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Parallel Preconditioners and Solvers for Modern Architectures

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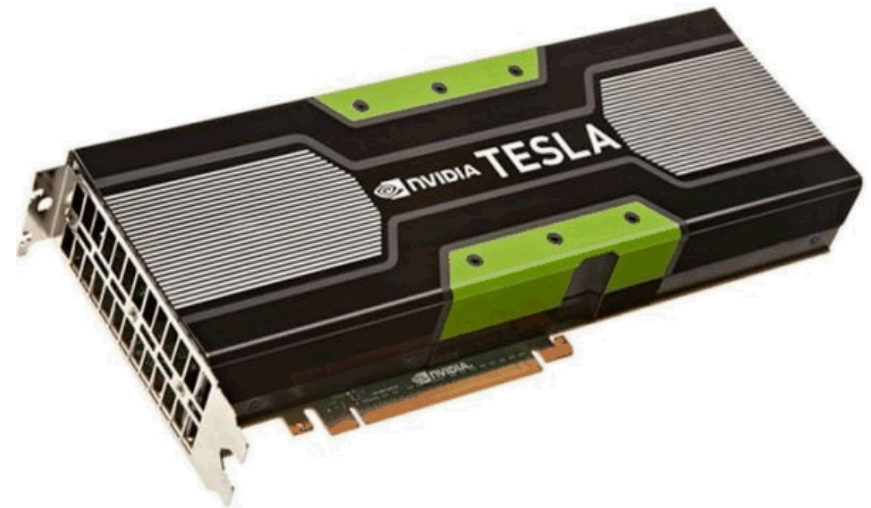
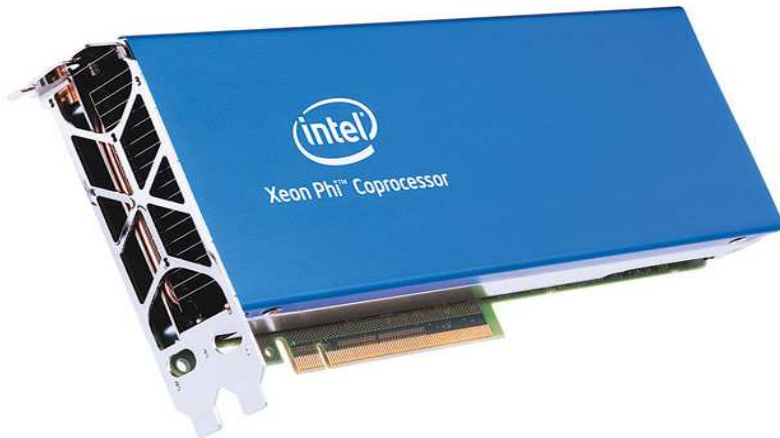
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Extreme-Scale Computing Challenges

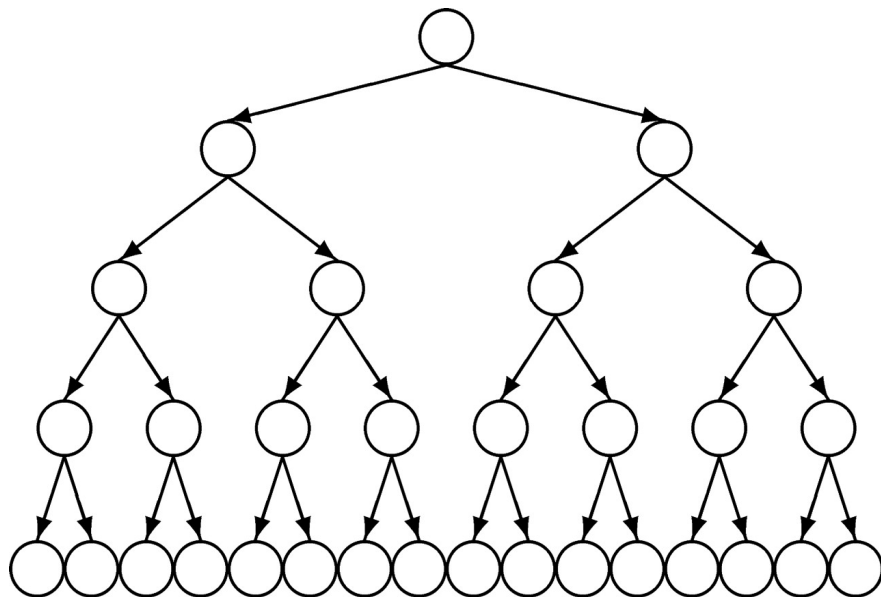
- Current parallel solvers on distributed memory systems are quite successful in many areas
 - Challenge: Reduce global communication (not in this talk)
- Architecture trend: Node count does not grow
- Main challenge is greater **concurrency on the node**
 - Accelerators: GPU, MIC
 - Concurrency on node is now $O(100-1000)$



