



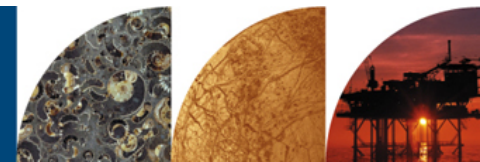
# Pathways Towards Unstructured Mesh Adaptivity for Reservoir Modelling

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**Applied Modelling and Computation Group  
Novel Reservoir Modelling and Simulation Group**

**Presentation to NORMS**

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## Outline

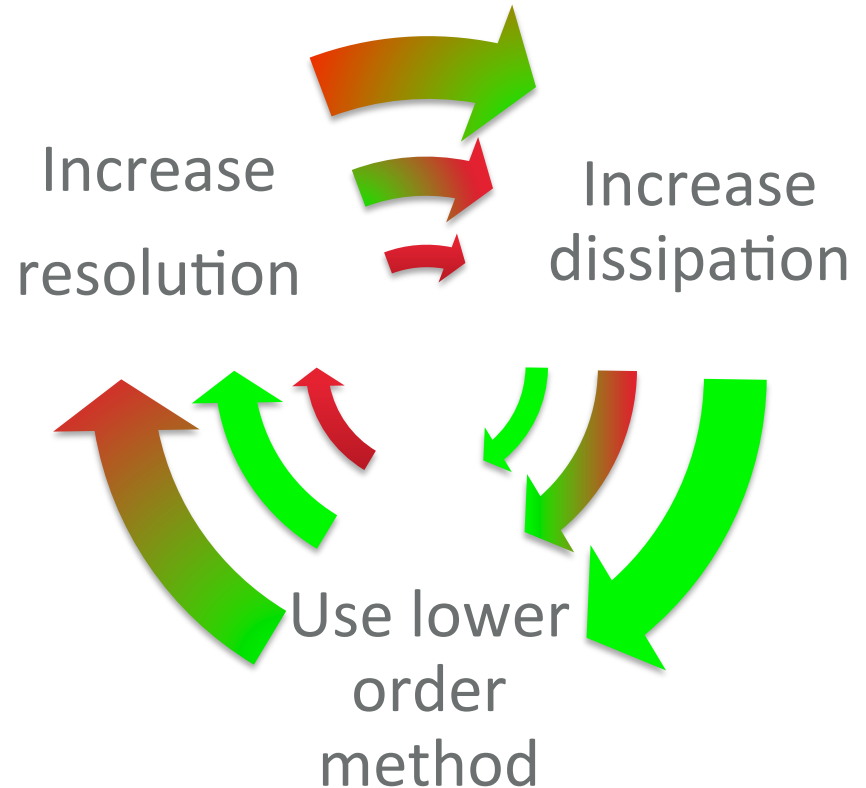
- **Background (What is Mesh Adaptivity?)**
- **Motivation (Why Adapt Meshes?)**
- **Current Methodology (How does IC-FERST adapt meshes?)**
- **Discussion (How to be better?)**

## Background – The goals of numerical simulation

In numerical simulation there are 3 key concerns:

- Accuracy – simulation is a good representation of real behaviour of the system
- Stability – answers remain physically relevant.
- Cost – Time/expense of generating solution

Often impossible to achieve all of fast, accurate & robust.



## Background – Why discretize?

Computers (& humans) have finite memories


PDEs contain information at infinite no. of scales

$$\frac{\partial \phi S_k}{\partial t} + \nabla \cdot \mathbf{u}^{\text{Darcy}} = 0$$

$$S := S(\mathbf{x}, t)$$

Physics in PDEs converted into linear algebra via a choice of discretization

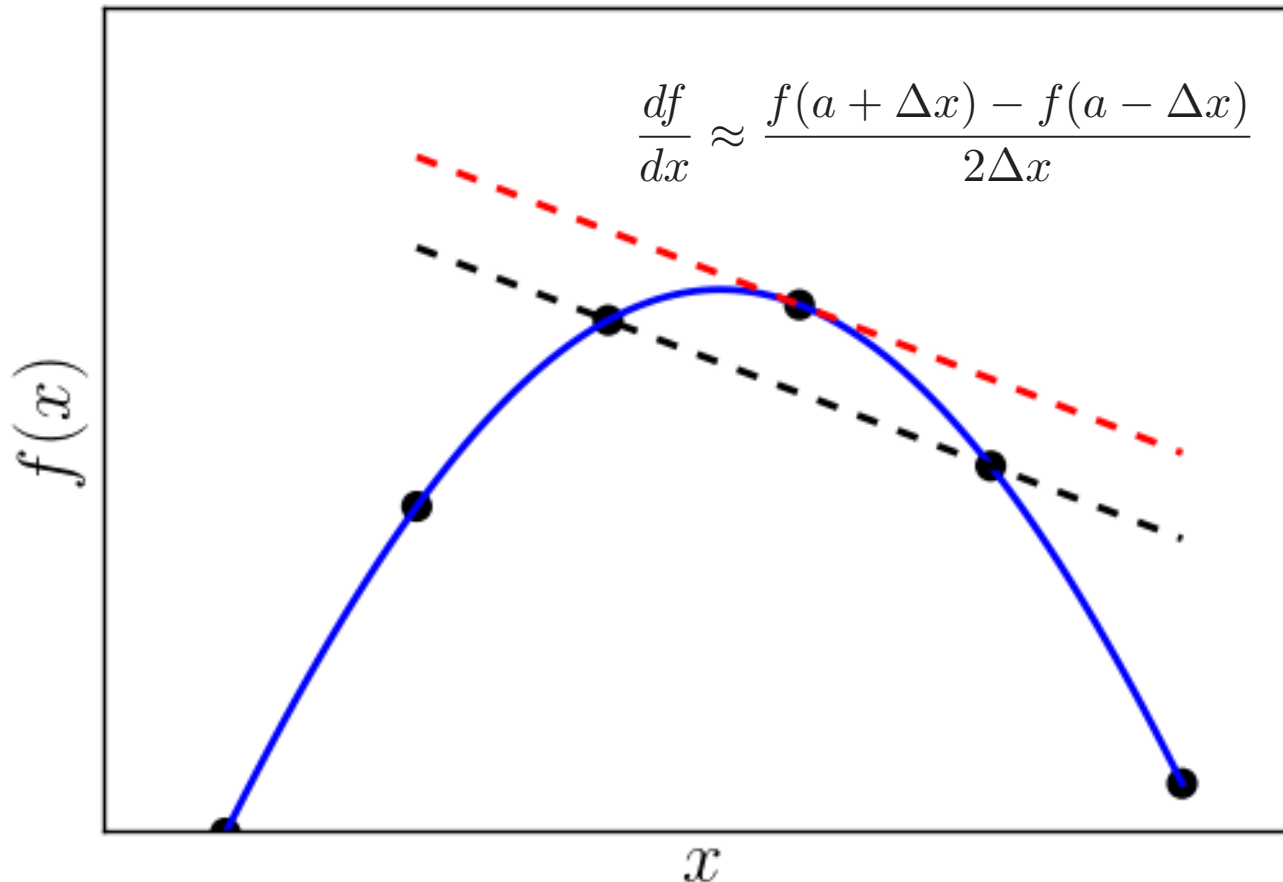
1. Finite difference
2. Finite Volume
3. Spectral Methods
4. Finite element
5. Mesh Free methods

	Integer		Floating point
16 bit	-32,768	 <p>Notice! Different discretization methods often combined.</p>	
32 bit	-2,147,483,648 to 2,147,483,647		$\pm 1.4 \times 10^{-45}$ to $3.4 \times 10^{38}$ (~7.2 sig. fig)
64 bit	-9,223,372,036,854,775,808 to +9,223,372,036,854,775,807		$10^{-308}$ to $10^{308}$ ( 16 sig. fig)

## Background – Finite Differences

Finite differences – Schoolboy calculus

$$f(a + \Delta x) = f(a) + \Delta x \left. \frac{df}{dx} \right|_{x=a} + \frac{\Delta x}{2} \left. \frac{d^2 f}{dx^2} \right|_{x=a} + \frac{\Delta x}{6} \left. \frac{d^3 f}{dx^3} \right|_{x=a} + \dots$$

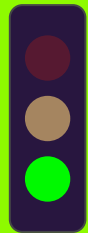
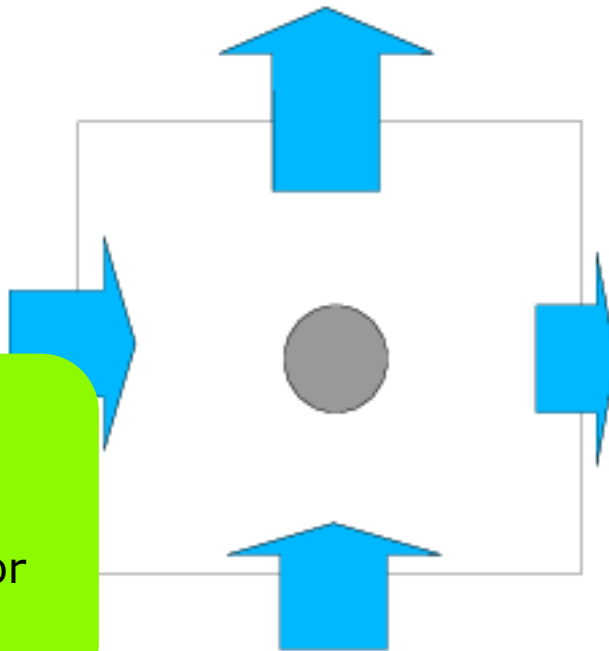
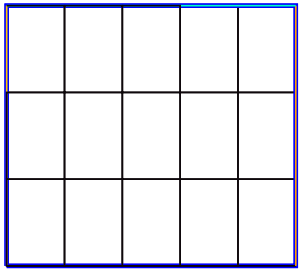


Notice!  
Implementation is usually simpler (& more accurate) on structured uniform meshes

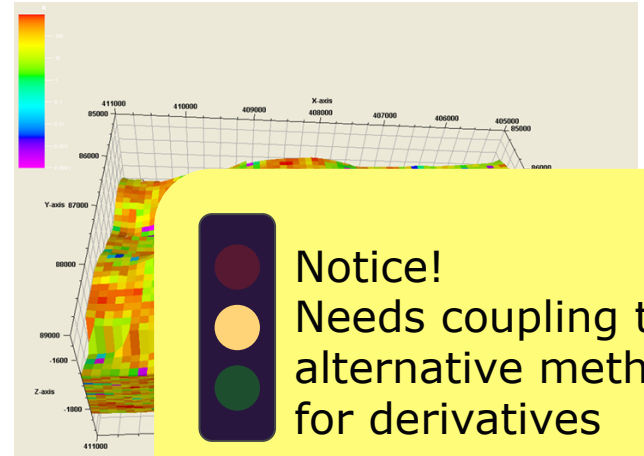
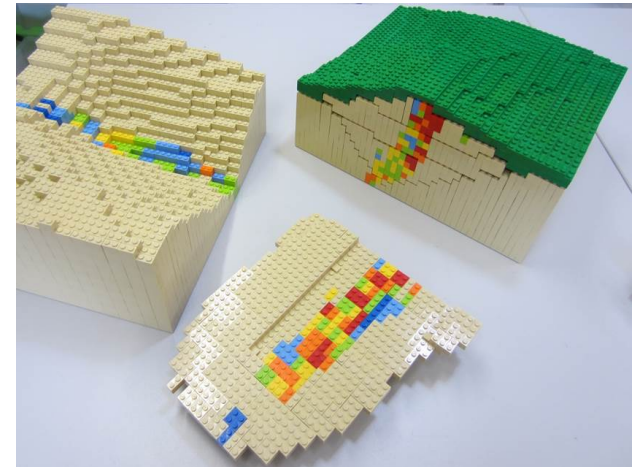
# Background – Finite Volumes

“Lego science”

$$\frac{d}{dt} \int_{\Omega_i} \rho dV = \sum_{\text{faces}} \int_{\delta\Omega_i^{(j)}} \rho \mathbf{u} \cdot \mathbf{n} dS$$



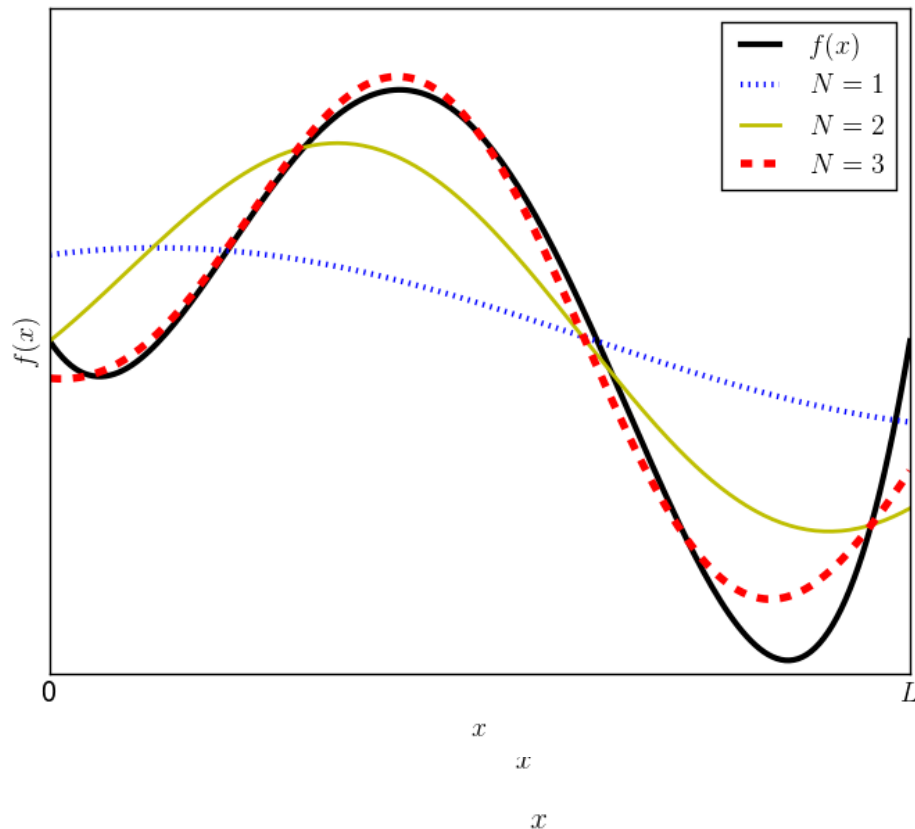
**Bonus!**  
Ensures local  
conservation for  
(almost) free



