Multiscale Methods for Capturing Geological Heterogeneity

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Physical Scales in Subsurface Modelling

The scales that impact fluid flow in oil reservoirs range from

- the micrometer scale of pores and pore channels
- via dm-m scale of well bores and laminae sediments
- to sedimentary structures that stretch across entire reservoirs.



Geological Models

Expressing the geologists' preception of the reservoir:

- here: geo-cellular models
- describe the reservoir geometry (horizons, faults, etc)
- typically generated using geostatistics (or process simulation)
- give rock parameters (permeability and porosity)





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Rock parameters:

- have a multiscale structure
- details on all scales impact flow
- permeability spans many orders of magnitude





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Gap in resolution:

- Geomodels: $10^7 10^9$ cells
- Simulators: $10^5 10^6$ cells





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 \longrightarrow sector models and/or upscaling of parameters



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Many alternatives:

- Harmonic, arithmetic, geometric,
 ...
- Local methods (K or T)
- Global methods
- Local-global methods
- Pseudo methods
- Ensemble methods
- Steady-state methods



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Simulation on Seismic/Geologic Grid

Why do we want/need it?

- Upscaling is a bottleneck in workflow,
- gives loss of information/accuracy,
- is not sufficiently robust,
- extensions to multiphase flow are somewhat shaky



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Simulation on seismic/geologic grid:

- best possible resolution of the physical processes,
- faster model building and history matching,
- makes inversion a better instrument to find remaining oil,
- better estimation of uncertainty by running alternative models



Example: Gullfaks Field (North Sea)

Bypassed oil (4D inversion vs simulation):





Arnesen, WPC, Madrid, 2008



Difference in resolution (10 million vs 1 billion cells):



From Dogru et al., SPE 119272



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Simplified flow physics:

Can often tell a lot about the fluid movement. "Full physics" is typically only required towards the end of a workflow



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Operator splitting:

- fully coupled solution is slow..
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Use of sparsity / (multiscale) structure:

- effects resolved on different scales
- small changes from one step to next
- small changes from one simulation to next

From Upscaling to Multiscale Pressure Solvers



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Multiscale method:



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From Upscaling to Multiscale Pressure Solvers

Standard upscaling:



Multiscale method:





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1) Automated coarsening: uniform partition in index space for corner-point grids





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2) Detect all adjacent blocks





1) Automated coarsening: uniform partition in index space for corner-point grids



3) Compute basis functions

$$\nabla \cdot \psi_{ij} = \begin{cases} w_i(x), \\ -w_j(x), \end{cases}$$
 for all pairs of blocks

2) Detect all adjacent blocks



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1) Automated coarsening: uniform partition in index space for corner-point grids



3) Compute basis functions



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4) Block in coarse grid: component for building global solution



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- Reuse of computations, key to computational efficiency

Operations vs. upscaling factor:







Inhouse code from 2005: Multiscale: 2 min and 20 sec Multigrid: 8 min and 36 sec

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- Fine-scale velocity —> different grid for flow and transport —> dynamical adaptivity

Flow-based gridding:



with and without dynamic Cartesian refinement

Research by: Vera Louise Hauge, Shell scholarship



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- Method for model reduction:
 - adjoint simulations → approximate gradients
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Reservoir geometry from a Norwegian Sea field



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History matching 1 million cells:



7 years: 32 injectors, 69 producers

Generalized travel-time inversion + multiscale: 7 forward simulations, 6 inversions

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

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- Multiphysics applications

Stokes-Brinkmann:



How well do these methods handle complex physics?

- No fully-implicit formulation available
- Compressibility, gravity, $\ldots \longrightarrow$ correction functions
- \bullet Strong coupling \longrightarrow more iterations and updates of basis and correction functions
- To force residual to zero, multiscale methods start to look like multigrid/domain decomposition
- Not yet applied to compositional/thermal/...



Other issues:

- How to choose good coarse grids for unstructured grids?
- Need for global information or iterative procedures?
- A posteriori error analysis (resolution or fine-scale junk)?
- More than two levels in hierarchical grid?
- How to include pore-scale models?



Example: Fracture Corridors



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Example: SPE10 with Fracture Corridors



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- Streamline methods
 - intuitive visualization + new data
 - subscale resolution
 - good scaling, known to be efficient







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Time-of-flight (timelines):







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- Optimal ordering
 - same assumptions as for streamlines
 - utilize causality $\longrightarrow \mathcal{O}(n)$ algorithm, cell-by-cell solution
 - local control over (non)linear iterations

Topological sorting





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Local iterations:



Johansen formation: 27437 active cells

Global vs local Newton-Raphson solver

Δt	global		local	
days	time	iter	time (sec)	iter
125	2.26	12.69	0.044	0.93
250	2.35	12.62	0.047	1.10
500	2.38	13.25	0.042	1.41
1000	2.50	13.50	0.042	1.99

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- Flow-based coarsening
 - agglomeration of cells → simple and flexible coarsening
 - hybrid griding schemes
 - heterogeneous multiscale method?
 - efficient model reduction

Cartesian grid:



Triangular grids:



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Different partitioning:



Uniform coarsening + Cartesian/NUC refinement



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Model reduction by coarsening:



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Keys to enable fast simulation on seismic/geological grids:

- Simplified physics
- Operator splitting
- Sparsity / (multiscale) structure

In the future: fit-for-purpose rather than one-simulator-solves-all..?



Current and Future Research



Geological representation



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