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Dynamic Coarsening for Geothermal Applications

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Presentation outline

Motivation

Governing equations and discretization

The MATLAB Reservoir Simulation Toolbox

Dynamic coarsening

Numerical examples

Concluding remarks



Motivation

- Geothermal heat is an appealing resource for energy production and storage
 - Renewable ✓ Always on ✓ Available anywhere ✓ Low carbon footprint ✓
- Viability depends a number of factors (Glassley 2010; Stober and Bucher 2013)
 - Efficiency, storage capacity, operational and drilling costs, legal regulations, ...
- Assessment requires solid system knowledge (Andersson 2007)
 - Aquifer/aquiclude geology, groundwater chemistry, flow properties, ...

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Complexity and size typically renders numerical simulations the only viable option
(O'Sullivan, Pruess, and Lippmann 2000; K. S. Lee 2010; Stober and Bucher 2013)



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Governing equations and discretization

Single-phase conservation of mass on residual form

$$R_f = \partial_t(\phi \rho_f) + \nabla \cdot (\rho_f \vec{v}_f) - \rho_f q_f = 0$$

- Velocity given by Darcy's law: $\vec{v}_f = -\frac{1}{\mu_f} \mathbf{K}(\nabla p - \rho_f g \nabla z)$

ϕ	Pore volume	\mathbf{K}	Permeability	Λ	Thermal conductivity	\vec{g}	Gravity
ρ	Density	μ	Viscosity	u	Internal energy	h	Enthalpy
p	Pressure	T	Temperature	q	Sources/sinks	r/f	Fluid/rock

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Governing equations and discretization

Conservation of energy on residual form

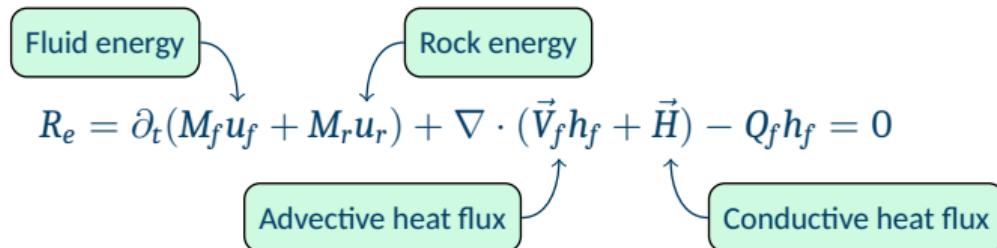
$$R_e = \partial_t(\phi \rho_f u_f + [1 - \phi] \rho_r u_r) + \nabla \cdot (\rho_f \vec{v}_f h_f + \vec{H}) - \rho_f q_f h_f = 0$$

- Conductive heat flux from Fourier's law: $\vec{H} = -(\Lambda_f + \Lambda_r) \nabla T$

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Conservation of energy on residual form

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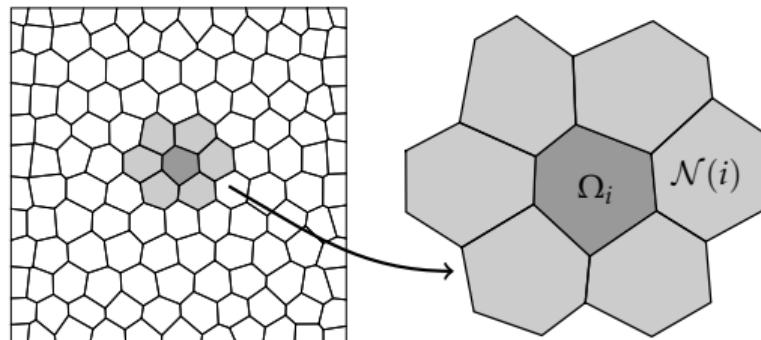
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Governing equations and discretization

Finite volumes in space, implicit backward Euler in time

$$\mathbf{R}^{n+1} = \frac{1}{\Delta t^n} (\mathbf{M}^{n+1} - \mathbf{M}^n) + \operatorname{div}(\mathbf{V}^{n+1}) - \mathbf{Q}^{n+1} = 0$$



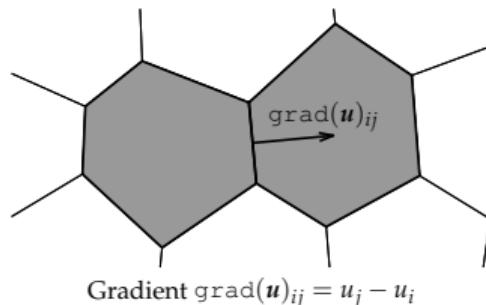
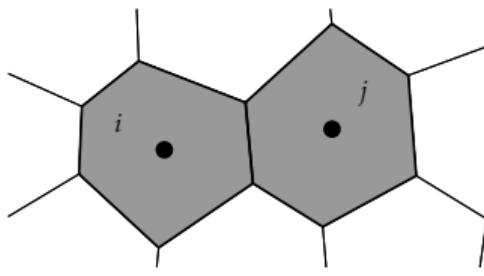
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$$\mathbf{V} = -\operatorname{upw}(\rho/\mu) [\mathbf{Kgrad}(\mathbf{p}) - g \operatorname{favg}(\rho) \mathbf{Kgrad}(\mathbf{z})]$$

- \mathbf{Kgrad} : discrete operator $\mathbf{K}\nabla$ (linear/nonlinear two-point, multipoint, mimetic, etc.)
 - In this work: linear two-point flux approximation (comparison: Ø. Klemetsdal et al. 2020)



$$\mathbf{Kgrad} = \mathbf{Tgrad}$$

$$\mathbf{T}_{ij} = \left(\mathbf{T}_{i,j}^{-1} + \mathbf{T}_{j,i}^{-1} \right)^{-1}$$

$$\mathbf{T}_{i,j} = |F_{ij}| \frac{\vec{c}_{i,j} \mathbf{K}_i \vec{n}_{i,j}}{|\vec{c}_{i,j}|^2}$$

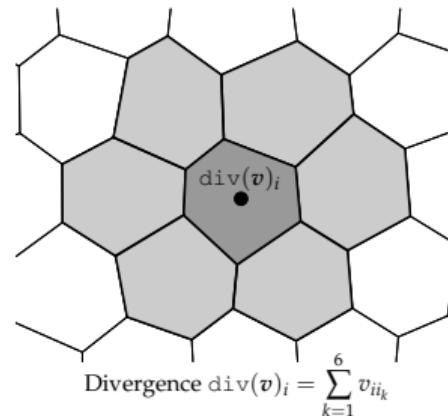
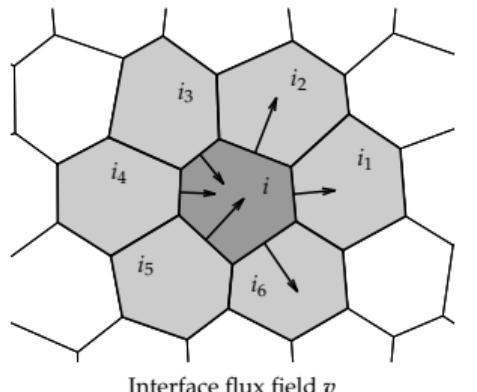
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- div : discrete divergence operator

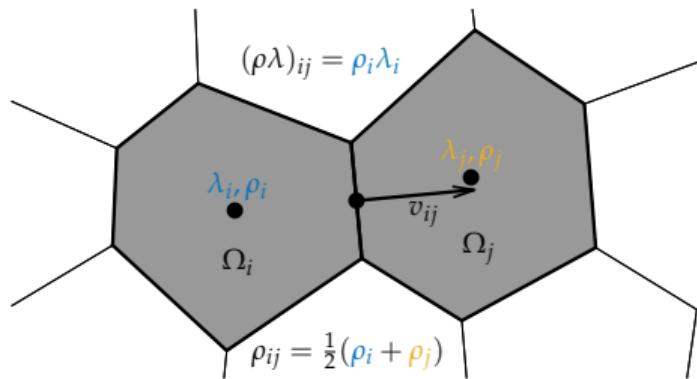


Governing equations and discretization

Finite volumes in space, implicit backward Euler in time

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- **upw**: Upwind discretization (single-point here); **favg**: Face average operator



Governing equations and discretization

Finite volumes in space, implicit backward Euler in time

$$\mathbf{R}^{n+1} = \frac{1}{\Delta t^n} (\mathbf{M}^{n+1} - \mathbf{M}^n) + \operatorname{div}(\mathbf{V}^{n+1}) - \mathbf{Q}^{n+1} = 0$$

Newton's method: make system $\mathbf{R}(\mathbf{x}) = \mathbf{0}$, linearize, neglect higher-order terms

$$\mathbf{x}^{k+1} = \mathbf{x}^k + \Delta \mathbf{x}, \quad -\frac{\partial \mathbf{R}}{\partial \mathbf{x}} \Delta \mathbf{x} = \mathbf{R}(\mathbf{x}^k)$$

Sequential implicit formulation

1. Form pressure equation as weighted sum of \mathbf{R}_f and \mathbf{R}_e

$$\mathbf{R}_p = \omega_f \mathbf{R}_f + \omega_e \mathbf{R}_e, \quad \partial_{\mathbf{x}} (\omega_f \mathbf{M}_f^{n+1}) + \partial_{\mathbf{x}} (\omega_e [\mathbf{M}_f \mathbf{u}_f + \mathbf{M}_r \mathbf{u}_r]^{n+1}) = \mathbf{0}, \quad \mathbf{x} \neq \text{pressure}$$

2. Solve $\mathbf{R}_p = \mathbf{0}$ with fixed temperature and transport variables \rightarrow pressure + intercell fluxes
3. Solve $\mathbf{R}_f = \mathbf{0}$ and $\mathbf{R}_e = \mathbf{0}$ with fixed pressure and intercell fluxes \rightarrow temperature + transport

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Transport formulation: solve for temperature \mathbf{T} and total saturation \mathbf{S}_t
 \rightarrow allow total saturation to be $\neq 1$, multiply densities by total saturation

$$\rho_f \rightarrow \mathbf{S}_t \rho_f, \quad \rho_r \rightarrow \mathbf{S}_t \rho_r$$



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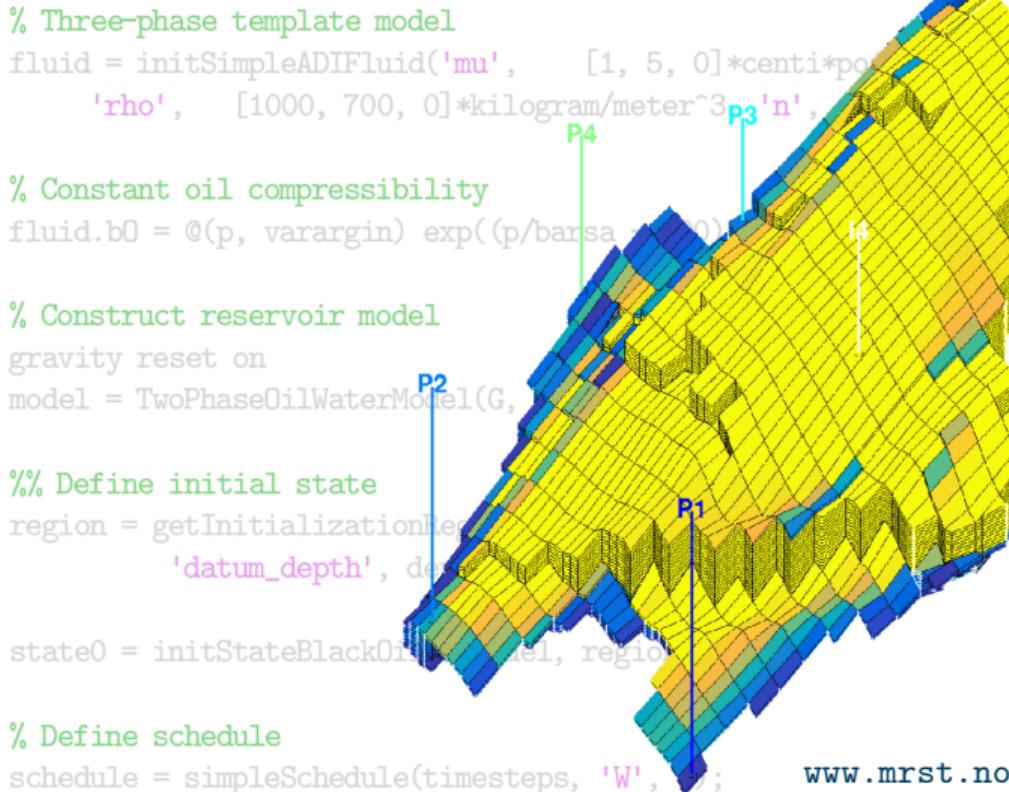


MATLAB Reservoir Simulation Toolbox (MRST)