**Notice!**  
Needs coupling to  
alternative method  
for derivatives

# Background – Spectral Methods

Spectral methods – basis functions



$$f(x) = \sum_{n=1}^{\infty} a_n \cos\left(n \frac{\pi x}{L}\right) + b_n \sin\left(n \frac{\pi x}{L}\right)$$

$$\approx \sum_{n=1}^N a_n \cos\left(n \frac{\pi x}{L}\right) + b_n \sin\left(n \frac{\pi x}{L}\right)$$

$$\int_{\Omega} \cos(ax) \cos(bx) dx = C \delta_{ab}$$

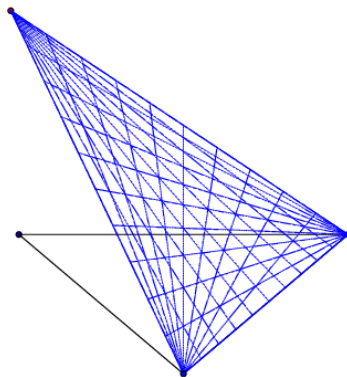



**Notice!**  
Doesn't much like  
boundary conditions  
or complex  
geometries

## Background – Finite Elements

Couple basis functions  
+ partition into simple shapes

$$\sum_j \int_{\Omega} N_i N_j \psi_j dV = \int_{\Omega} N_i f(\mathbf{x}) dV$$



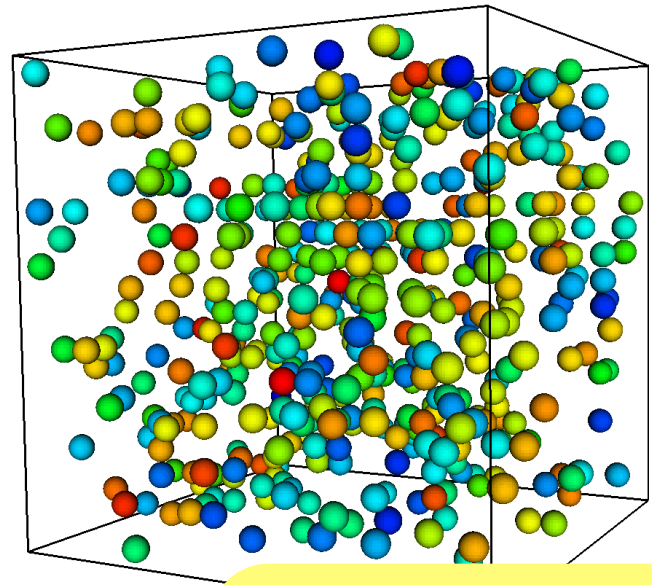
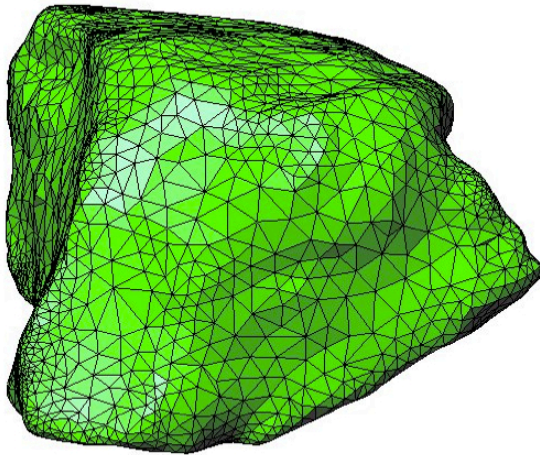
 **Warning!**  
Lots of maths  
being skipped  
(weak forms, test  
functions, etc.)

## Background - Mesh Free Methods - DEM

FEM

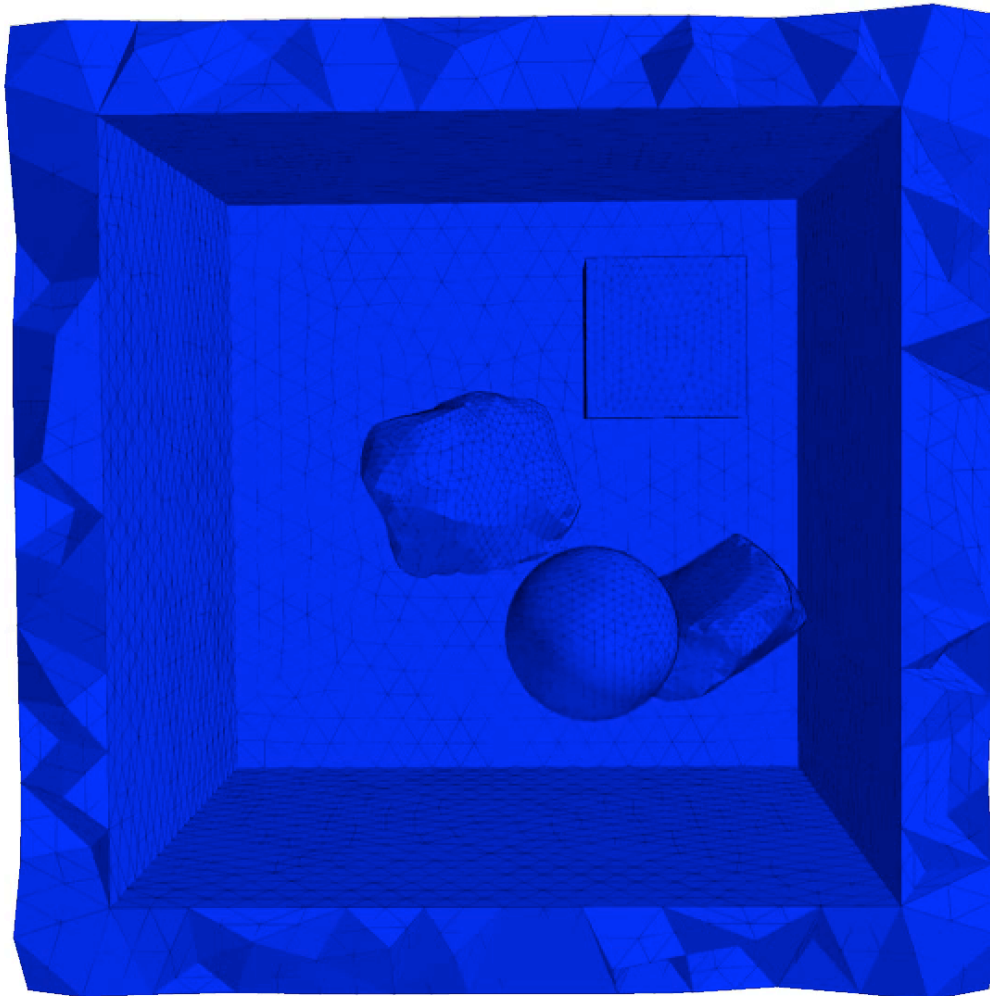
+

DEM



Notice!  
Slides from (and  
questions to)  
Jiansheng Xiang

# What is FEMDEM?

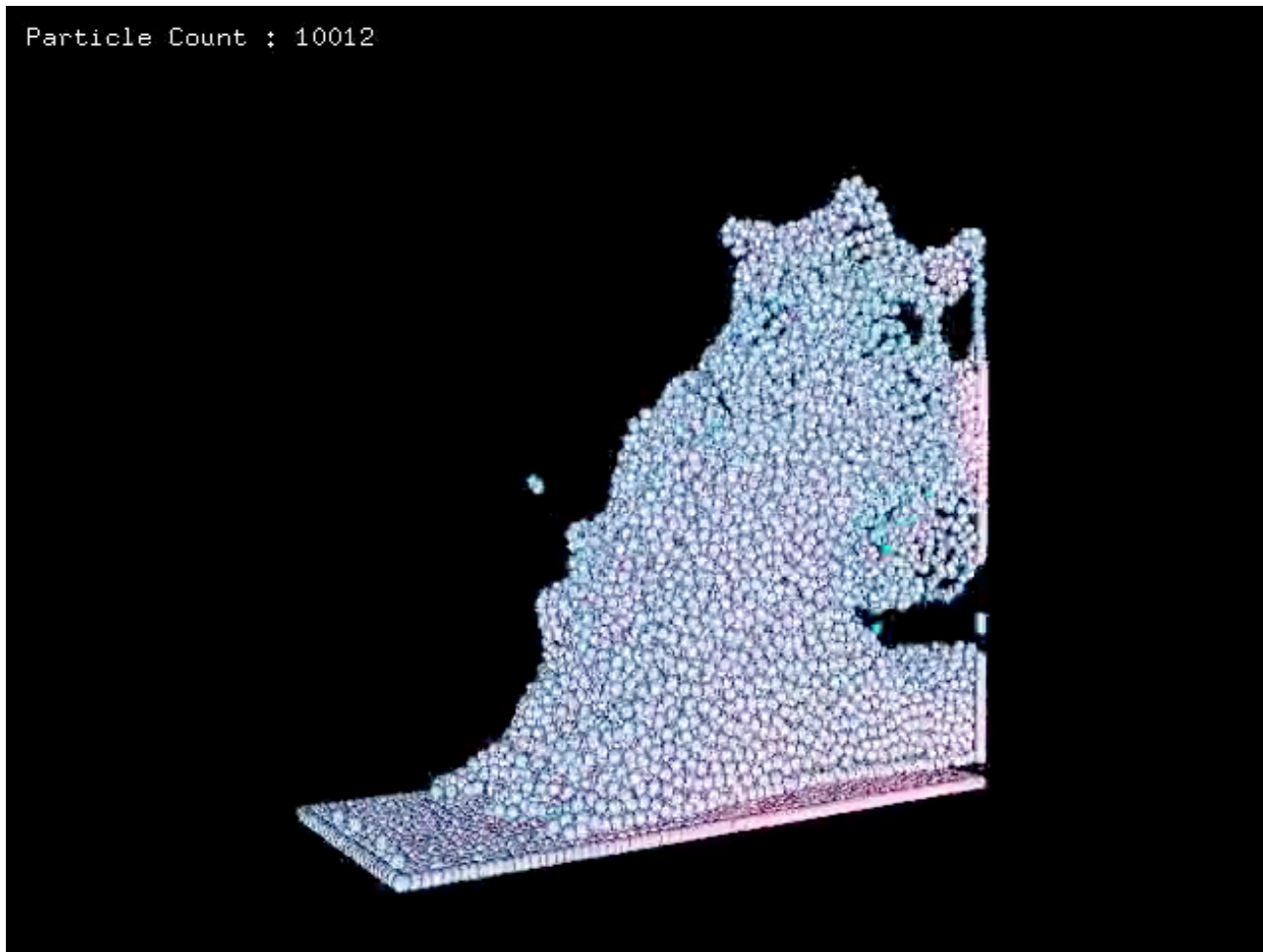


**Website:**  
<http://www.vgest.net>

Xiang, J. et al. 2009. Finite strain, finite rotation quadratic tetrahedral element for the combined finite-discrete element method. *Int Journal for Numerical Methods in Engineering* DOI: 10.1002/nme.2599

## Mesh Free Methods

### Smoothed Particle Hydrodynamics (SPH)



Slide  
from  
Youtube

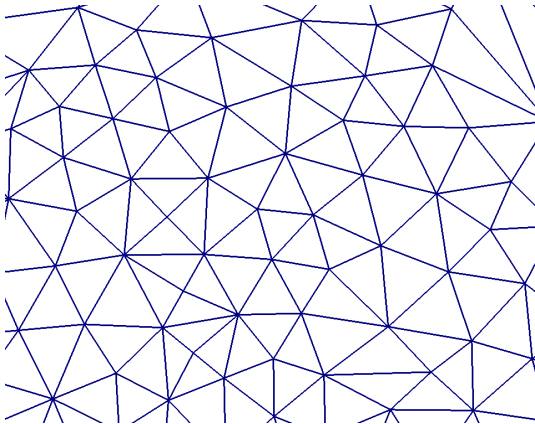
M. Müller

## Background – Hybrid Methods

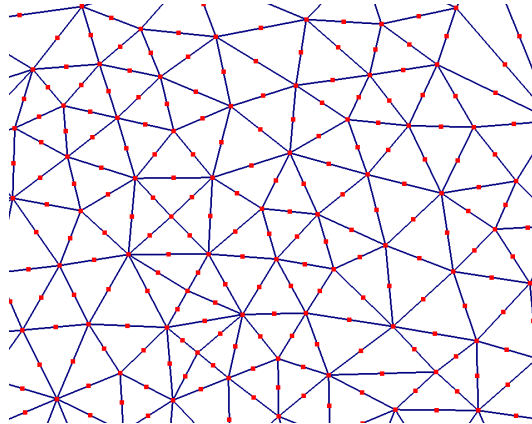
Methods combined.

