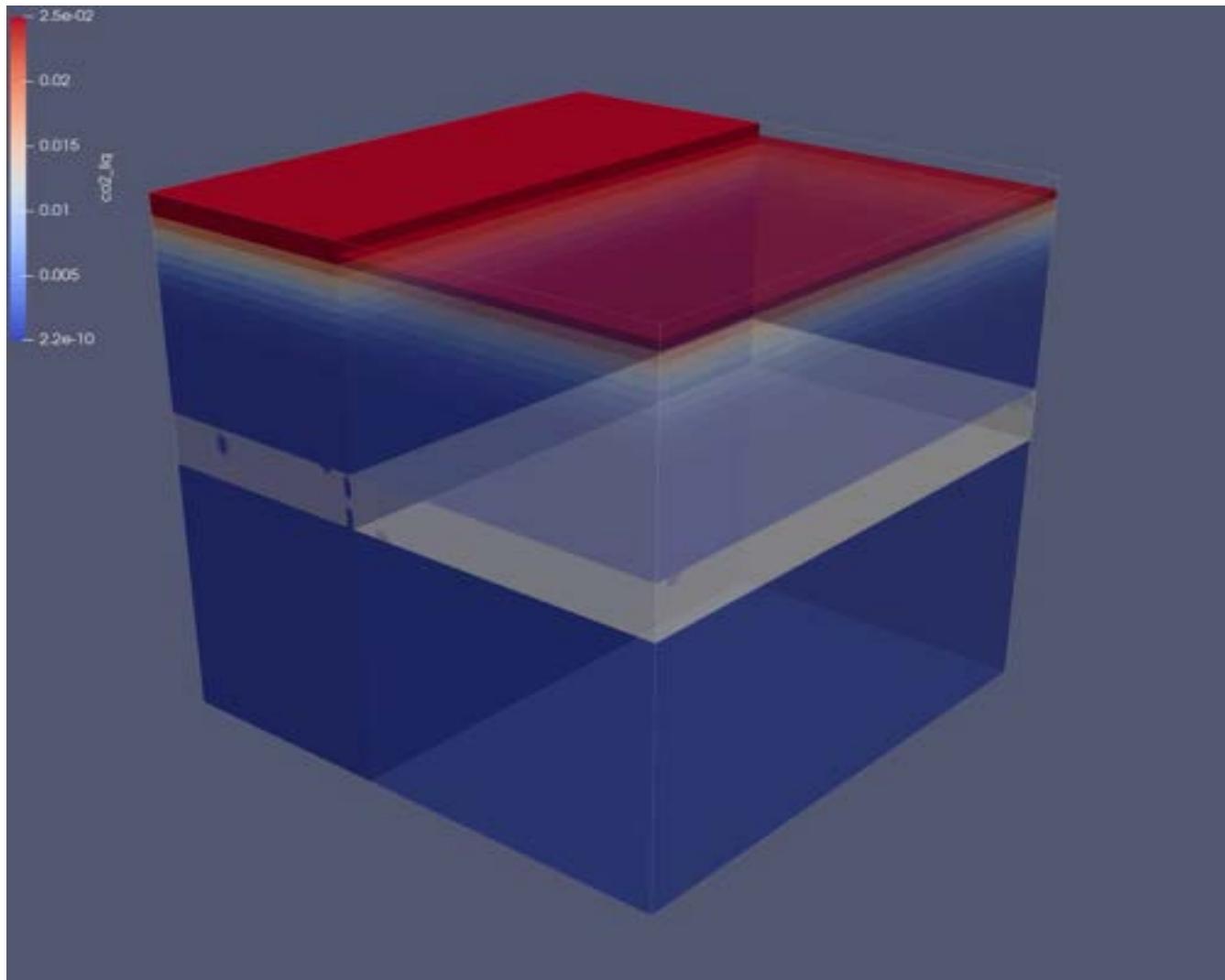
Two-dimensional benchmark



Comparison of different simulators



Combined dissolution



Acknowledgments

- DARTS team: Xiaocong Lyu, Mark Khait, Michiel Wapperom, Yang Wang, Aleks Novikov, Stephan de Hoop, Xiaoming Tian, Kiarash Mansour Pour



Time for Questions and Answers

