

# Multigrid at Extreme scales: Communication Reducing Data Models and Asynchronous Algorithms

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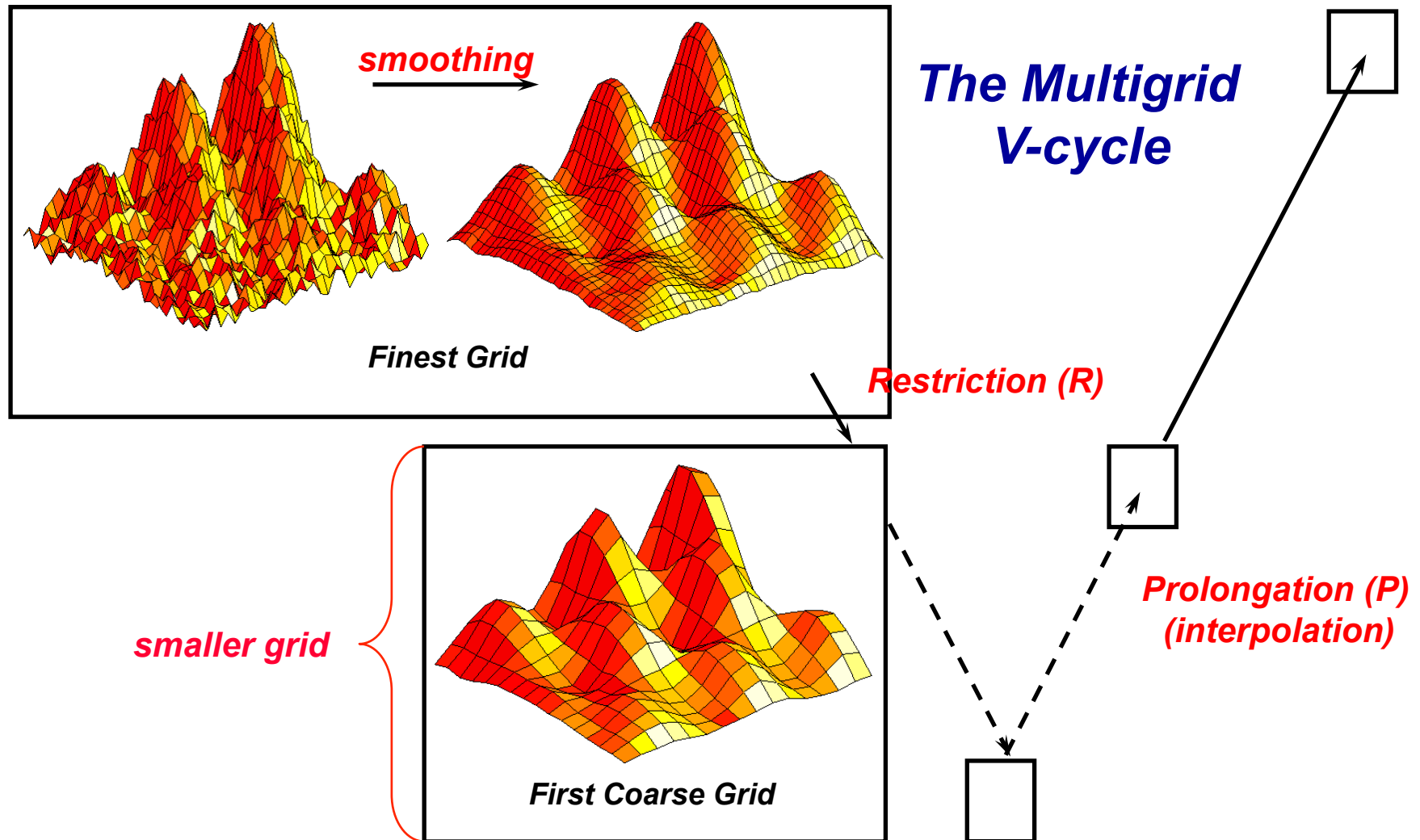
# Outline

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- Establish a lower bound on solver complexity
  - Apply ideas to Magnetohydrodynamics (MHD)
- Distributed memory & communication avoiding MG
  - Asynchronous unstructured Gauss-Seidel
- New algebraic multigrid (AMG) in PETSc
  - Application to 3D elasticity and 2D Poisson solves
- Data centric MG: cache aware & communication avoiding
  - Application to 2D 5-point stencil  $V(1,1)$  cycle

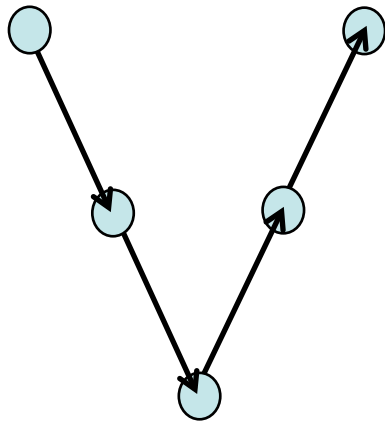


# Multigrid motivation: smoothing and coarse grid correction

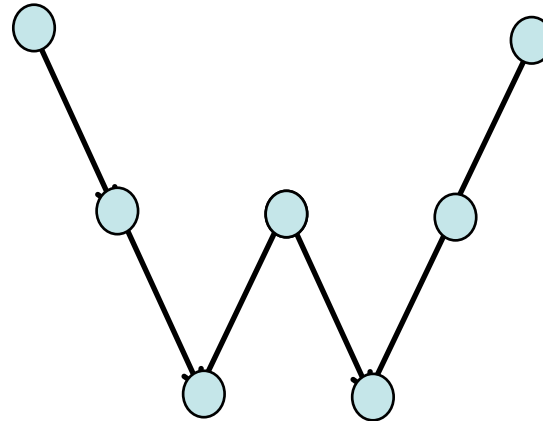


## Multigrid Cycles

V-cycle

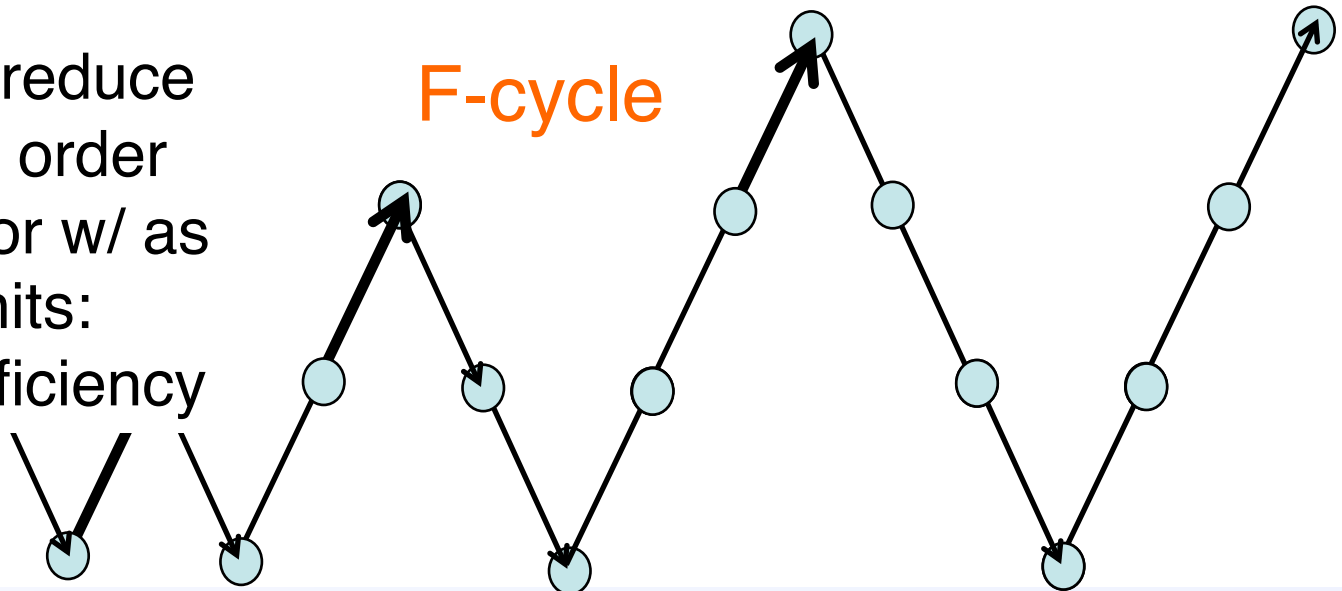


W-cycle



One F-cycle can reduce algebraic error to order discretization error w/ as little as 5 work units:  
“textbook” MG efficiency

F-cycle



## Discretization error in one F-cycle (Bank, Dupont, 1981)

- Define error:  $E(x) \leq E_d(x) + E_a(x)$  (discrete. + algebraic)
- Assume error  $E_d(x) \leq Ch^p$  (point-wise theory)
- Example: 2<sup>nd</sup> ( $p=2$ ) order discretization & coarsening factor of 2.
- **Induction** hypothesis: require  $r \geq E_a/E_d$  (eg,  $r=1/2$ )
- Define  $\Gamma$  rate error reduction of solver (eg, 0.1 w/ a V-cycle)
  - Can *prove* this or *determine experimentally*
  - No  $\Gamma$  w/defect correction – can use  $\Gamma$  of low order method.
- Use induction: Error from coarse grid:  $C(2h)^2 + r \cdot C(2h)^2$ 
  - Alg. Err. Before V-cycle:  $E_a < C(2h)^2 + r \cdot C(2h)^2 - Ch^2$ 
    - Actually should be  $+Ch^2$  but sign of error should be same
  - And we want  $\Gamma \cdot E_a = \Gamma \cdot (C(2h)^2 + r \cdot C(2h)^2 - Ch^2) < r \cdot E_d \leq r \cdot Ch^2$
  - $\Gamma = r/(4r + 3)$ , 1 equation, 2 unknowns ... fix one:
    - eg,  $r = 1/2 \rightarrow \Gamma = 0.1$
    - If you want to use  $+ Ch^2$  term then its  $\Gamma = r/(4r + 5)$

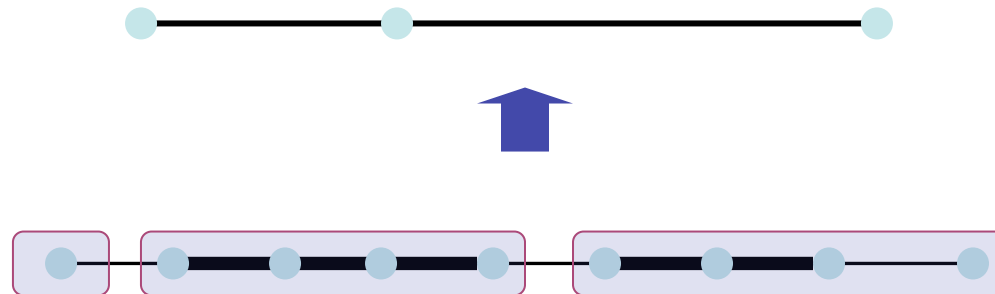
## Multigrid $V(v_1, v_2)$ & $F(v_1, v_2)$ cycle

