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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:  $\vec{v}_f = -\frac{1}{\mu_f} \mathbf{K}(\nabla p - \rho_f g \nabla \vec{z})$

$\phi$	Pore volume	$\mathbf{K}$	Permeability	$\Lambda$	Thermal conductivity	$\vec{g}$	Gravity
$\rho$	Density	$\mu$	Viscosity	$u$	Internal energy	$h$	Enthalpy
$p$	Pressure	$T$	Temperature	$q$	Sources/sinks	$r/f$	Fluid/rock

# Governing equations and discretization

Single-phase conservation of mass on residual form

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Mass

Flux

Sources/sinks

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Conservation of energy on residual form

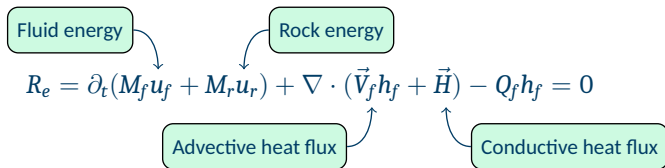
$$R_e = \partial_t(\phi \rho_f u_f + [1 - \phi] \rho_r u_r) + \nabla \cdot (\rho_f \vec{v}_f h_f + \vec{H}) - \rho_f q_f h_f = 0$$

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

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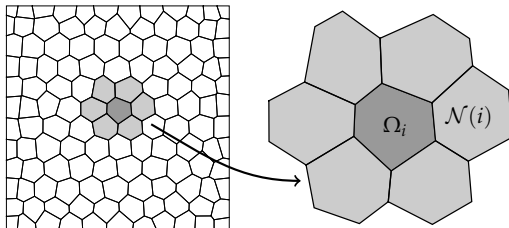
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# Governing equations and discretization

Finite volumes in space, implicit backward Euler in time

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



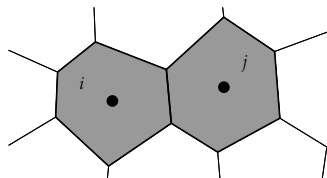
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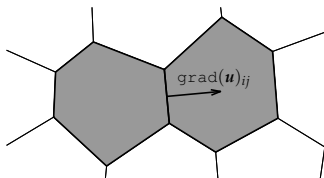
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$$\mathbf{V} = -\text{upw}(\rho/\mu) [\mathbf{K} \text{grad}(\mathbf{p}) - g \text{fav}(\rho) \mathbf{K} \text{grad}(\mathbf{z})]$$

- $\mathbf{K} \text{grad}$ : 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)



Scalar field  $u$



Gradient  $\text{grad}(u)_{ij} = u_j - u_i$

$$\mathbf{K} \text{grad} = \mathbf{T} \text{grad}$$

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

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

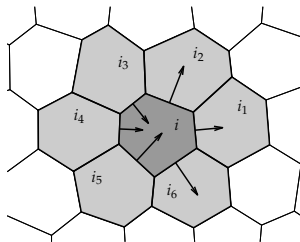
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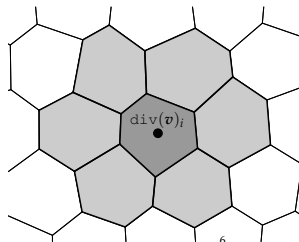
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- **div**: discrete divergence operator



Interface flux field  $\mathbf{v}$



$$\text{Divergence } \text{div}(\mathbf{v})_i = \sum_{k=1}^6 v_{ii_k}$$



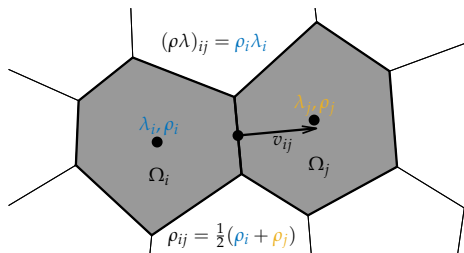
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- **upw**: Upwind discretization (single-point here); **favg**: Face average operator



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Finite volumes in space, implicit backward Euler in time

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**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  $T$**  and **total saturation  $S_t$**   
 $\rightarrow$  allow total saturation to be  $\neq 1$ , multiply densities by total saturation

$$\rho_f \rightarrow S_t \rho_f, \quad \rho_r \rightarrow S_t \rho_r$$



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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

```
% Three-phase template model
```

```
fluid = initSimpleADIFluid('mu', [1, 5, 0]*centi*po  
'rho', [1000, 700, 0]*kilogram/meter^3, 'n',
```

```
% Constant oil compressibility
```

```
fluid.b0 = @(p, varargin) exp((p/barsa - 10)
```

```
% Construct reservoir model
```

```
gravity reset on
```

```
model = TwoPhaseOilWaterModel(G,
```

```
%% Define initial state
```

```
region = getInitializationKey
```

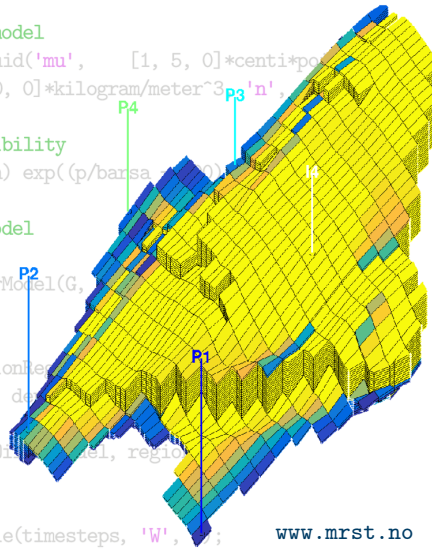
```
'datum_depth', de
```

```
state0 = initStateBlackOil(model, region
```

```
% Define schedule
```

```
schedule = simpleSchedule(timesteps, 'W',
```

[www.mrst.no](http://www.mrst.no)



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### Differentiation operators

