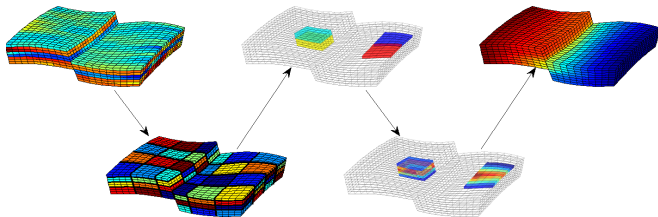


# Flow-Based Coarsening for Multiscale Simulation of Transport in Porous Media

Knut-Andreas Lie, SINTEF, Norway



Workshop on Averaging, Upscaling, and New Theories  
in Porous Media Flow and Transport  
Bergen, October 14-15, 2010

# What is multiscale simulation?

## Generally:

Methods that incorporate fine scale information into a set of coarse scale equations in a way which is consistent with the local property of the differential operator

## Herein:

Multiscale pressure solver (upscaling + downscaling in one step)

$$\nabla \cdot \vec{v} = q, \quad \vec{v} = -\lambda(S)\mathbf{K}\nabla p$$

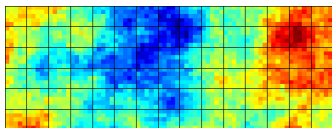
+ Transport solver (on fine, intermediate, or coarse grid)

$$\phi \frac{\partial S}{\partial t} + \nabla \cdot (\vec{v}f(S)) = q$$

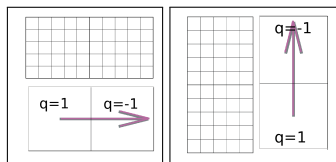
= Multiscale simulation of models with higher detail

# What is multiscale simulation?

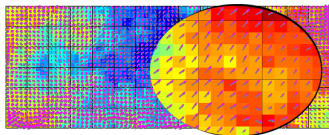
Coarse partitioning:



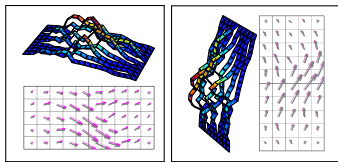
Local flow problems:



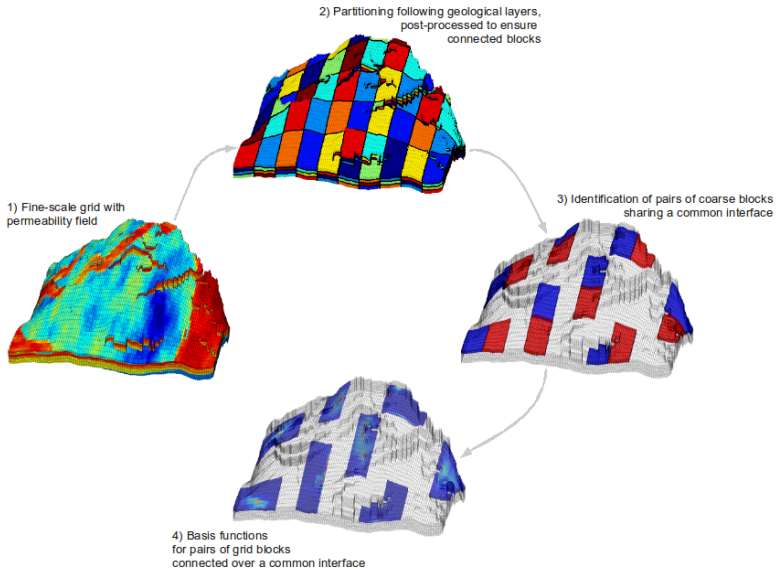
Flow field with subresolution:



Flow solutions  $\rightarrow$  basis functions:

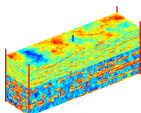


# What is multiscale simulation?



# What can you do with it?

## Example 1: Model 2 of SPE 10



60 × 220 × 85 cells

Inhouse code from 2005 (TPFA):

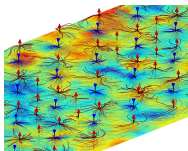
multiscale: 2 min and 20 sec

multigrid: 8 min and 36 sec

Matlab/C solver (2010):

ms-mimetic: 5–6 min

## Example 2: History matching



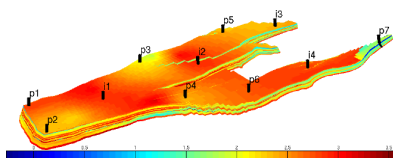
7 years: 32 injectors, 69 producers, 1 mill cells