- function  $u = \text{MGV}(A, f)$ 
  - If  $A$  coarsest grid
    - $u \leftarrow A^{-1}f$
  - else
    - $u \leftarrow S^{v_1}(f, 0)$                       -- Smoother (pre)
    - $r_H \leftarrow P^T(f - Au)$
    - $e_H \leftarrow \text{MGV}(P^TAP, r_H)$               -- recursion (Galerkin)
    - $u \leftarrow u + Pe_H$
    - $u \leftarrow S^{v_2}(f, u)$                       -- Smoother (post)
  
- function  $u = \text{MGF}(A_i, f)$ 
  - if  $A_i$  is coarsest grid
    - $u \leftarrow A_i^{-1}f$
  - else
    - $r_H \leftarrow R f$
    - $e_H \leftarrow \text{FGV}(A_{i-1}, r_H)$               -- recursion
    - $u \leftarrow Pe_H$
    - $u \leftarrow u + \text{MGV}(A_i, f - A_i u)$

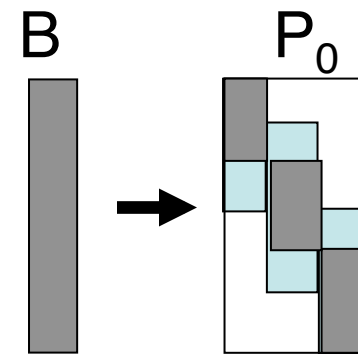


## Algebraic multigrid (AMG) - Smoothed Aggregation

- MG requires a smoother and coarse grid space
  - Columns of  $P$
- Piecewise constant functions are easy
  - “Plain” aggregation
- Nodal aggregation, or partitioning
- Example: 1D 3-point stencil



Kernel vectors of operator ( $B$ )



“Smoothed” aggregation: lower energy of functions

For example: one Jacobi iteration:  $P \leftarrow (I - \omega D^{-1} A) P_0$

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## Compressible resistive MHD equations in strong conservation form

$$\frac{\partial U}{\partial t} + \frac{\partial F_j(U)}{\partial x_j} = \frac{\partial \tilde{F}_j(U)}{\partial x_j} \rightarrow \begin{array}{l} \text{Diffusive} \\ \text{Hyperbolic} \end{array}$$

$$U = \{\rho, \rho u_i, B_i, e\}^T$$

$$F_j(U) = \left\{ \begin{array}{l} \rho u_j \\ \rho u_i u_j + p \delta_{ij} + \frac{1}{2} B_k B_k \delta_{ij} - B_i B_j \\ u_j B_i - B_j u_i \\ (e + p + \frac{1}{2} B_k B_k) u_j - B_i u_i B_j \end{array} \right\}$$

$$\tau_{ij} = \rho \nu \left( \frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} - \frac{2}{3} \delta_{ij} \frac{\partial u_k}{\partial x_k} \right)$$

$$e = \frac{p}{\gamma - 1} + \frac{1}{2} \rho u_i u_i + \frac{1}{2} B_i B_i$$

$$\tilde{F}_j(U) = \left\{ \begin{array}{l} 0 \\ Re^{-1} \tau_{ij} \\ S^{-1} \eta \left( \frac{\partial B_i}{\partial x_j} + \frac{\partial B_j}{\partial x_i} \right) \\ S^{-1} \eta \left( \frac{1}{2} \frac{\partial B_i B_i}{\partial x_j} - B_i \frac{\partial B_j}{\partial x_i} \right) + Re^{-1} \tau_{ij} u_i + Pe^{-1} \kappa \frac{\partial T}{\partial x_j} \end{array} \right\} \rightarrow \begin{array}{l} \text{Reynolds no.} \\ \text{Lundquist no.} \\ \text{Peclet no.} \end{array}$$

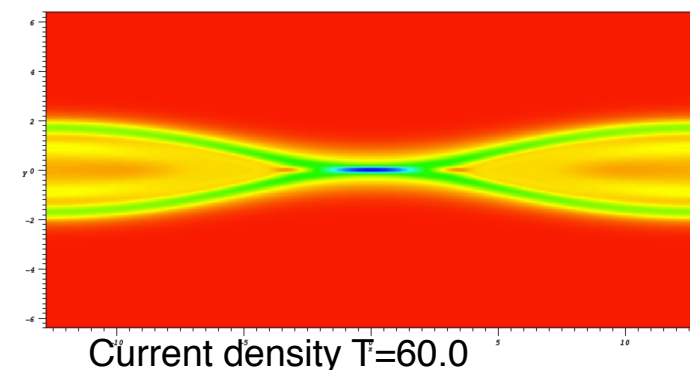
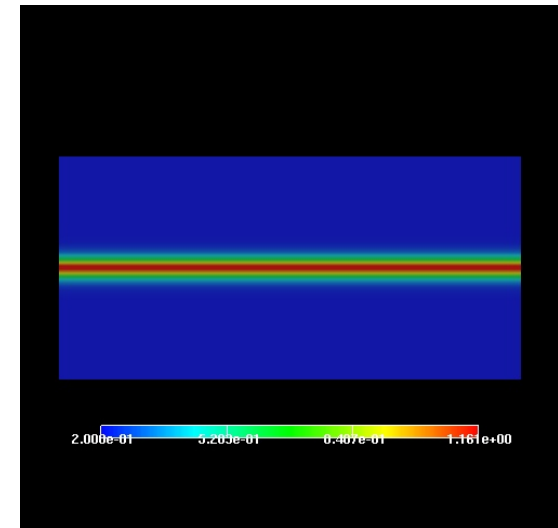
## Fully implicit resistive compressible MHD Multigrid – back to the 70's

- Geometric MG, Cartesian grids
  - Piecewise constant restriction R, linear interpolation (P)
- Red/black point Gauss-Seidel smoothers
  - Requires inner G-S solver be coded
- F-cycle
  - Two V(1,1) cycles at each level
  - Algebraic error < discretization error in one F-cycle iteration
- Matrix free - more flops less memory
  - Memory increasingly bottleneck - Matrix free is way to go
  - processors (cores) are cheap
    - memory architecture is expensive and slow (relative to CPU)
- Non-linear multigrid
  - No linearization required
- Defect correction for high order ( $L_2$ ) methods
  - Use low order discretization ( $L_1$ ) in multigrid solver (stable)
  - Solve  $L_1 x^{k+1} = f - L_2 x^k + L_1 x^k$



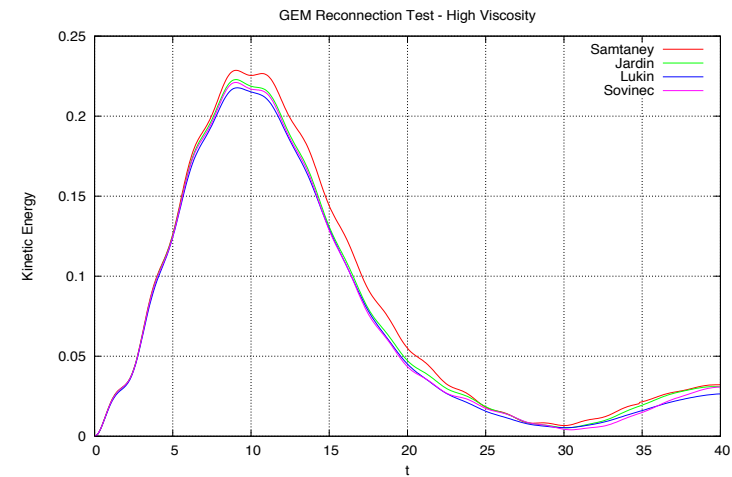
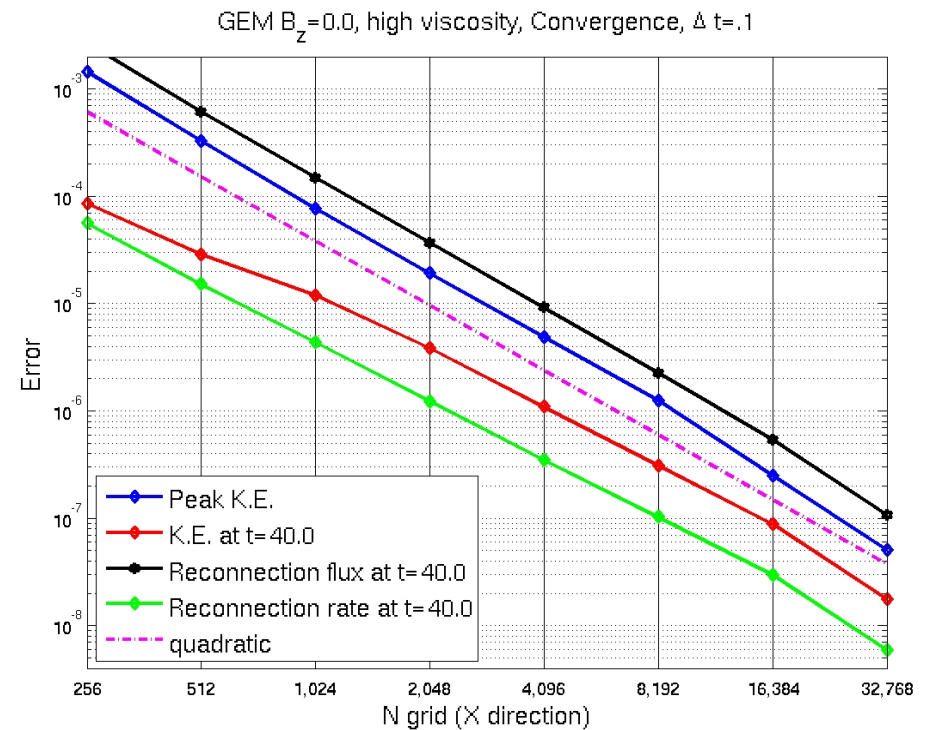
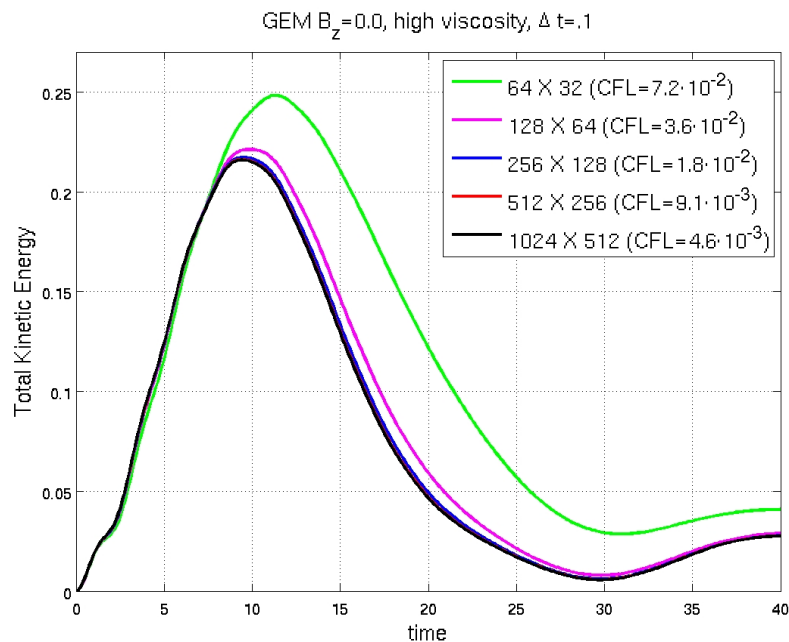
# Magnetic reconnection problem