*Write discrete equations on form very close to continuous equations*

$$\nabla \cdot \vec{H}$$

$$\text{div}(\mathbf{H})$$

$$\vec{H} = -(\lambda_f + \lambda_r) \nabla T$$

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

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

### 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]}

$$\frac{\partial x}{\partial x}$$

$$\frac{\partial x}{\partial y}$$

y = ADI Properties:  
val: 2  
jac: {[0] [1]}

$$\frac{\partial y}{\partial x}$$

$$\frac{\partial y}{\partial y}$$

z = ADI Properties:  
val: 0.4060  
jac: {[ -0.8120] [-0.4060]}

$$\frac{\partial z}{\partial x} \Big|_{x=1,y=2}$$

$$\frac{\partial z}{\partial y} \Big|_{x=1,y=2}$$





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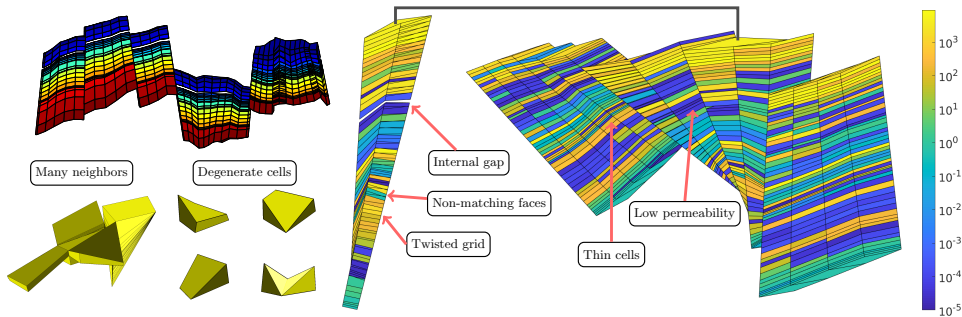
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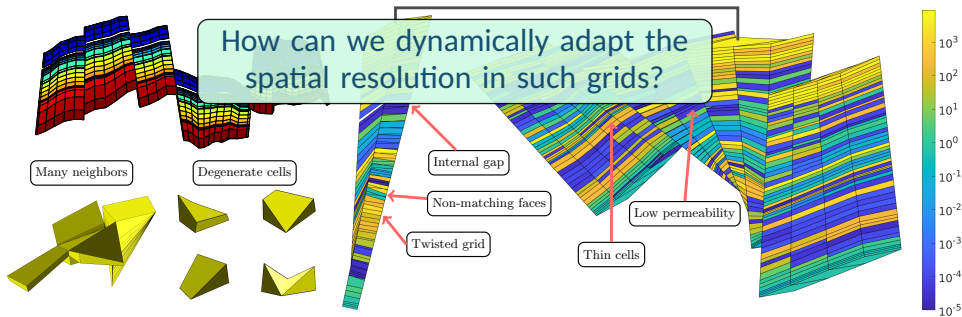
# Reservoir simulation grids

- Subsurface reservoirs are complex: layers, faults, fractures, erosion, wells, ...
- Simulation models often upscaled  $\rightarrow$  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

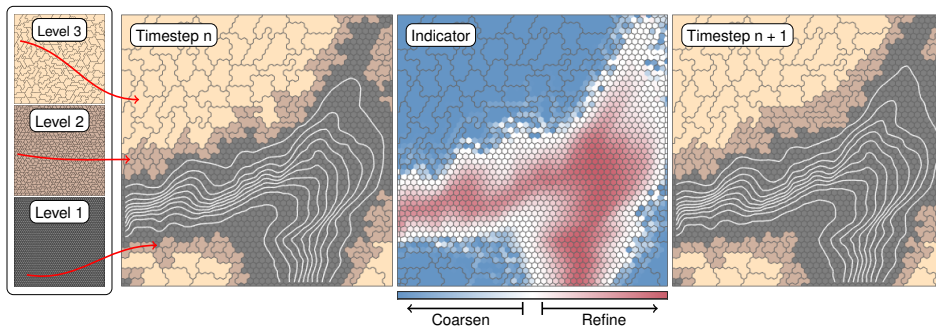
## Dynamic coarsening

- Transport of geothermal heat chiefly confined to proximity of wells
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- Reservoir engineering applications:

# cells in simulation grid  $\ll$  # cells in geocellular model

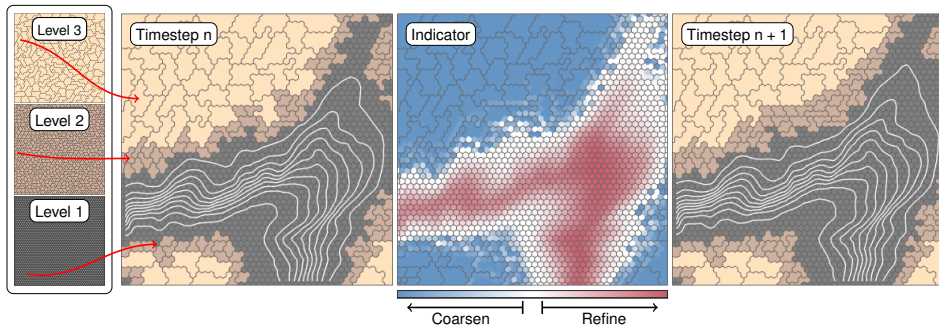
- State-of-the-art multiscale methods (attempt to) bridge gap for pressure problems (Jenny, S. H. Lee, and Tchelepi 2006; Møyner and Lie 2016; Lie et al. 2017, etc.)
- Here: attempt to bridge this gap for transport problems by **dynamic coarsening**

# Dynamic coarsening



Quandalle and Besset 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



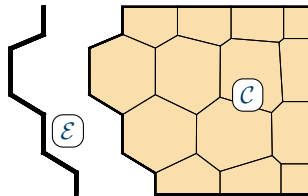
Keep track of which cells to refine/coarsen using coarsening indicator  $\mathcal{I}(u) \in \mathbb{R}_+^N$