Generalized travel-time inversion + multiscale:

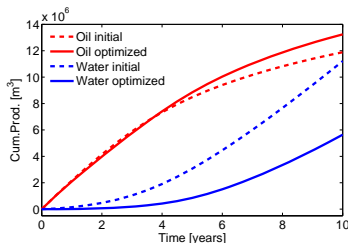
7 forward simulations, 6 inversions

Solver	CPU-time (wall clock)		
	Total	Pres.	Transp.
Multigrid	39 min	30 min	5 min
Multiscale	<b>17 min</b>	7 min	6 min

## Example 3: Rate optimization



Reservoir geometry from a Norwegian Sea field



Forward simulations:

44 927 cells, 20 time steps, < 5 sec in Matlab

~ 100 times speedup

## Capabilities:

- ✓ Incompressible (two-phase) flow
- ✓ Cartesian / unstructured grids
- ✓ Realistic flow physics  $\Rightarrow$  iterations
  - ▶ Correction functions + smoothing
  - ▶ Residual formulation + domain decomposition
- ✓ Pointwise accuracy  $\Rightarrow$  iterations

## Capabilities:

- ✓ Incompressible (two-phase) flow
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## Not yet there:

- ▶ Compressible three-phase black-oil + non-Cartesian grids
- ▶ Fully implicit formulation
- ▶ Parallelization
- ▶ Compositional, thermal, . . .
- ▶ **Efficient and robust transport solvers**

# Transport solvers on coarse grids

## Goal:

Given the ability to model velocity on geomodels and transport on coarse grids: Find a suitable coarse grid that best resolves fluid transport and minimizes loss of accuracy.

Formulated as the minimization of two measures:

- 1 the *projection error* between fine and coarse grid
- 2 the *evolution error* on the coarse grid

Difficult to formulate a practical and well-posed minimization problem for optimal coarsening → ad hoc algorithms

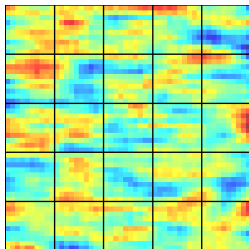


# Coarsening by amalgamation

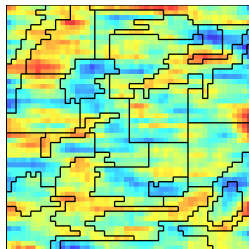
## Amalgamation of cells:

- ▶ coarse grid represented as partition vector: cell  $c_i$  in the fine grid is in coarse block  $B_j$  if  $p_i = j$
- ▶ coarsening process steered by a set of admissible and feasible amalgamation directions

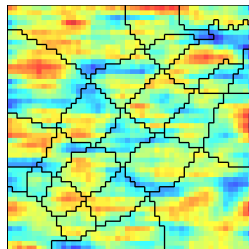
## 50 × 50 lognormal permeability:



regular: 25 blocks



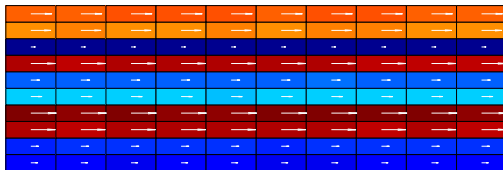
nonuniform: 26 blocks



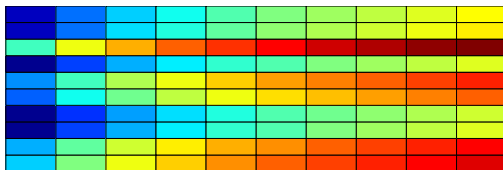
isocontours [p]: 26 blocks

# Motivation: layered reservoir

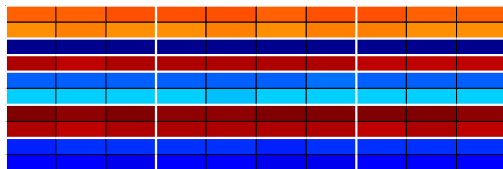
Permeability and velocity



Time-of-flight



Partition



# Heuristic minimization algorithms

Formulated using a set of:

*sources*

create a partition vector based on grid topology, geometry, flow-based indicator functions, error estimates, or expert knowledge supplied by the user, thereby introducing the feasible amalgamation directions

*filters*

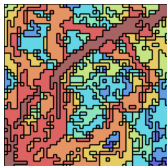
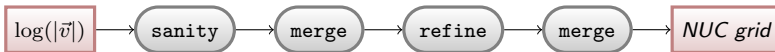
take a set of partition vectors as input and create a new partition as output, by

- ▶ combining/intersecting different partitions
- ▶ performing sanity checks, ensuring connected partitions etc
- ▶ modifying partition by merging small blocks or splitting large blocks

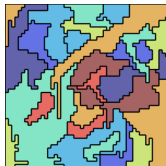
# Non-uniform coarsening

## Aarnes, Efendiev & Hauge:

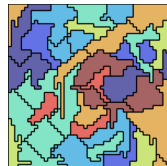
Use flow velocities to make a nonuniform grid in which each coarse block admits approximately the same total flow.