- GEM reconnection test
  - 2D Rectangular domain, Harris sheet equilibrium
  - Density field along axis: (fig top)
  - Magnetic (smooth) step
  - Perturb B with a “pinch”
- Low order preconditioner
  - Upwind - Rusanov method
- Higher order in space: C.D.
- Solver
  - 1 F-cycle w/ 2 x V(1,1) cycles per time step
    - Nominal cost of 9 explicit time steps
    - ~18 work units per time step
- Viscosity:
  - Low:  $\mu=5.0\text{D-}04$ ,  $\eta=5.0\text{D-}03$ ,  $\kappa=2.0\text{D-}02$
  - High:  $\mu=5.0\text{D-}02$ ,  $\eta=5.0\text{D-}03$ ,  $\kappa=2.0\text{D-}02$
- $B_z$ :  $B_z=0$  and  $B_z=5.0$ 
  - Strong guide field  $B_z$  (eg, 5.0)
  - critical for tokamak plasmas



## $B_z = 0$ , High viscosity

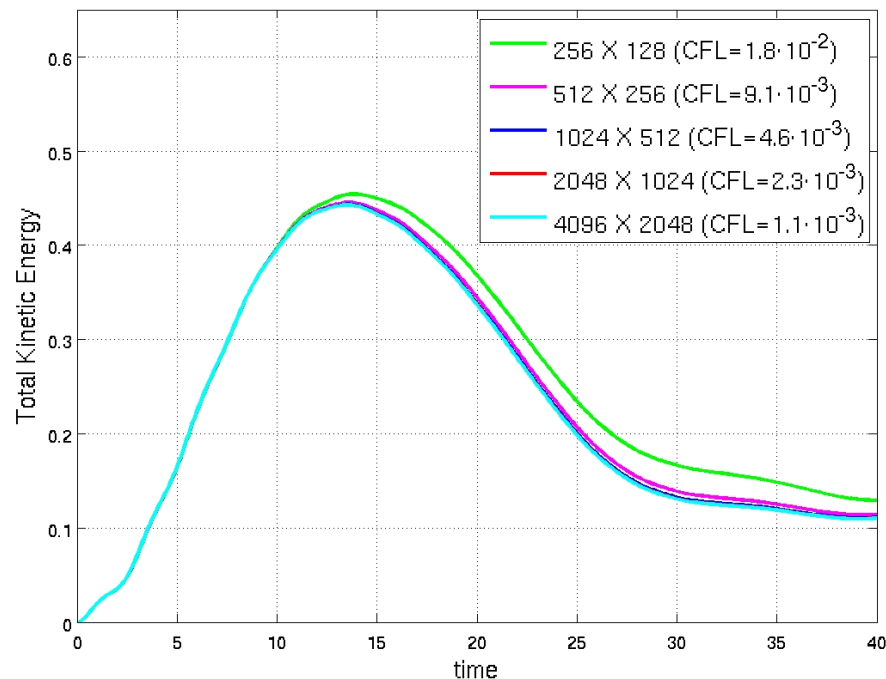
- Time = 40.0,  $\Delta t = 1$ .
  - ~100x CFL on 512 X 256 grid
- 2<sup>nd</sup> order spatial convergence
- Backward Euler in time
- Benchmarked w/ other codes
- Convergence studies (8B eqs)



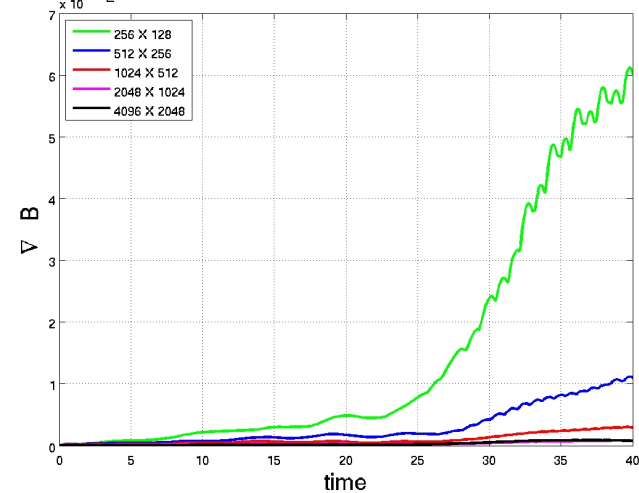
## $B_z = 0$ , Low viscosity, $\nabla \cdot \mathbf{B} = 0$

- Time = 40.0,  $\Delta t = .1$
- 2<sup>nd</sup> order spatial convergence
- $\nabla \cdot \mathbf{B} = 0$  converges
- Kin. E compares well w/ other codes

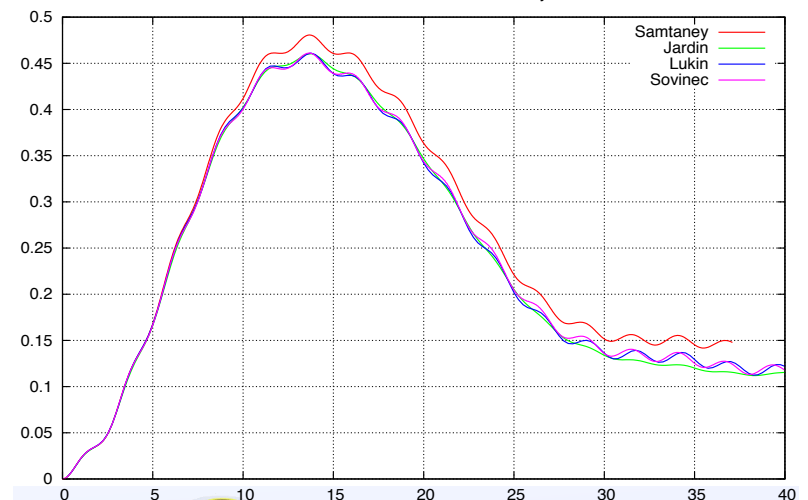
GEM  $B_z = 0.0$ , low viscosity, convergence (space),  $\Delta t = .1$ , 1 F-cycle w/  $2 \times V(1,1)$



GEM  $B_z = 0.0$ , low viscosity,  $\nabla \cdot \mathbf{B}$ ,  $\Delta t = .1$ , 1 F-cycle w/  $2 \times V(1,1)$



GEM Reconnection Test : Low Viscosity Case



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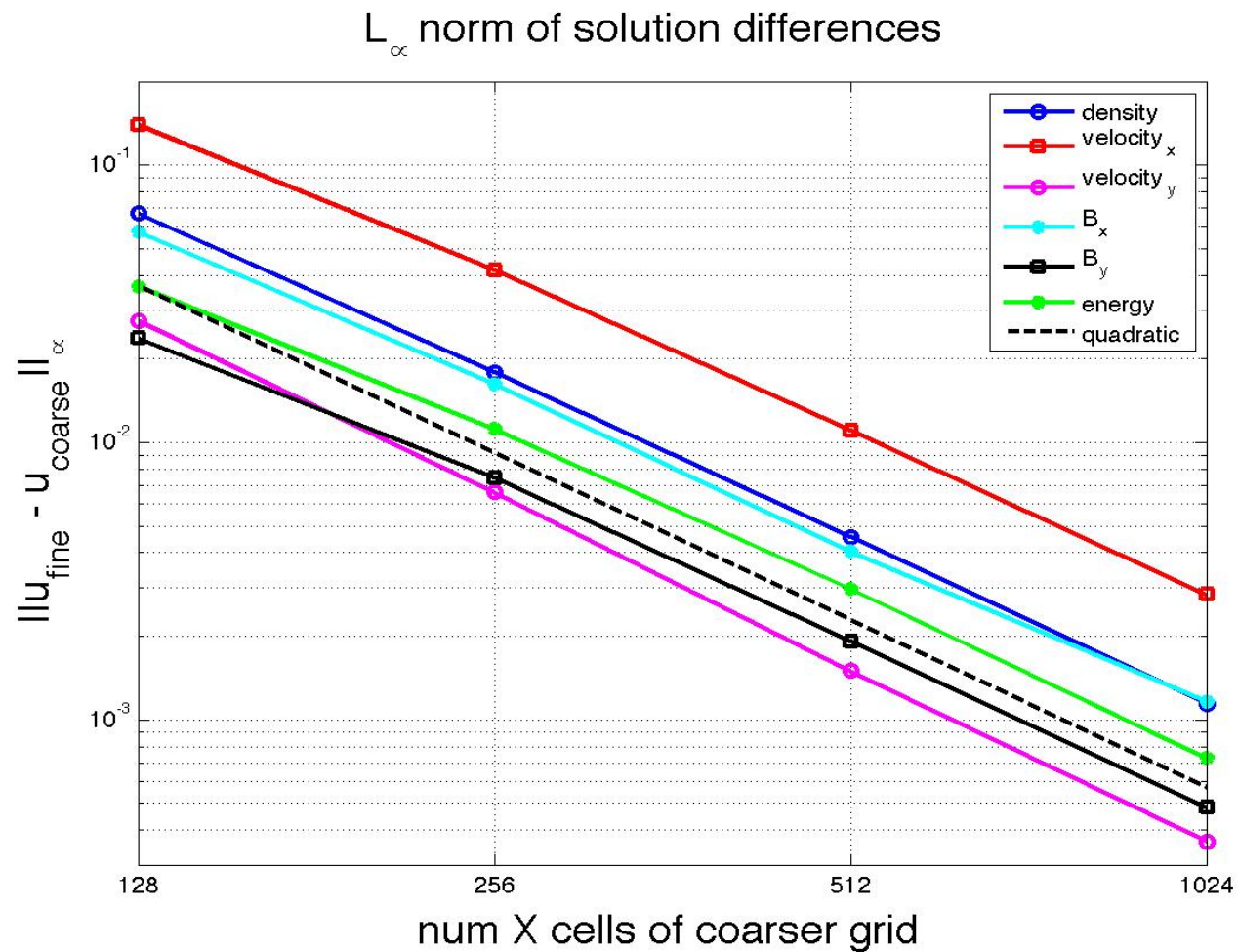