Common Types of Parallelism

Coarse-grained:

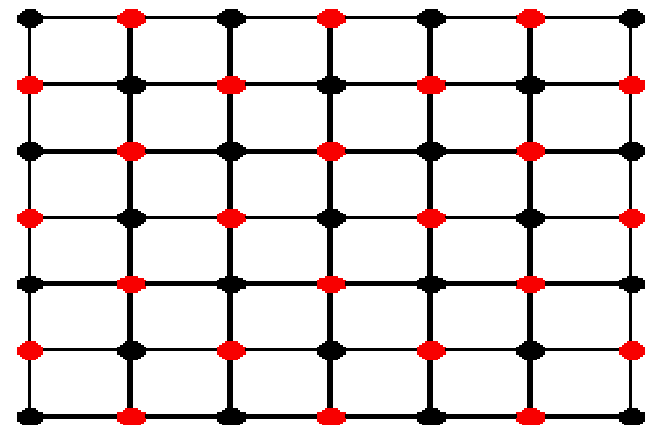
- Divide-and-conquer
- Nested Dissection



Fine-grained:

- Independent Sets
- Graph Coloring

Red-Black Ordering of Grid Points



Black points have only Red neighbors

Red points have only Black neighbors

Concurrency vs. convergence

Trade-off: Reordering for greater concurrency often gives slower convergence.

Example: Multicolor Gauss-Seidel.

- Update all vertices within each color class in parallel.
- #synchronization points = #colors
- Slow convergence, use as preconditioner in Krylov.

Matrix	Method	Natural	Multicolor
Wathen120	CG/SGS	20 iter.	23 iter.
Bcsstk18	CG/SGS	323 iter.	322 iter.
Torso2	GMRES/GS	22 iter.	19 iter.
Audikw1	CG/SGS	2512 iter.	2780 iter.

Multicolor Gauss-Seidel

CPU:

Compare serial vs. parallel preconditioner.

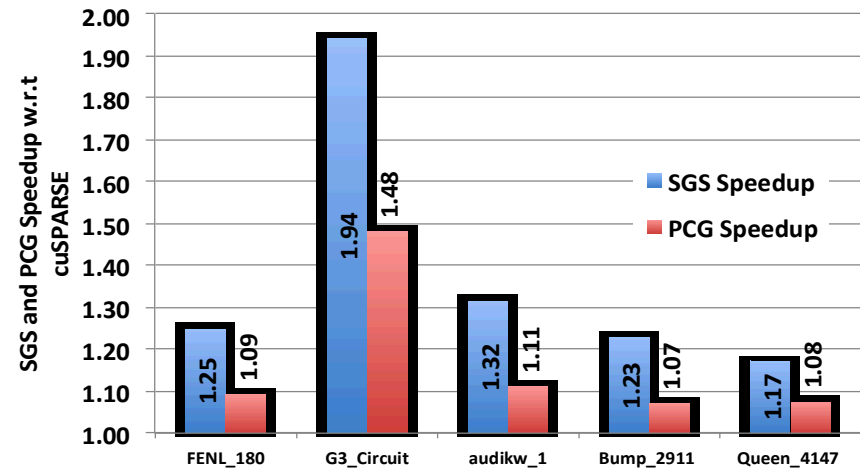
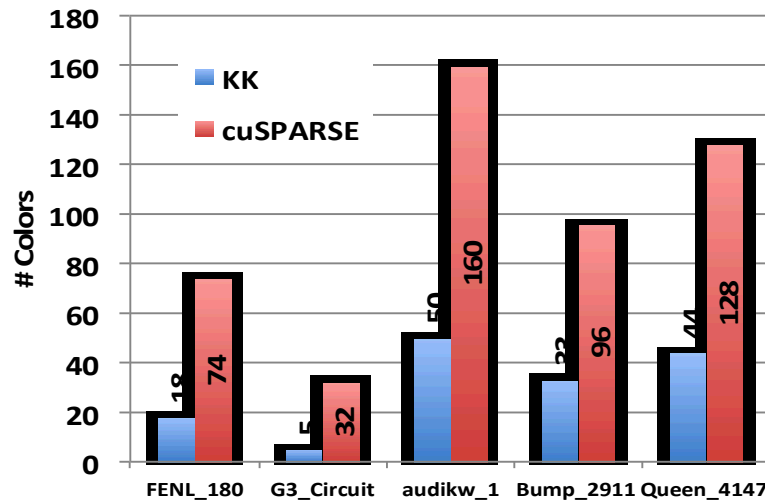
<i>SGS</i>	<i>1 thr.</i>	<i>4 thr.</i>	<i>16 thr.</i>
<i>Serial</i>	<i>2.84</i>	<i>1.31</i>	<i>1.07</i>
<i>Parallel</i>	<i>4.66</i>	<i>1.43</i>	<i>0.73</i>

Multithreading pays off after only ~4 threads.