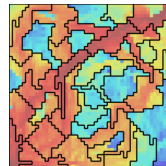
Partition: 304 blocks



Merging: 29 blocks

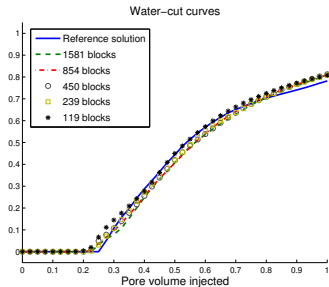
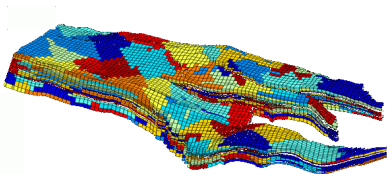
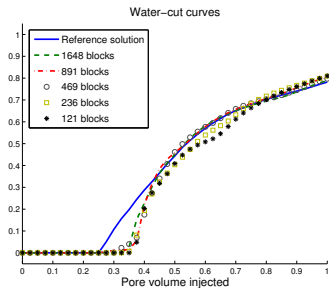
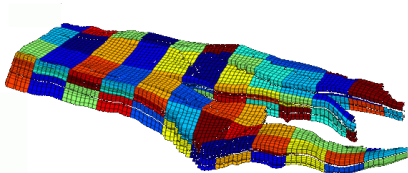


Refinement: 47 blocks



Merging: 39 blocks

# Example: reservoir model from Norwegian Sea



## Underlying principles:

- ▶ Minimize heterogeneity of flow field inside each block

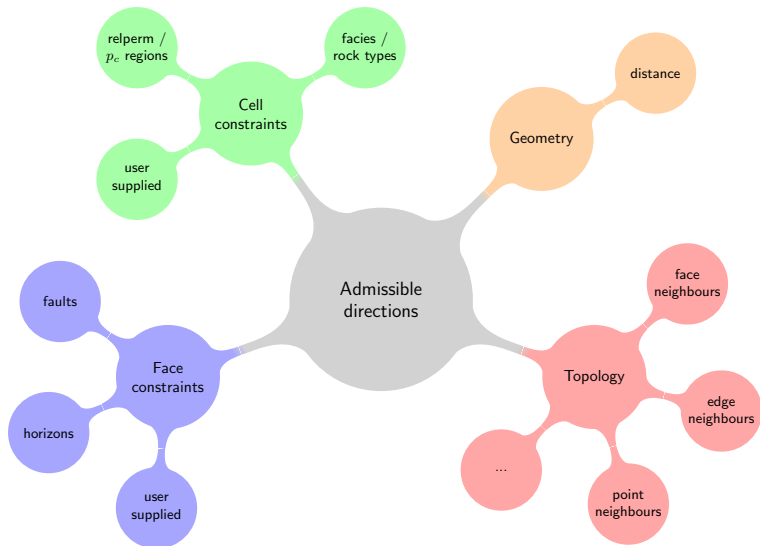
$$\min_{B_j} \left( \sum_{p_i=j} |I_1(c_i) - I_1(B_j)|^p |c_i| \right)^{\frac{1}{p}}, \quad 1 \leq p \leq \infty,$$

- ▶ Equilibrate indicator values over grid blocks

$$\min \left( \sum_{j=1}^N |I_2(B_j) - \bar{I}_2(\Omega)|^p |B_j| \right)^{\frac{1}{p}}, \quad 1 \leq p \leq \infty,$$

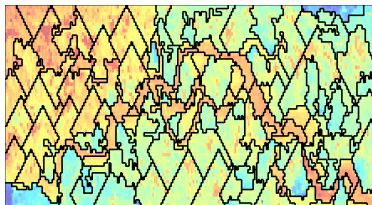
- ▶ Block size within prescribed lower and upper bounds