SciDAC

Scientific Discovery through Advanced Computing

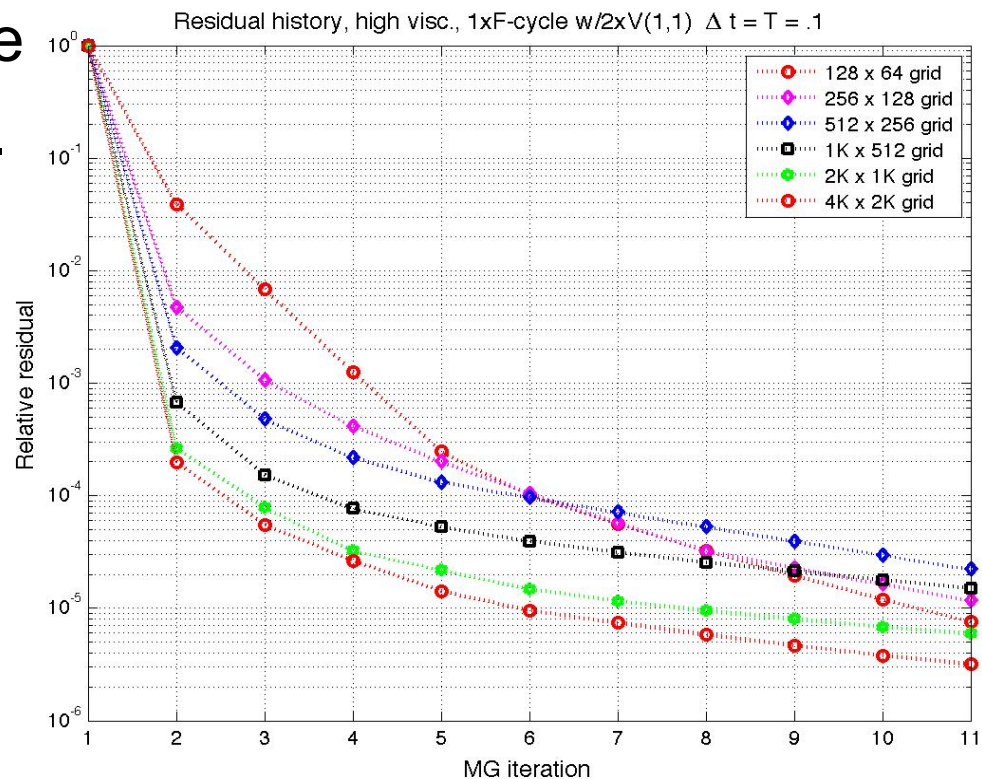
# Solution Convergence

$\mu=1.0\text{D-}03$ ,  $\eta=1.0\text{D-}03$ ,  $B_z=0$



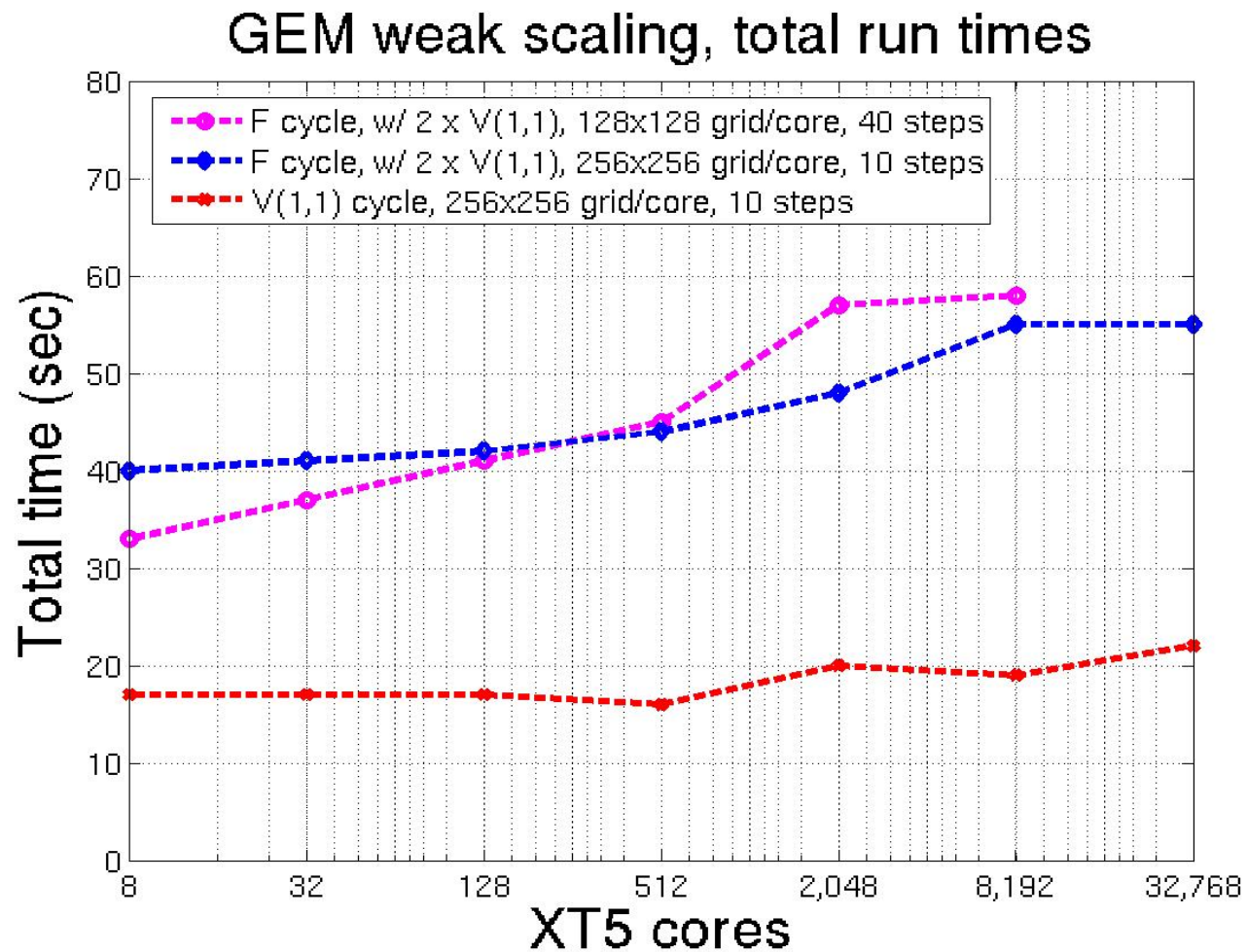
## Residual history

- Residual history (1<sup>st</sup> time step), high viscosity,  $B = 0$
- F cycles achieve discretization error
  - Super convergence
- No  $\Gamma$  w/defect correct.
- Use  $\Gamma$  for  $L_1$





## Weak scaling – Cray XT-5





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# What do we need to make multigrid fast & scalable at exa-scale?

- Architectural assumptions:
  - Distributed memory message passing is here for a while
  - Future growth will be primarily on the “node”
  - Memory bandwidth to chip can not keep up with processing speed
    - Need higher computational intensity - “flops are free” ...
- Multigrid issues:
  - Distributed memory network (latency) is still critical (if not hip)
    - Growth is on the node but distributed memory dictates data structures, etc.
      - Node optimizations can be made obsolete after distributed data structures added
    - Applications must **use good distributed data models and algorithms**
    - **Coarse grids must be partitioned carefully - especially with F-cycles**
      - **Coarse grids put most pressure on network**
    - Communication avoiding algorithms are useful here
      - But tedious to implement – *need support compilers, source-to-source, DSLs, etc.*
  - Computational intensity is low - increase with loop fusion (or streaming HW?)
    - Textbook V(1,1) multigrid does as few as 3 work units per solve
      - Plus a restriction and interpolation.
    - Can *fuse* one set of 2 (+restrict.) & one set of 1 (+interp.) of these loops
    - Communication avoiding can be added ... **data centric multigrid**



# Outline

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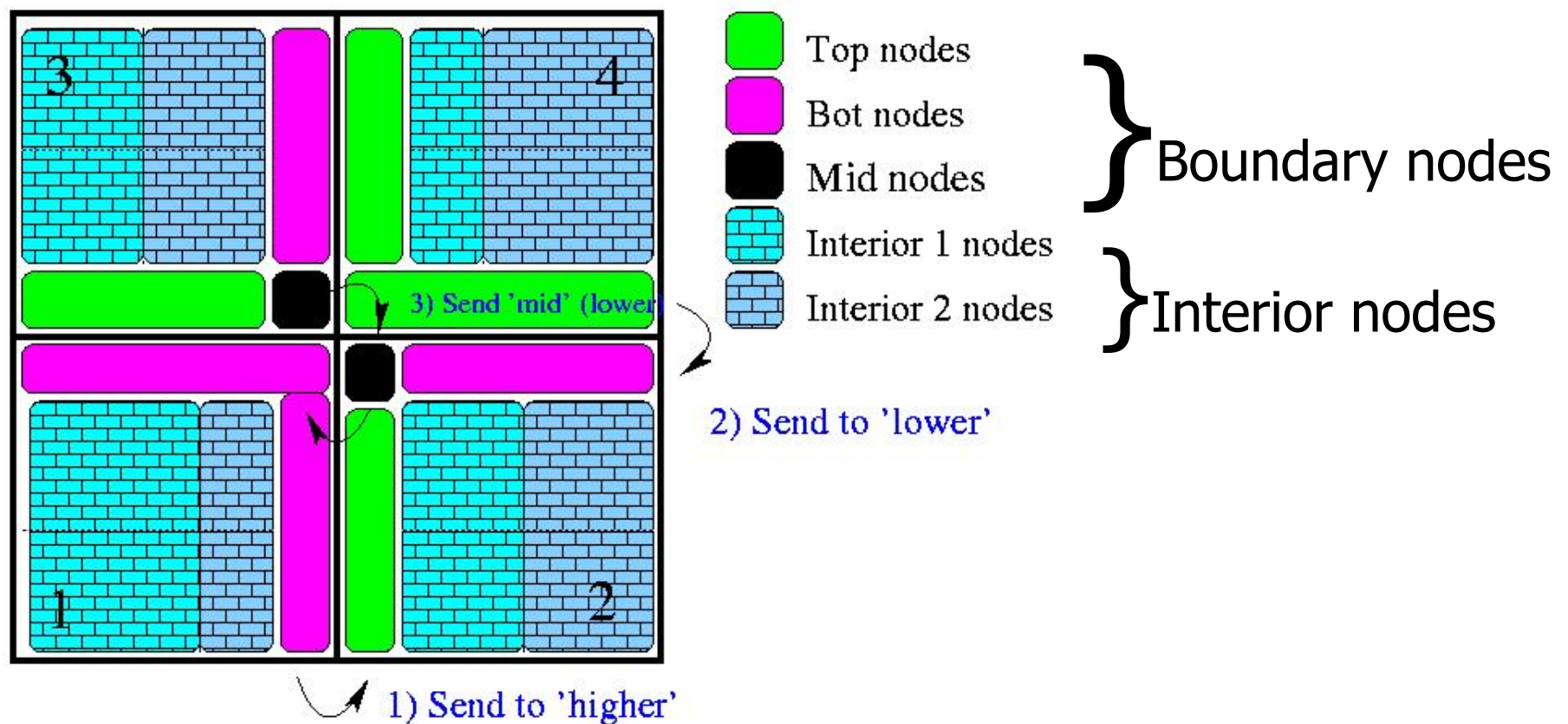
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## Case study: Parallel Gauss-Seidel Algorithm