Parallel Graph Coloring

- Optimal coloring is NP-hard
- Fast greedy coloring works well in practice
 - But is sequential
- Parallel greedy methods
 - Jones-Plassmann '90: independent sets
 - Gebremedhin-Manne '99: speculative/optimistic method
 - May have conflicts, need several rounds to resolve
- Deveci, B. Devine, Rajamanickam '16:
 - Edge-based speculative/optimistic method for manycore
 - Faster and better quality (4X geom. mean) than cuSparse

GPU: Coloring-based Symmetric Gauss-Seidel



- New coloring algorithm and Gauss-Seidel are now available in the KokkosKernels package
- Reduces the overall CG solve time by ~33% compared to cuSparse.
 - Shows that coloring quality matters!

Other Methods: Fine-Grain Parallel ILU

Chow & Patel 2015: New way to compute incomplete $A \sim LU$.

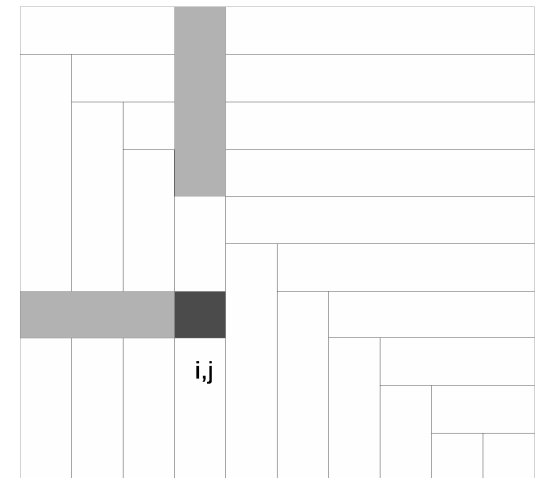
- Solve larger, bilinear system for each entry in L, U
- “Gauss-Seidel” on factors converges quickly (5-10 sweeps)
 - Asynchronous version works fine, no coloring needed.
 - Highly parallel: Can use one thread per nonzero.
- Triangular solves become the bottleneck, but can solve iteratively.
- Well suited for GPU: Anzt et al. (2015)
- Software implementations in Magma and Trilinos/ShyLU.

Algorithm 1: Fine-Grained Parallel Incomplete Factorization

```

Set unknowns  $l_{ij}$  and  $u_{ij}$  to initial values ;
for  $sweep = 1, 2, \dots$  until convergence do
  parallel for  $(i, j) \in S$  do
    if  $i > j$  then
       $l_{ij} = (a_{ij} - \sum_{k=1}^{j-1} l_{ik}u_{kj}) / u_{jj}$ 
    else
       $u_{ij} = a_{ij} - \sum_{k=1}^{i-1} l_{ik}u_{kj}$ 
    end
  end
end
end

```



Row Projections: Kaczmarz

Row projection methods are very old:

- Kaczmarz (1937)
- Cimmino (1938)

Rediscovered several times, e.g. as ART and CIRT.

Can be used as solver or preconditioner. Several advantages:

- Work on single row at a time, so fine grained.
- No global communication.

Kaczmarz: For each row i

$$x^k = x^k + \lambda \frac{b_i - \langle a_i, x^k \rangle}{|a_i|^2} a_i$$

Parallel Kaczmarz

Traditional method is sequential, but two options:

1. Multicolored Kaczmarz
 - Update independent rows in parallel.
2. Asynchronous Kaczmarz
 - Use most recent values, changes the algorithm
 - Essentially becomes hybrid Kaczmarz-Cimmino

Randomized Kaczmarz:

- Pick random row to update (no “sweeps”)
- Convergence is linear (“exponential”) [Strohmer & Vershynin]
- Gives hope asynchronous execution will behave similarly.

Software: Kokkos

Challenge: How to write portable multi-threaded code for many different platforms?

- CPU, Xeon Phi (MIC), GPU, etc.

Kokkos is a C++ library for performance-portable manycore programming.

