

# Algebraic Multigrid

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# Problem?

- Consider solving the  $n \times n$  system:

$$Ax = b$$

- How do we solve these systems optimally?

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- Consider solving the  $n \times n$  system:

$$Ax = b$$

- How do we solve these systems optimally?
- We have two options **Direct** or **Iterative**
- Direct**: flops  $\approx \mathcal{O}(n^3)$ , large memory cost
- Iterative**: flop  $\approx \mathcal{O}(n)$ ???, low memory cost

# Generic iterative method form

- Most iterative methods have the following form, where  $r_k$  is the residual at iteration  $k$

$$x_{k+1} = x_k + M^{-1}r_k$$

- Let  $e_k = x - x_k$  be the error, and note that  $r_k = Ae_k$
- The error propagation for the iterative method is

$$e_{k+1} = (I - M^{-1}A)e_k$$

# Why parallel?

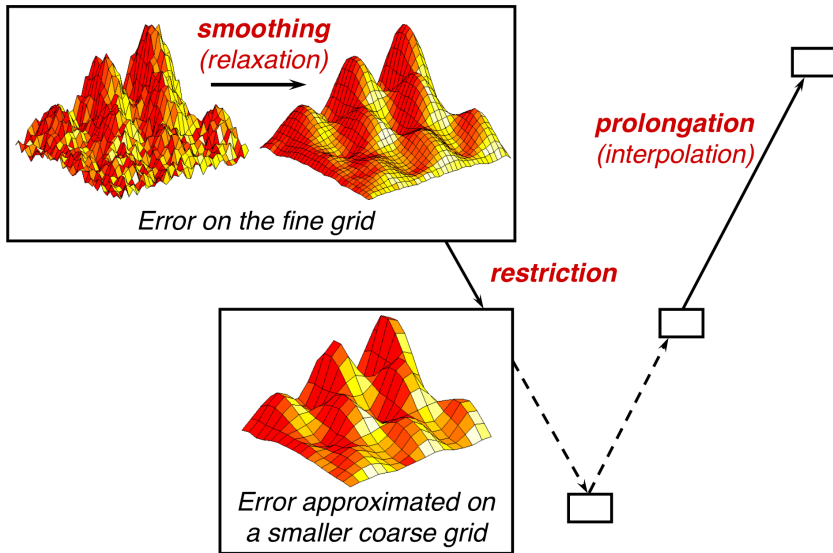
# Why parallel?

Basic answer: real world problems are huge!!!

# Why parallel?

- Modeling earth's mantle - grid cells about 1km across
- Geophysical surveys - seismic, EM, DC
- Weather forecasting
- Multi-phase porous media flow
- ...

# Geometric multigrid





# Geometric multigrid (continued)

General multigrid algorithm

**Step 1:** smooth error (residual)

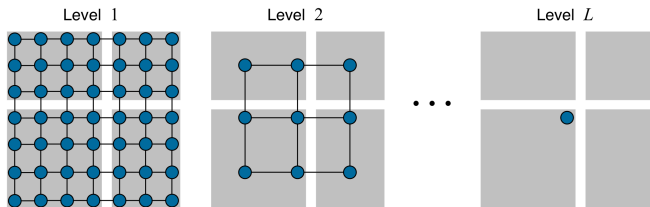
**Step 2:** restrict to coarse grid

**Step 3:** Solve on coarse grid

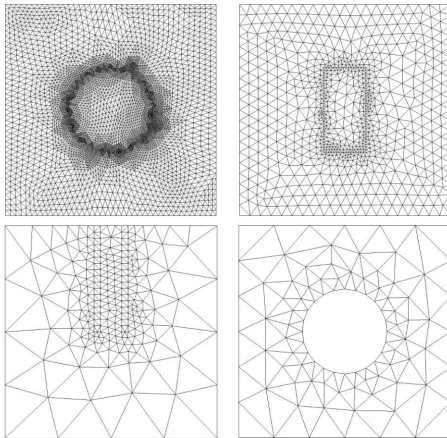
**Step 4:** prolongate and correct

# Parallel geometric multigrid

Basic idea: split up discretisation grid



# Unstructured grids



# Algebraic multigrid

How do we do this in without defining a sequence of grids

- Algebraic smoothness
- Algebraic restriction (how to define a sequence of grids)

# Algebraic multigrid: smoothness

Smoothness:

$$e^T Ae = \lambda \ll 1$$

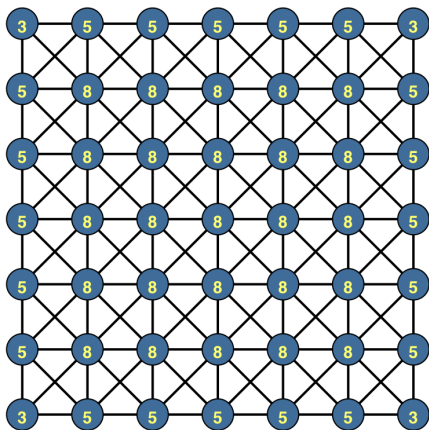
$$e^T Ae = \sum_{i < j} (-a_{ij})(e_i - e_j)^2 \ll 1$$