Transforming research on reservoir modelling

Unique prototyping platform:

- Standard data formats
- Data structures/library routines
- Fully unstructured grids
- Rapid prototyping:
 - Differentiation operators
 - Automatic differentiation
 - Object-oriented framework
 - State functions
- Industry-standard simulation



www.mrst.no

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Differentiation operators

Write discrete equations on form very close to continuous equations

$$\nabla \cdot \vec{H} \quad \vec{H} = -(\lambda_f + \lambda_r) \nabla T$$
$$\text{div}(H) \quad H = -(lambdaF + lambdaR) .* \text{grad}(T)$$

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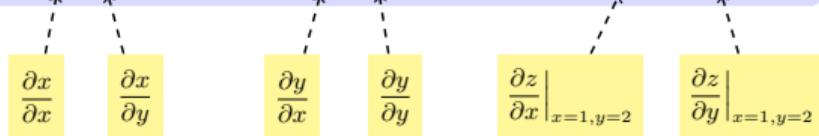
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Automatic differentiation

Combine chain rule and elementary differentiation rules by means of operator overloading to analytically evaluate all derivatives
→ Computing Jacobians amounts to writing down residual equations.

```
[x,y] = initVariablesADI(1,2); z = 3*exp(-x*y)
```

```
x = ADI Properties:  
val: 1  
jac: {[1] [0]}  
y = ADI Properties:  
val: 2  
jac: {[0] [1]}  
z = ADI Properties:  
val: 0.4060  
jac: {[[-0.8120] [-0.4060]}}
```





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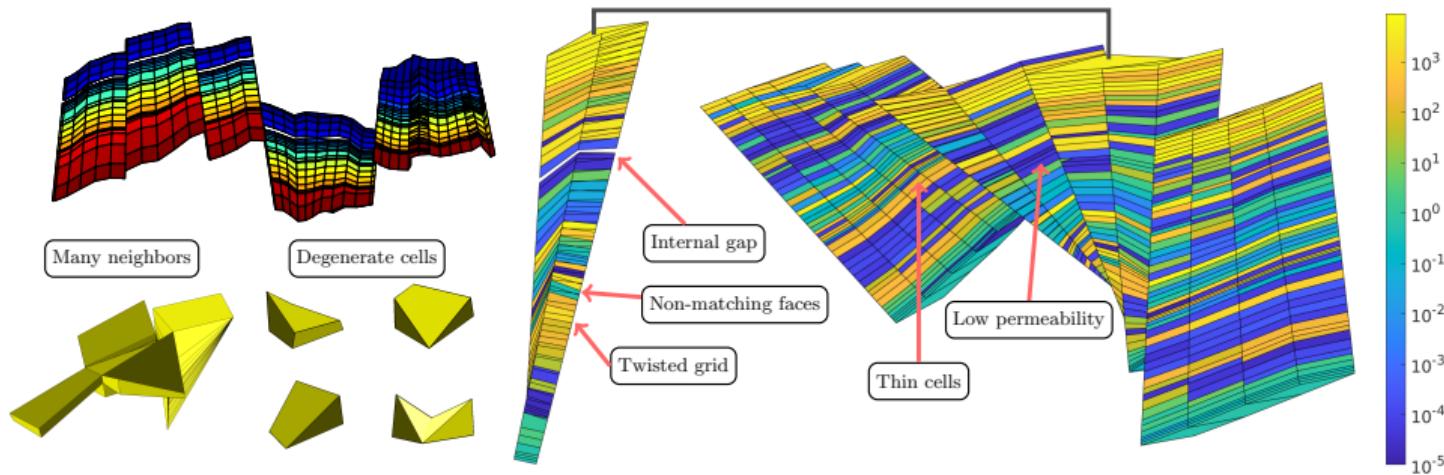
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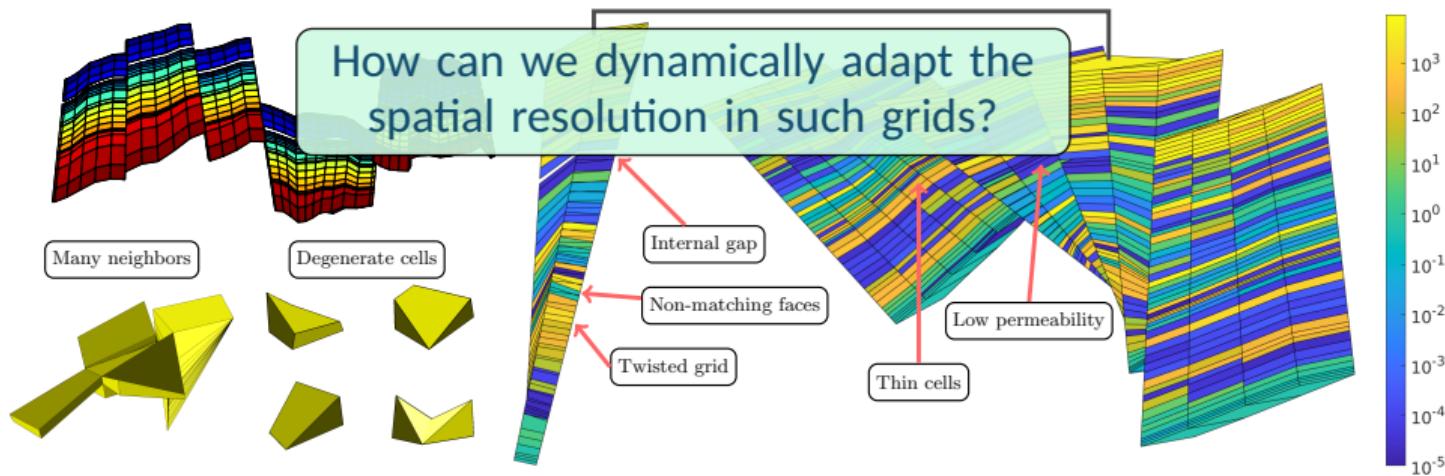
Reservoir simulation grids

- Subsurface reservoirs are complex: layers, faults, fractures, erosion, wells, ...
- Simulation models often upscaled → polyhedral cells with full-tensor permeability



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Dynamic coarsening

- Transport of geothermal heat chiefly confined to proximity of wells
- Difficult to determine appropriate grid resolution a priori
- Many geomodels are not suitable for conventional grid refinement methods

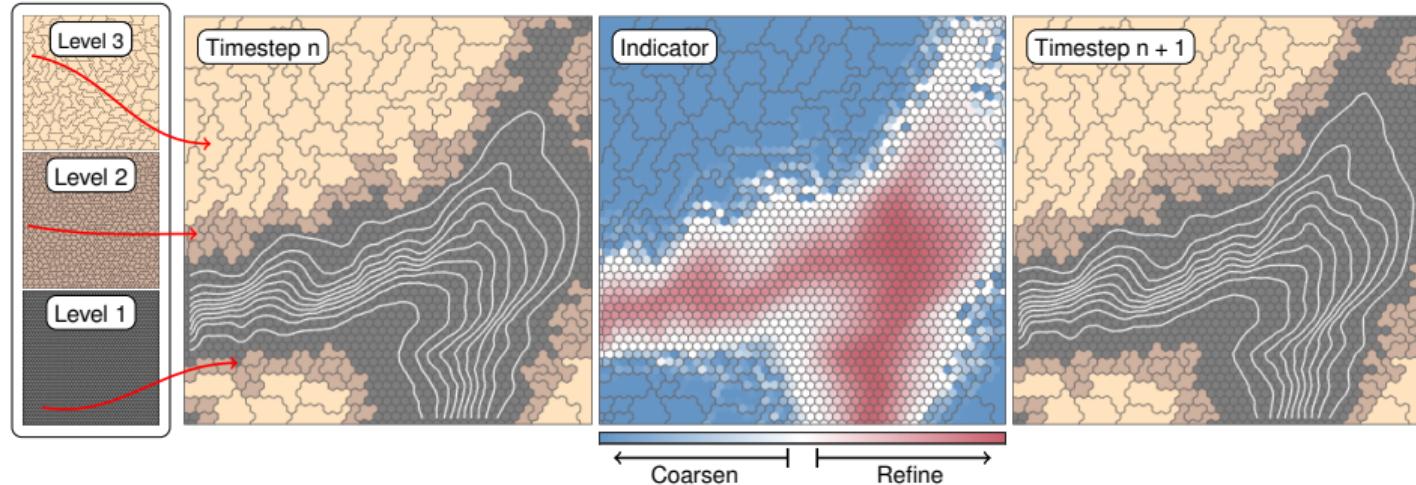
Dynamic coarsening

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- Difficult to determine appropriate grid resolution a priori
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- Reservoir engineering applications:

$$\# \text{ cells in simulation grid} \quad \ll \quad \# \text{ cells in geocellular model}$$

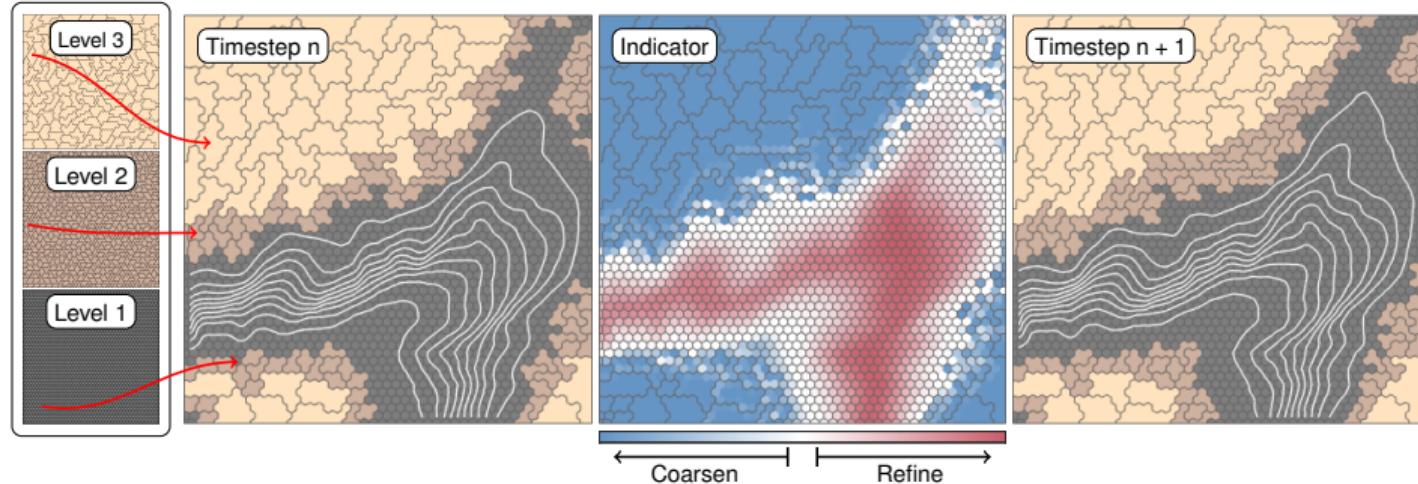
- State-of-the-art multiscale methods (attempt to) bridge gap for pressure problems (Jenny, S. H. Lee, and Tchelepi 2006; Møyner and Lie 2016; Lie et al. 2017, etc.)
- Here: attempt to bridge this gap for transport problems by **dynamic coarsening**

Dynamic coarsening



Quandalle and Basset 1983; Christensen et al. 2004; Batenburg et al. 2011; Hoteit and Chawathe 2016; Cusini and Hajibeygi 2018; Ø. S. Klemetsdal and Lie 2020; Ø. S. Klemetsdal, Møyner, et al. 2021

Dynamic coarsening



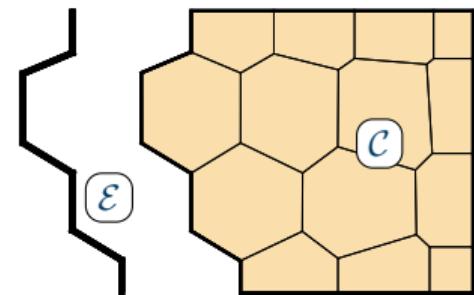
Keep track of which cells to refine/coarsen using coarsening indicator $\mathcal{I}(u) \in \mathbb{R}_+^N$

Coarse block comprising fine-scale cells \mathcal{C}

coarsen if $\mathcal{I}_i < \epsilon_c$ for all $i \in \mathcal{C}$, refine if $\mathcal{I}_i > \epsilon_r$ for any $i \in \mathcal{C}$

Dynamic coarsening – Mapping quantities

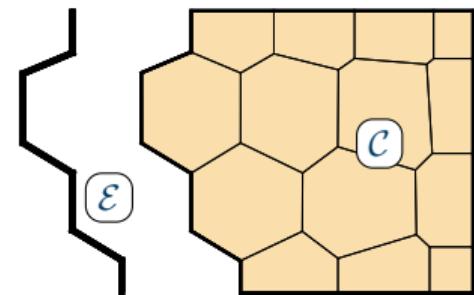
Mapping should be inexpensive and **energy conservative**