# Amalgamation: admissible directions (neighbourship)

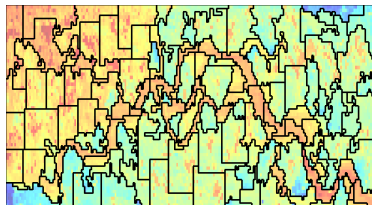


# Example: extended neighbourhood

## Structured grid:



5-neighborhood



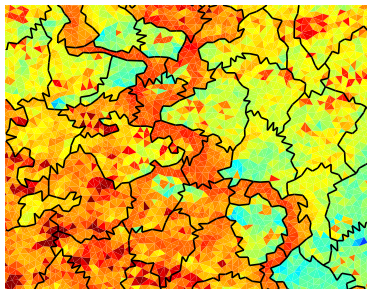
9-neighborhood



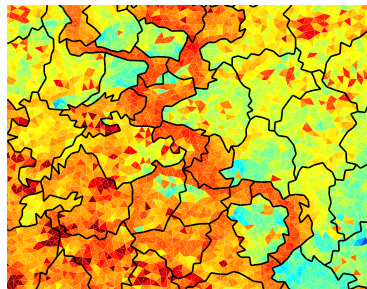


# Example: extended neighbourhood

## Triangular grid:



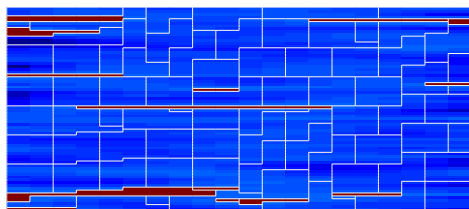
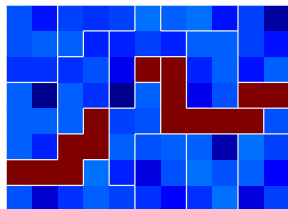
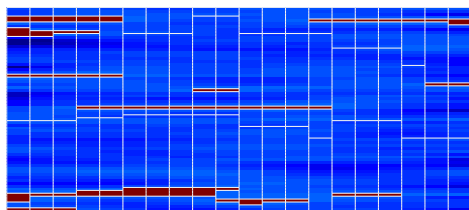
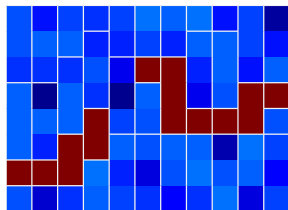
face neighbors



extended



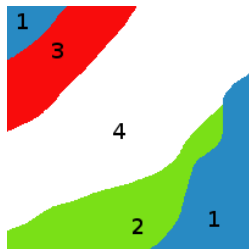
# Example: restricted neighbourhood (topology)



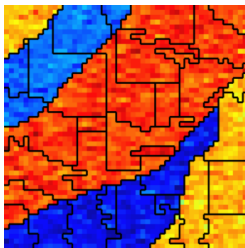
Upper row:  $\mathcal{N}(c_{ij}) = \{c_{i,j-1}, c_{i,j+1}\}$

Lower row:  $\mathcal{N}(c_{ij}) = \{c_{i,j\pm 1}, c_{i\pm 1,j}, c_{i\pm 1,j\pm 1}\}$

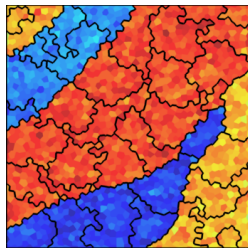
# Example: restricted neighbourhood (facies)



Facies distribution



Cartesian

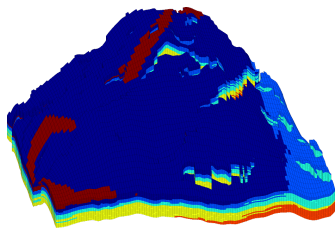


PEBI

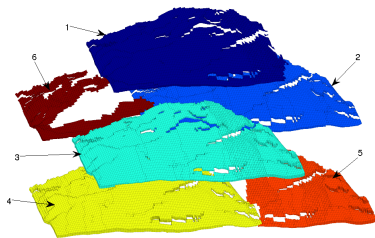
Constraining to facies / saturation regions:

- ▶ useful to preserve heterogeneity
- ▶ useful to avoid upscaling  $k_r$  and  $p_c$  curves

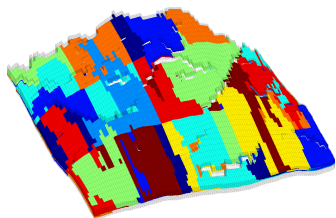
# Example: restricted neighbourhood (saturation regions)