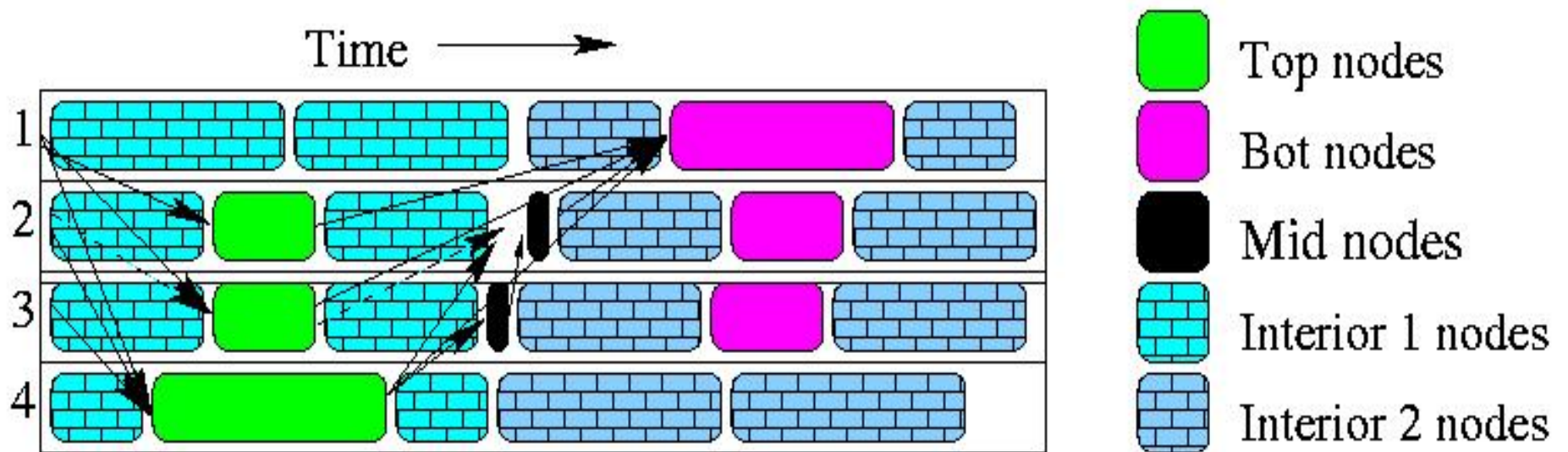
- Standard CS algorithm (bulk synchronous) graph coloring:
  - Color graph and for each color:
    - Gauss-Seidel process vertices
    - communicate ghost values (**soft synchronization**)
- 3, 5, 7 point stencil (1D, 2D, 3D) just two colors (not bad)
- 3D hexahedra mesh: 13+ colors (**lots of synchronization**)
  - General coloring also has pathological cache behavior
- Exploit domain decomposition + nearest neighbor graph property (data locality) + static partitioning
- Instead of computational depth 13+
  - have computational depth about 4+ (3D)
    - The number of processors that a vertex talks to
      - Corners of tiling
- Completely asynchronous algorithm

# Locally Partition (classify) Nodes



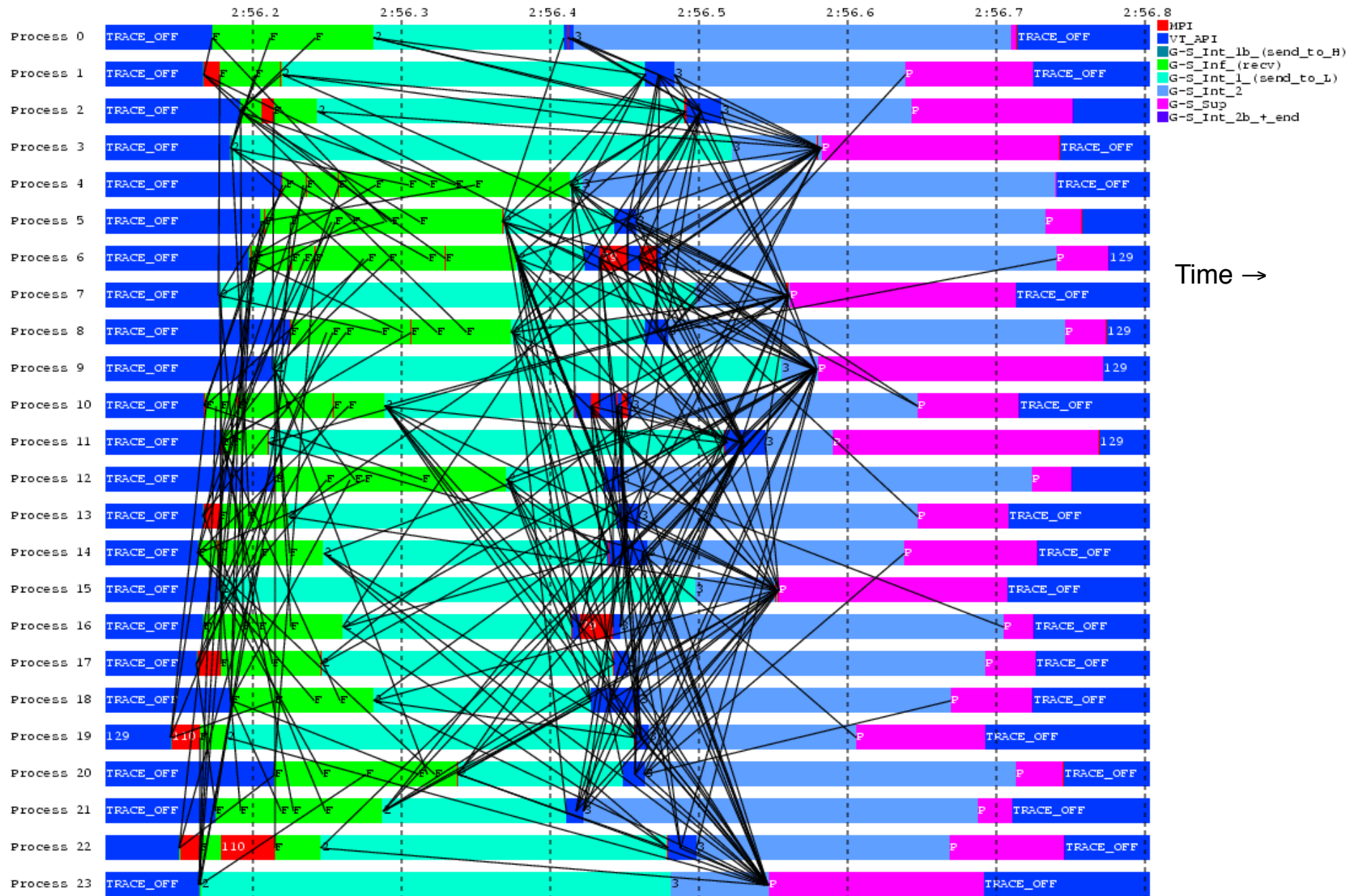
# Schematic Time Line

Note: reversible

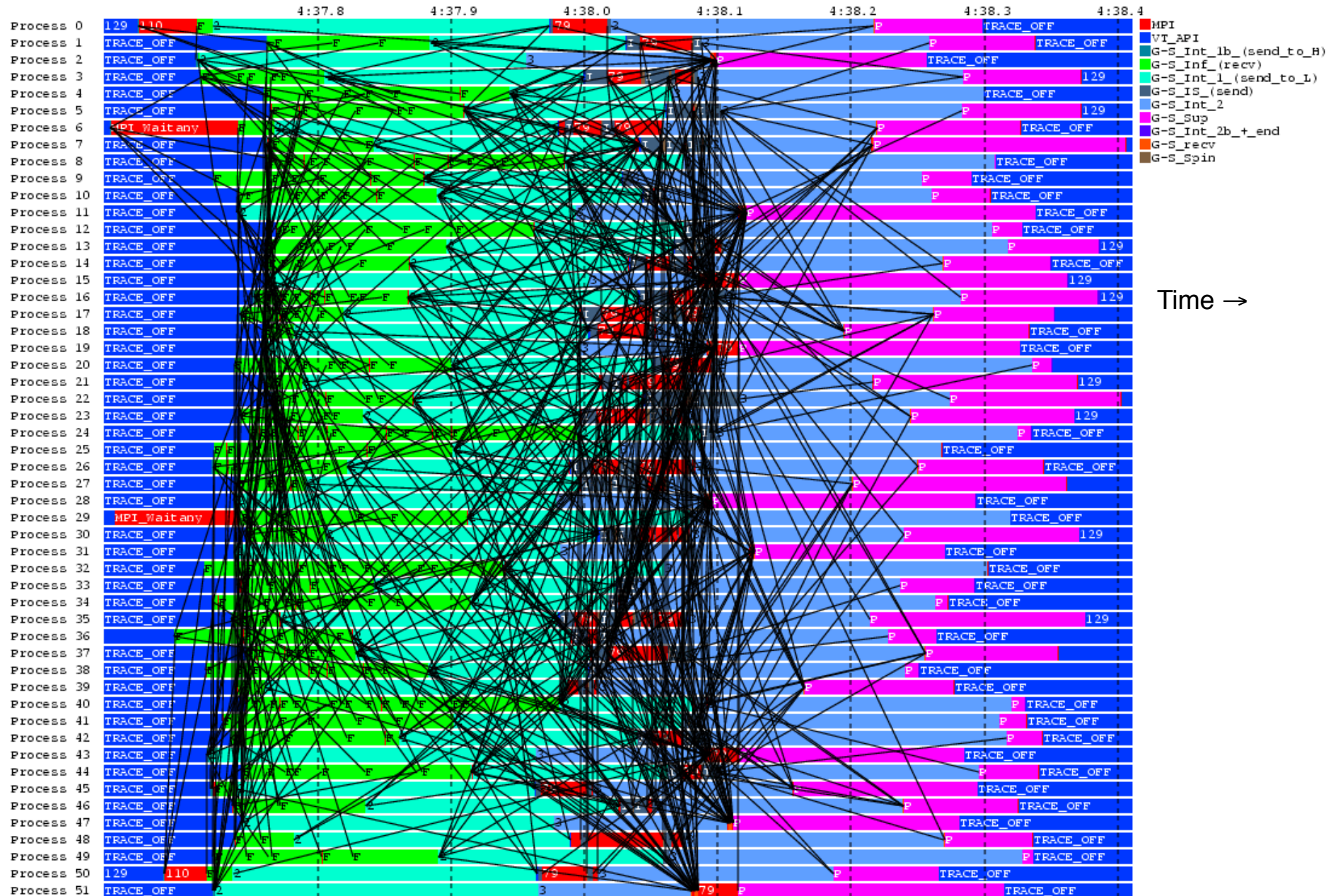




## Cray T3E - 24 Processors – About 30,000 dof Per Processor



**Cray T3E - 52 processors – about 10,000 nodes per processor**





## Lesson to be learned form parallel G-S

- Exploit finite sized domains
  - Domains of order stencil width
- Exploit static partitioning to coordinate parallel processing
- Technique applicable to any level of memory hierarchy
- Overlap communication and computation
- Exploit “surface to volume” character of PDE graphs



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## Implementations

- These ideas implemented in parallel FE framework Olympus & AMG solver Prometheus
  - Gordon Bell prize 2004.
- And in new unstructured geometric MG & smoothed aggregation AMG implementation in PETSc (PC GAMG):
  - -pc\_type gamg –pc\_gamg\_type sa
  - Rely on common parallel primitives to
    - Reduce code size
    - Amortize cost of optimization & of porting to new architectures/PMs
  - PETSc has rich set of common parallel primitives:
    - GAMG ~2,000 lines of code
    - Prometheus ~25,000 lines of code
      - About 20K of this implements GAMG functionality

## New aggregation algorithm for SA

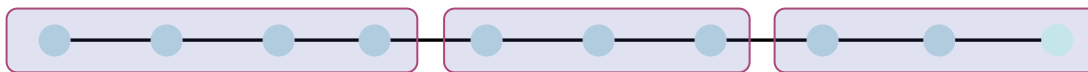
- My old aggregation algorithm is complex, don't want to reimplement, want to use standard PETSc primitives if possible
- Independent sets are useful in coarsening
  - Independent set: set of vertices w/o edges between each other
  - Maximal: can not add a vertex and still be independent
- MIS(k) (MIS on  $A^k$ ) algorithm is well defined & good parallel algorithms
  - “Greedy” MIS algorithms naturally create aggregates
- Rate of coarsening critical for complexity
  - Slow coarsening helps convergence at expense of coarse grid complexity
  - Optimal rate of coarsening for SA for 2<sup>nd</sup> order FEM is 3x
    - Recovers geometric MG in regular grid
    - Results in no stencil growth on regular grids
- MIS(2) provides a decent coarsening rate for unstructured grids
- MIS/greedy aggregation can lead to non-uniform aggregate sizes
- New “aggregation smoothing” with precise parallel semantics and use of MIS primitives.