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**

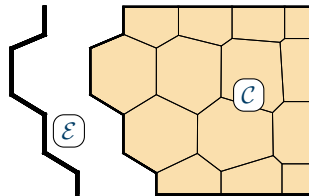




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1. Accumulate coarse-block energy:  $\sum_{i \in \mathcal{C}} (\mathbf{M}_f \mathbf{u}_f + \mathbf{M}_r \mathbf{u}_r)_i$

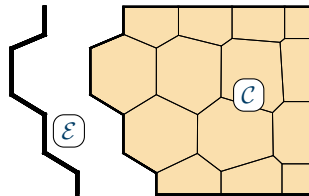


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$$\mathbf{p}^a = \frac{1}{\Phi^a} \sum_{j \in \mathcal{C}} (\Phi \mathbf{p})_j, \quad \mathbf{T}^a = \frac{1}{\Phi^a} \sum_{j \in \mathcal{C}} (\Phi \mathbf{T})_j, \quad \mathbf{v}^a = \sum_{(m,n) \in \mathcal{E}} \mathbf{v}_{mn}$$

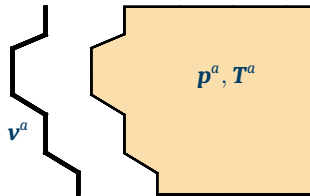


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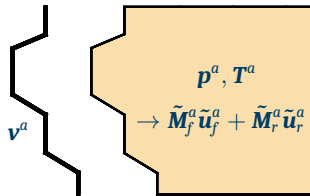
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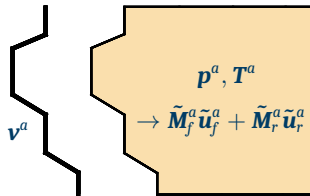


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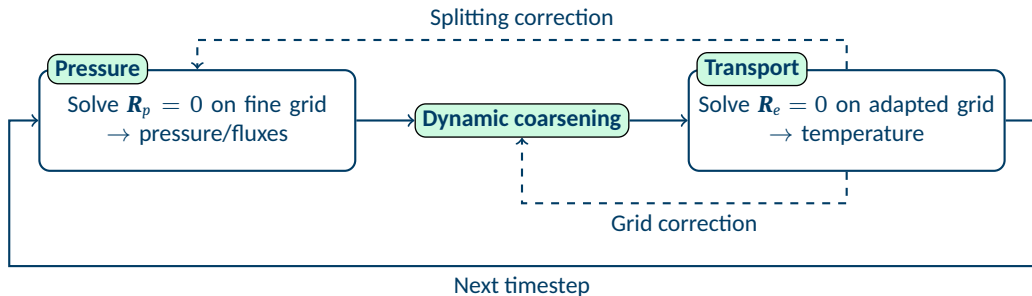
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3. Compute energy in coarse block with adapted properties
4. Ensure conservation of energy through **energy discrepancy**  $\rightarrow$  density correction

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

## Dynamic coarsening – Solution procedure





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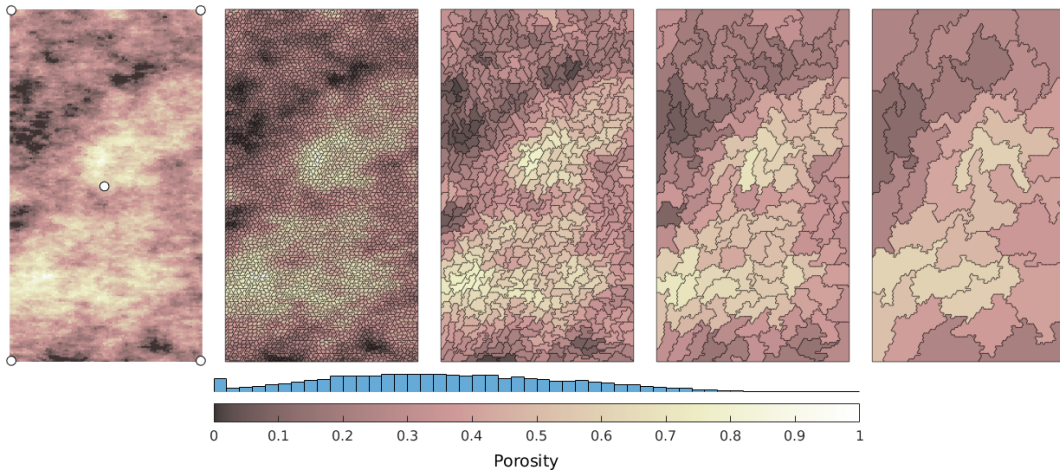
Concluding remarks

## Example: SPE10 Model 2

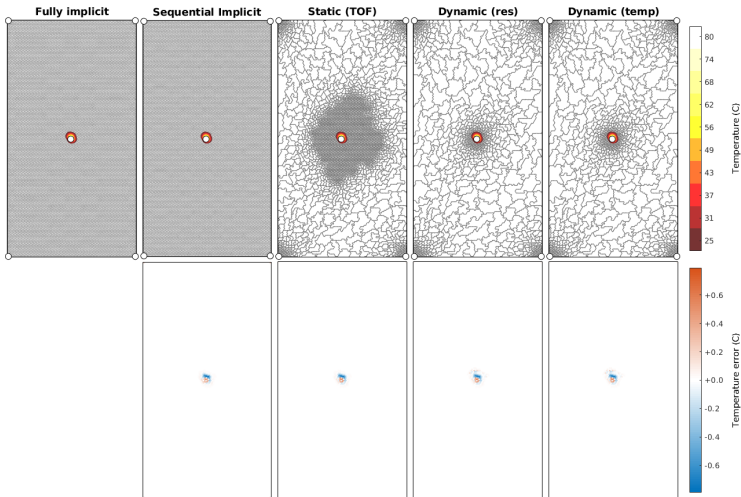
- Heat storage in two different layers of SPE10 Model 2
- Three one-year cycles of storage in center well with pressure support in corner wells
  1. **Charge:** 3 months of injection at 80 °C, bhp = 70 bar
  2. **Rest:** 3 months with no driving forces
  3. **Discharge:** 3 months of extraction, bhp = 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