Dynamic coarsening – Mapping quantities

Mapping should be inexpensive and **energy conservative**

1. Accumulate coarse-block energy: $\sum_{i \in \mathcal{C}} (\mathbf{M}_f \mathbf{u}_f + \mathbf{M}_r \mathbf{u}_r)_i$

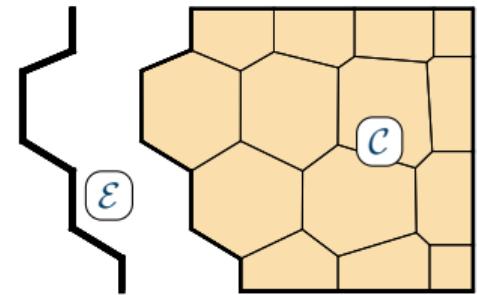


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$$\mathbf{p}^a = \frac{1}{\Phi^a} \sum_{j \in \mathcal{C}} (\Phi \mathbf{p})_j, \quad \mathbf{T}^a = \frac{1}{\Phi^a} \sum_{j \in \mathcal{C}} (\Phi \mathbf{T})_j, \quad \mathbf{v}^a = \sum_{(m,n) \in \mathcal{E}} \mathbf{v}_{mn}$$

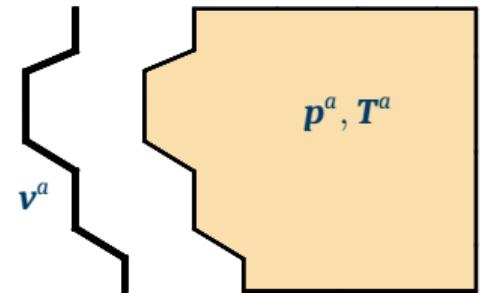


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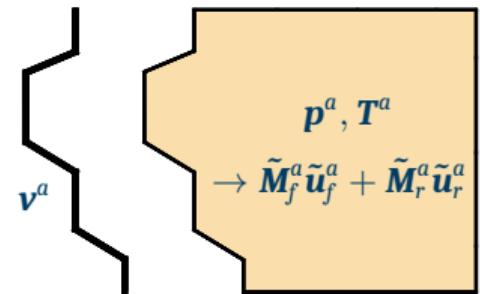
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3. Compute energy in coarse block with adapted properties



Dynamic coarsening – Mapping quantities

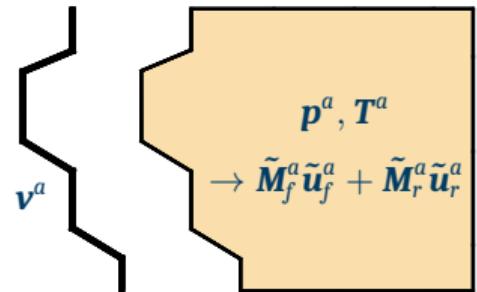
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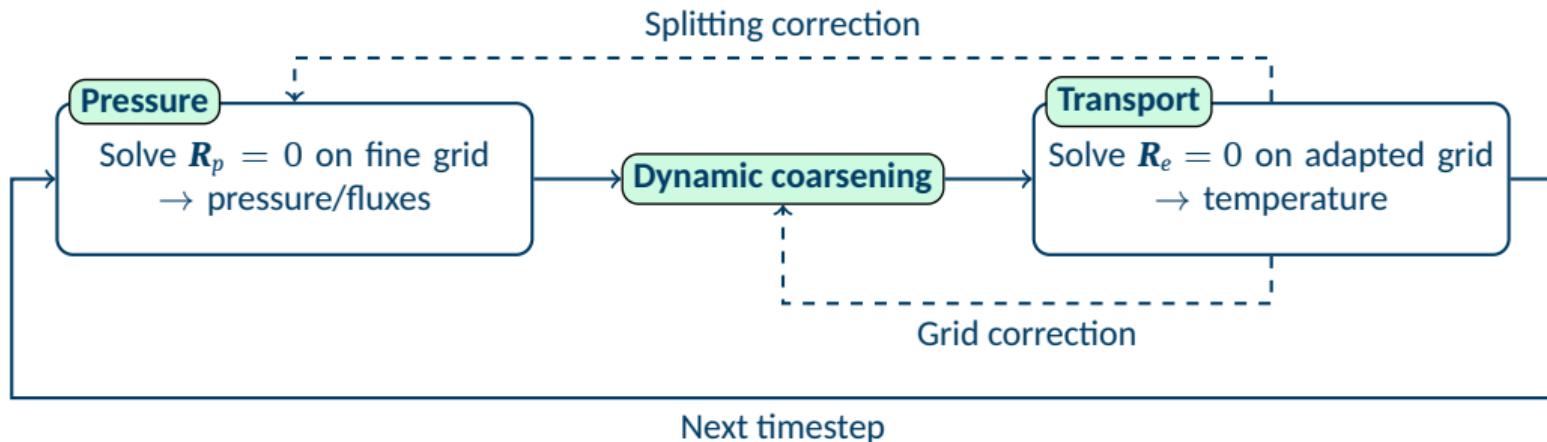
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3. Compute energy in coarse block with adapted properties
4. Ensure conservation of energy through **energy discrepancy** \rightarrow density correction

$$S_t = \frac{\sum_{i \in \mathcal{C}} (\mathbf{M}_f \mathbf{u}_f + \mathbf{M}_r \mathbf{u}_r)_i}{\tilde{\mathbf{M}}_f^a \tilde{\mathbf{u}}_f^a + \tilde{\mathbf{M}}_r^a \tilde{\mathbf{u}}_r^a} = \frac{\text{accumulated energy from fine grid}}{\text{energy on adapted grid}}$$



Dynamic coarsening – Solution procedure





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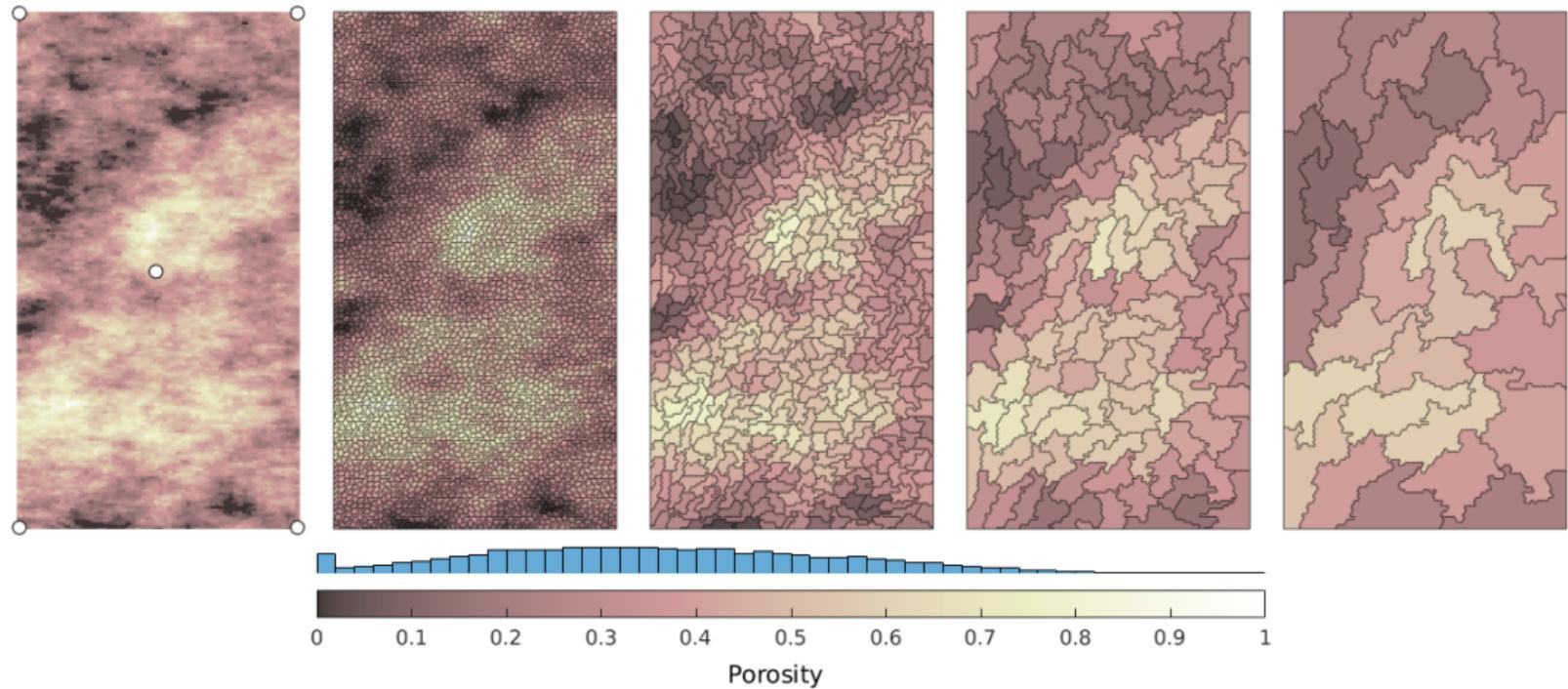
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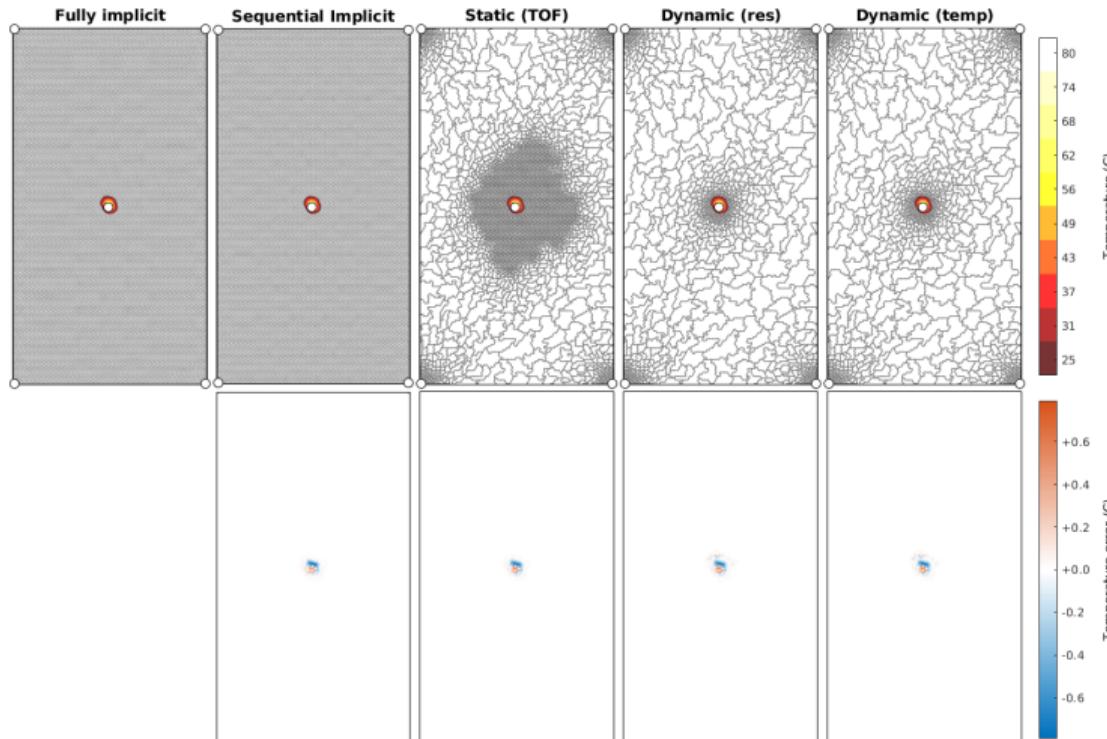
Example: SPE10 Model 2

- Heat storage in two different layers of SPE10 Model 2
- Three one-year cycles of storage in center well with pressure support in corner wells
 1. **Charge:** 3 months of injection at 80 °C, bhp = 70 bar
 2. **Rest:** 3 months with no driving forces
 3. **Discharge:** 3 months of extraction, bhp = 1500 bar
 4. **Rest:** 3 months with no driving forces
- Three coarsening approaches
 1. Static based on incompressible time-of-flight
 2. Dynamic with residual-based indicator
 3. Dynamic with temperature indicator

