

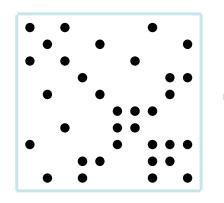
Empirical Complexity of Laplacian Linear Solvers: Discussion

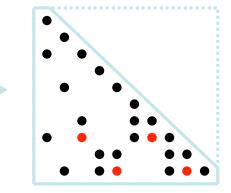
Erik Boman, Sandia National Labs Kevin Deweese, UC Santa Barbara John R. Gilbert, UC Santa Barbara

Simons Institute Workshop on Fast Algorithms via Spectral Methods December 2, 2014

Support: Intel, Microsoft, DOE Office of Science, NSF

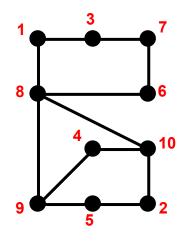
Graphs and sparse matrix computation (philosophical digression)



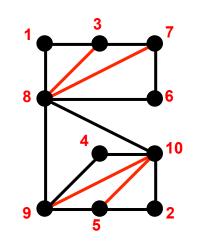


Symmetric Gaussian elimination:

for j = 1 to n add edges between j's higher-numbered neighbors



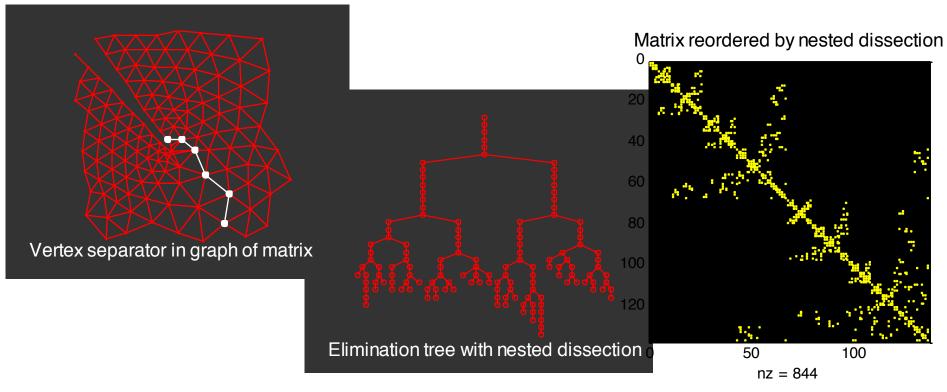
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Nested dissection and graph partitioning [George 1971, then many papers]



- Find a small vertex separator, number it last, recurse on subgraphs
- Approx optimal separators => approx optimal fill & flop count

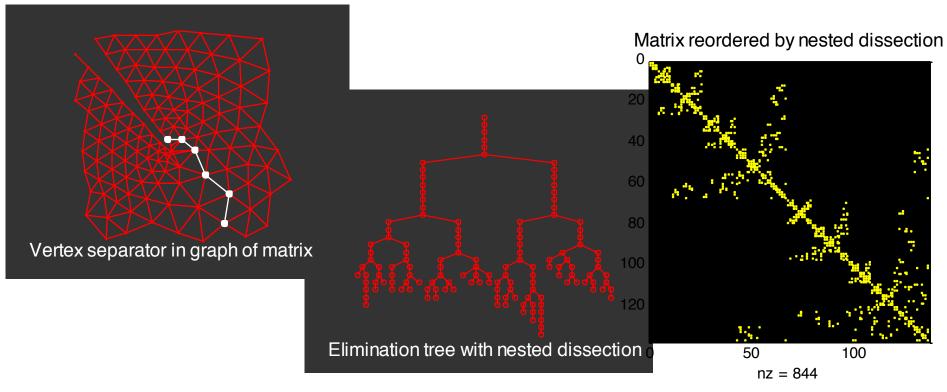


Many, many graph algorithms have been used, invented, and implemented at large scale for sparse matrix computation:

- Symmetric problems: elimination tree, nonzero structure prediction, sparse triangular solve, sparse matrix-matrix multiplication, min-height etree, ...
- Nonsymmetric problems: sparse topological solve, bipartite matching (weighted and unweighted), Dulmage-Mendelsohn decomposition / strong components, …
- Iterative methods: graph partitioning again, independent sets, low-stretch spanning trees, ...



Nested dissection and graph partitioning [George 1971, then many papers]



- Find a small vertex separator, number it last, recurse on subgraphs
- Approx optimal separators => approx optimal fill & flop count
- It took more than 20 years for nested dissection to become the method of choice for sparse GE *in practice*.