Eg. Finite Element Method+ Discrete Element Method = FEMDEM

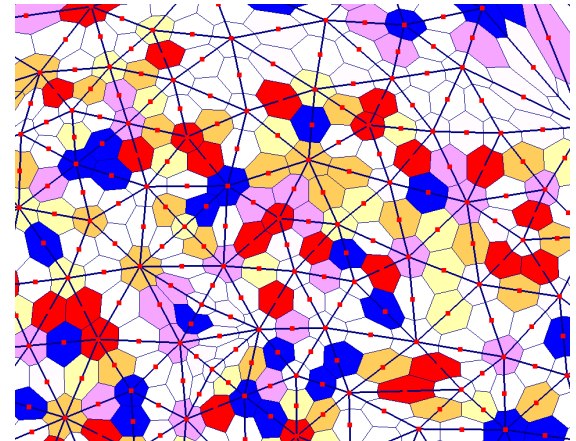
Control volumes: - Just finite volumes piggybacking on finite elements



Finite Element Mesh



Finite element "nodes"



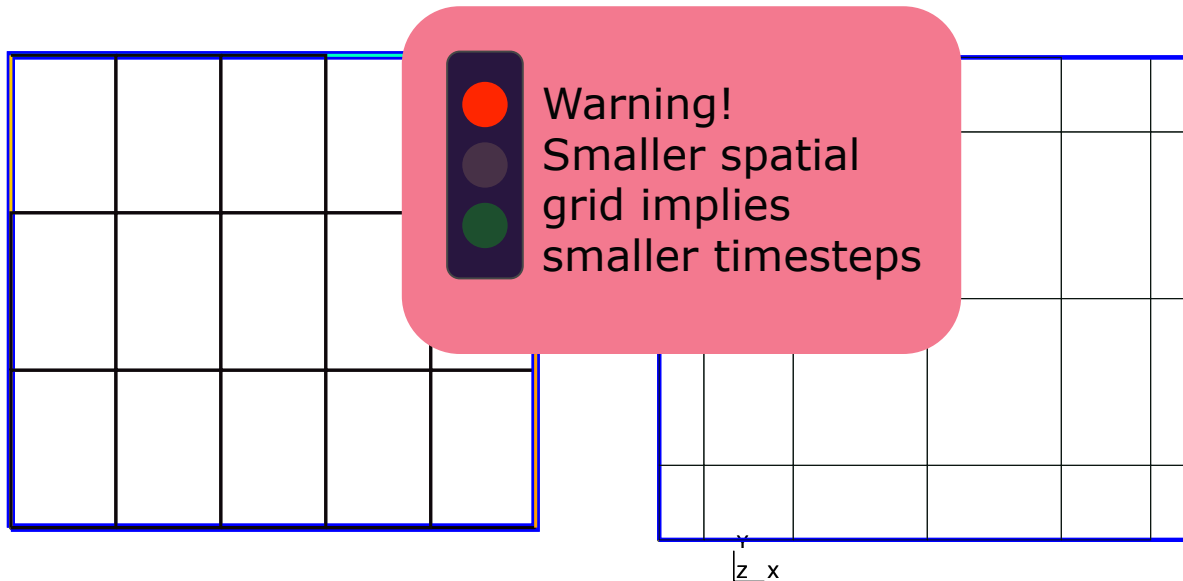
Voronoi dual mesh  
Control volumes

## Motivation – Meshes & Grids

Meshs/grids are ubiquitous  
in computational science

Discretization imposes length  
scale on the problem

Trade-off: high resolution  
discretizations often more  
accurate, but take more  
time to simulate



Idea: Only put high  
resolution “where  
it needs to be”  
dynamically.

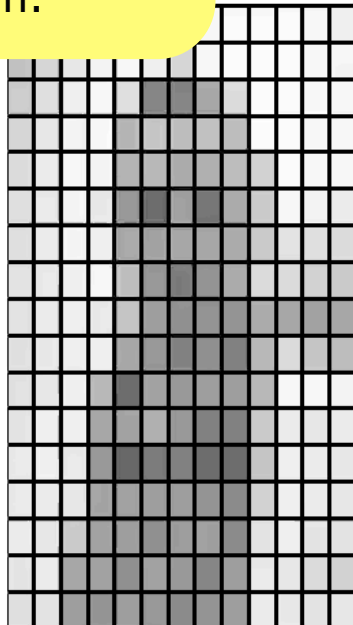
## Motivation - AMR

Idea has been applied on structured meshes:  
Adaptive Mesh Refinement (AMR)

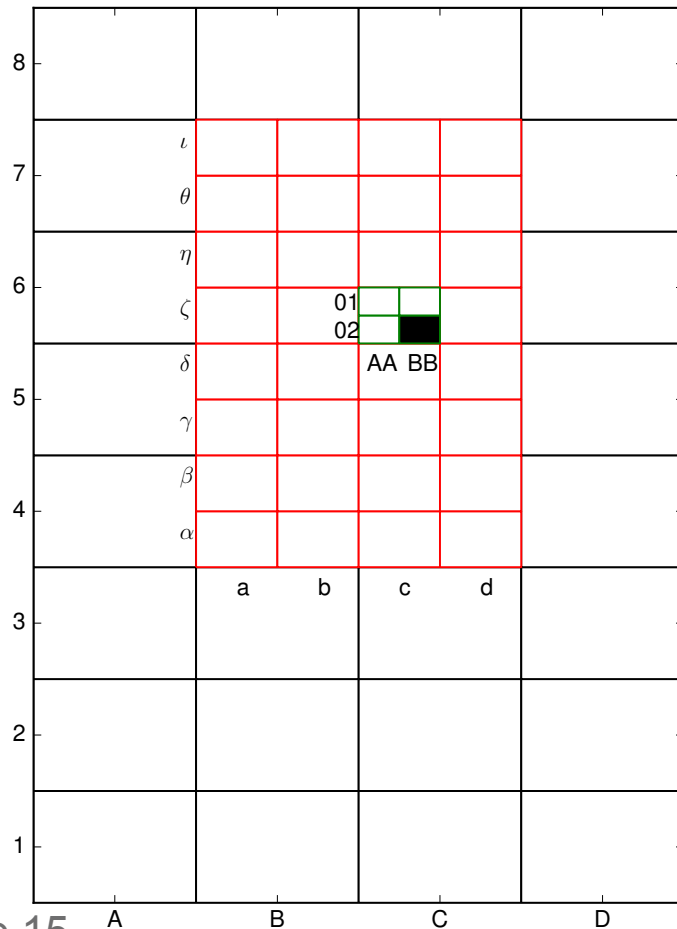


Notice!

The video is created  
by hand, coarsening  
the high res photo.  
No algorithm.



# Motivation - AMR



Adaptive Mesh Refinement:

Refinement forms a tree:  
Cell BB02 is in cell cy is in Cell C6

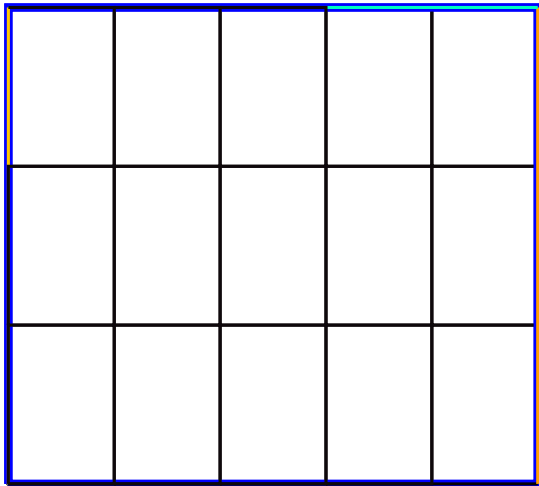
Convenient for  
implementation

Less good for physics

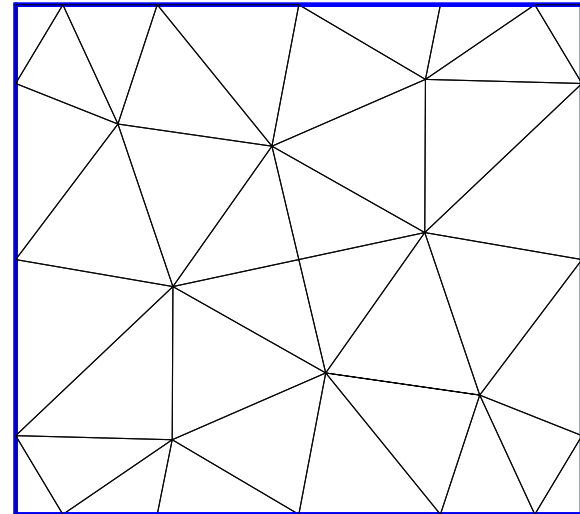
- Lots of wasted resolution (especially in 3D)
- Can only coarsen on predefined scales
- Factors of 2 everywhere.

## Motivation/Methodology

Fluidity uses unstructured meshes. Can we do better?



Y  
|  
Z x

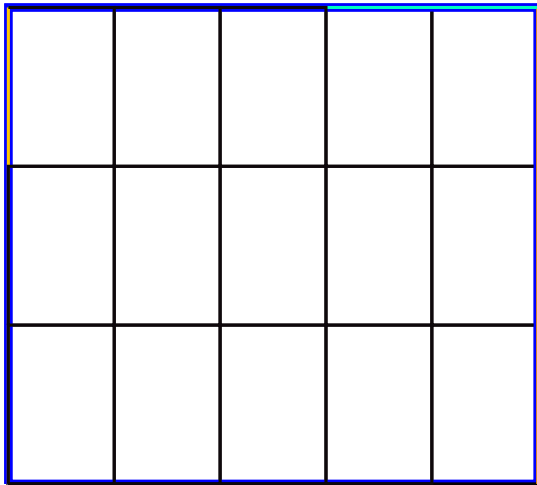


Y  
|  
Z

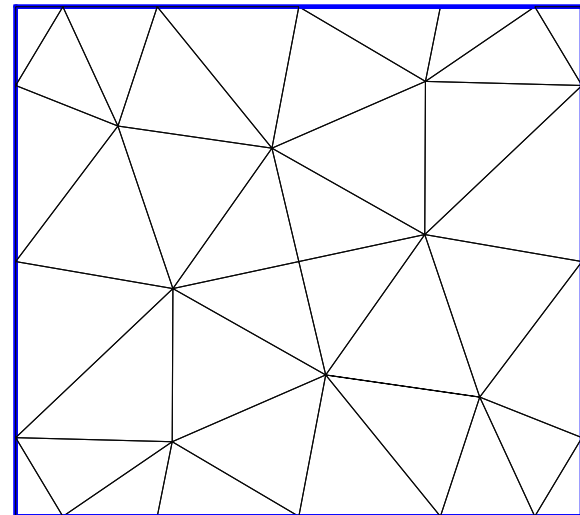
## Motivation/Methodology

Yes:

- Fewer wasted regions of high resolution
- Not forced to preserve initial orientation or grid lengths
- Length scales can differ along/across structures (anisotropy)
- Fewer factors of 2.