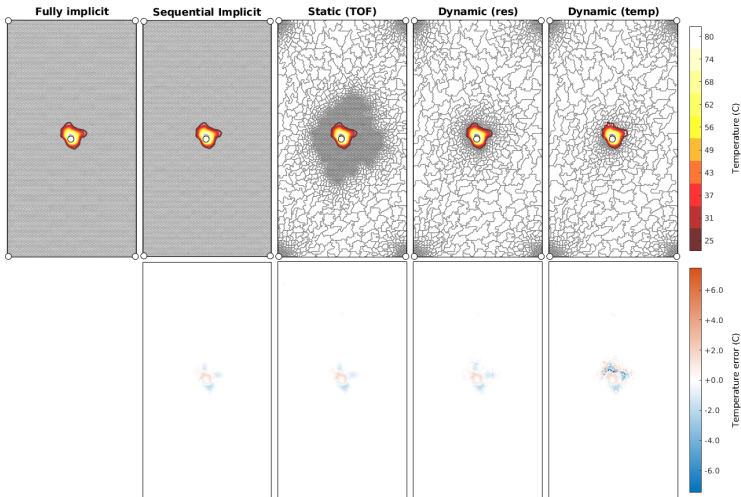
## Example: SPE10 Model 2 – Tarbert (layer 10)



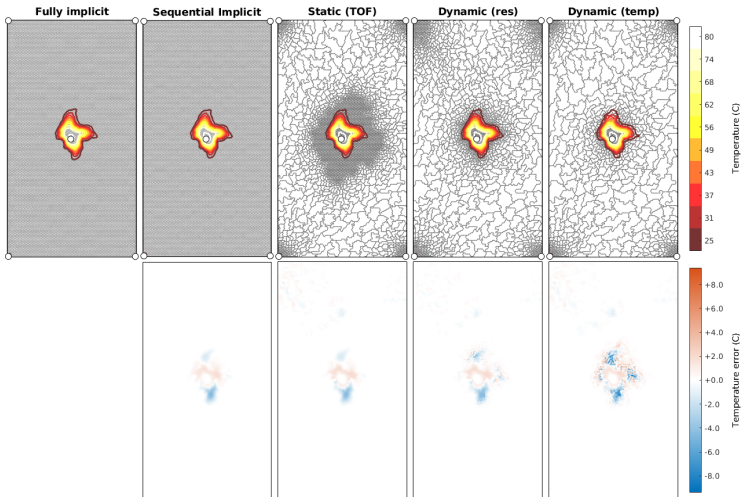
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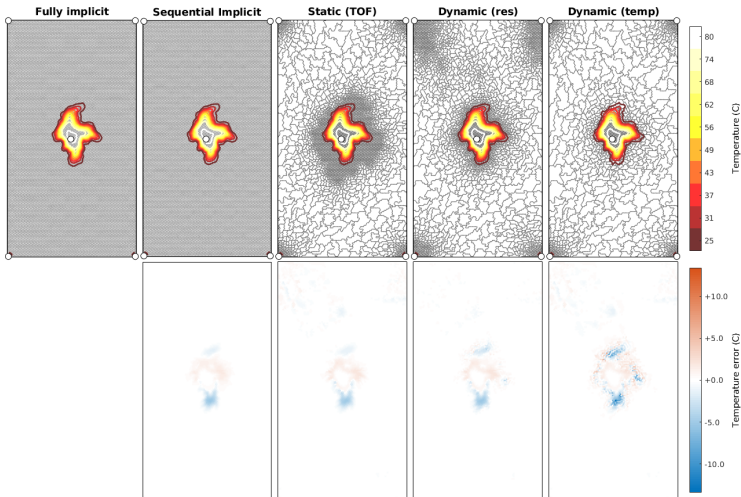
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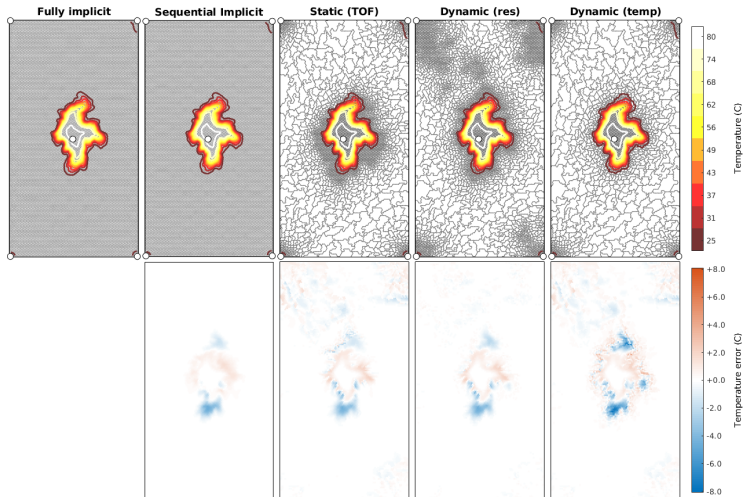
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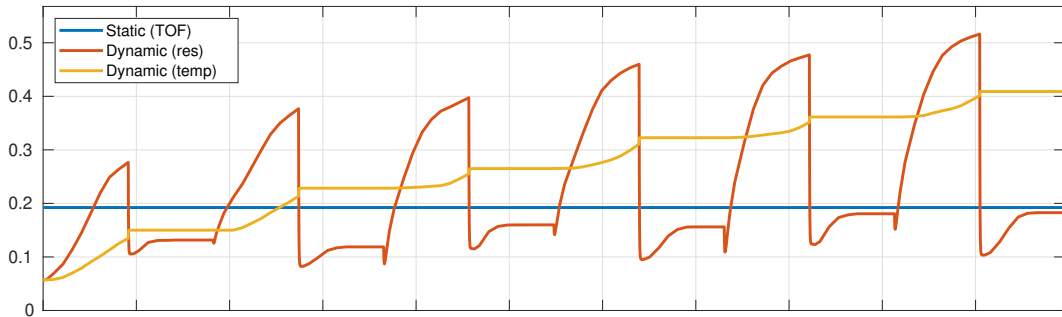