- Vaidya 1990: O(n^{1.75})
- Spielman/Teng 2004: O(n log^c n)
- Koutis/Miller/Peng 2010: O(n log n log log n)
- Kelner/Orecchia/Sidford/Zhu 2013: O(n log² n log log n)

(for sparse graphs, fixed ε)



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Okay, it's been more than 20 years now ...



The Laplacian World Championships

- Why should you participate?
- What would make you more likely to?
- How can we incent team leaders, members, HPC experts, ... ?
- What are the right metrics?
 Single-core time; parallel time; anything not based on timings?
- Can "you implement, we measure" work for anything besides just single-core time?
- Thoughts on test graphs? Collections, generators, graphs supplied by contestants?

- Should code from entries be available for wider use?
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- Other comments? Questions?
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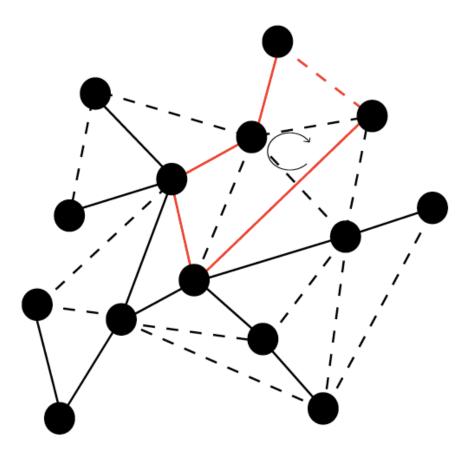
Comparing empirical complexity to theory

- Just counting things, not measuring time
 - a stepping stone between O() and running time
- Here: K/O/S/Z SimpleSolver (henceforth "RK")
 - log n off of their FullSolver
- A whole bunch of sample matrices
 - UF collection, DIMACS, BTER, etc.

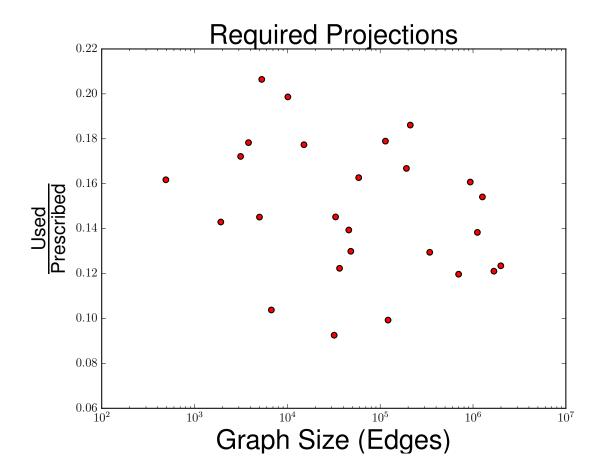


Kelner/Orecchia/Sidford/Zhu algorithm

- Select cycle (with probability proportional to stretch) from a fundamental cycle basis.
- Update flows around cycle.

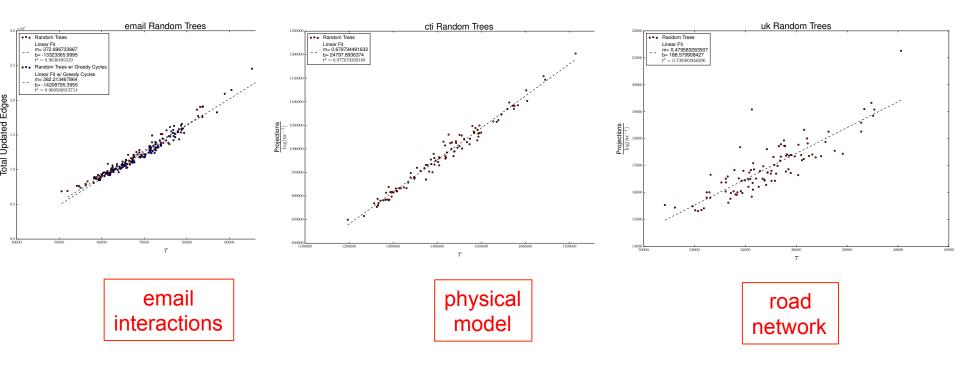


Number of projections to convergence



More optimistic than the bound by a factor of 5 or 10.