## New aggregation algorithm for SA

- Drop small edges from graph  $G$  induced by matrix
  - $G = D^{-1/2}(AA^T)D^{-1/2}$
  - If  $G_{ij} < \theta$ , then drop from Graph (eg,  $\theta = 0.05$ )
- Use MIS(2) on  $G$  to get initial aggregates



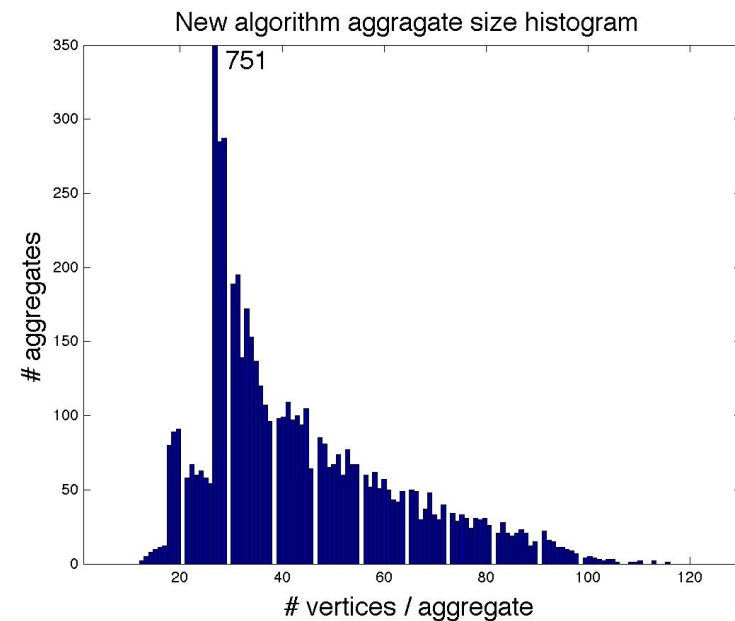
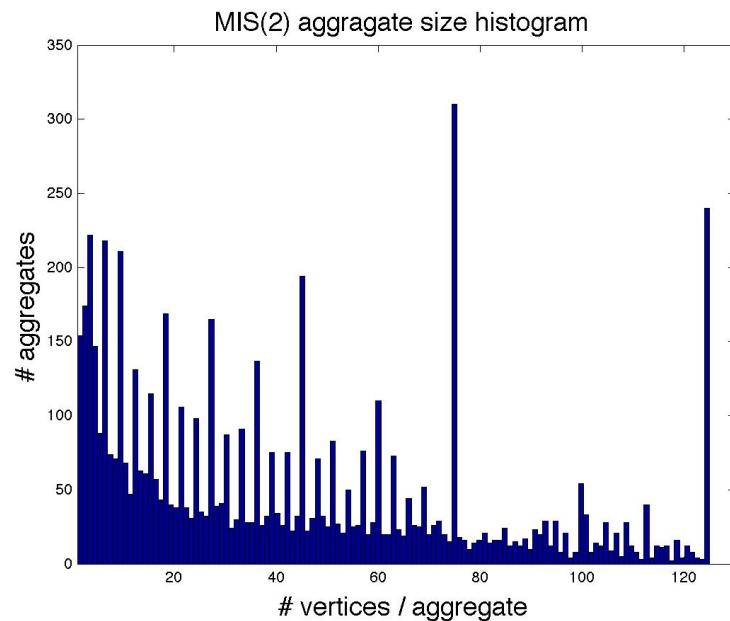
- Greedy (MIS(1) like algorithm) modified aggregates



## Results of new algorithm Histogram of aggregate sizes

**$64^3$  Mesh (262144 nodes)**

**First order hex mesh of cube**



## Weak Scaling of SA on 3D elasticity

### Cray XE-6 (Hopper)

- Weak scaling of cube
  - 81,000 eqs / core
- 8 node “brick” elements
- F-cycles
- Smoothed aggregation
- 1 Chebyshev pre & post smoothing
- Dirichlet on one face only
- Uniform body force parallel to Dirichlet plane

### Performance

Cores	27	216	1,728	13,824
N ( $\times 10^6$ )	2.2	17.5	140	1,120
Solve Time	4.1	4.9	5.6	7.0
Setup (1)	5.2	6.1	13	28
S (2) partit.	9.2	11	21	155
Iterations	11	12	12	14
Mflops/s/ core	334	314	276	257

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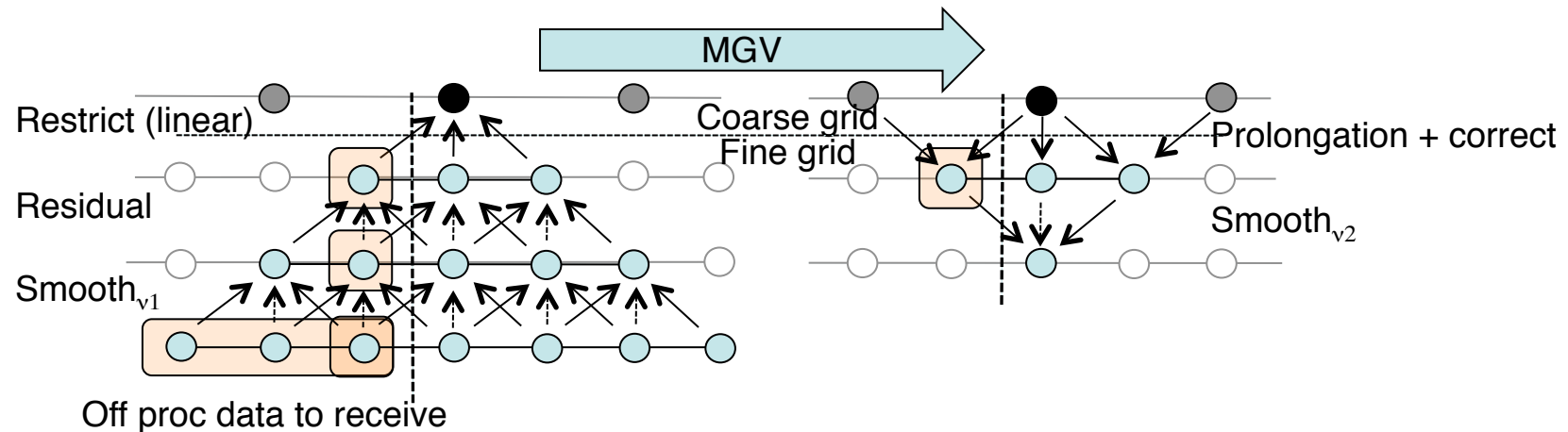
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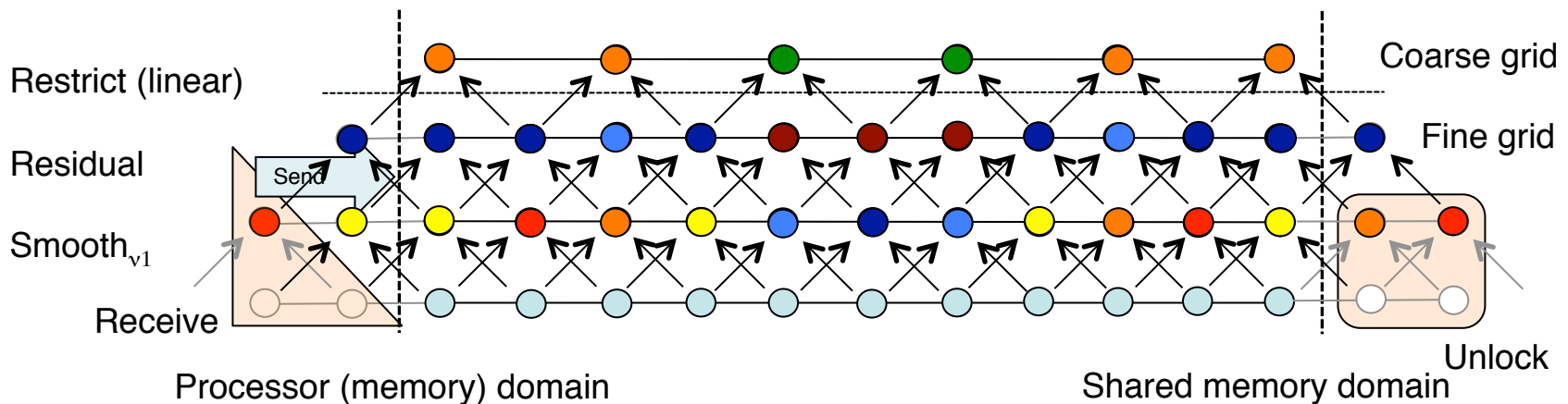
## Data Centric Multigrid - V(1,1)

- MG algorithm: Sequential with parallel primitives
  - Common way to think and code.
- Problem: poor data reuse, low comp. intensity, much data movement
- A Solution: loop fusion (eg, C. Douglas et. al.)
  - “Vertical” partitioning of processing instead of (pure) “horizontal”
    - Vertex based method with linear restriction & prolongation
    - Fuse: one loop; course grid correction; 2<sup>nd</sup> loop
    - **Data dependencies** of two level MG, 1D, 3-point stencil:



# Hierarchical memory (cache & network) optimization - fusion

- Approach to fusing 1<sup>st</sup> leg of V-cycle, 1D, 3-point stencil
  - One smoothing step with simple preconditioner (ie, no new data dependencies)
  - Residual
  - Restriction
- Overlap communication and computation & aggregate messages *w/ multiple states*
  - Communication avoiding
- Multiple vectors (lhs, rhs, res, work) and vector ops (AXPY, etc.) not shown
- Arrows show data dependencies (vertical, self, arrows omitted)
- Processor domain boundary (left) w/ explicit message passing
- Shared memory domain (right) “unlocks” memory when available
- Boundary processing could be asynchronous
- Multiple copies of some data required (not shown) at boundaries and ghost regions