Y  
|  
Z x

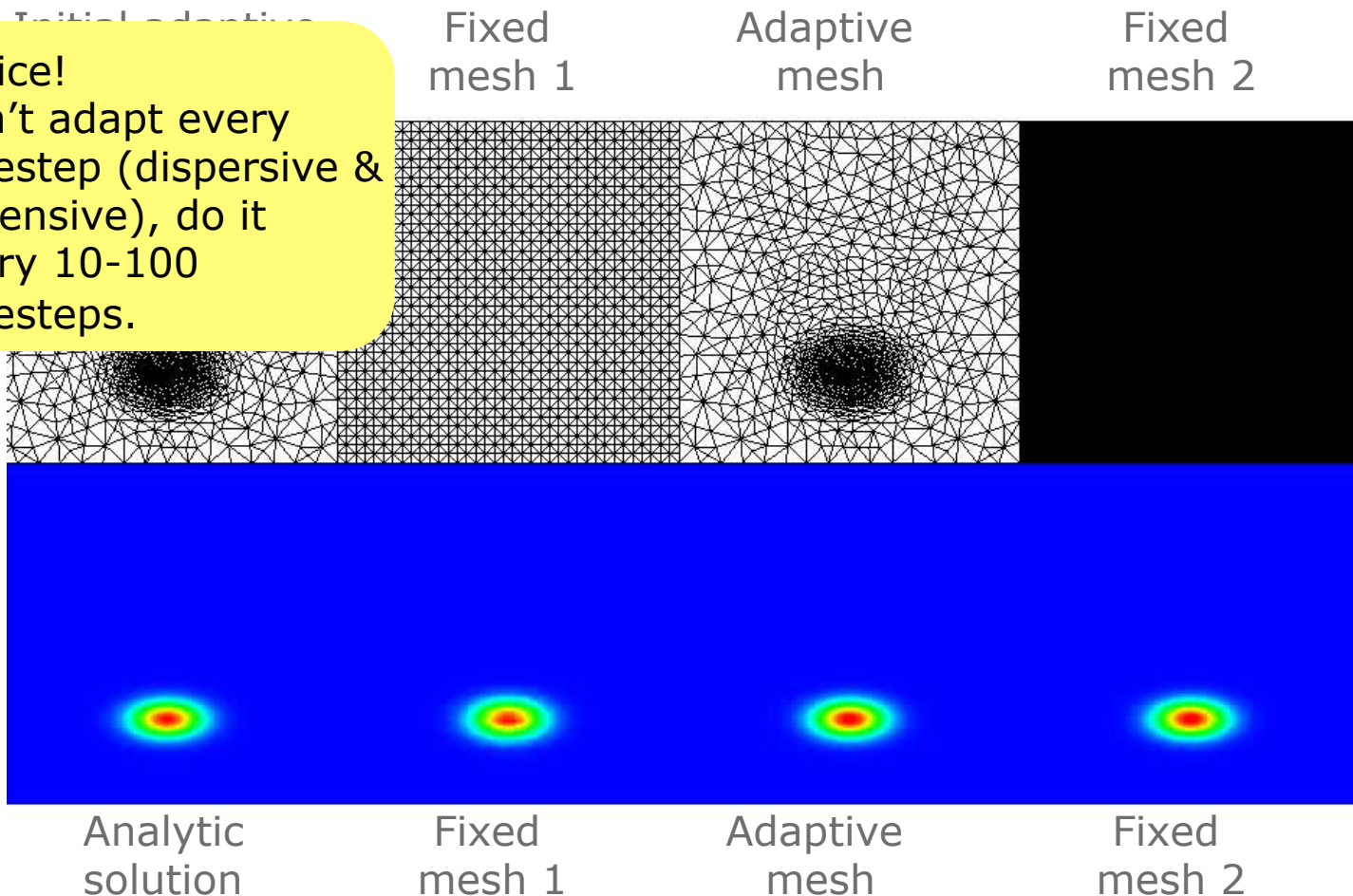


Y  
|  
Z x

## Methodology – An example of Mesh Adaptivity

### Advection of a Gaussian bubble

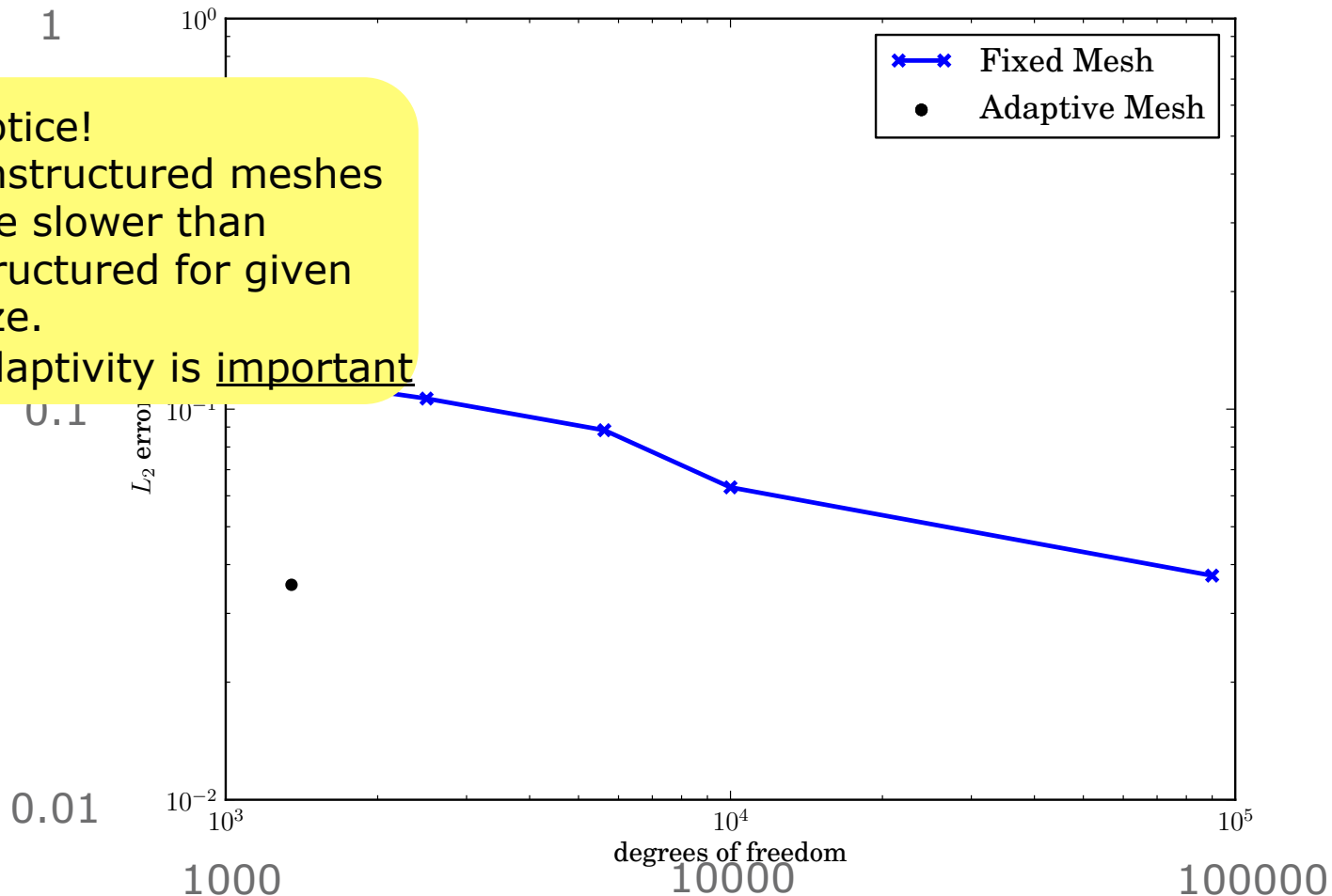
Notice!  
Don't adapt every timestep (dispersive & expensive), do it every 10-100 timesteps.



# Mesh Adaptivity: Faster & more accurate solutions

2 orders of magnitude smaller problem/half error

30x speed up

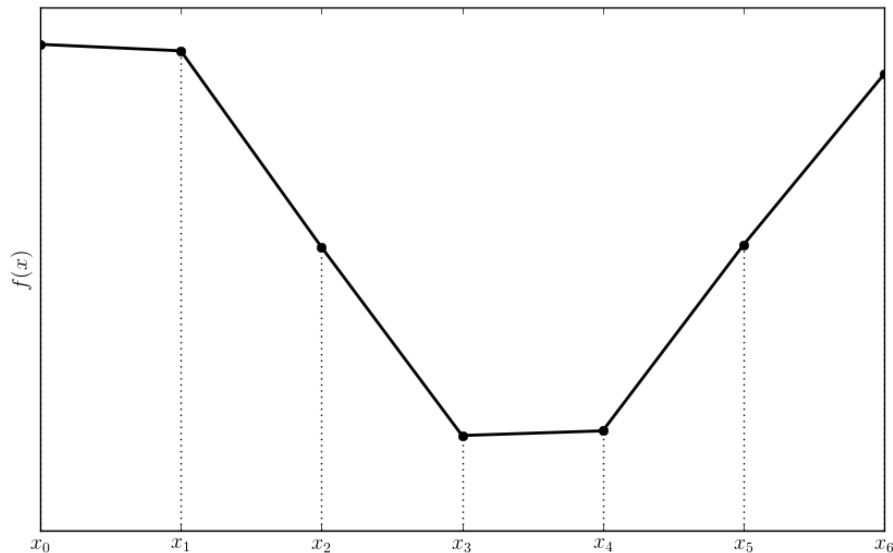


Notice!

Unstructured meshes  
are slower than  
structured for given  
size.

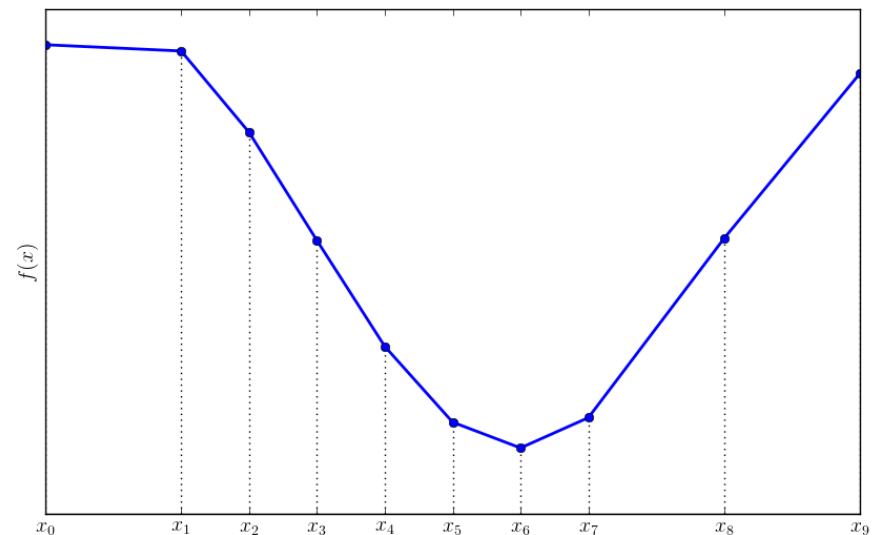
Adaptivity is important

# Current Methodology – $h$ adaptivity

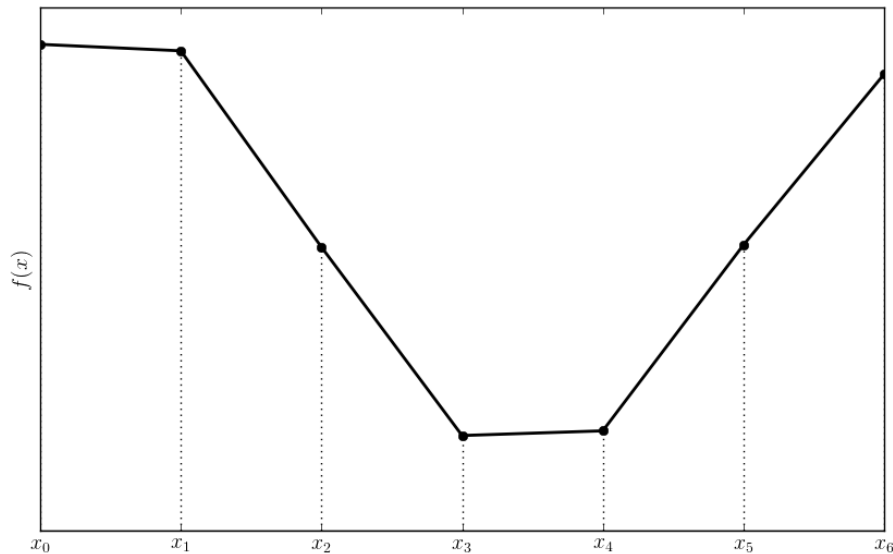


Just like AMR for elements

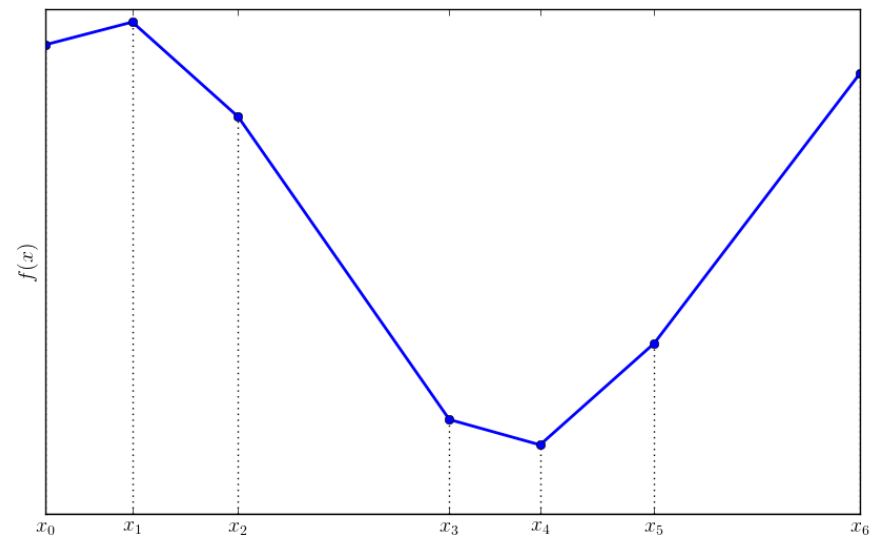
$h$  – for step length  
like  $\Delta x$  in  
finite difference notation



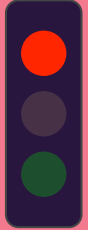
## Current Methodology – $r$ adaptivity



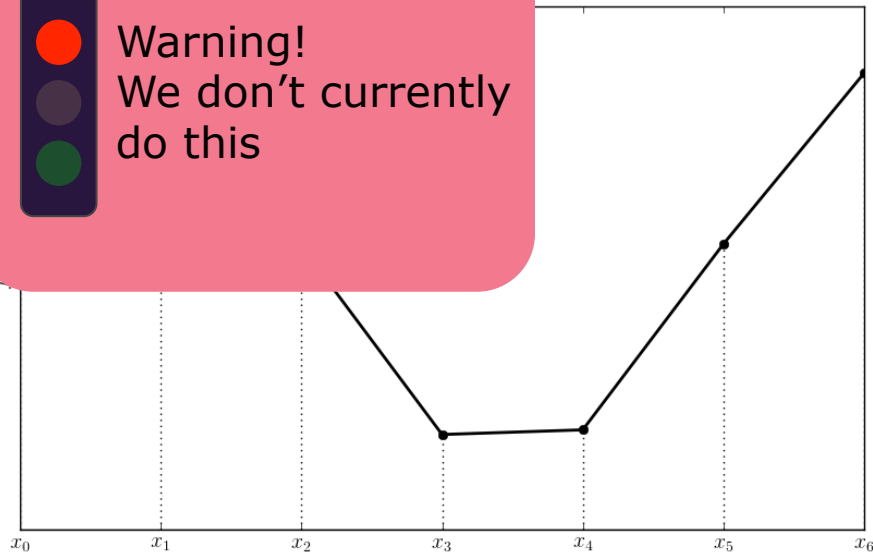
$r$  for  
redistribution/  
repositioning



## Current Methodology – $p$ adaptivity

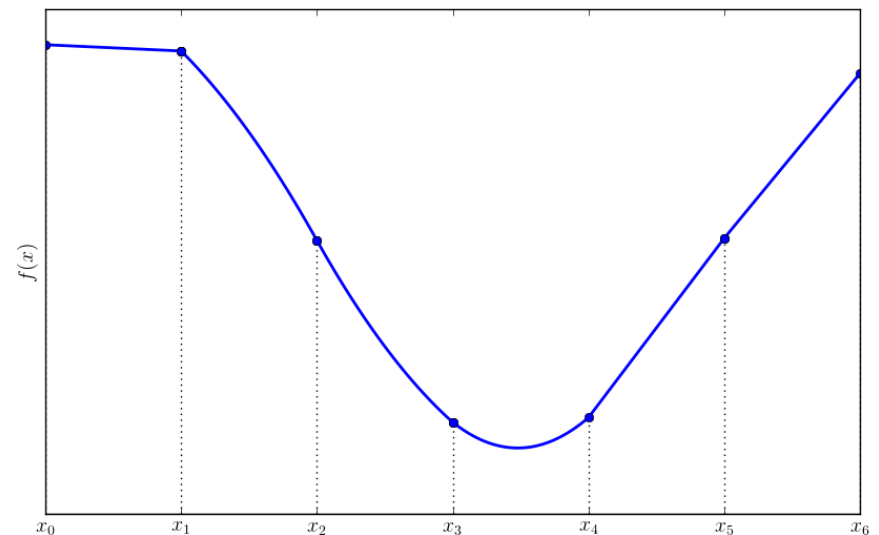


Warning!  
We don't currently  
do this



$p$  - adaptivity

P for polynomial  
order

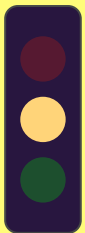


# Current Methodology – Error estimates

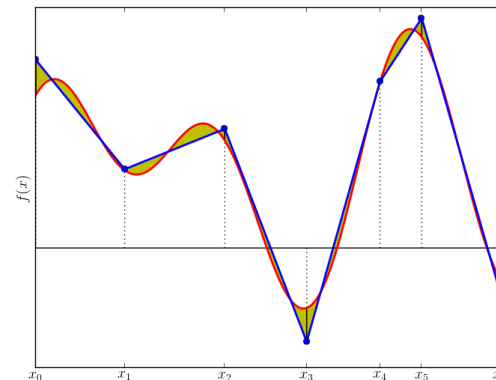
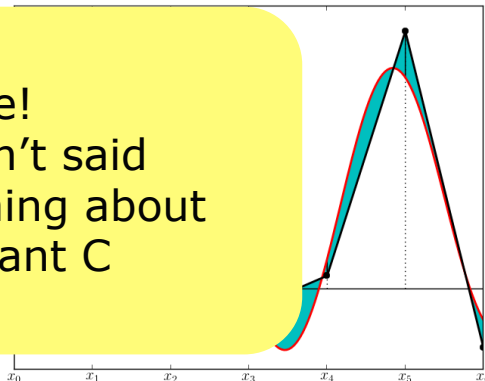
Motivation: Céa's Lemma