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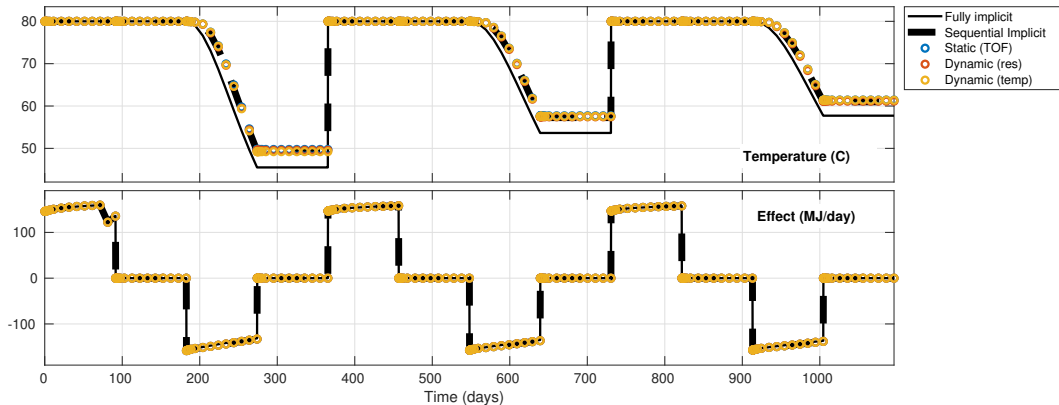
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Dynamic grid relative cell count



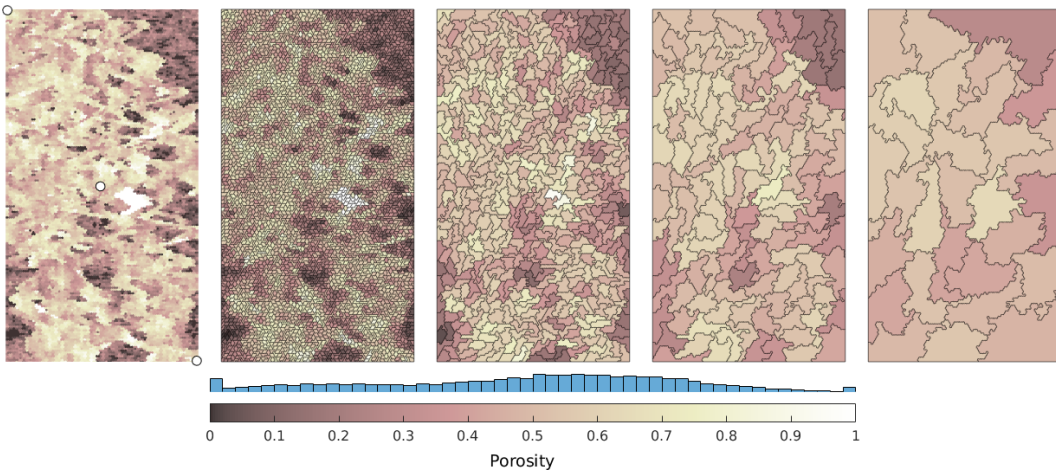
## Example: SPE10 Model 2 – Tarbert (layer 10)

### Injection well output

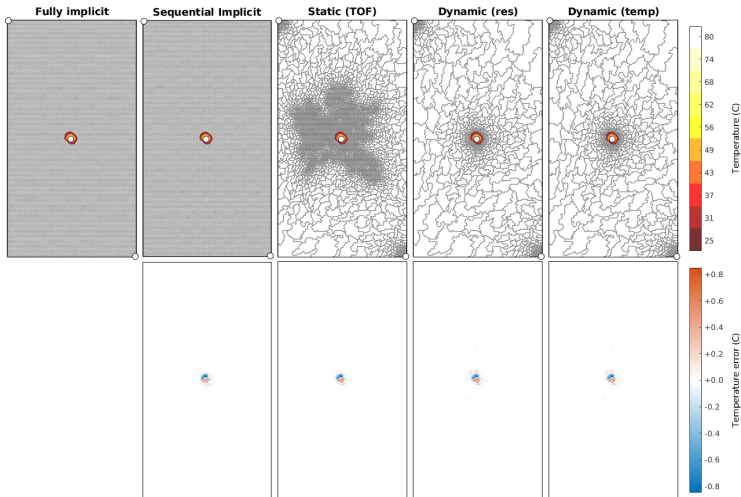




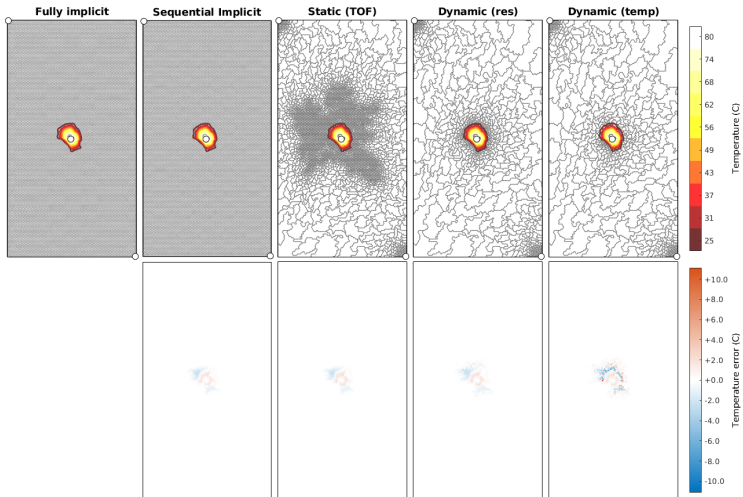
## Example: SPE10 Model 2 – Upper Ness (layer 85)