- Supports parallel patterns (for, reduce, etc.)
- Optimized layout of multidimensional arrays
- Hierarchical parallelism (leagues, teams, threads)
- Developer productivity: Write once, run anywhere!
- Talk by Christian Trott, Wed. AM

Solvers using Kokkos: Rajamanickam talk, Wed. PM

Future Work: Block Row Projections

Many row projection options:

- Kaczmarz, Cimmino, CAV, CARP, ...

Block Row Projection Methods are superior:

- Better convergence
- Better memory access?

Challenge: For each block, we need to solve a rectangular system

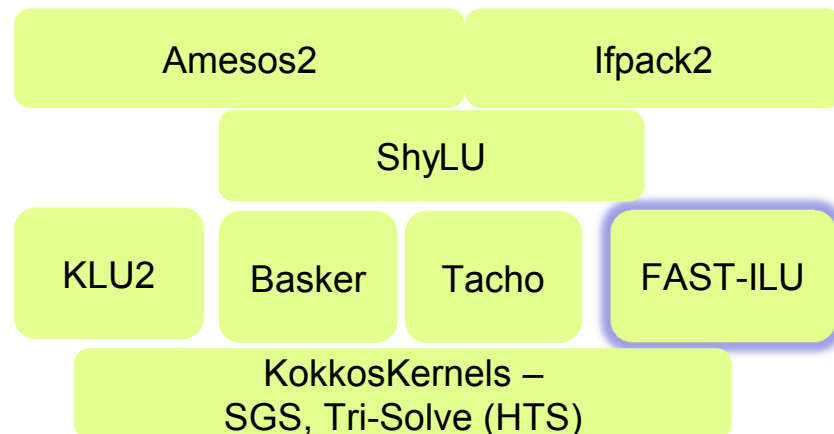
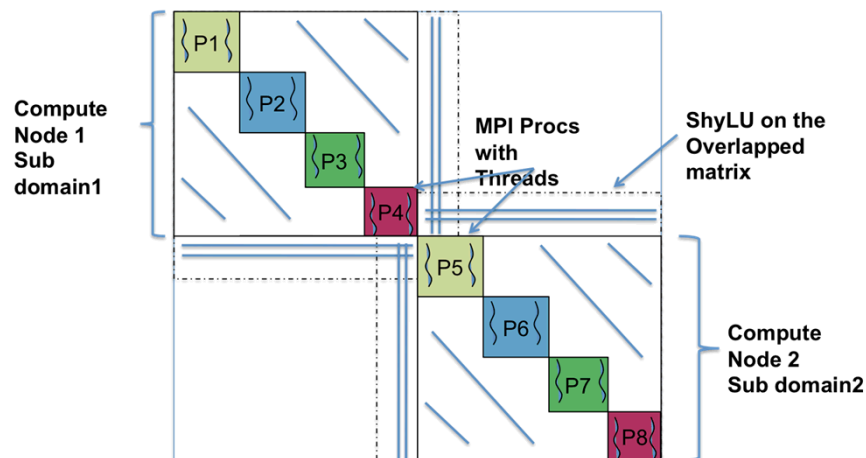
- Underdetermined system, want min-norm solution
- Options: sparse QR, normal equations, augmented system
- Suitable for 2-level parallelism
 - No sparse QR that works at the “thread team” level
 - ABCD (Zenadi et al.) uses MUMPS on augmented system

Conclusions

- Highly parallel architectures are disruptive
 - Require us to revisit algorithms.
 - Old ideas may come back to life?
- Coloring is powerful concept
 - Exposes fine-grain parallelism
 - Reordering slows convergence but still a win compared to sequential approach
- On-node performance critical for exascale.
 - Rewrite of solver software is needed.

Backup

ShyLU and Subdomain Solvers : Overview



- MPI+X based subdomain solvers
 - Decouple the notion of one MPI rank as one subdomain: Subdomains can span multiple MPI ranks each with its own subdomain solver using X or MPI+X
 - Epetra based solver, Tpetra interface still being developed
 - Trilinos Solver Factory a big step forward to get this done (*Thanks to M. Hoemmen*)
- **Subpackages of ShyLU:** Multiple Kokkos-based options for on-node parallelism
 - **Basker** : LU or ILU (t) factorization (J. Booth)
 - **Tacho**: Incomplete Cholesky - IC (k) (*See Kyungjoo's talk*)
 - **Fast-ILU**: Fast-ILU factorization for GPUs (A. Patel)
- **KokkosKernels**: Coloring based Gauss-Seidel (M. Deveci), Triangular Solves (*See Andrew's talk for HTS*)
- **Experimental code base under active development.** Jointly funded by ASC, ATDM, FASTMath, LDRD.