For sufficiently nice PDEs and linear (or better) elements

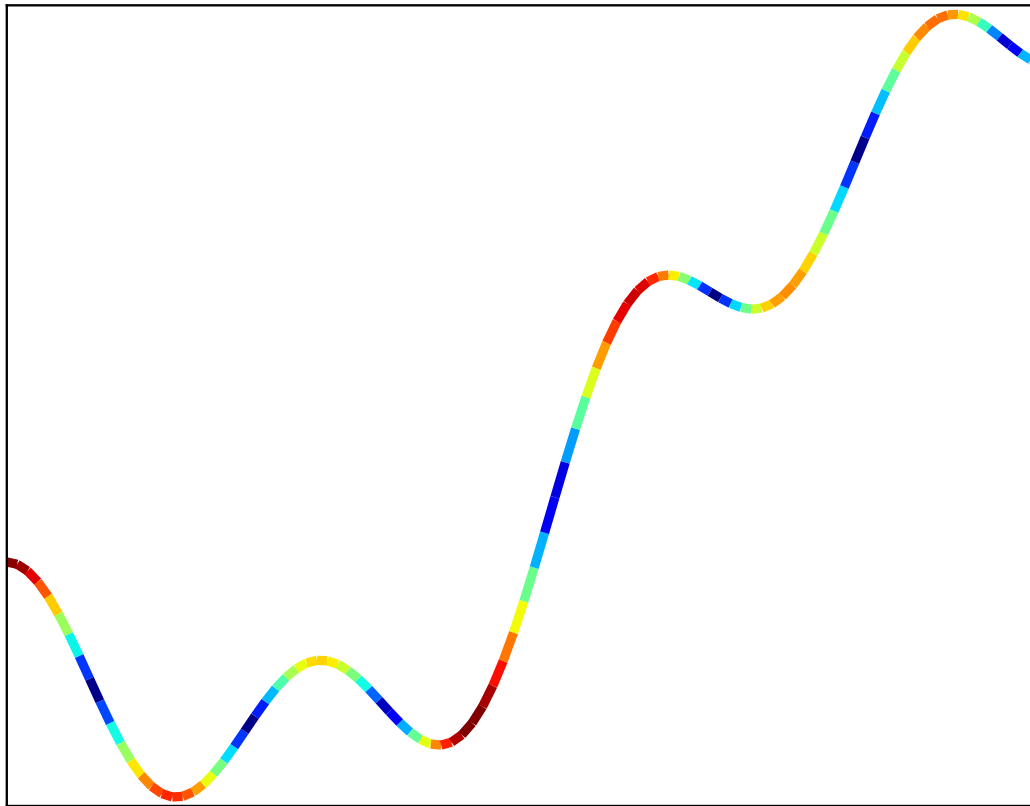
$$\begin{aligned} \text{error} &:= \left\| \psi^{\text{exact}} - \psi^\delta \right\| \leq C_1 \left\| \psi^{\text{exact}} - \psi^{\text{proj}} \right\| \\ &\leq C_2 \sum_i h_i^2 \max_{x \in \Omega} \left| \frac{\partial^2 \psi}{\partial x^2} \right| \end{aligned}$$



Notice!  
Haven't said  
anything about  
constant C



## Current Methodology – Interpolation error estimates



Plotting a linear function  
with areas of high  
curvature coloured red

Places where a  
linear approximation  
is bad.

## Current Methodology – Interpolation error estimates

In higher dimensions, error estimate is a function of the Hessian and the edge vectors.

$$\mathcal{H}(\psi) = \begin{pmatrix} \frac{\partial^2 \psi}{\partial x^2} & \frac{\partial^2 \psi}{\partial x \partial y} & \frac{\partial^2 \psi}{\partial x \partial z} \\ \frac{\partial^2 \psi}{\partial x \partial y} & \frac{\partial^2 \psi}{\partial y^2} & \frac{\partial^2 \psi}{\partial y \partial z} \\ \frac{\partial^2 \psi}{\partial x \partial z} & \frac{\partial^2 \psi}{\partial y \partial z} & \frac{\partial^2 \psi}{\partial z^2} \end{pmatrix}$$

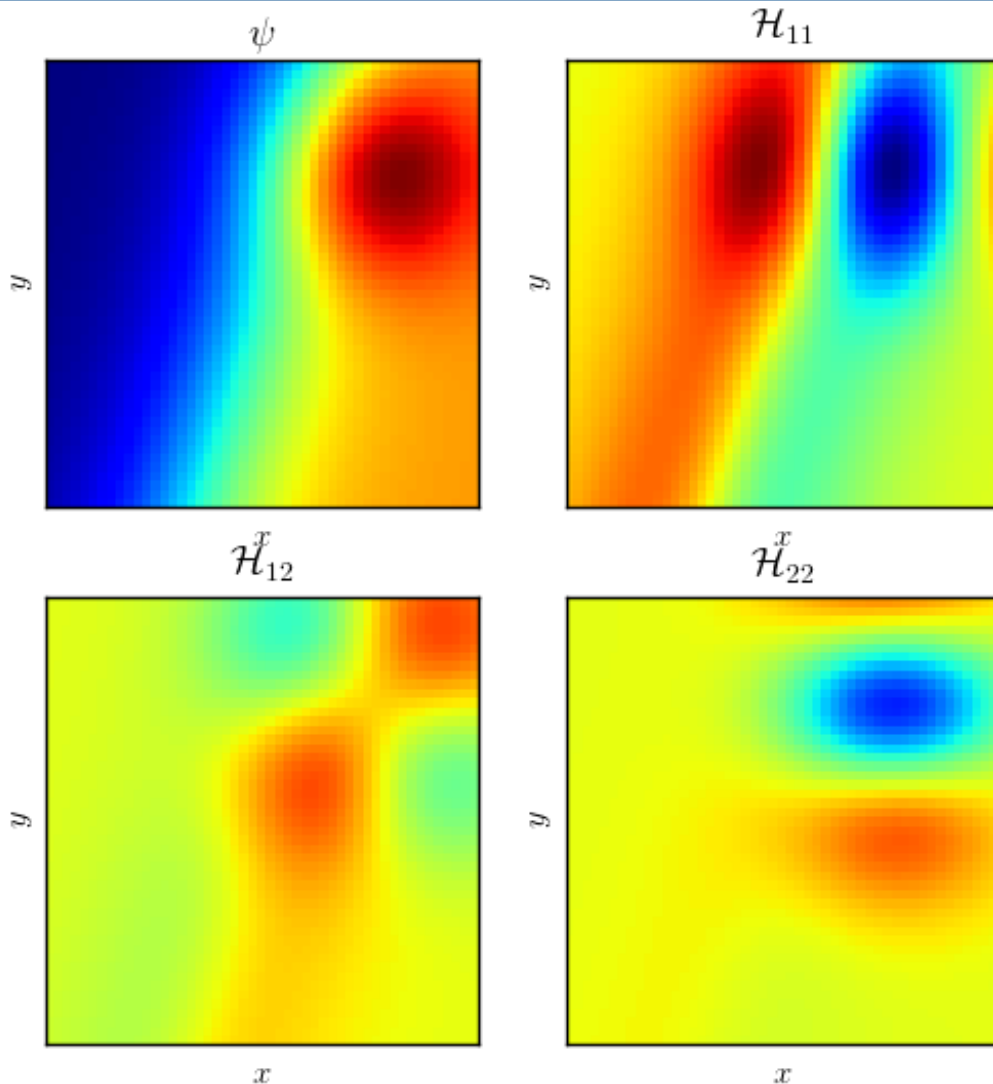
$$\text{error} \sim \sum_k \mathbf{v}_k^T \mathcal{M} \mathbf{v}_k$$

$$\mathcal{M} = \mathcal{M}(\mathcal{H}(\psi))$$

The  $\mathbf{v}$ s are the edges of the mesh

“Mesh metric”

## Current Methodology – Interpolation error estimates



Pick out areas of curvature in physical space or in value.

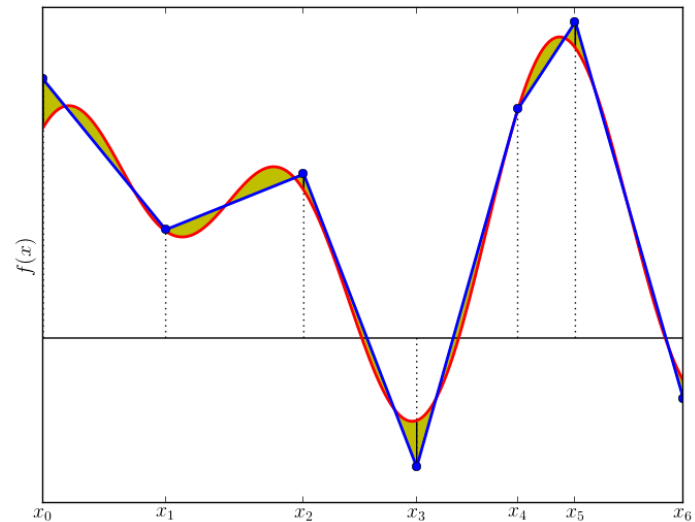
Anisotropic:  
Allowable scales  
along a front larger  
than across it.

# Current Methodology – Interpolation Error Estimates

**Problem:** we don't know  $\psi^{\text{exact}}$   
**Answer:** Use  $\psi^\delta$  instead.

Estimate second derivatives  
 from the finite element data

Need a signal present in the  
 data before we can adapt to it



Try to extremize

$$I(\mathbf{v}) = \sum_k \frac{1}{\epsilon} \mathbf{v}_k^T \mathcal{M} \mathbf{v}_k - 1$$

The  $\mathbf{v}$ s are the edges of the mesh

$$\mathcal{M} = \mathcal{M}(\mathcal{H}(\psi^\delta))$$

## Current Methodology – Optimization

Two possibilities once metric is calculated:

1. Global remeshing: "Slash and burn"  
Delete old mesh structure and create new mesh from blank sheet of paper
2. Local remeshing: "Patch and fix"  
Preserve the old mesh in regions where it is adequate

IC-FERST impliments the second option.

- Advantages in speed, data retention and parallelism

Apply *hr* adaptivity in the regions where fixing is necessary: Add, remove nodes, reconnect nodes and move nodes.

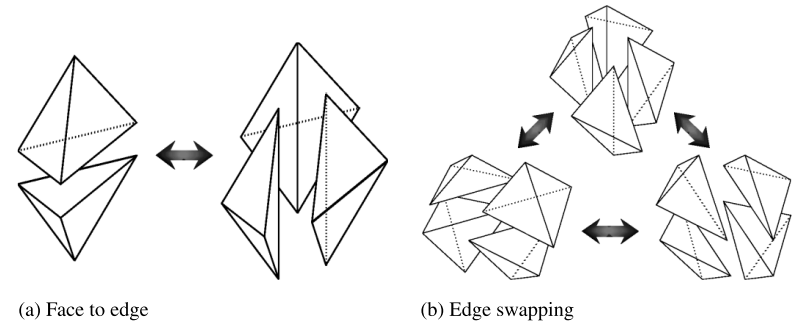


Fig. 1. Digram showing: (a) edge to face and face to edge swapping; (b) edge to edge swapping with four elements.

Pain et al. 2001

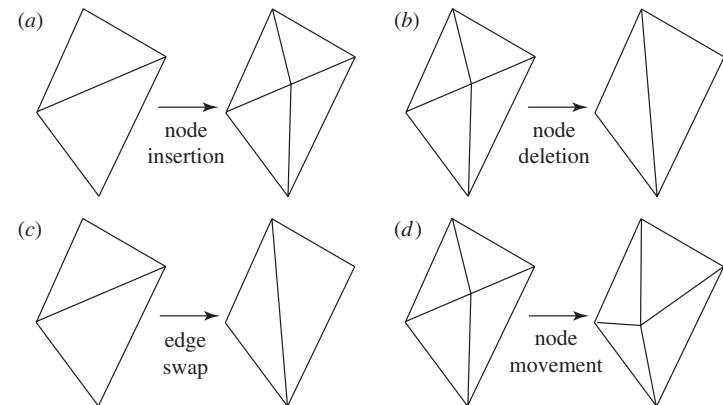


Figure 1. Local element operations used to optimize the mesh in two dimensions. (a) Node insertion or edge split. (b) Node deletion or edge collapse. (c) Edge swap. (d) Node movement.

Piggott et al 2009

## Mesh metrics - Metric Advection

Adaptivity algorithm attempts to minimize error for given data at given time level.

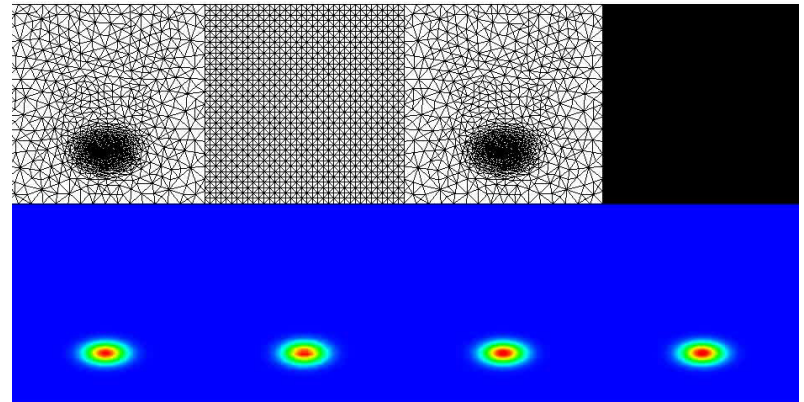
Want a mesh that remains good in the future too.