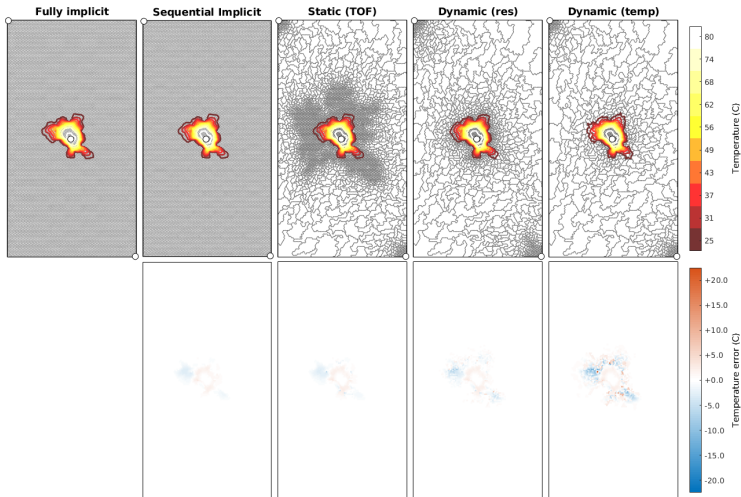
## Example: SPE10 Model 2 – Upper Ness (layer 85)



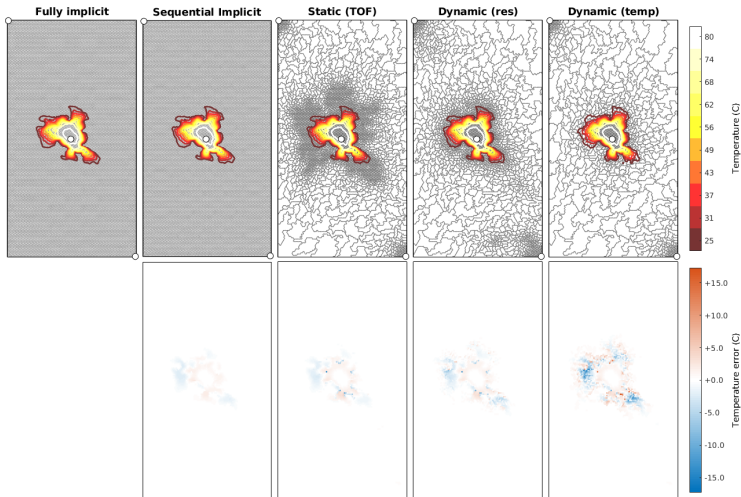
## Example: SPE10 Model 2 – Upper Ness (layer 85)



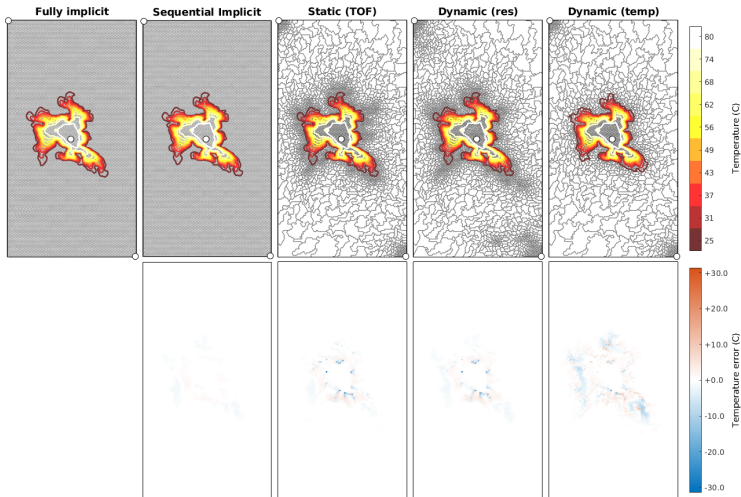
## Example: SPE10 Model 2 – Upper Ness (layer 85)



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

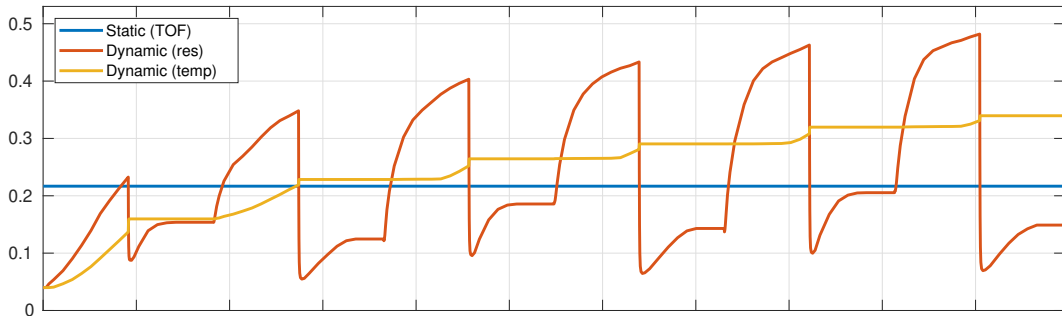


## Example: SPE10 Model 2 – Upper Ness (layer 85)



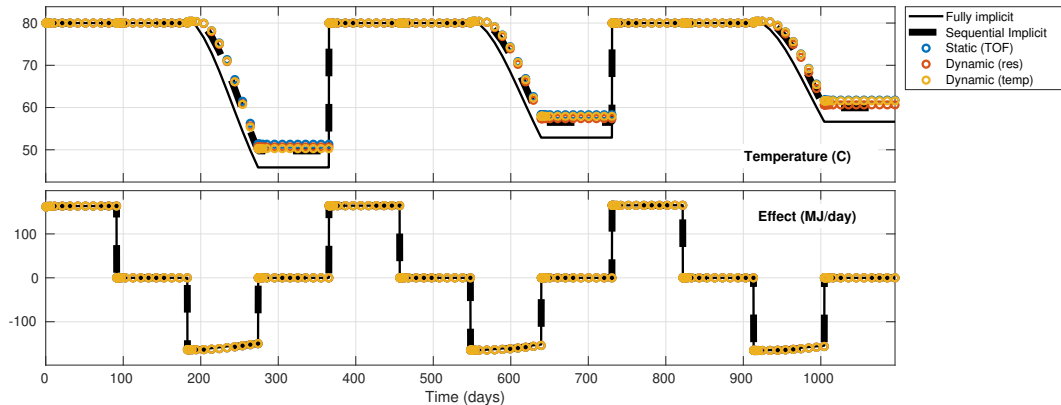
## Example: SPE10 Model 2 – Upper Ness (layer 85)