# Algebraic multigrid: grid selection

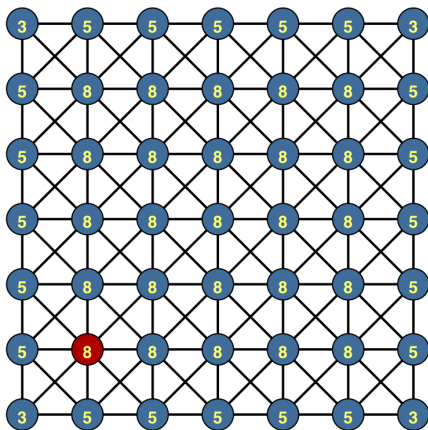
## Strength of Connection

$$-a_{ij} \geq \theta \max_{k \neq i} \{-a_{ik}\} \quad \text{where } \theta \in (0, 1]$$

1. Define strength matrix  $A_s$
2. Choose set of fine points based on  $A_s$
3. Choose extra points to satisfy interpolation requirements

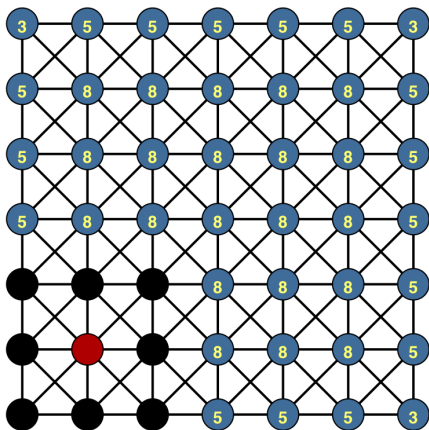


- select C-pt with maximal measure
- select neighbours as F-pts
- update measures of F-pt neighbours

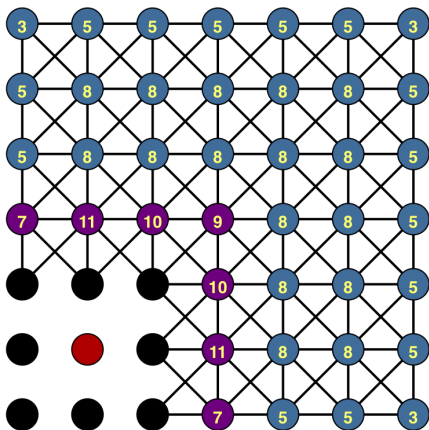


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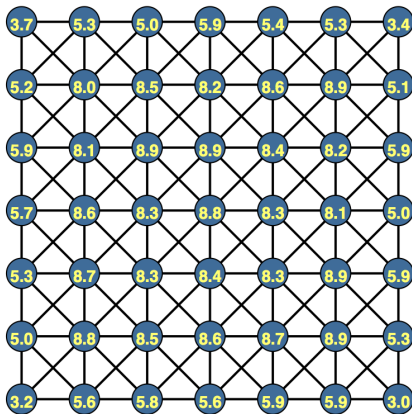
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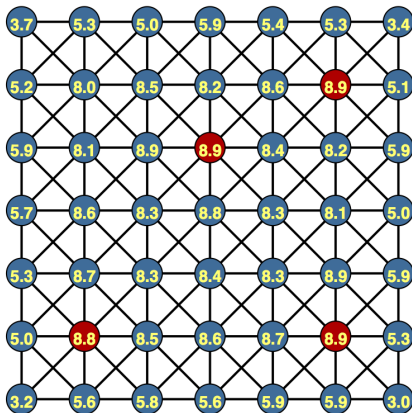
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# Parallel algebraic multigrid

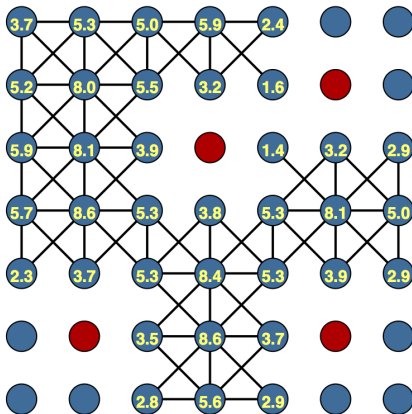
- This process is inherently sequential.....
- Parallel algorithms:
  - CLJP (Cleary-Luby-Jones-Plassmann) – one-pass approach with random numbers to get concurrency
  - Falgout – C-AMG on processor interior, then CLJP to finish
  - PMIS – CLJP without the ‘C’; parallel version of C-AMG first pass
  - HMIS – C-AMG on processor interior, then PMIS to finish
  - ...



- select C-pts with maximal measure locally
- remove neighbour edges
- update neighbour measures



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# Summary

- Multigrid methods are optimal and have good scaling potential
- GM: relies on a sequence of predetermined geometric grids
- AMG: uses matrix coefficients to determine a sequence of “grids”
- P-AMG: additional restrictions on AMG algorithmic development

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THANK YOU