For advection dominated problems, treat metric as another advected quantity.

Extremise for

$$\sum_k \frac{1}{\epsilon} \max_{\tau \in [t_0, t_f]} \left( \mathbf{v}_k^T \mathcal{M}(\tau) \mathbf{v}_k \right) - 1$$

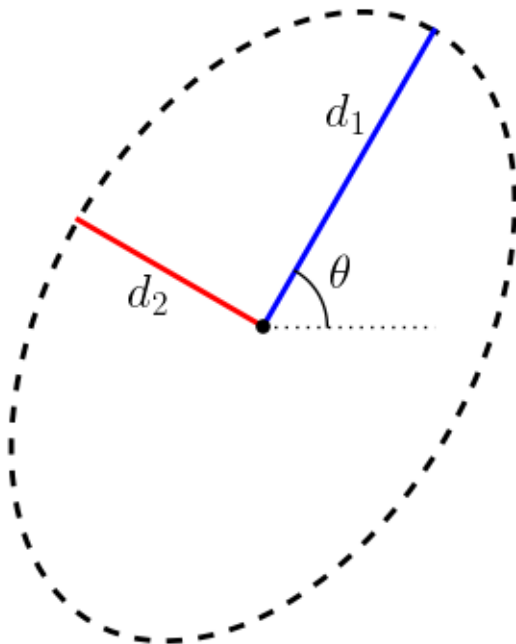
$$\frac{\partial \mathcal{M}}{\partial t} + \mathbf{u}(t_0) \cdot \nabla \mathcal{M} = 0, \quad t \in [t_0, t_f]$$



## Mesh metrics – bounding edge length

Usually have constraints on:

- Minimum edge length [cost]
- Maximum edge length [accuracy]



Simple – do surgery on specified metric after it's calculated

$$\Phi = \begin{pmatrix} a_{11} & a_{21} \\ a_{21} & a_{22} \end{pmatrix}$$

$$= R \begin{pmatrix} d_1 & 0 \\ 0 & d_2 \end{pmatrix} R^T$$

$$R = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}$$

$$\mathbf{v}^T \Phi^{-1} \Phi^{-1} \mathbf{v}$$

## Mesh metrics - Gradation

Often a bad idea to have length scales changing very rapidly between neighbouring elements. (conditioning)

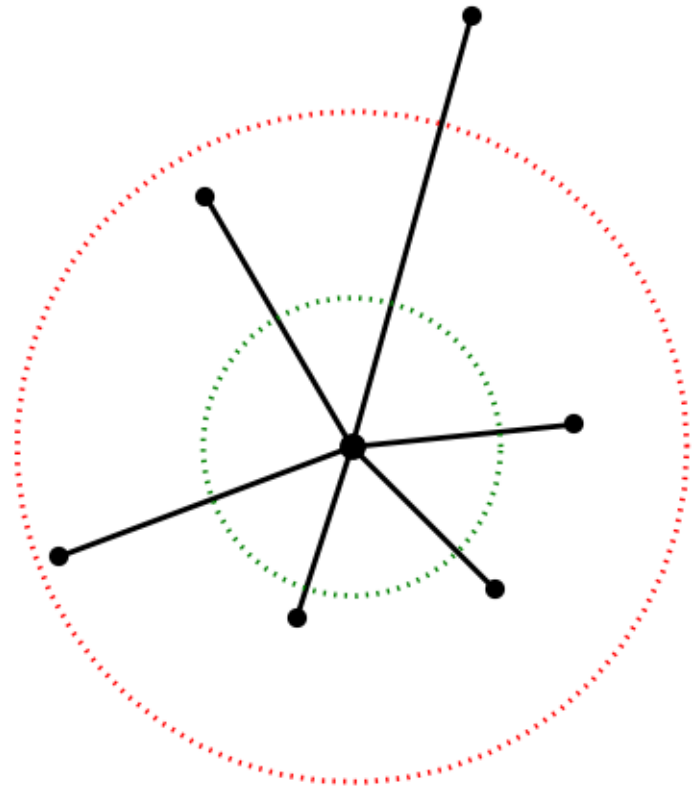
Post-processing of  $M$  can enforce smooth increases in the metric away from minima:

Isotropic

$$\|v_i\|/\|v_j\| \leq 1.5$$

Anisotropic

$$v_i \cdot \underline{\Gamma} \cdot v_j \leq \|v_j\|^2$$



## Mesh Metrics - Other knobs and levers

Aspect ratio –like gradation, but edges of a single element

“Fixed” surfaces

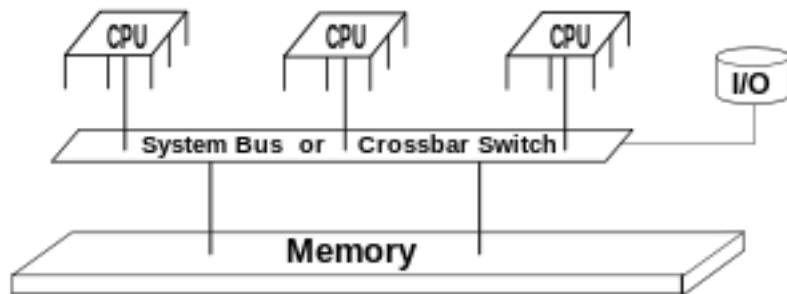


## Mesh – Metrics – Aside on parallelism

### Shared Memory

Multiple processors shared memory  
(RAM/Hard disk etc)  
Threading

“Painting a wall/Piano duet”



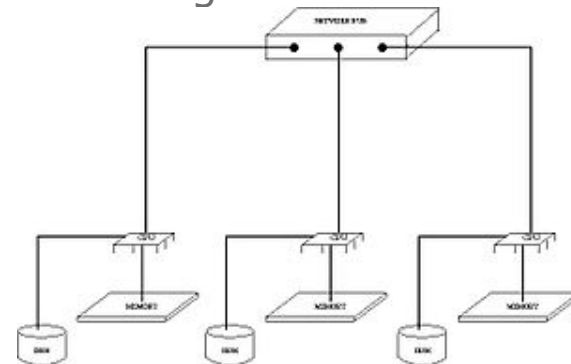
Communication fast, but hard to scale

Not currently implemented in IC-FERST  
but in use in single-phase Fluidity

### Distributed Memory

Multiple systems each have  
their own RAM/hard disk and  
communicate over a LAN.

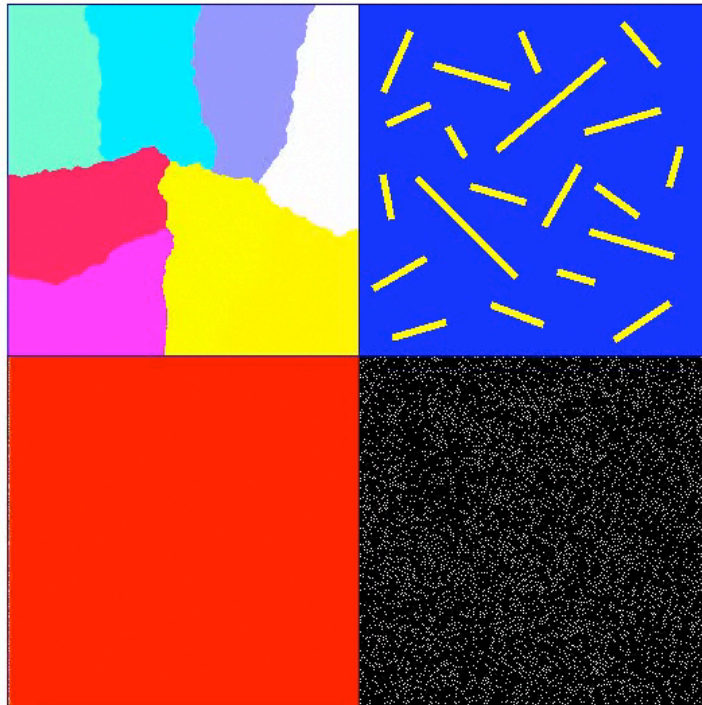
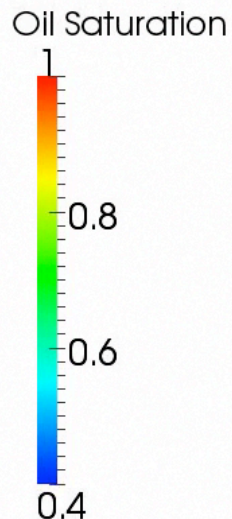
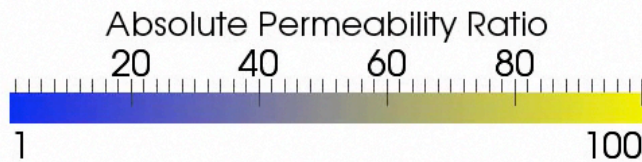
“Filming a movie



Scales well, but  
communication is slow.

Implimented and under  
testing in IC-FERST

## Mesh Metrics - Parallel



IC FERST uses distributed memory paradigm for parallelism.

Each process acts independently and communicates information as needed.

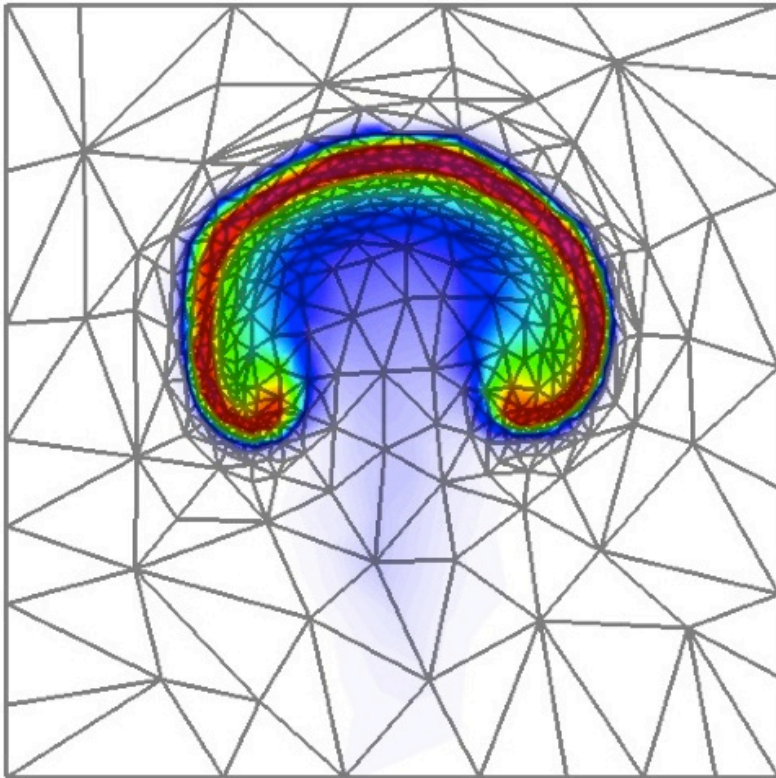
For adaptivity this means locking boundary nodes to other processes, then adapting the rest of the domain as in serial.

Once new mesh is found, nodes are redistributed (hopefully shifting the ownership boundaries)

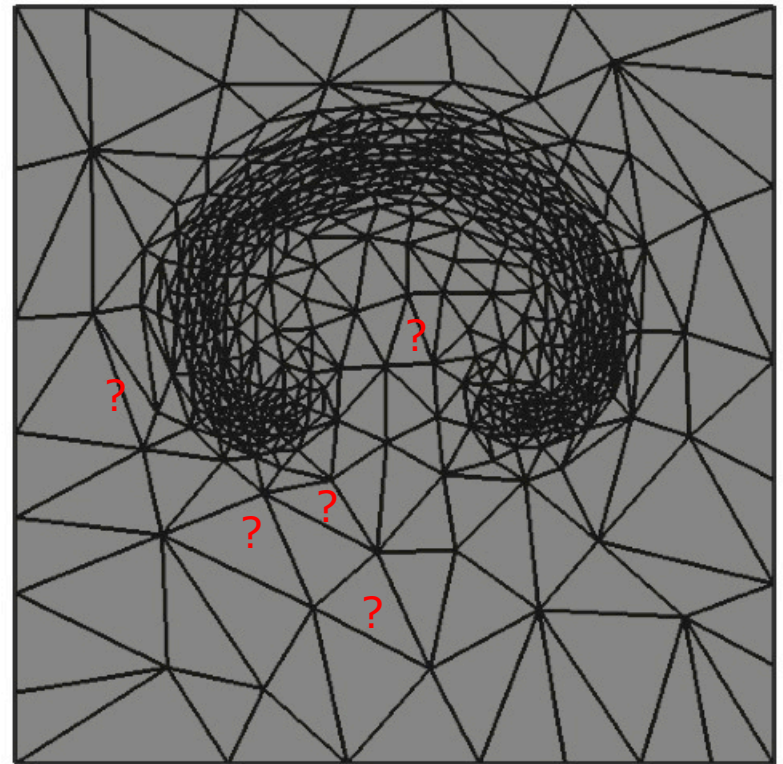
This loop is then repeated several times.

## Current Methodology – mesh to mesh interpolation

Data on old mesh



Data on new mesh?