## Multigrid $V(v_1, v_2)$ with fusion

- function  $u = \text{MGV}(A, f)$

- If  $A$  coarsest grid

- $u \leftarrow A^{-1}f$

- else

- $u \leftarrow S^{v_1}(f, u)$

-- Smoother (pre)

- $r \leftarrow f - Au$

C. Douglas et.al.

- $r_H \leftarrow Rr$

Chombo

- $e_H \leftarrow \text{MGV}(RAP, r_H)$

-- recursion (Galerkin)

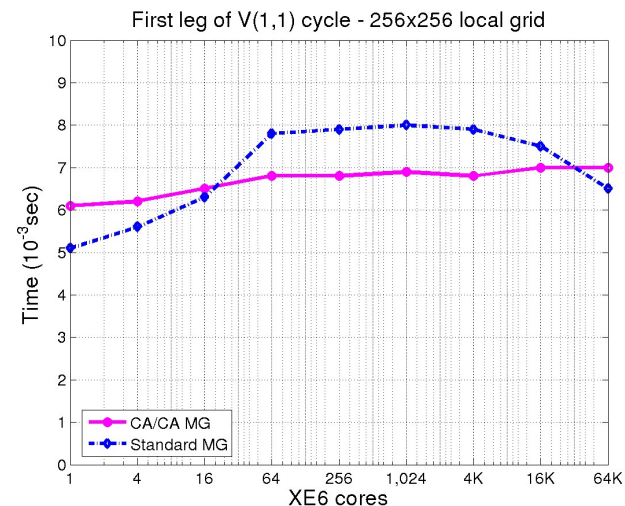
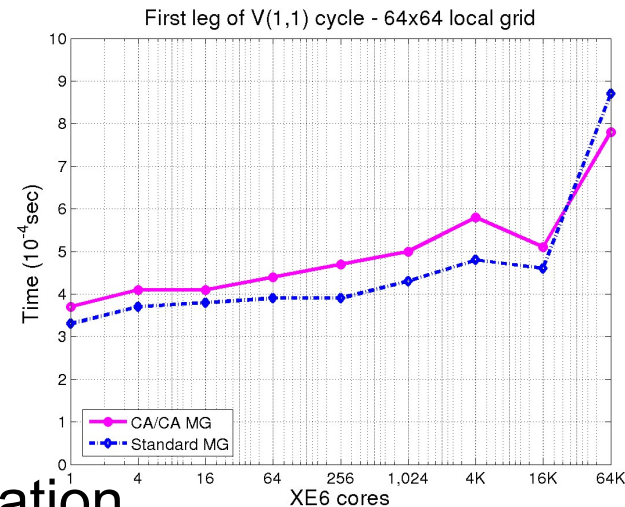
- $u \leftarrow u + Pe_H$

- $u \leftarrow S^{v_2}(f, u)$

-- Smoother (post)

## Numerical tests

- Reference Implementation of first leg of V(1,1) cycle
  - 2D 5-point FV stencil
  - Linear interp./prol.
  - ~800 lines of FORTRAN
  - Horrible to code!
- Compare with standard implementation
  - Non-blocking send/recv
  - Overlap comm. & comp.
  - ~400 lines of FORTRAN
- Cray XE-6 at NERSC
  - Four levels of MG
  - 256 x 256 and 64 x 64 fine grid
- I am not a good compiler!



## Conclusion

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- Equations solvers are too big to fail
- Multigrid is a shovel ready algorithm
- Good distributed memory implementations are hard and getting harder with deep memory architectures
- Many-core node, data centric algorithms (loop fusion, GPUs,...) are not well suited to FORTRAN/C
- Need compiler/tools/language support
  - of some sort ...

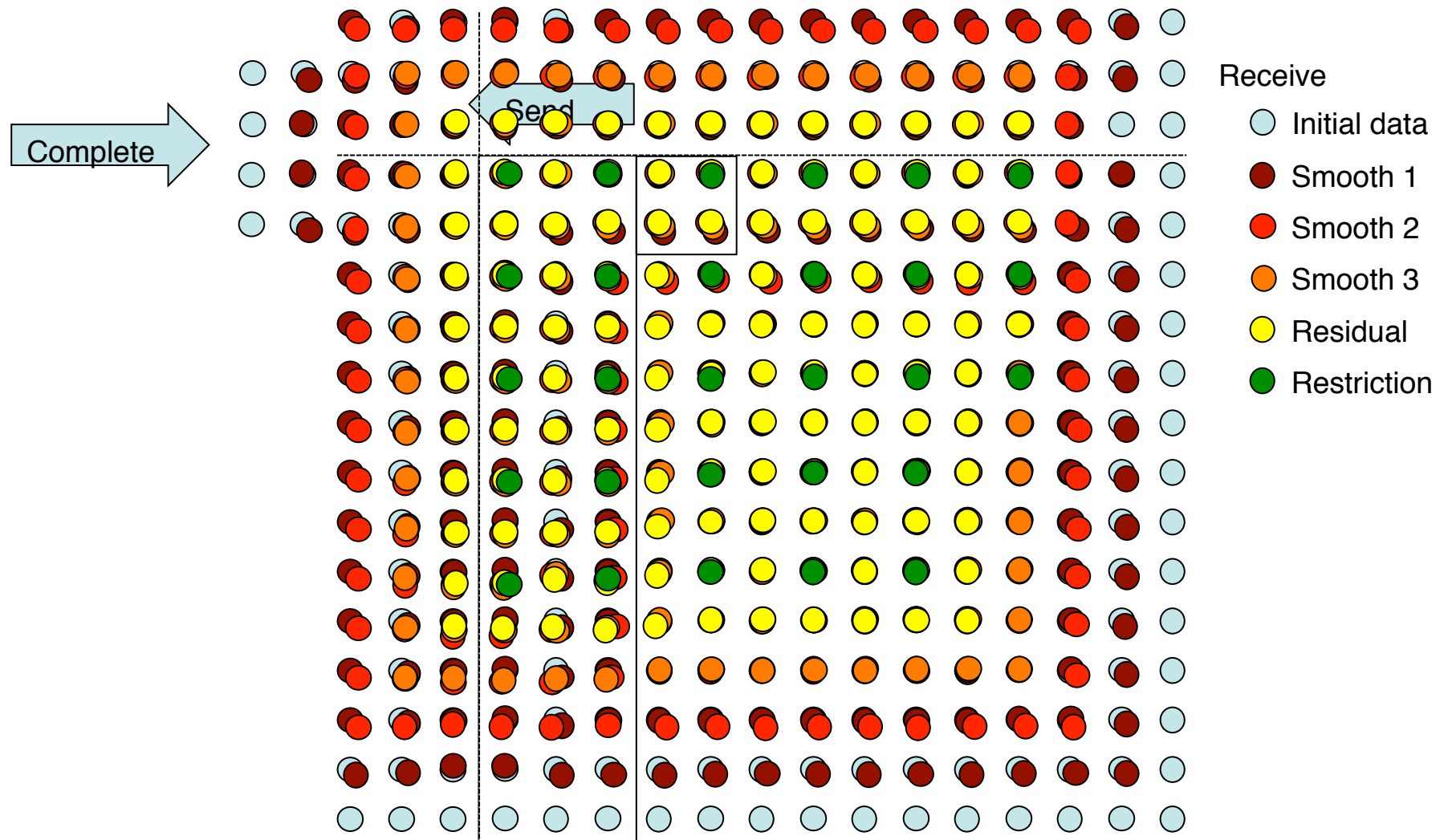




# Thank you

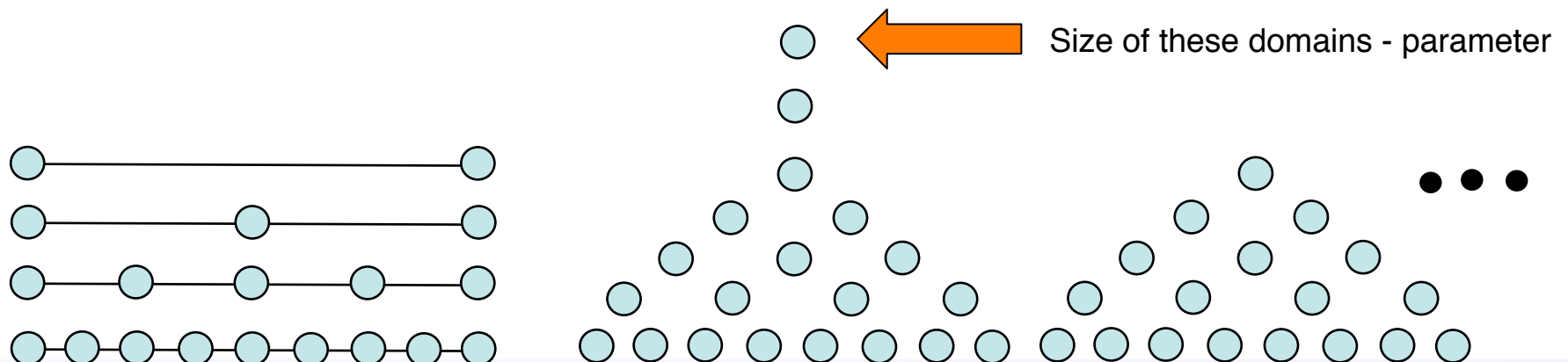


## 2D, 9-point stencil, 1<sup>st</sup> leg of V(3,3) w/ bilinear restriction



## A word about parallel complexity

- Solver work complexity:
  - $M$  iterations \* flops/iteration
  - All components of MG can have  $O(N)$  work complexity
    - Optimal – it takes  $O(N)$  work to print the solution
  - 1D C-cycle work complexity:  $C*N*(1+1/2+1/4+1/8\dots) < 2*C*N = O(N)$
- Parallel complexity – work depth
  - V-cycle has  $O(\log(N))$  work depth
    - Optimal – Laplacian is fully coupled
      - ie, Green's function has global support
    - Same as a dot product
  - F-cycles:  $O(\log^2(N))$