Dynamic grid relative cell count



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

### Injection well output

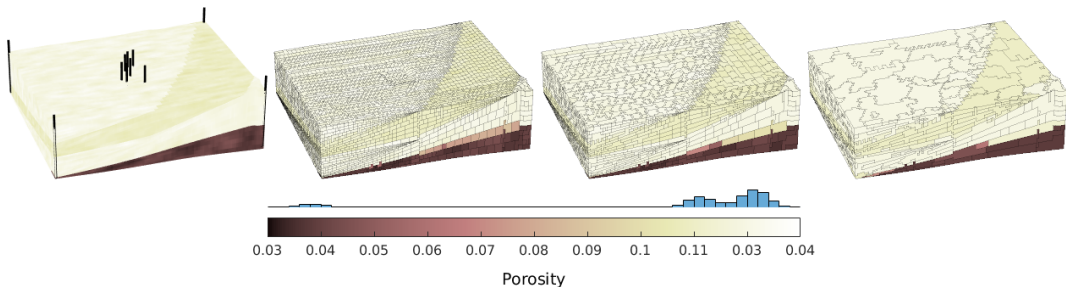




## Example: SPE10 Model 2

- 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

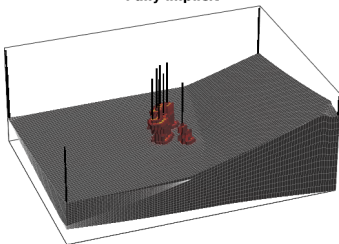


- Model of real geothermal storage site, provided by Ruden AS
- Corner-point grid with four geological layers
- Group of center wells inject at 73 °C over four months, pressure support in corner wells
- Dynamic coarsening with residual-based indicator

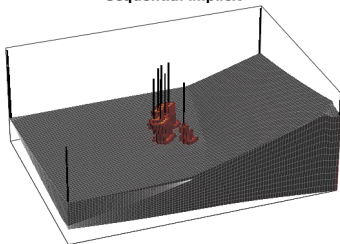
## Example: Real(istic) Model

### Reservoir temperature

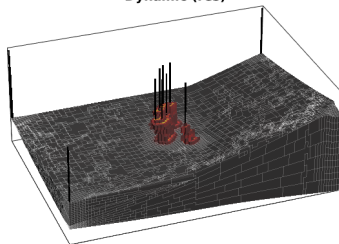
Fully implicit



Sequential implicit



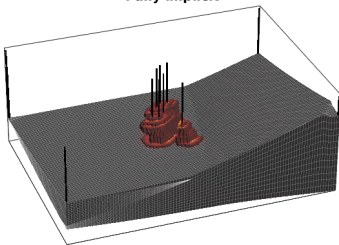
Dynamic (res)



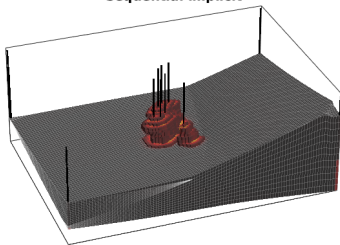
## Example: Real(istic) Model

### Reservoir temperature

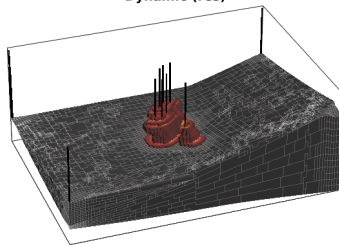
Fully implicit



Sequential implicit



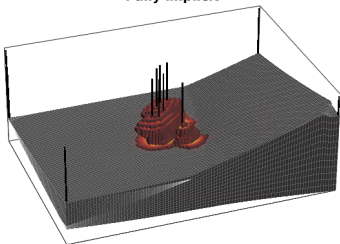
Dynamic (res)



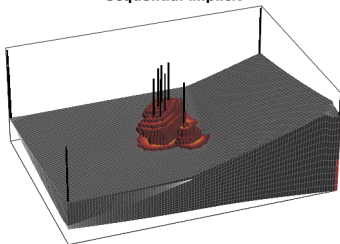
## Example: Real(istic) Model

### Reservoir temperature

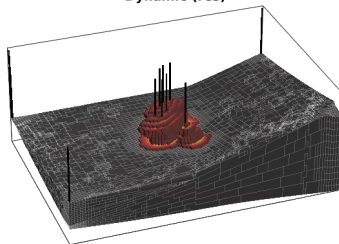
Fully implicit



Sequential implicit



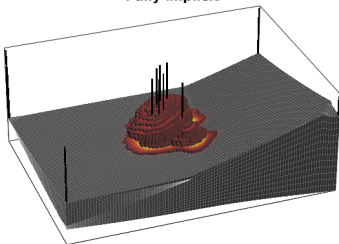
Dynamic (res)



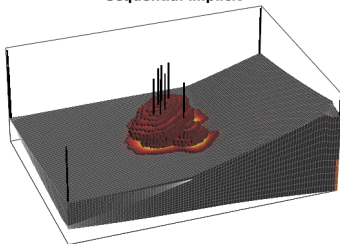
## Example: Real(istic) Model

### Reservoir temperature

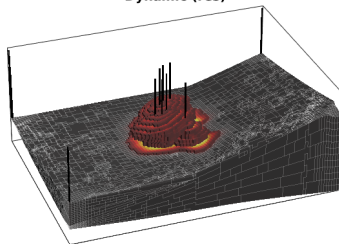
Fully implicit



Sequential implicit

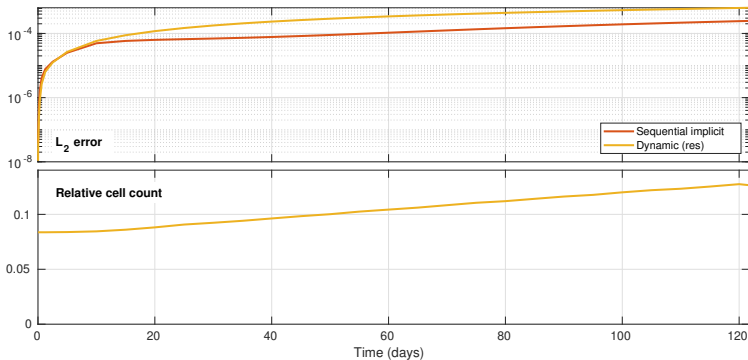


Dynamic (res)