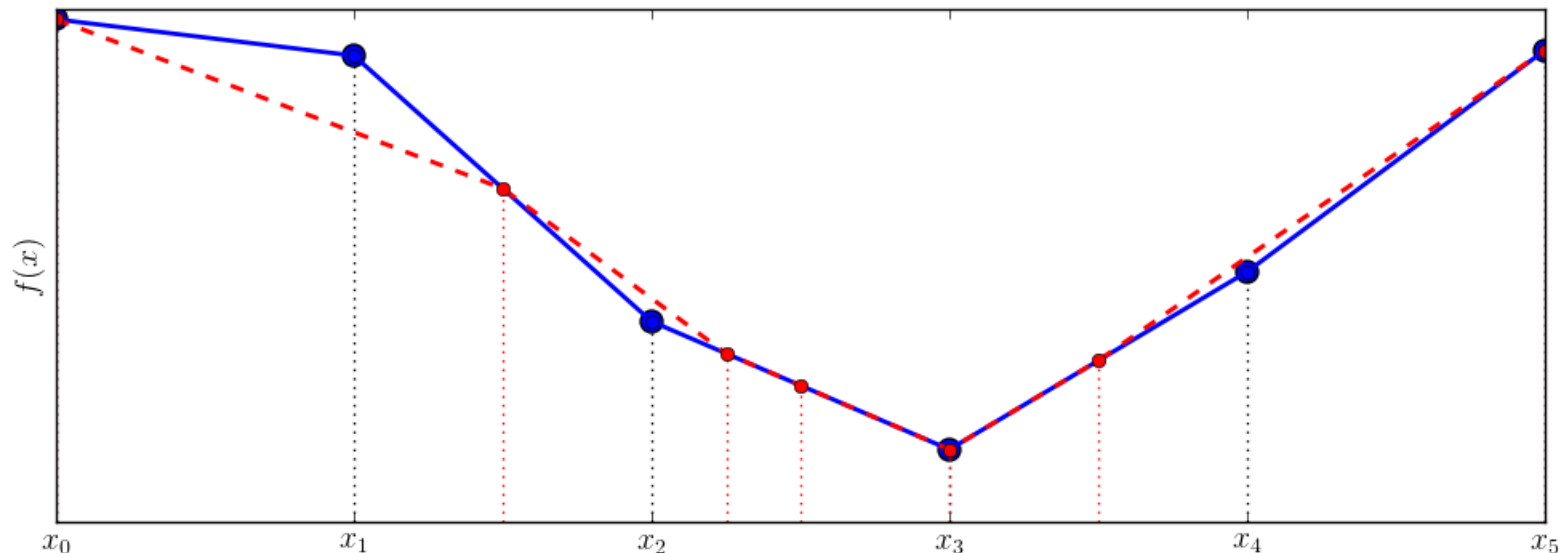
# Interpolation – Method 1 “Consistent” Interpolation

Set values at new nodes to be spatial value on old finite element representation

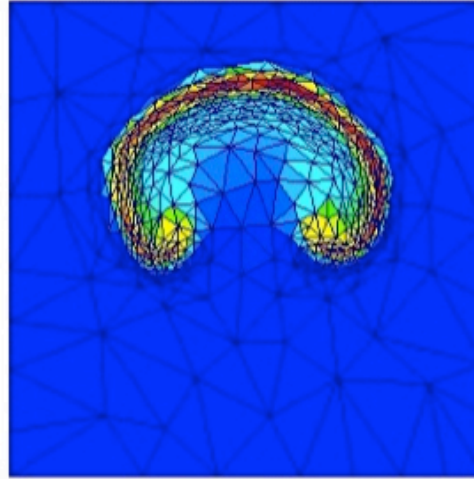
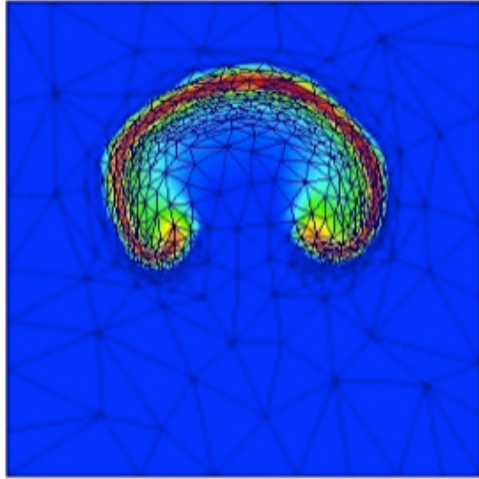
Pros:  
Cheap  
Bounded  
Gives original data back if mesh doesn't change

$$\psi_i^{\text{new}} = \psi^{\text{new}}(\mathbf{p}_i) = \sum_j N_j^{\text{old}}(\mathbf{p}_j) \psi_j^{\text{old}}$$

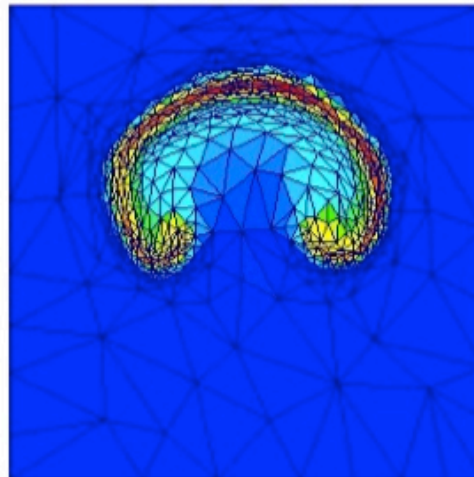
Cons:  
Nonconservative



## Interpolation – Method 2 - Grandy Interpolation

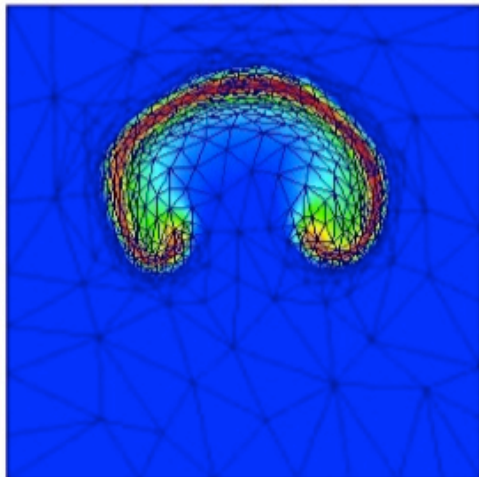


1. Average mesh data onto old elements as flat functions.
2. Calculate average on new elements
3. Project back to original shapes.



Pros:  
Cheap(ish)  
Bounded  
Local

Cons:  
Dissipative  
especially for  
discontinuous  
fields

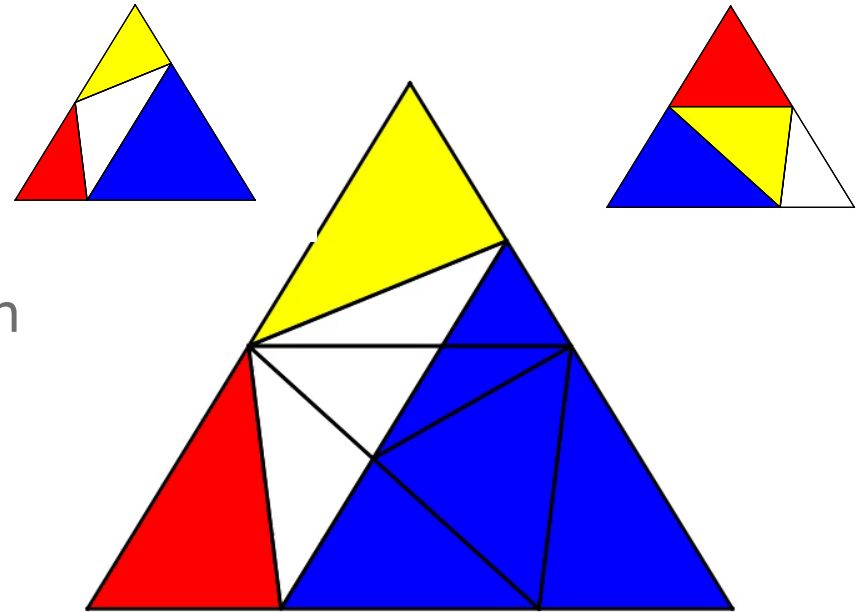


# Supermeshing

FE solutions/test functions  
piecewise smooth over mesh  
elements

Elements of supermesh: old  
variables and new test fns both  
smooth.  
No jumps.

Allows efficient  
conservative mesh to  
mesh interpolation via  
projection methods



P. E. Farrell & J. R. Maddison (2011)  
Computer Methods in Applied  
Mechanics and Engineering

## Interpolation - Galerkin Projection

Galerkin Finite Element solution to  $\psi^{\text{new}} = \psi^{\text{old}}$

$$\sum_j \int_{\Omega} N_i^{\text{new}} N_j^{\text{new}} \psi_j^{\text{new}} dV = \sum_k \int_{\Omega} N_i^{\text{new}} N_k^{\text{old}} \psi_k^{\text{old}} dV$$

Left hand side is usual mass matrix

Right hand side is trickier  
Supermeshing to the rescue!

Pros

Conservative

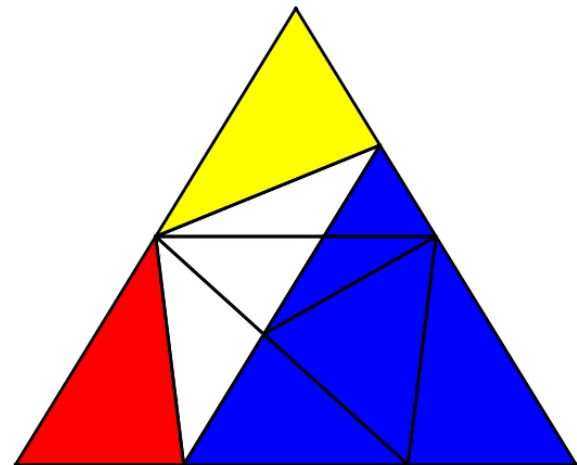
Local

Cons

No bounding

(more) expensive

Questions over control volumes



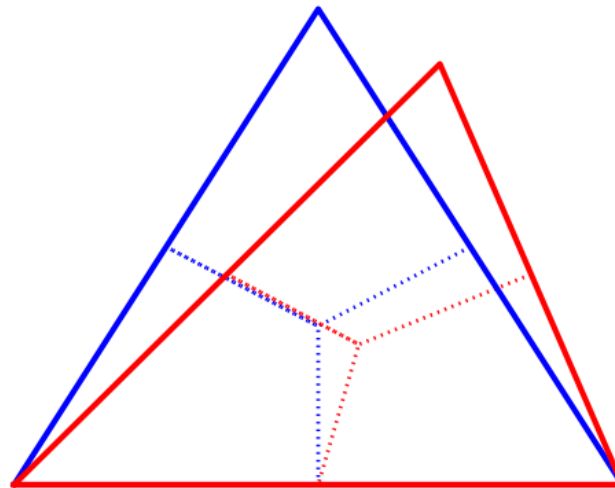
## Discussion

### Questions to the room:

- What variables are worthwhile to adapt to?
  - Saturation? Pressure?
- Do the length-scales inside the absolute permeability tensor matter?
- Interpolation methods for control volumes in the fully discontinuous formulation