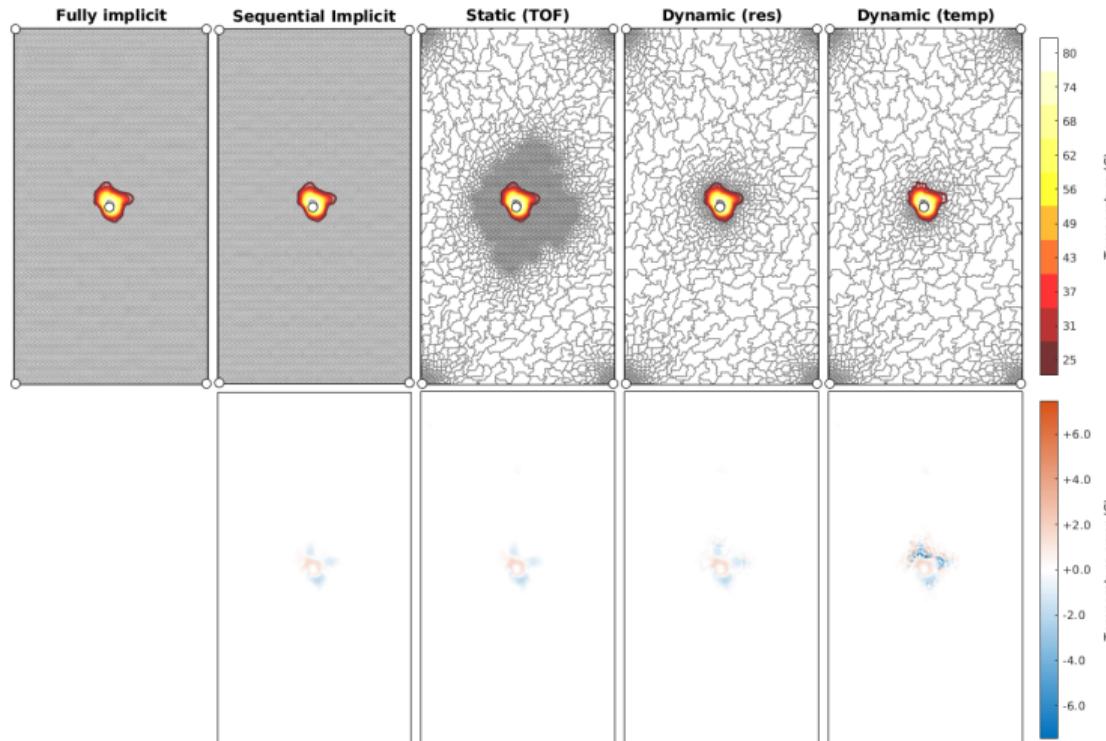
Example: SPE10 Model 2 – Tarbert (layer 10)



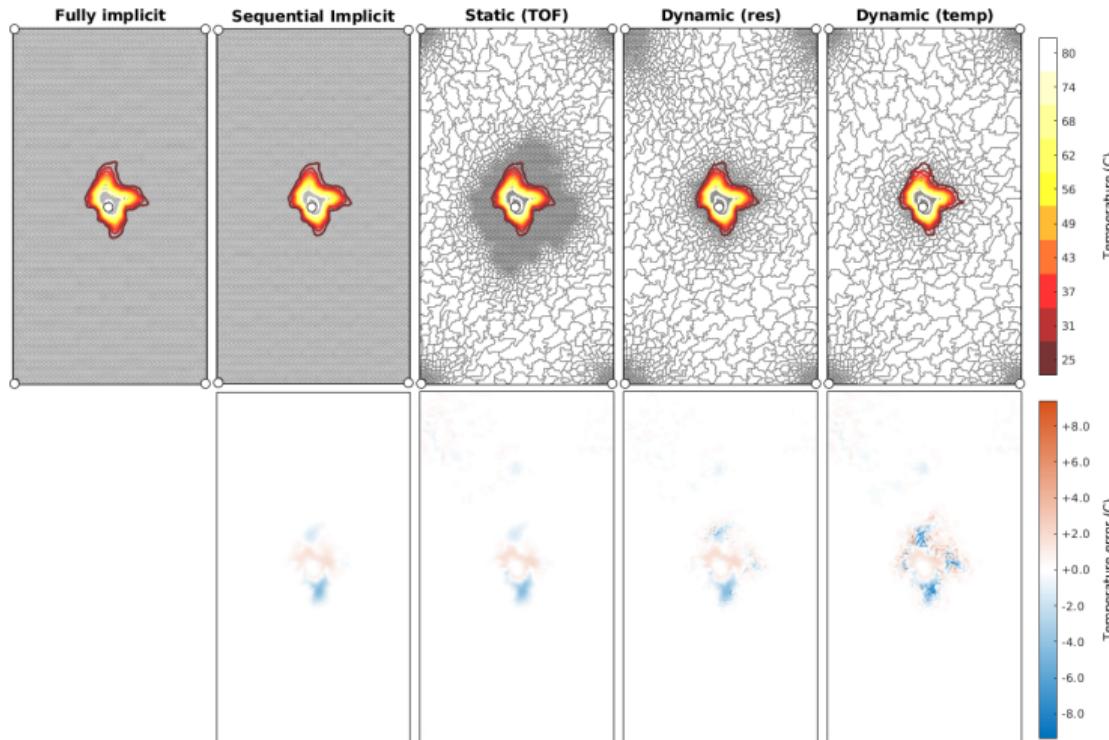
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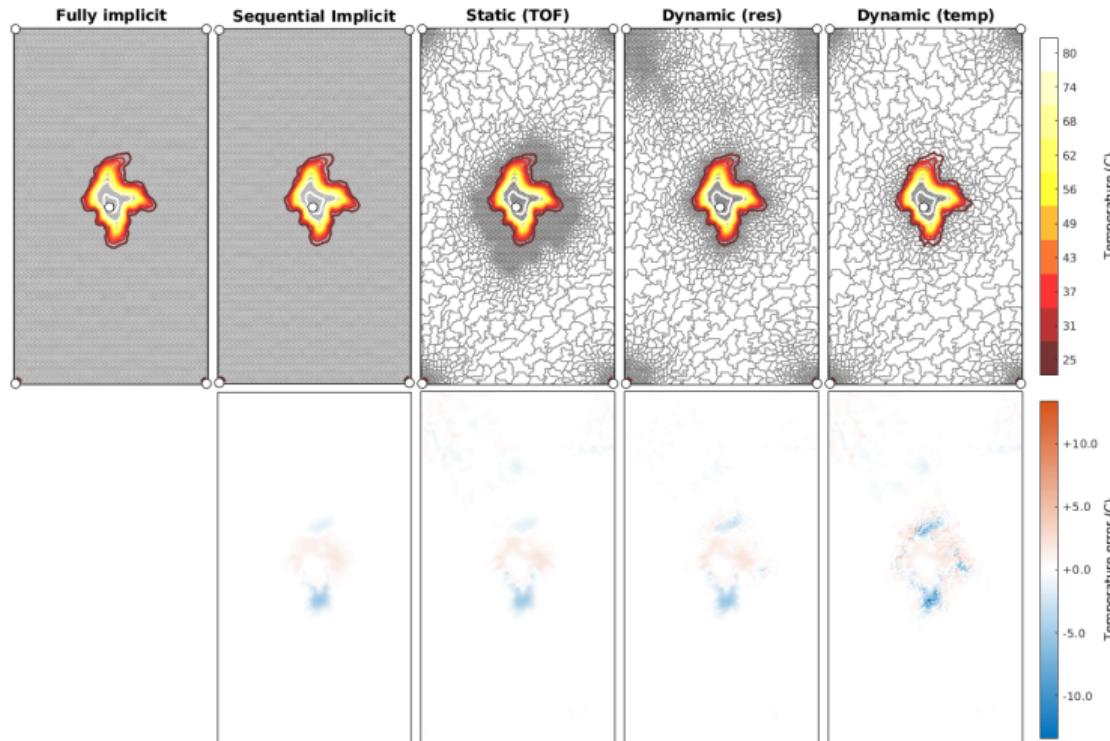
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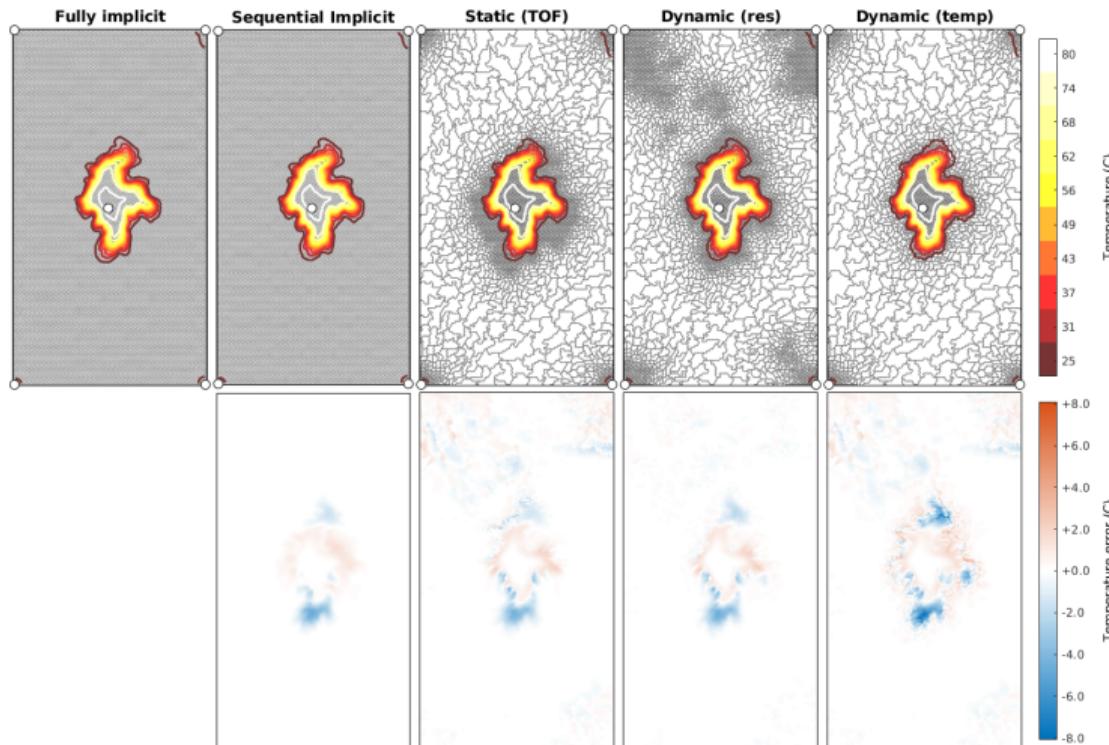
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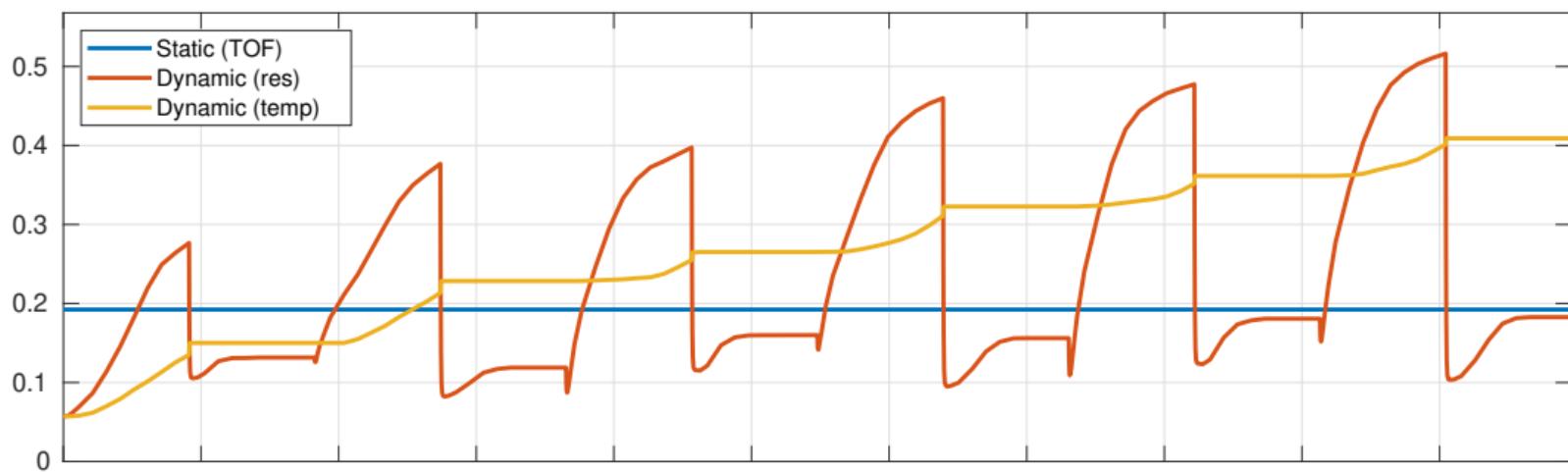