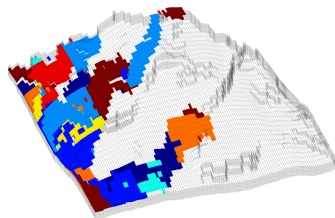
facies



facies separated



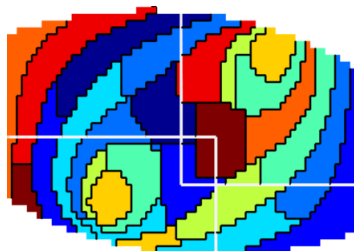
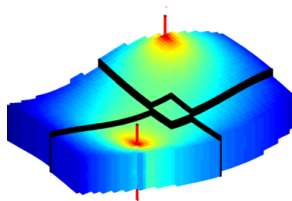
facies # 3



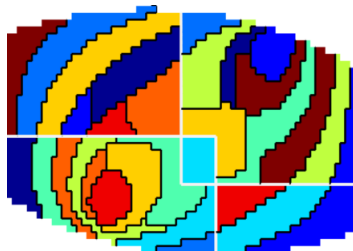
facies #6

Realization from SAIGUP study, coarsening within six different saturation regions

# Example: restricted neighbourhood (faults)

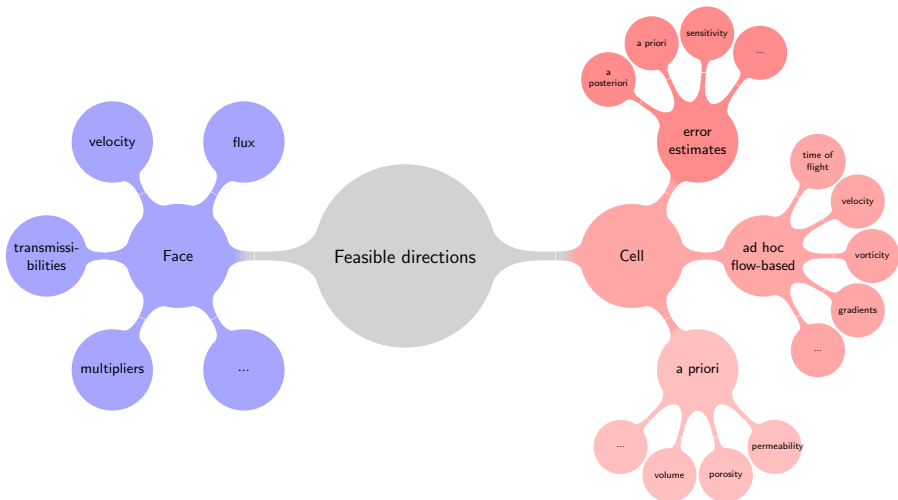


unconstrained: 26 blocks

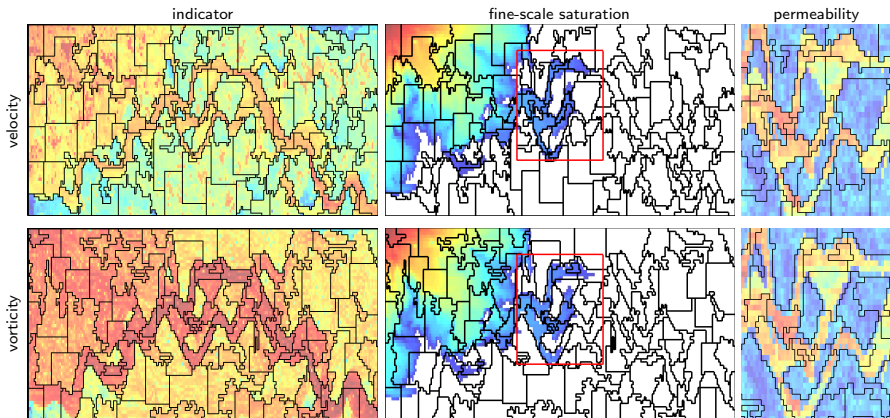


constrained: 31 blocks

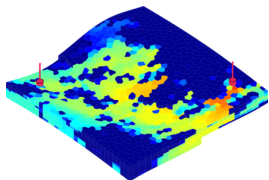
# Amalgamation: feasible directions (indicators)