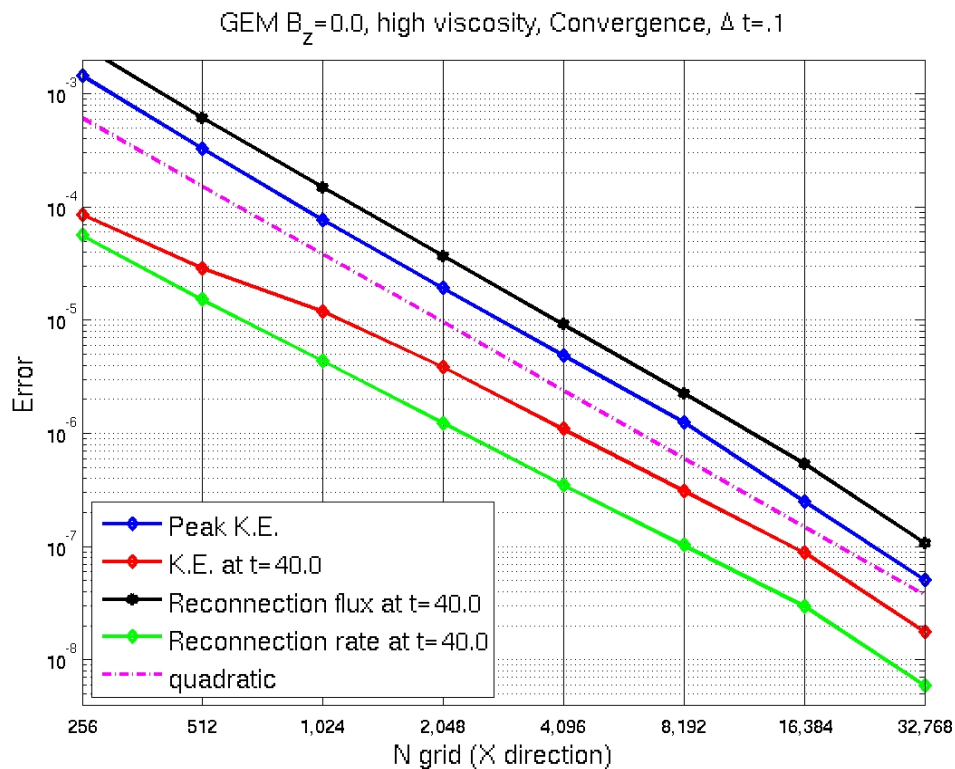
# Solver Algorithm issues past and future

- Present and future: memory movement limited
- 70's had similar problems as today, and what we see as the future
  - Then: couldn't afford memory – matrix free
  - Now: can't afford to architect it and use it
- 80's were pernicious:
  - Ubiquitous uniform access memory and big hair ...
  - Big memory did allow AMG and direct solvers to flourish
- Solutions that work on exa-scale machines ... look to the 70's
  - Low memory, matrix free, algorithms
  - Perhaps more regular grids as well
- Multigrid can solve with spatial/incremental truncation error accuracy
  - With work complexity of as low as ~6 residual calculations (work units)
  - On the model problem: low order discretization of Laplacian
    - Proven 30 years ago
  - “Textbook” multigrid efficiency
- No need to compute a residual (*no synchronous* norm computations)
- *No* need for CG's *synchronous* dot products
- MG is weakly synchronized but this comes from the elliptic operator complexity
  - no way around it
- MG has  $O(N)$  work complexity in serial,  $O(\log(N))$  work depth in parallel
  - F-cycles, required for truncation accurate solutions, is  $O(\log^2(N))$
- Work complexity looks less relevant now – “memory movement” complexity?

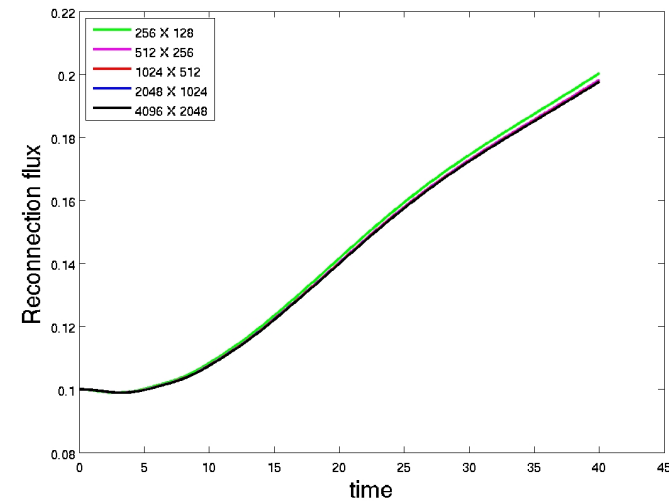


# Verify 2nd order convergence

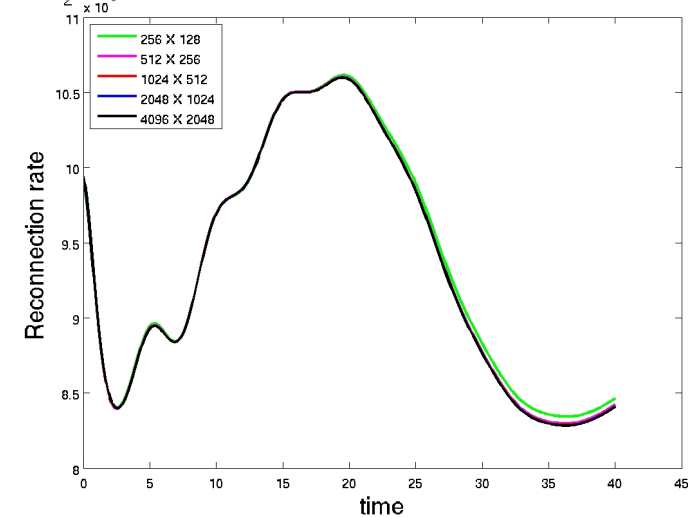
- 2nd order spatial accuracy
  - Achieved with F-cycle MG solve
- $B_z = 0$ , high viscosity
- Up to 1B cells (8B equations)



GEM  $B_z=0.0$ , high viscosity, Reconnection Flux,  $\Delta t=.1$ , 1 F-cycle w/  $V(1,$



GEM  $B_z=0.0$ , high viscosity, Reconnection Rate,  $\Delta t=.1$ , 1 F-cycle w/  $V(1,$



## Multigrid performance - smoothers

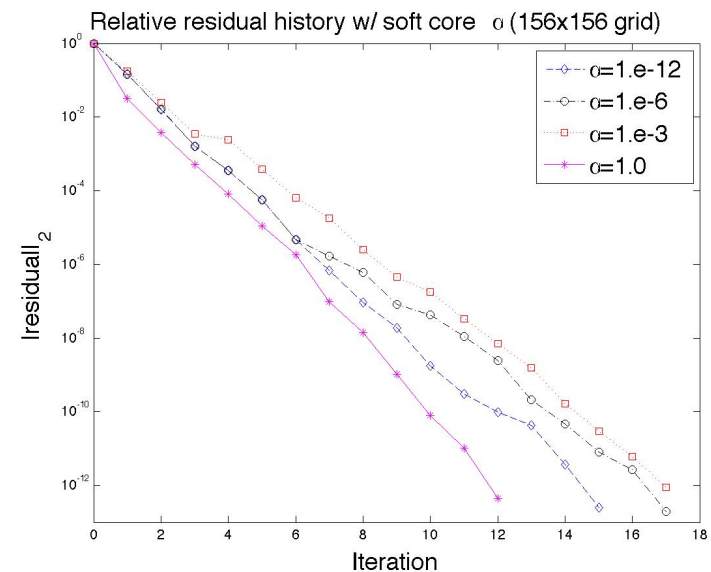
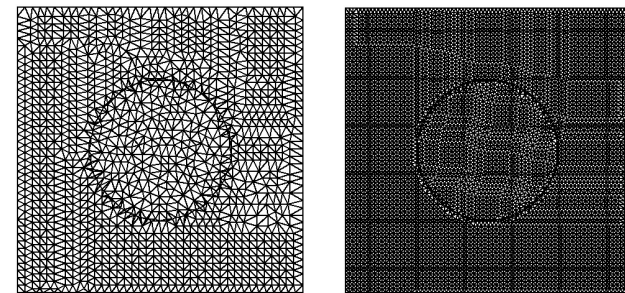
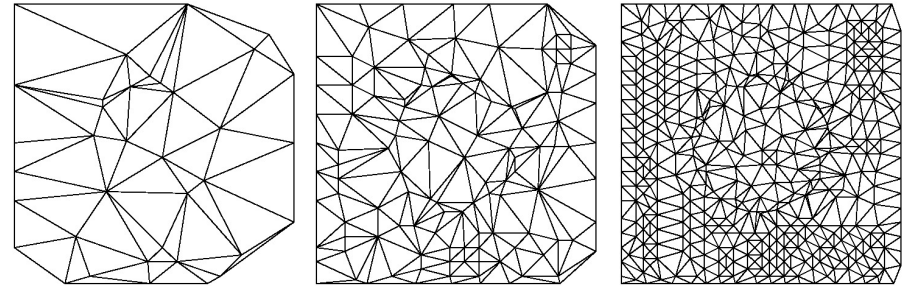
- Multigrid splits the problem into two parts
  - Coarse grid functions (MG method proper) - takes care of **scaling**
  - Smoother (+ exact coarse grid solver) - takes care of **physics**
- Smoothers, where most of flops are – important for performance opt.
- Additive MG smoother requires damping
  - $Be = (I - (B_1 + B_2 + \dots B_m)A)e$
  - Good damping parameter not always available
    - eg, non-symmetric problems
  - Krylov methods automatically damp
    - But not stationary & have hard synchronization points
- Multiplicative smoothers (eg, Gauss-Seidel)
  - $Be = (I - B_1A) (I - B_2A) \dots (I - B_mA)e$
  - Excellent MG smoother in theory
  - Distributed memory algorithm is a hard problem
    - Exploit nature of FE/FD/FV graphs ...

## Common parallel primitives for AMG

- Matrix matrix products:
  - $A_{i+1} = P^T A_i P$
  - $P = (I - \omega D^{-1}A)P_0$
- Computing (re)partitioning (ParMetis)
- Moving matrices (repartitioning)
- Maximal Independent Sets of  $A^k$  - MIS(k)
  - Useful mechanism for aggregation
  - Want coarsening factor of about 3
    - This is perfect on regular hexahedra mesh

# Unstructured geometric multigrid

- Select coarse points
  - MIS(1)
- Remesh (TRIANGLE)
- Use finite element shape functions for restriction/prolongation
- Example: 2D square scalar Laplacian with “soft” circle



# Coarse grid complexity at extreme scales

- Multigrid has theoretically optimal parallel complexity
  - “Data movement” complexity?
- $\log(N)$  computational depth - not enough parallelism available on coarse grids
- **Coarse grid complexity is main source of inefficiency at extreme scales**
- AMG issues: Support of coarse grid functions tend to grows
  - Independent sets are useful in coarsening
    - Independent set: set of vertices w/o edges between each other
    - Maximal: can not add a vertex and still be independent
  - The *maximum* independent set give  $3^3$  (27) aggs, every 3<sup>rd</sup> point on 3D cart. grid
    - This is perfect for SA - no support growth on coarse grids & recovers geo. MG
  - But support grows on unstructured problems, for example consider
    - stencil grows from 27 to 125 points (extra layer)
    - One vertex/proc – communicate with ~124 procs
    - $3^3$  vertex/proc – communicate with ~26 procs
- Thus, coarse grid memory complexity increases communication
- Amelioration strategy: use same basic idea as in parallel G-S:
  - Keep processor sub-domains from getting tiny (at least a few “stencils”)
  - Reduce active processors (eg, keep ~500 equations per processor)
    - This leads to need to repartition if original data was not recursively partitioned
    - No data locality with randomly aggregating sub-domains