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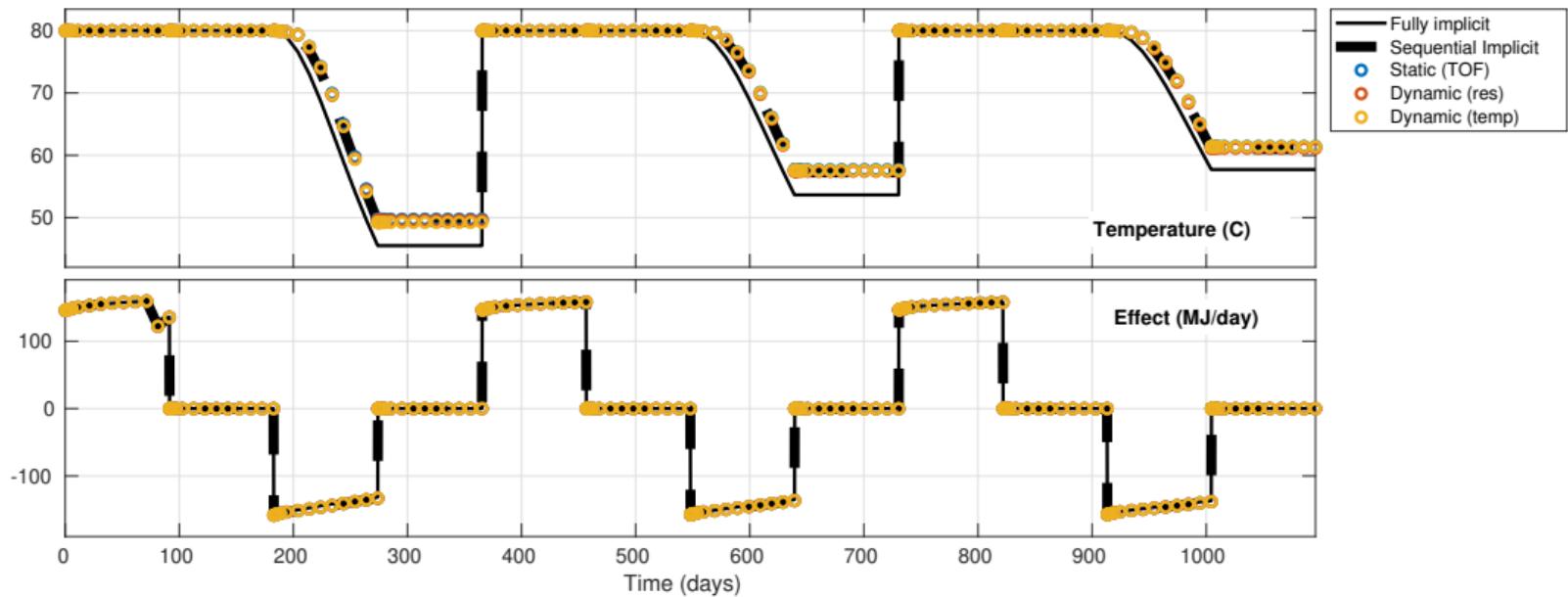
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Dynamic grid relative cell count

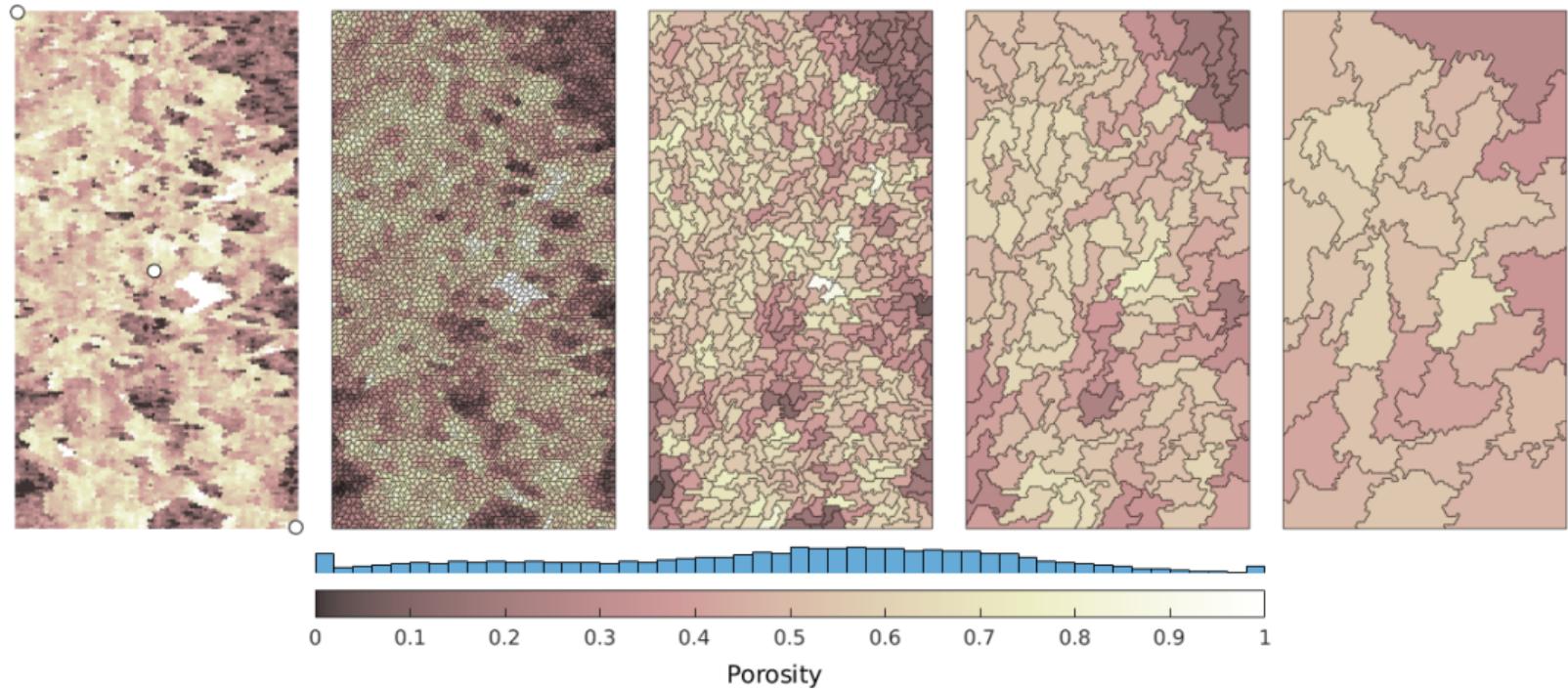


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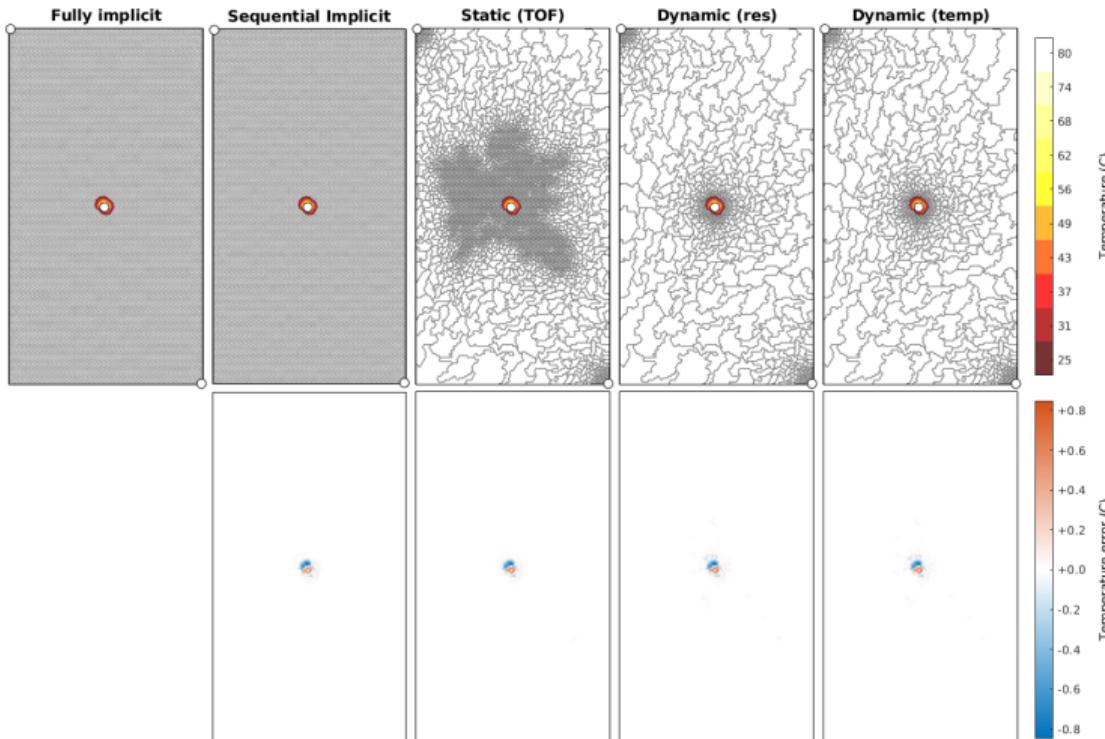
Injection well output



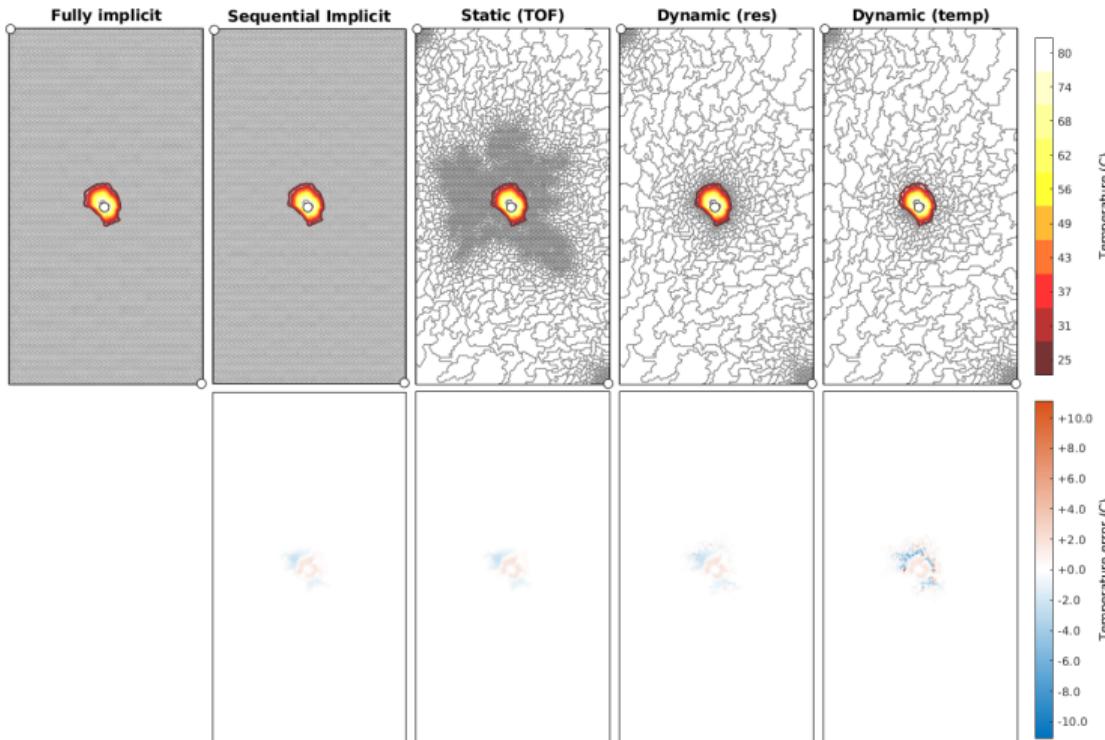
Example: SPE10 Model 2 – Upper Ness (layer 85)