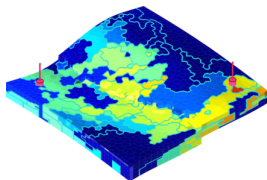
# Example: flow-based indicators



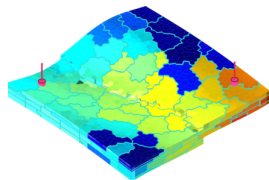
# Example: flow-based indicators



reference solution  
11 864 cells



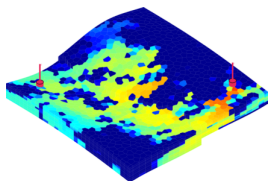
time-of-flight grid  
127 blocks



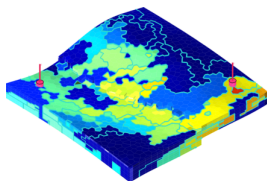
METIS grid  
175 blocks



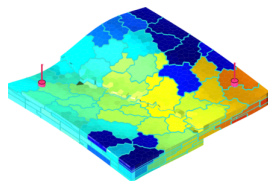
# Example: flow-based indicators



reference solution  
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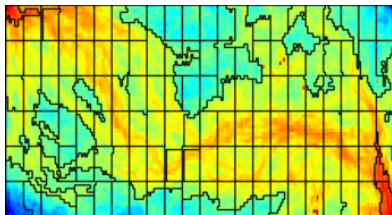
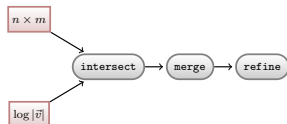
## General observations:

- ▶ Time-of-flight is a better indicator than velocity
- ▶ Velocity is a better indicator than vorticity
- ▶ Vorticity is a better indicator than permeability
- ▶ ...

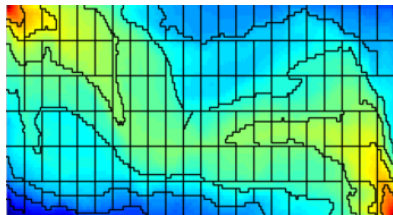
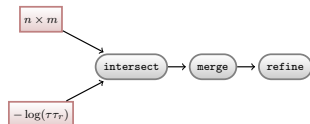
However, for smooth heterogeneities, the indicators tend to overestimate the importance of flow.

# Example: hybrid methods

## Velocity + Cartesian partition:

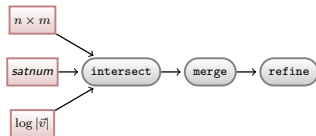
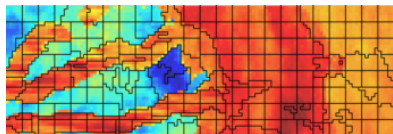
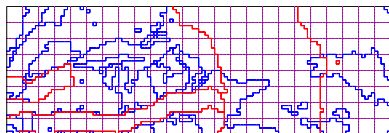


## Time-of-flight + Cartesian partition:



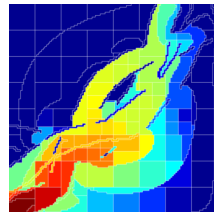
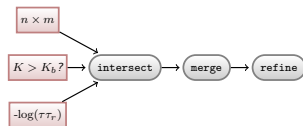
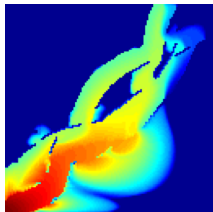
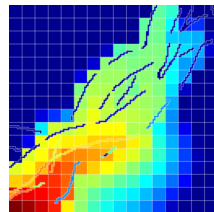
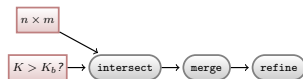
# Example: hybrid methods

Satnum + velocity + Cartesian:



# Example: hybrid methods

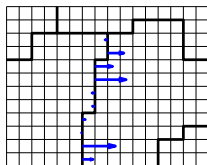
Adapting to barriers:



# Coarse-grid discretisation

**Bi-directional fluxes (upwind on fine scale):**

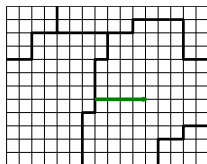
$$S_\ell^{n+1} = S_\ell^n - \frac{\Delta t}{\phi_\ell |B_\ell|} \left[ f(S_\ell^{n+1}) \sum_{\partial B_\ell} \max(v_{ij}, 0) - \sum_{k \neq \ell} \left( f(S_k^{n+1}) \sum_{\Gamma_{k\ell}} \min(v_{ij}, 0) \right) \right].$$



This gives a centred scheme on the coarse scale

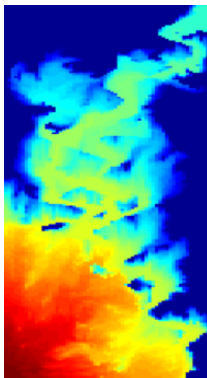
**Net fluxes:**

$$S_\ell^{n+1} = S_\ell^n - \frac{\Delta t}{\phi_\ell |B_\ell|} \sum_{k \neq \ell} \max \left( f(S_\ell^{n+1}) \sum_{\Gamma_{k\ell}} v_{ij}, -f(S_k^{n+1}) \sum_{\Gamma_{k\ell}} v_{ij} \right).$$