## The issue with control volumes

Supermeshing is much easier with triangles/tetrahedra than with control volumes

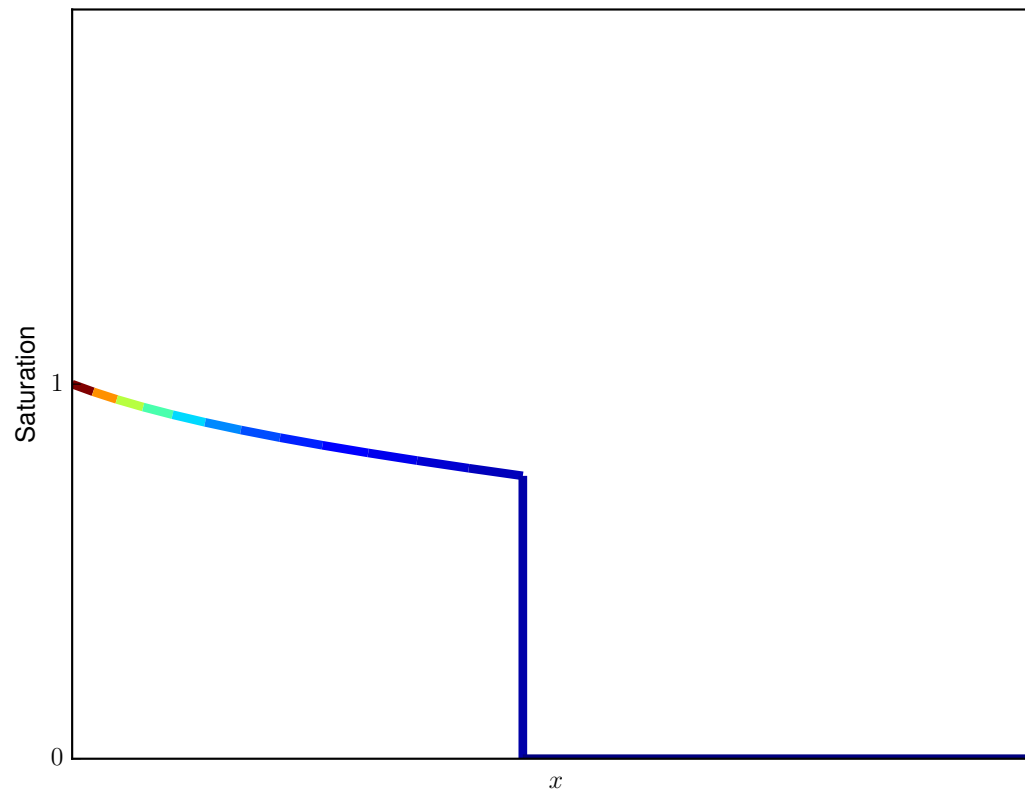


Moving one node creates 3 triangles of overlap for FE

Moving one node creates 7 quadrilaterals and one triangles of overlap for CV

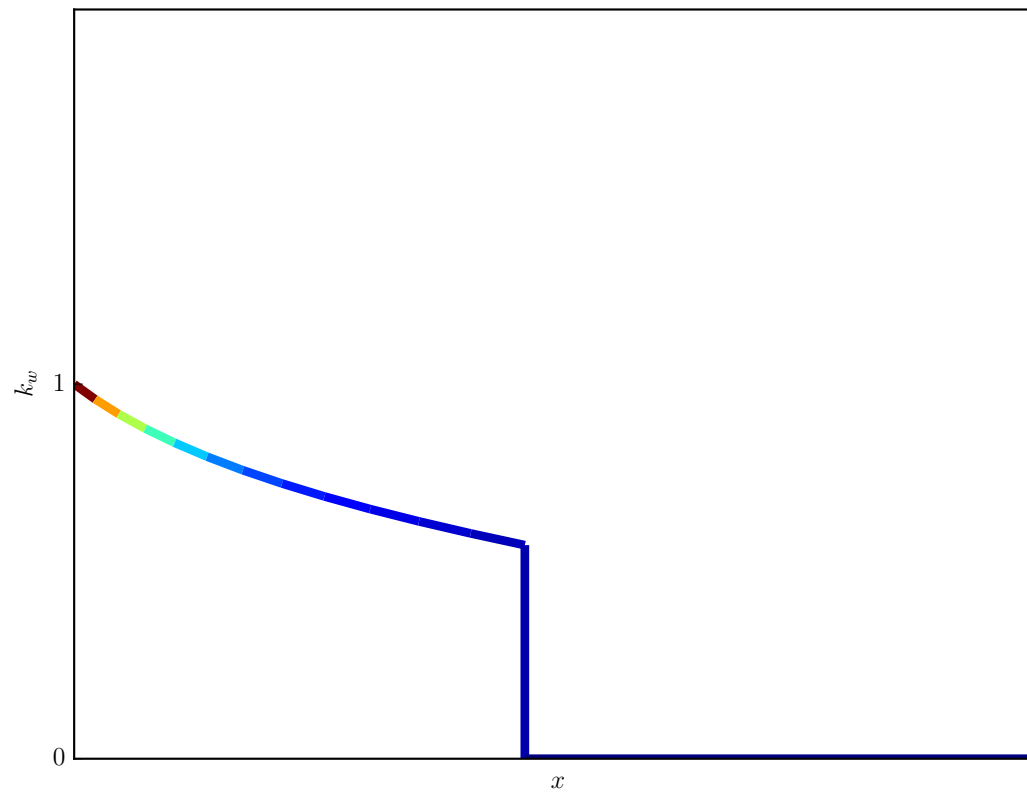
## Discussion: Choice of Variables to adapt to

Buckley Leverett solution : Saturation



## Discussion: Choice of Variables to adapt to

Buckley Leverett solution : Relative Permeability

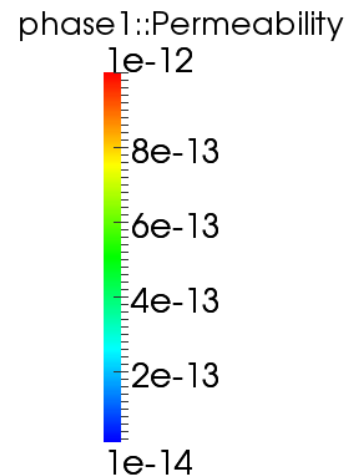
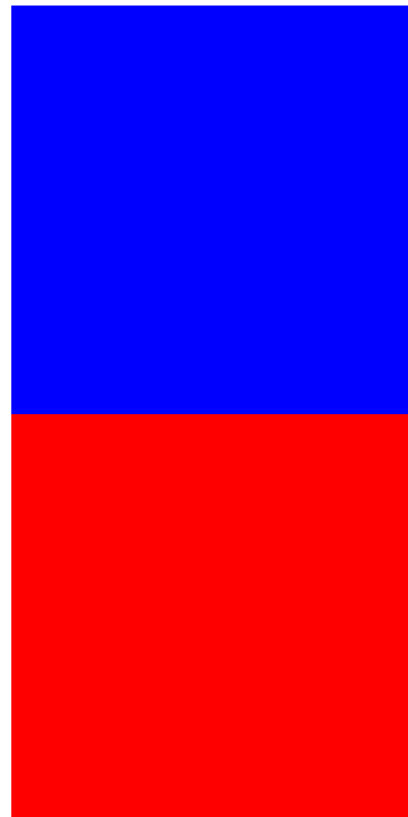


## Discussion: Choice of Variables to adapt to

Velocity or pressure?

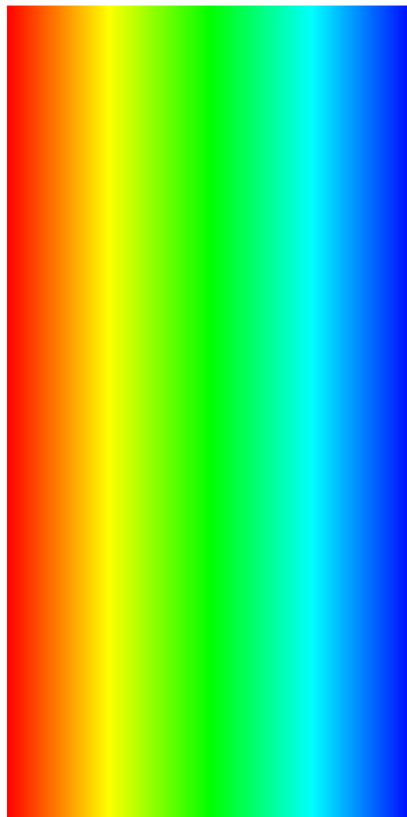
Single phase Darcy flow  
through two porous media

Prescribed pressure  
difference

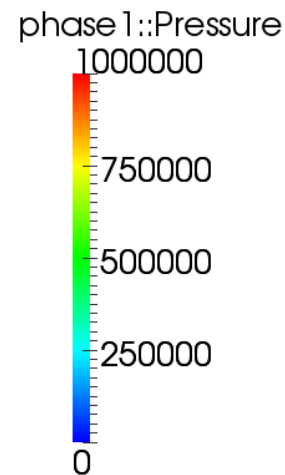


## Discussion: Choice of Variables to adapt to

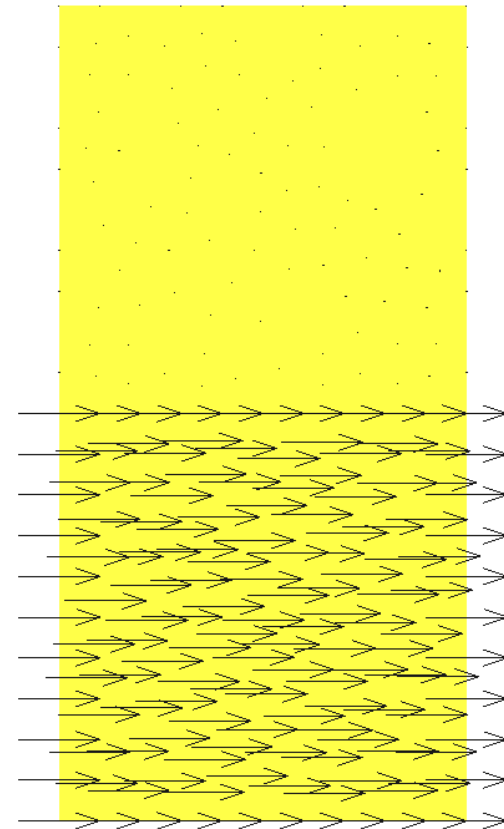
Pressure



Velocity or pressure?

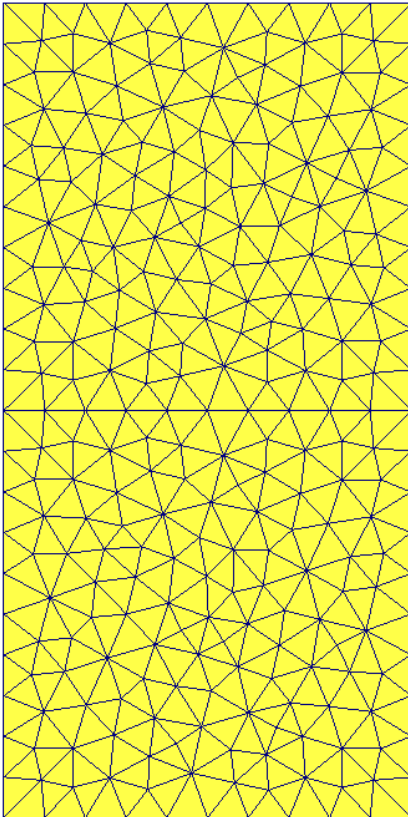


Velocity



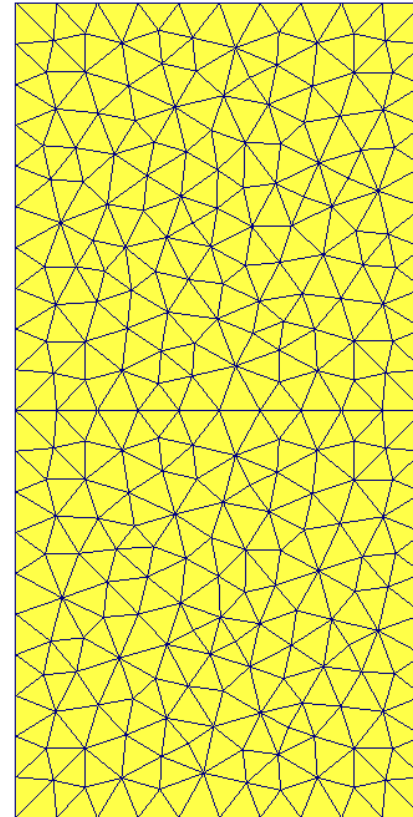
## Discussion: Choice of Variables to adapt to

Pressure



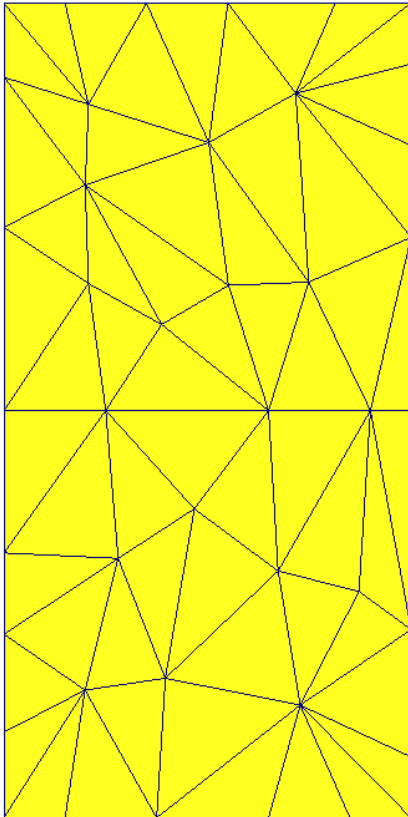
Velocity or pressure?

Velocity



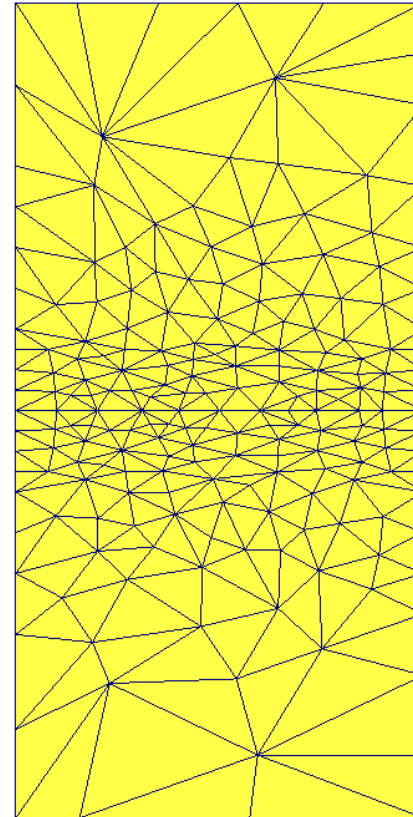
## Discussion: Choice of Variables to adapt to

Pressure



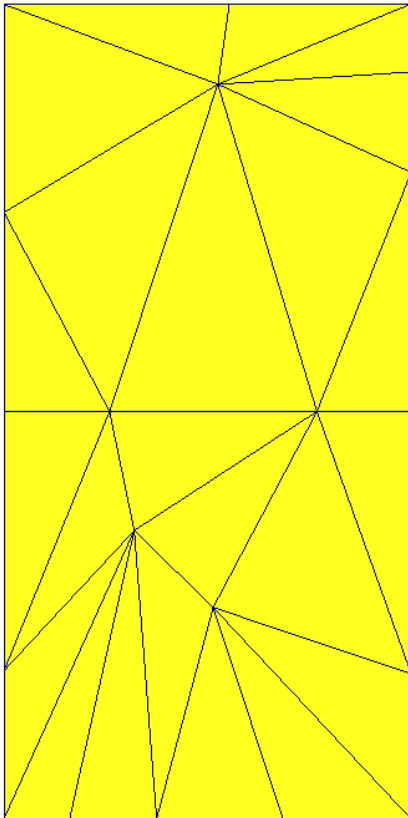
Velocity or pressure?

Velocity



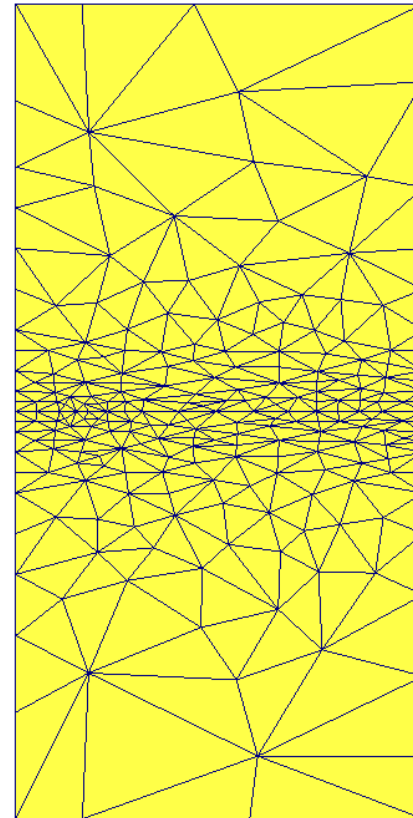
## Discussion: Choice of Variables to adapt to

Pressure



Velocity or pressure?

Velocity





**Any further points?**

**Thank you very much for your input**

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Room 4.85 RSM