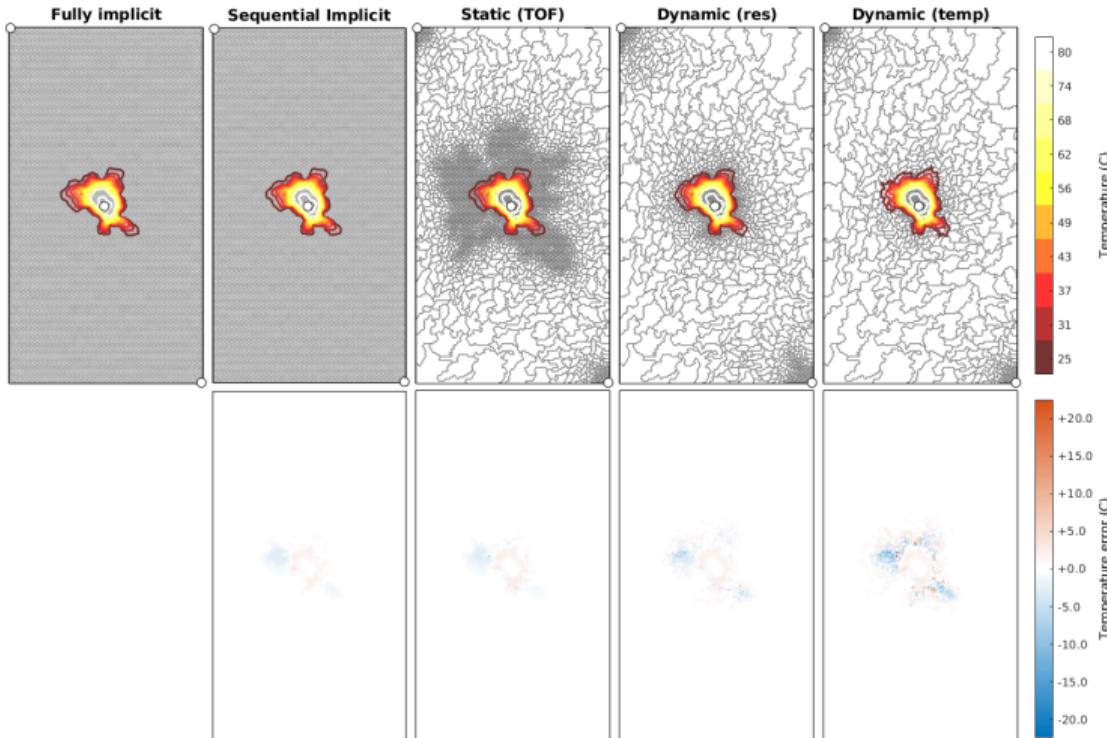
Example: SPE10 Model 2 – Upper Ness (layer 85)



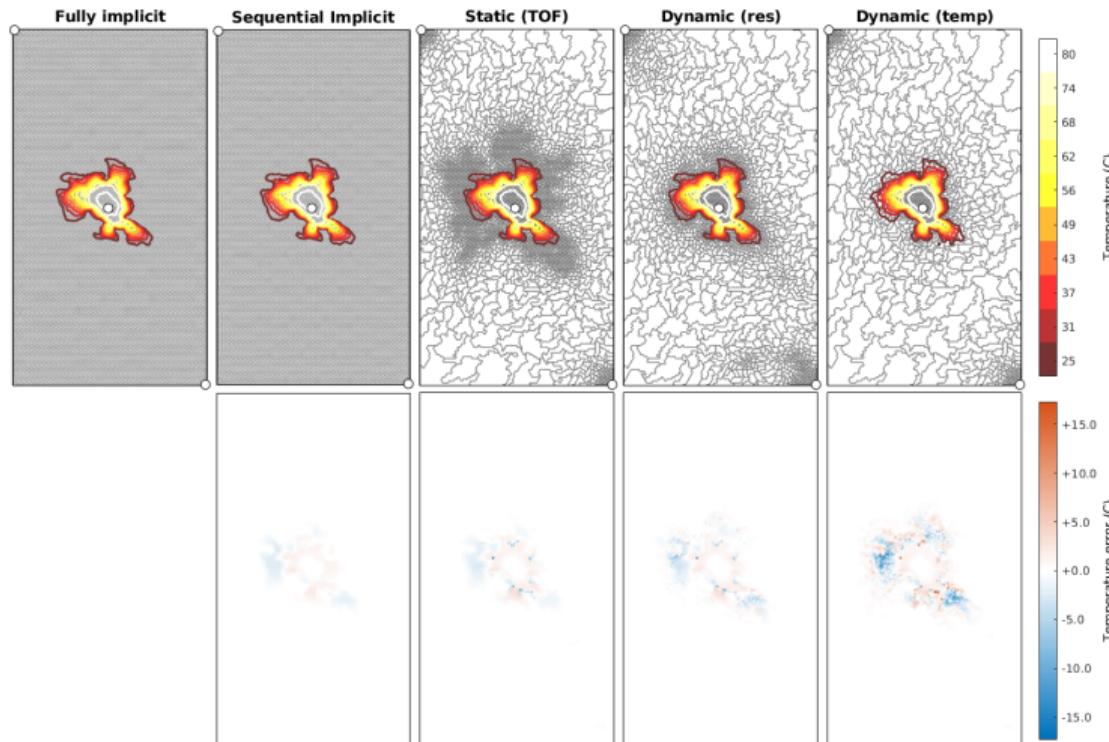
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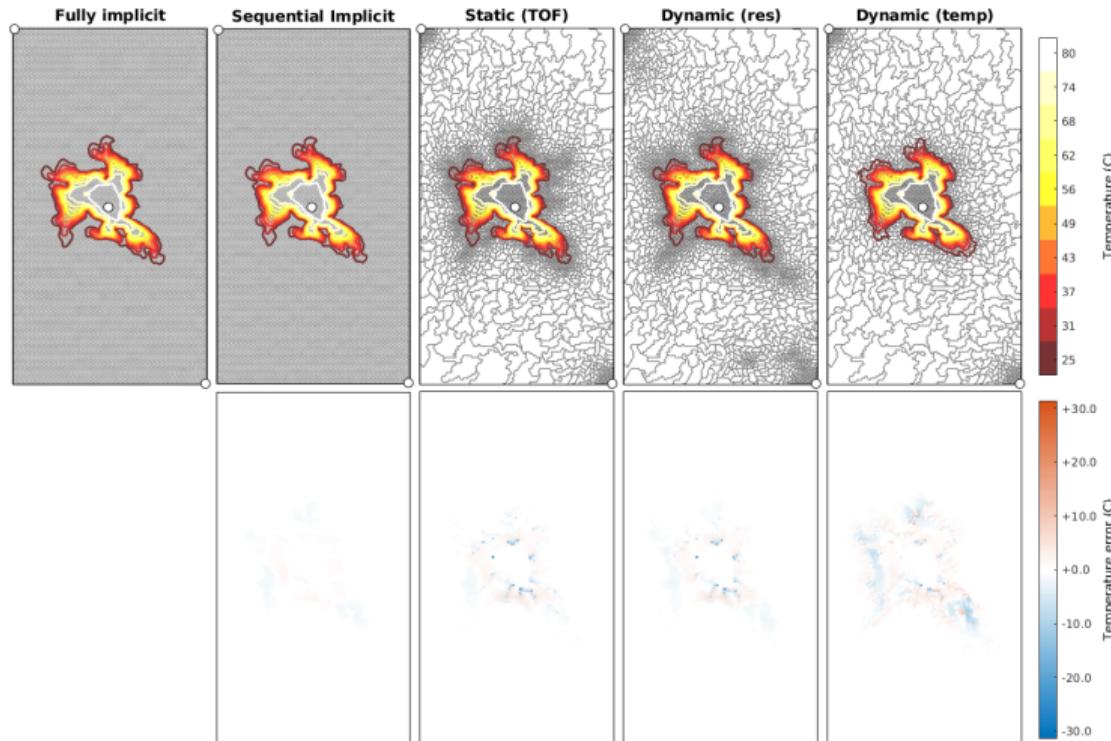
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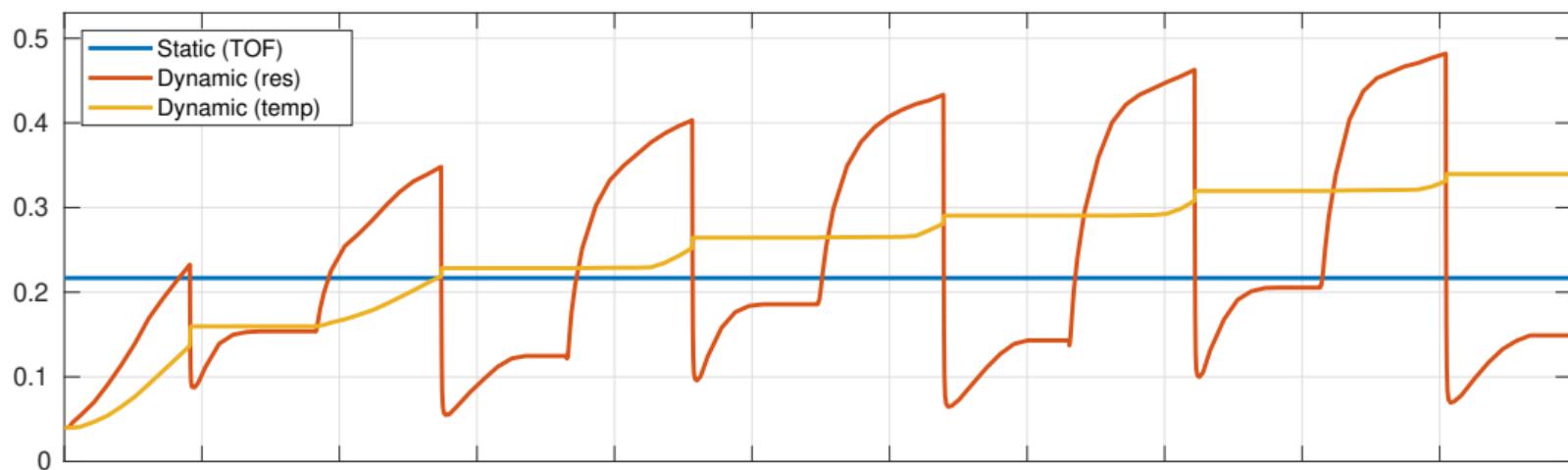


Example: SPE10 Model 2 - Upper Ness (layer 85)



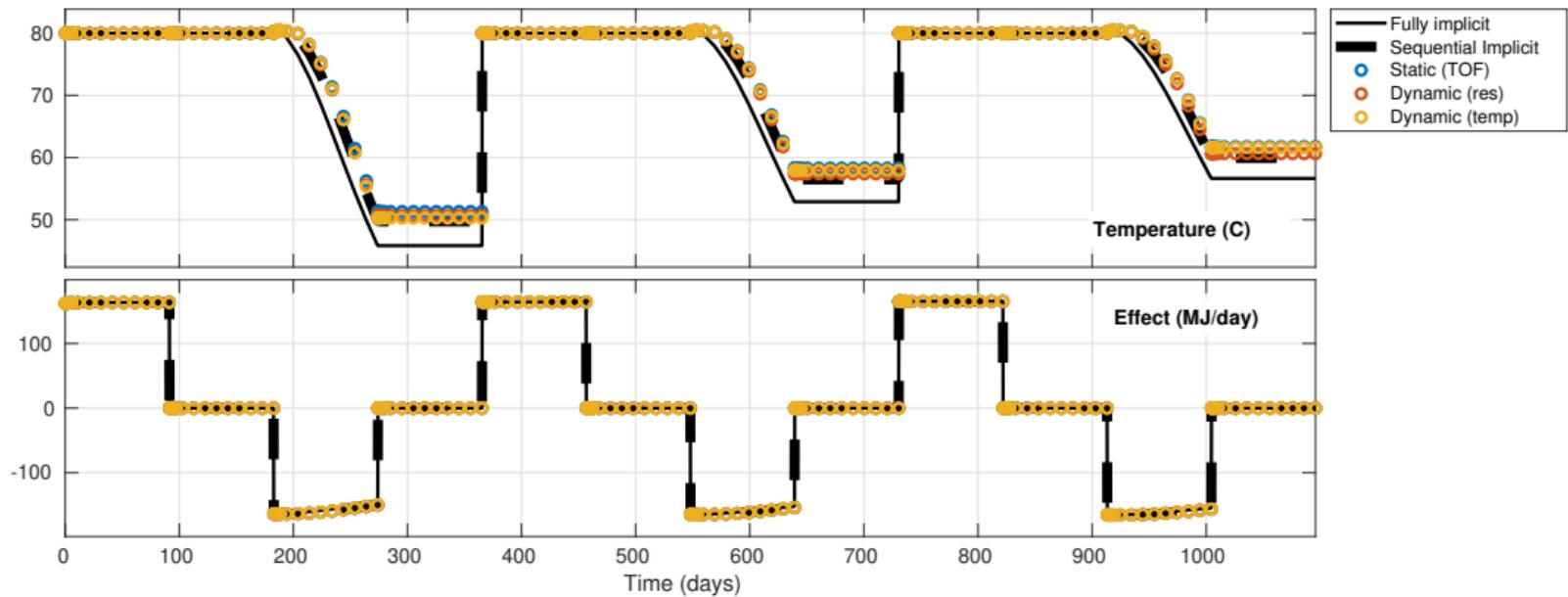
Example: SPE10 Model 2 – Upper Ness (layer 85)

Dynamic grid relative cell count



Example: SPE10 Model 2 – Upper Ness (layer 85)