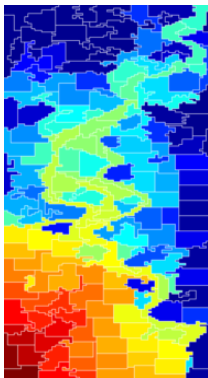


This gives an upwind scheme on the coarse scale

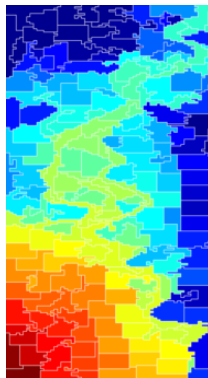
# Coarse-grid discretisation: numerical diffusion



reference



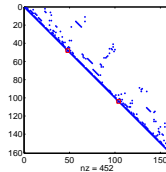
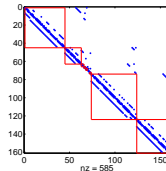
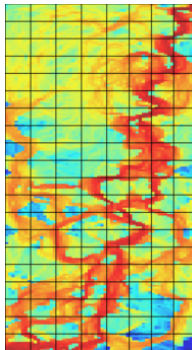
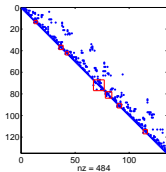
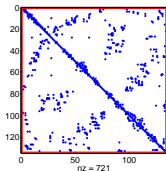
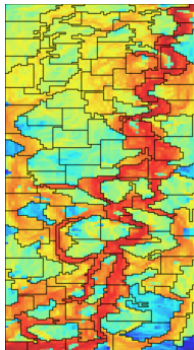
net fluxes



bi-directional fluxes

Layer 37 from SPE10

# Coarse-grid discretisation: matrix structure



Layer 68 from SPE10

# Coarse-grid discretisation: numerical errors

	Tarbert formation			Upper Ness formation		
	NUC/Cart	NUC	Cartesian	NUC/Cart	NUC	Cartesian
$E_s(\mathcal{P}\mathcal{R}S_f, S_f)$	0.0941	0.1042	0.0911	0.1371	0.1355	0.1772
$E_s(\mathcal{P}S_c, S_f)$	0.1910	0.2426	0.1687	0.2124	0.2243	0.2305
$E_s(S_c, \mathcal{R}S_f)$	0.1599	0.2100	0.1381	0.1522	0.1683	0.1604
$E_w(w_c, w_f)$	0.0695	0.0773	0.0701	0.0609	0.0668	0.0982
$E_s(\mathcal{P}S_c, S_f)$	0.1607	0.1875	0.1619	0.1795	0.1862	0.2191
$E_s(S_c, \mathcal{R}S_f)$	0.1237	0.1459	0.1302	0.1135	0.1225	0.1486
$E_w(w_c, w_f)$	0.0473	0.0444	0.0647	0.0237	0.0325	0.0844
# blocks	217–261	233–312	264	205–241	220–303	264
mean	236	275	264	222	264	264
# faces: mean	1069	1363	1090	1070	1309	1090

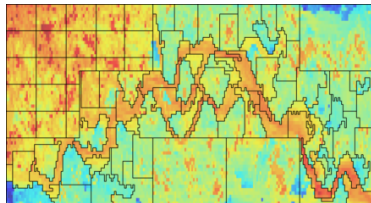
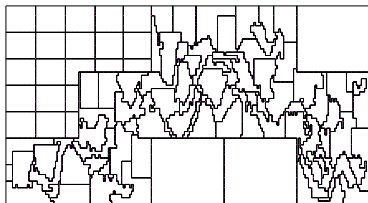
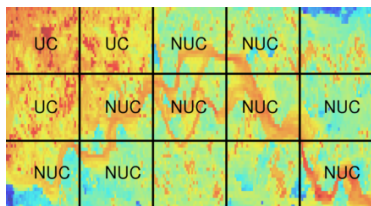
bi-directional fluxes