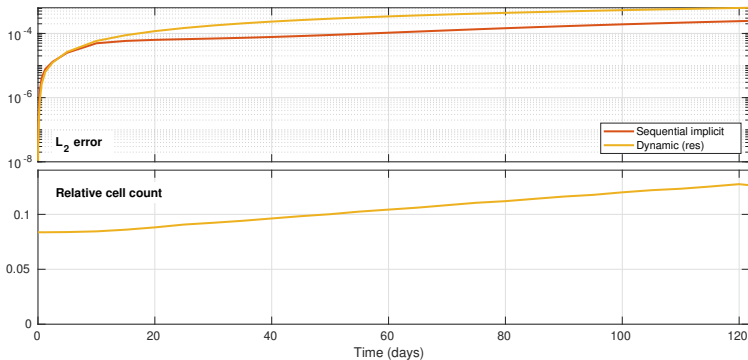
## Example: Real(istic) Model

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*



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# Technology for a better society

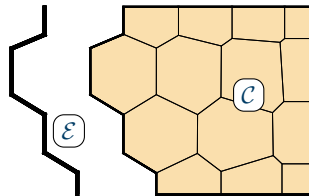


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# Extra slides



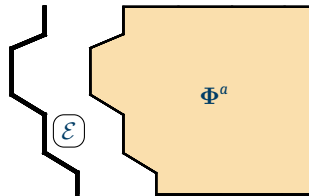
## Dynamic coarsening – Mapping parameters





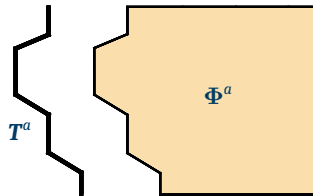
## Dynamic coarsening – Mapping parameters

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



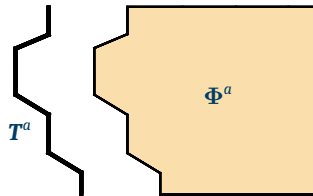
## Dynamic coarsening – Mapping parameters

- 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',
```

```
% Constant oil compressibility
```

```
fluid.b0 = @(p, varargin) exp((p/barsa - 10)
```

```
% Construct reservoir model
```

```
gravity reset on
```

```
model = TwoPhaseOilWaterModel(G,
```

```
%% Define initial state
```

```
region = getInitializationKey
```

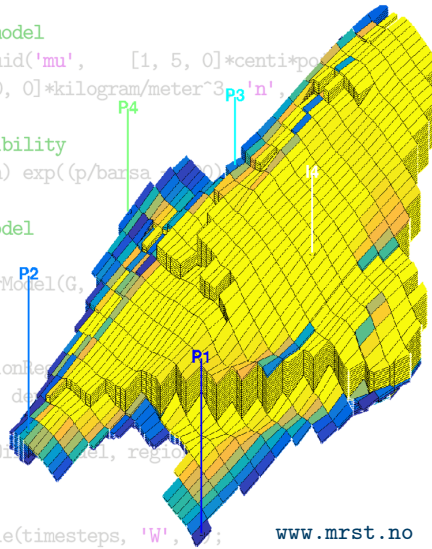
```
'datum_depth', de
```

```
state0 = initStateBlackOil(model, region
```

```
% Define schedule
```

```
schedule = simpleSchedule(timesteps, 'W',
```

[www.mrst.no](http://www.mrst.no)



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### Differentiation operators

*Writing discrete equations on form very close to continuous equations*

$$\nabla \cdot \vec{H}$$

$$\text{div}(\mathbf{H})$$

$$\vec{H} = -(\lambda_f + \lambda_r) \nabla T$$

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

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$$\vec{H} = -(\lambda_f + \lambda_r) \nabla T$$

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

### 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]}

$$\frac{\partial x}{\partial x}$$

$$\frac{\partial x}{\partial y}$$

y = ADI Properties:  
val: 2  
jac: {[0] [1]}

$$\frac{\partial y}{\partial x}$$

$$\frac{\partial y}{\partial y}$$

z = ADI Properties:  
val: 0.4060  
jac: {[ -0.8120] [-0.4060]}

$$\frac{\partial z}{\partial x} \Big|_{x=1,y=2}$$

$$\frac{\partial z}{\partial y} \Big|_{x=1,y=2}$$

## MATLAB Reservoir Simulation Toolbox (MRST)

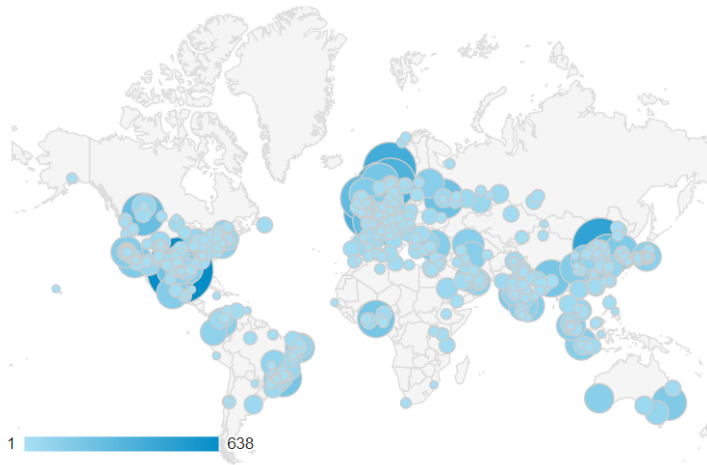
Transforming research on  
reservoir modelling

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