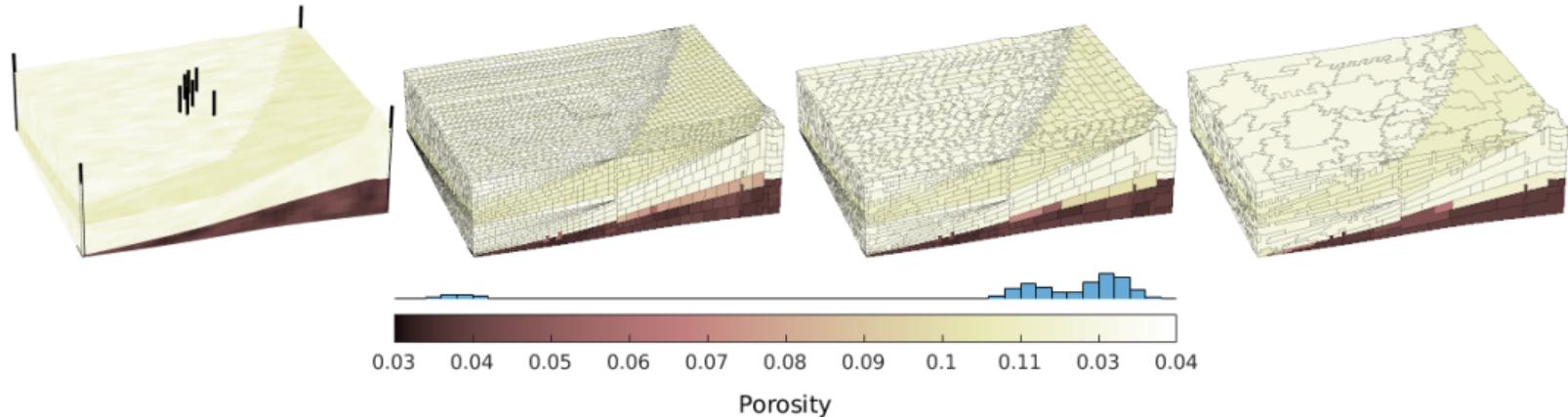
Injection well output



Example: SPE10 Model 2

- Very close match with fine-scale results for all indicators and coarsening strategies
 - Between 49% and 96% reduction in # transport problem dofs
- Point-wise large temperature differences
- Energy discrepancy correction ensures conservation of energy between scales

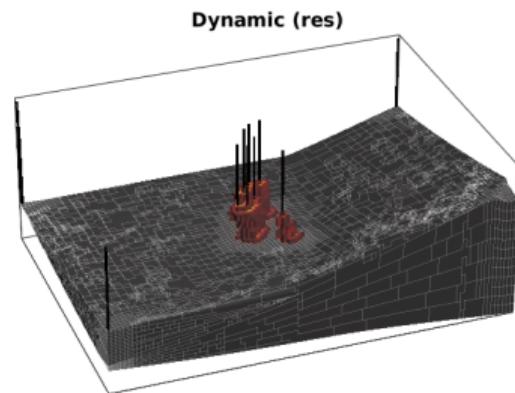
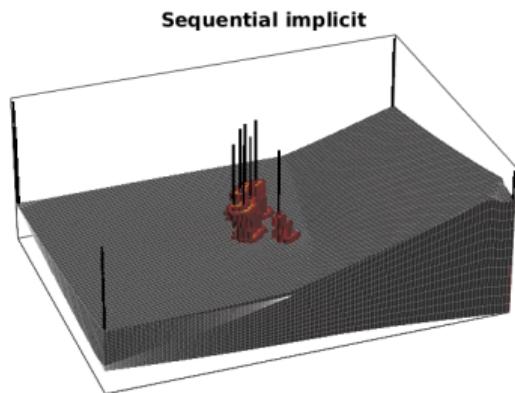
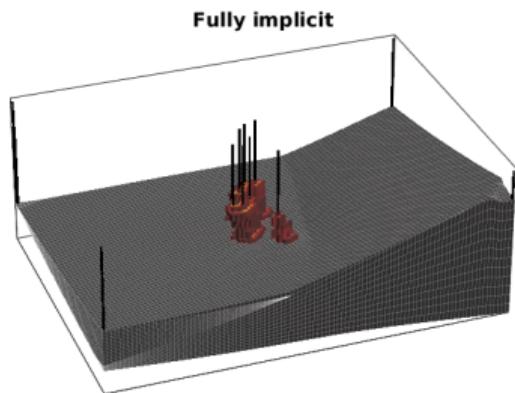
Example: Real(istic) Model



- Model of real geothermal storage site, provided by Ruden AS
- Corner-point grid with four geological layers
- Group of center wells inject at 73 °C over four months, pressure support in corner wells
- Dynamic coarsening with residual-based indicator

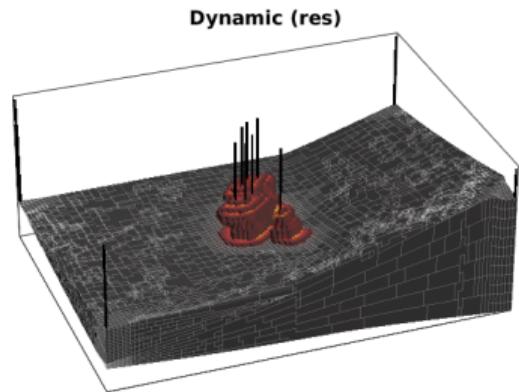
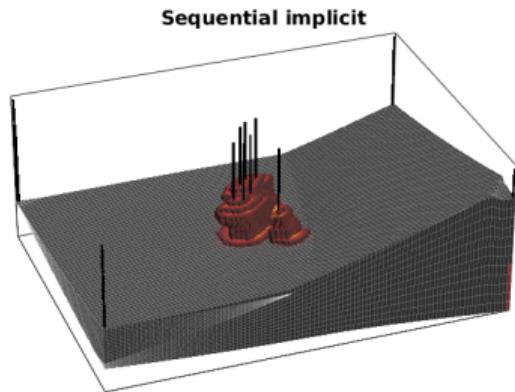
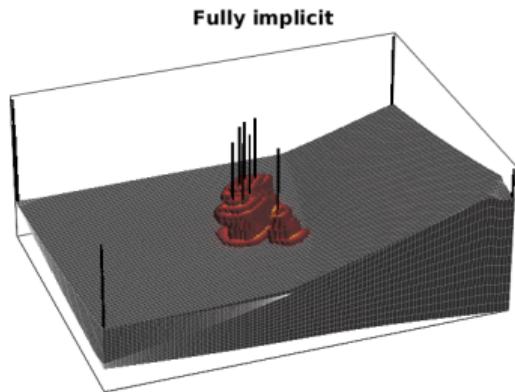
Example: Real(istic) Model

Reservoir temperature



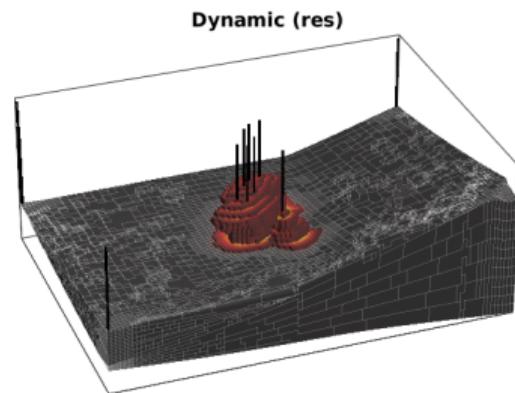
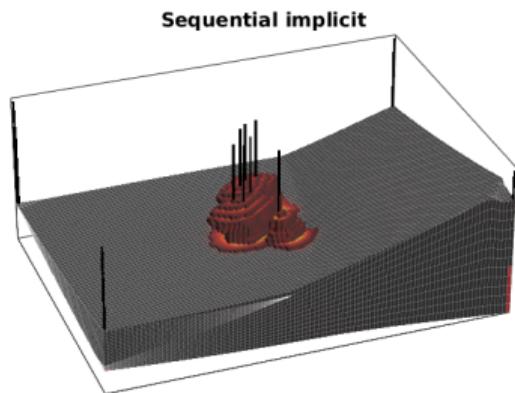
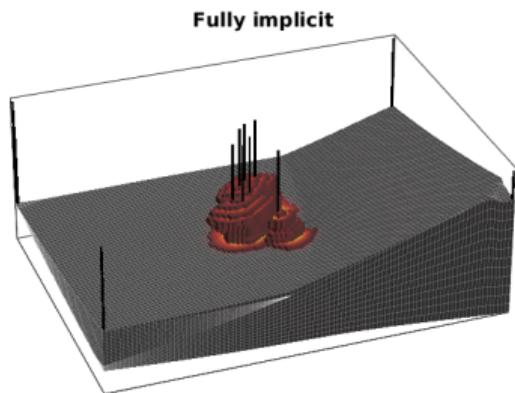
Example: Real(istic) Model

Reservoir temperature



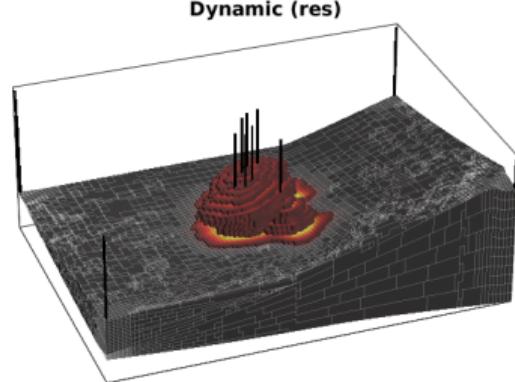
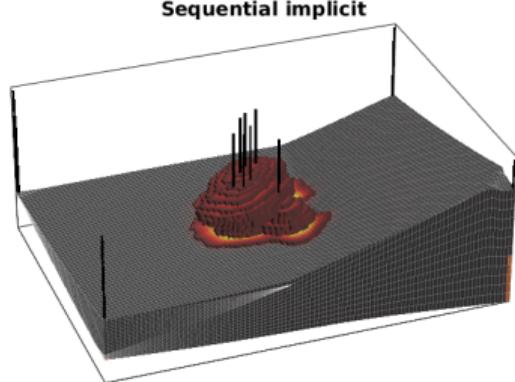
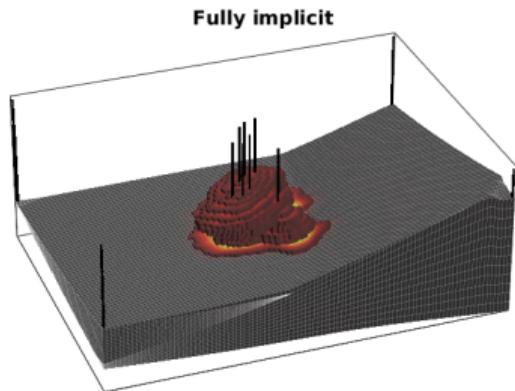
Example: Real(istic) Model

Reservoir temperature



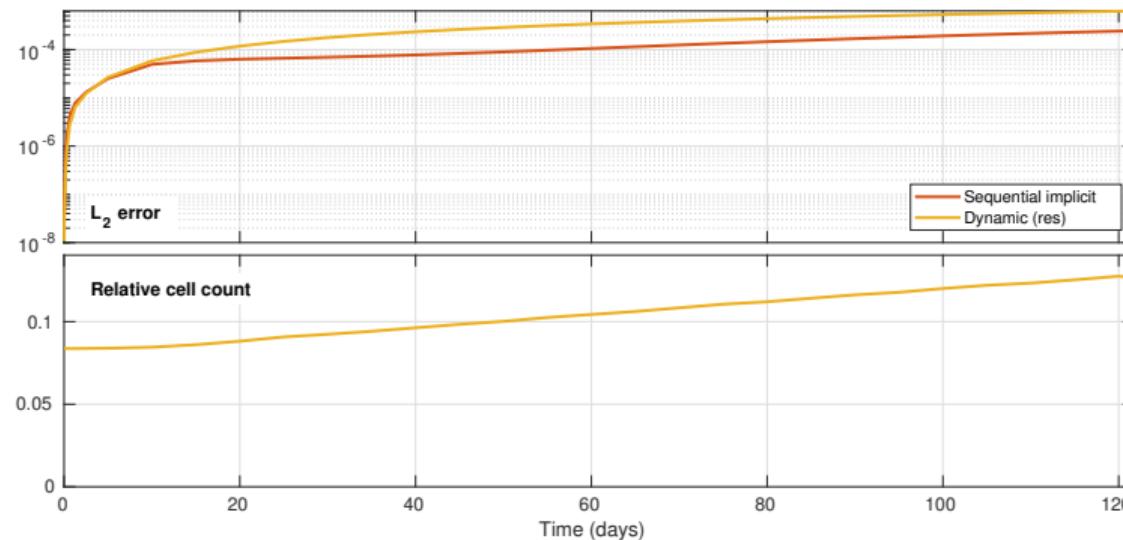
Example: Real(istic) Model

Reservoir temperature



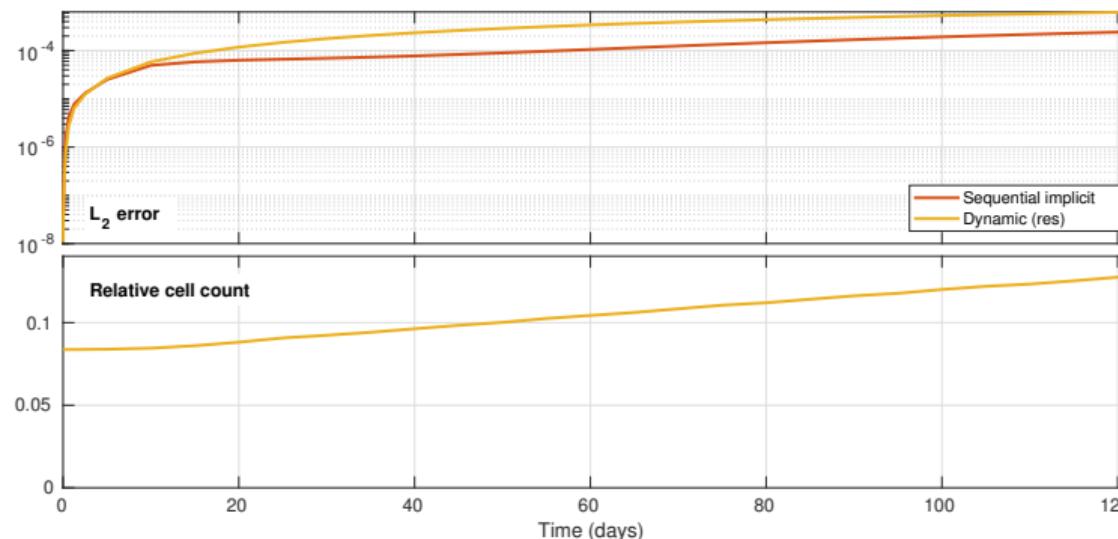
Example: Real(istic) Model

Relative L_2 energy difference from fully implicit and dynamic grid relative cell count