net fluxes

Average errors over all layers of the two formations in SPE10



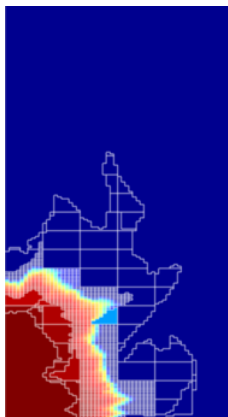
# Supervised coarsening



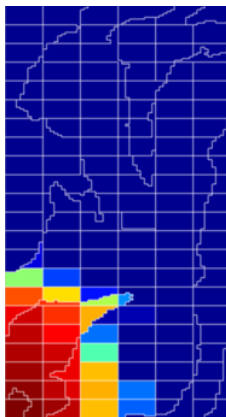
Layer 37 from SPE10

# Dynamical adaption

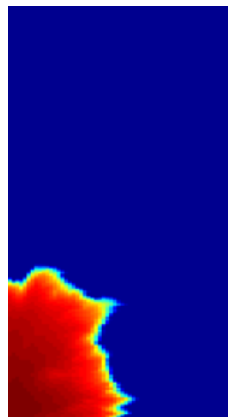
After injection of 0.1 PVI



adaptive



coarse

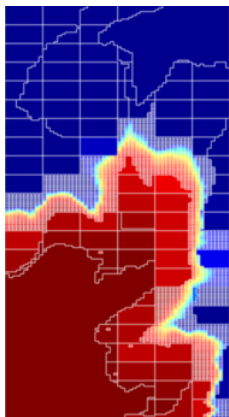


fine grid

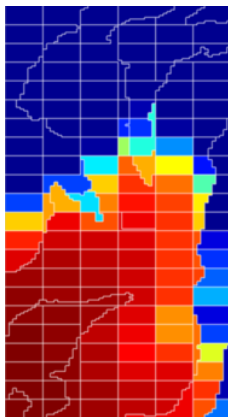
Layer 22 from SPE10

# Dynamical adaption

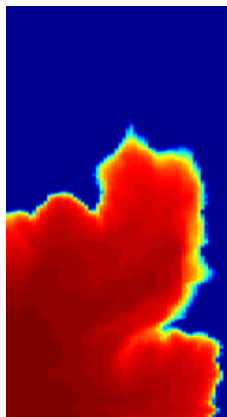
After injection of 0.5 PVI



adaptive



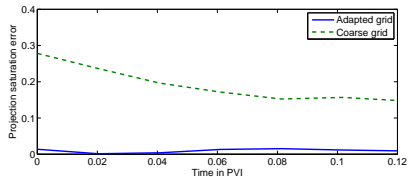
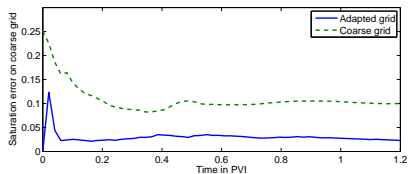
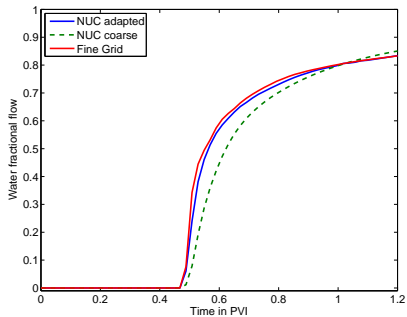
coarse



fine grid

Layer 22 from SPE10

# Dynamical adaption



Layer 37 from SPE10

How can these methods be useful? For what purpose would you apply them?

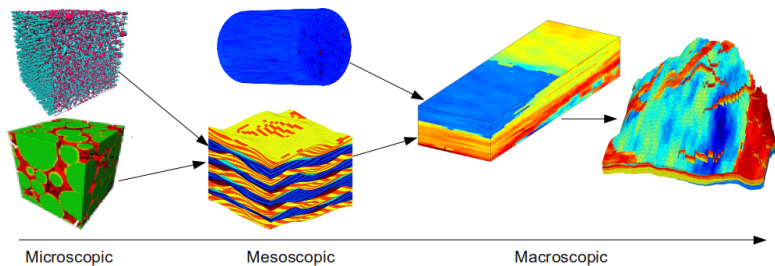
- ▶ As robust upscaling methods?
- ▶ As alternative to upscaling and fine-scale solution?
- ▶ To provide flow simulation earlier in the modelling loop?
- ▶ To get 90% of the answer in 10% of the time?
- ▶ Fit-for-purpose solvers in workflows for ranking, history matching, planning, optimization, ...

Which capabilities should we try to develop?

- ▶ More complex flow physics?
- ▶ Modelling of fault/fractures?
- ▶ Multiphysics formulations?
- ▶ Automated methods with goal-oriented error control?
- ▶ ...

What capabilities are sufficient for the methods to be more generally adopted?

# Questions



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Ultimate vision: 'truly' multiscale methods, bridging 'all' scales?

Is incorporation of pore/core/facies models realistic?