Example: Real(istic) Model

Relative L_2 energy difference from fully implicit and dynamic grid relative cell count



Less than 10^{-3} maximum relative L_2 difference with at least 87% reduction in # transport problem dofs



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Presentation outline

Motivation

Governing equations and discretization

The MATLAB Reservoir Simulation Toolbox

Dynamic coarsening

Numerical examples

Concluding remarks



Concluding Remarks

Conclusions

- Highly flexible **dynamic coarsening** method for geothermal simulations in MRST
 - Sequential splitting of flow and transport/energy
 - Applicable to unstructured, polytopal grids
 - Energy discrepancy correction ensures **conservation of energy**
- Method demonstrated on two examples (low/moderate enthalpy)
 - **Significant reduction** in # dofs in the transport subproblem
 - Very good match with fine-scale solution

Concluding Remarks

Further work

- Optimize implementation and investigate actual CPU speedup
- Test method for high-enthalpy systems (phase changes)
- Solve each subproblem at its appropriate timescale
 - Multiple transport steps for each pressure step
- Combine with a posteriori estimators for error control (Ahmed et al. 2021)

Concluding Remarks

Related talks

MS83B (16:30) *Using MRST for modeling and optimization of operational strategies for a geothermal storage plant in Asker, Norway*

MS50B (16:30) *Optimized graph-based methods for subsurface flow simulations*

Acknowledgments

*Thanks to Marine Collignon (University of Geneva),
Olav Møyner and Knut-Andreas Lie (SINTEF Digital) for fruitful discussions
Thanks to Ruden AS for allowing use of the field model in this work*

Technology for a
better society

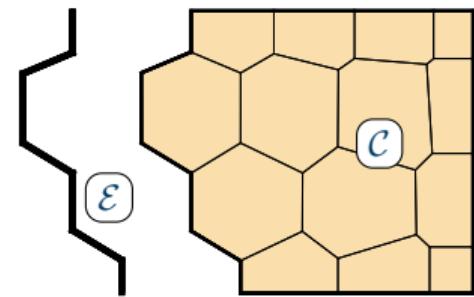


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Extra slides

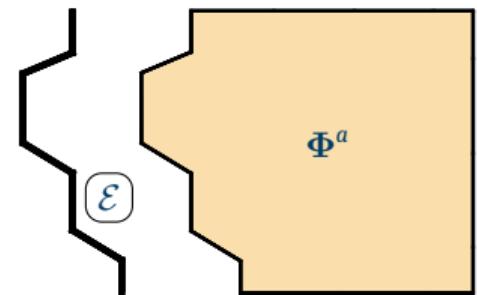


Dynamic coarsening – Mapping parameters



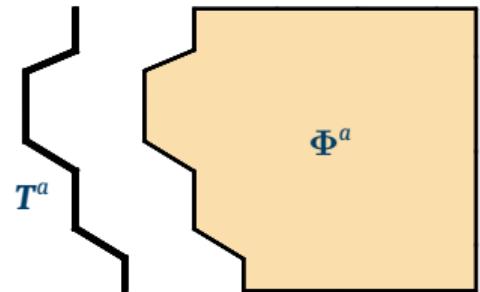
Dynamic coarsening – Mapping parameters

- Accumulate pore volumes: $\Phi^a = \sum_{i \in \mathcal{C}} \Phi_i$



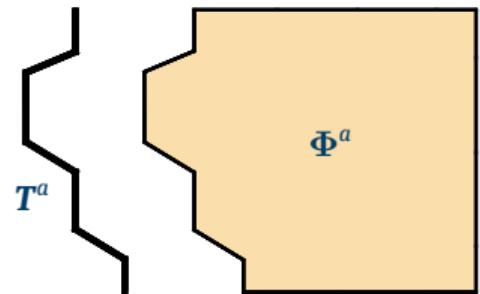
Dynamic coarsening – Mapping parameters

- Accumulate pore volumes: $\Phi^a = \sum_{i \in \mathcal{C}} \Phi_i$
- Compute transmissibilities (multiple options):
 1. Accumulation: $\mathbf{T}^a = \sum_{(m,n) \in \mathcal{E}} \mathbf{T}_{mn}$
 2. Upscale permeability and coarse geometry \rightarrow compute \mathbf{T}^a
 3. Compute representative transmissibility given flux $\mathbf{T}_{mn}^a = \mathbf{v}_{mn} / (\mathbf{p}_m - \mathbf{p}_n)$



Dynamic coarsening – Mapping parameters

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 3. Compute representative transmissibility given flux $\mathbf{T}_{mn}^a = \mathbf{v}_{mn} / (\mathbf{p}_m - \mathbf{p}_n)$
- These parameters can be computed in a preprocessing step (except transmissibility option 3)
 \rightarrow Adapting the grid amounts to looking up precomputed parameters



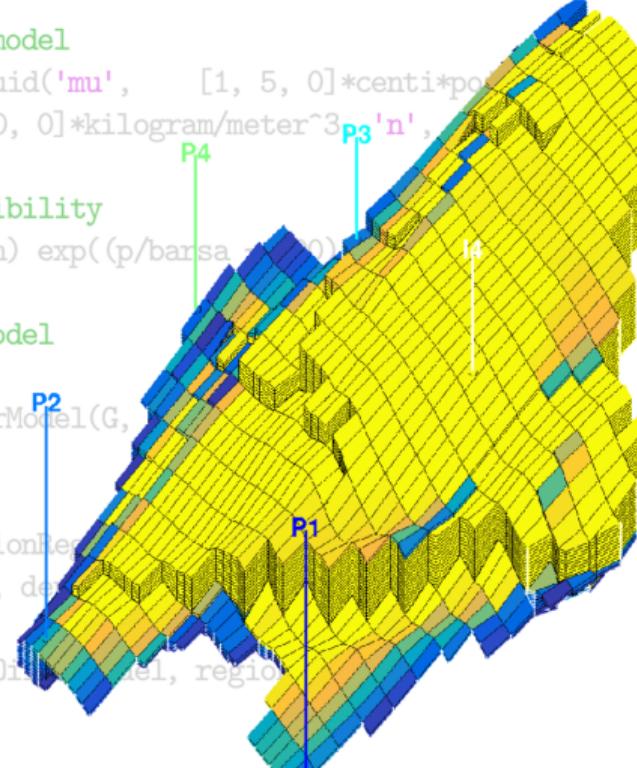
MATLAB Reservoir Simulation Toolbox (MRST)

Transforming research on reservoir modelling

Unique prototyping platform:

- Standard data formats
- Data structures/library routines
- Fully unstructured grids
- Rapid prototyping:
 - Differentiation operators
 - Automatic differentiation
 - Object-oriented framework
 - State functions
- Industry-standard simulation

```
% Three-phase template model
fluid = initSimpleADIFluid('mu', [1, 5, 0]*centi*po
                           'rho', [1000, 700, 0]*kilogram/meter^3
                           'n', [1, 1, 1]*centi*bar);
% Constant oil compressibility
fluid.b0 = @(p, varargin) exp((p/barsa - 100));
% Construct reservoir model
gravity = resetOn(model);
model = TwoPhaseOilWaterModel(G,
% Define initial state
region = getInitializationRegion(model);
'datum_depth', depth);
state0 = initStateBlackOilWater(model, region);
% Define schedule
schedule = simpleSchedule(timesteps, 'W', 1);
```



www.mrst.no

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Differentiation operators

Writing discrete equations on form very close to continuous equations

$$\nabla \cdot \vec{H} \quad \vec{H} = -(\lambda_f + \lambda_r) \nabla T$$
$$\text{div}(H) \quad H = -(lambdaF + lambdaR) .* \text{grad}(T)$$

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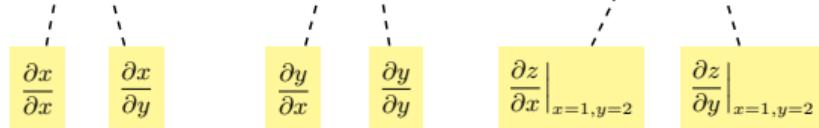
$$\text{div}(H) \quad H = -(lambdaF + lambdaR) .* \text{grad}(T)$$

Automatic differentiation

Combine chain rule and elementary differentiation rules by means of operator overloading to analytically evaluate all derivatives
→ Computing Jacobians amounts to writing down residual equations.

```
[x,y] = initVariablesADI(1,2); z = 3*exp(-x*y)
```

```
x = ADI Properties:  
val: 1  
jac: {[1] [0]}  
y = ADI Properties:  
val: 2  
jac: {[0] [1]}  
z = ADI Properties:  
val: 0.4060  
jac: {[[-0.8120] [-0.4060]}}
```



MATLAB Reservoir Simulation Toolbox (MRST)

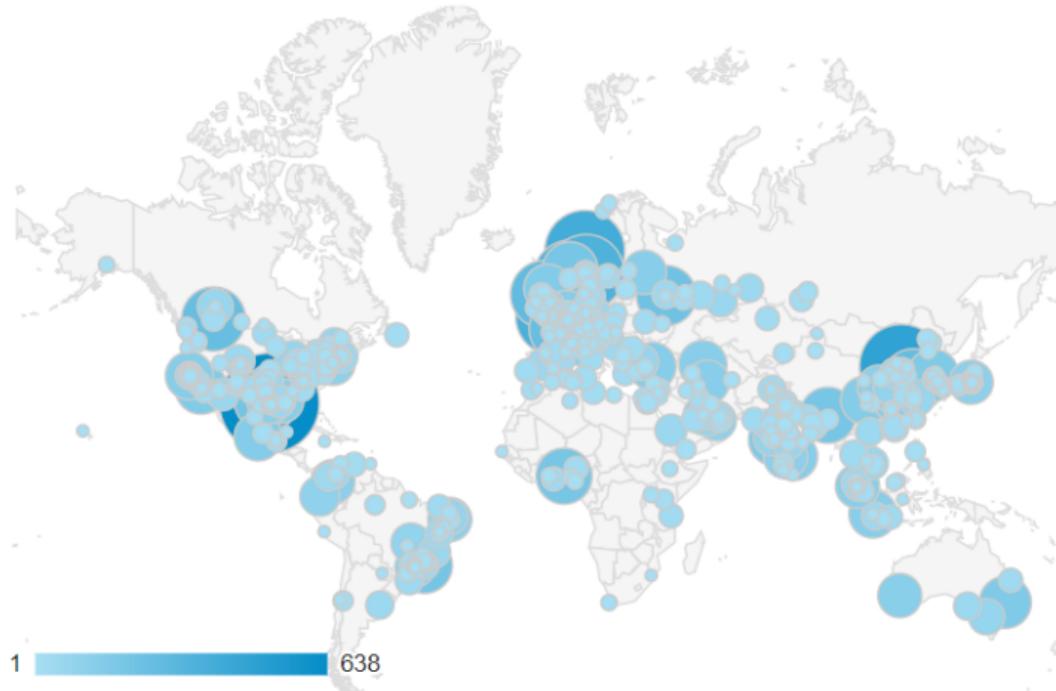
Transforming research on
reservoir modelling

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