Scaling of work with tree stretch

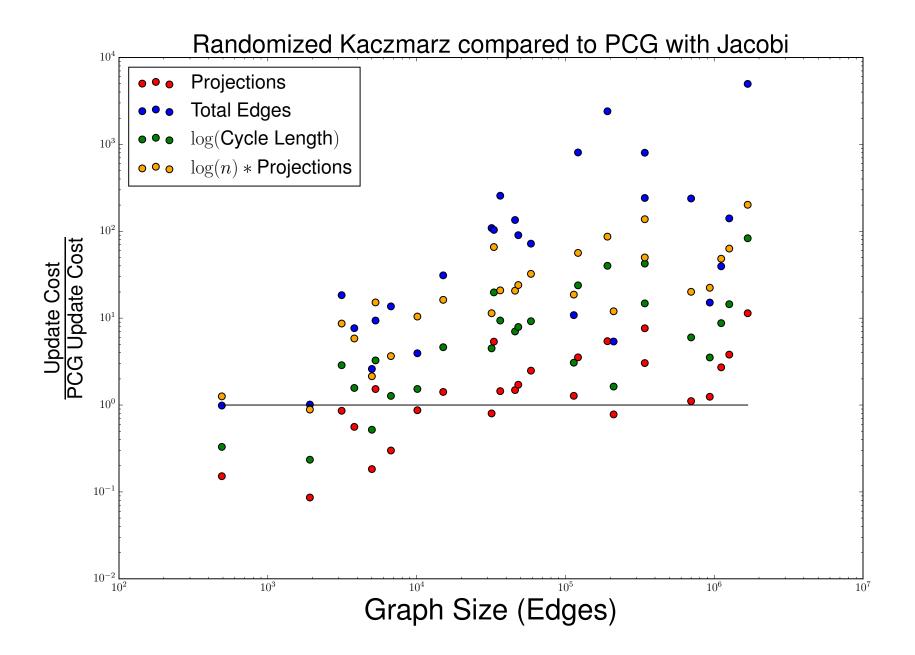


• Good correlation, better for some types of graphs than others.

Empirical complexity of RK compared to PCG

- PCG = conjugate gradient preconditioned by Jacobi
- PCG does about iters * (m+n) "edge touches"
- Count the cost of an RK projection in four different ways:
 - cycle length (naive)
 - Ig n (fast fundamental cycle data structure)
 - Ig (cycle length) (unwarranted optimism)
 - 1 (surely a lower bound)





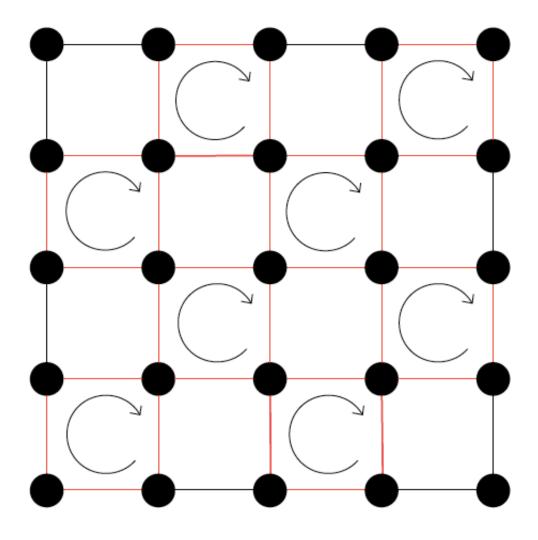
How about expanding the available projections?

- Add some cycles to basis
- Potentially gain parallelism in updates
- Maybe add flexibility with non-fundamental cycles
- More of the projection cost metrics are fantasy now, since we don't have fast update algorithms in general.



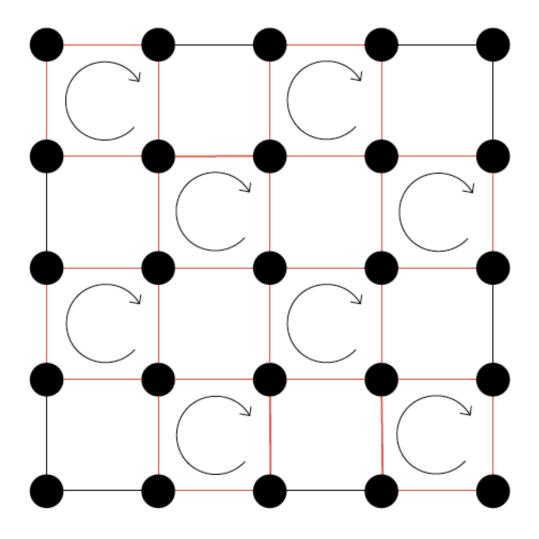
Example: non-fundamental cycles on a square grid

 Edge-disjoint cycles can be updated in parallel.

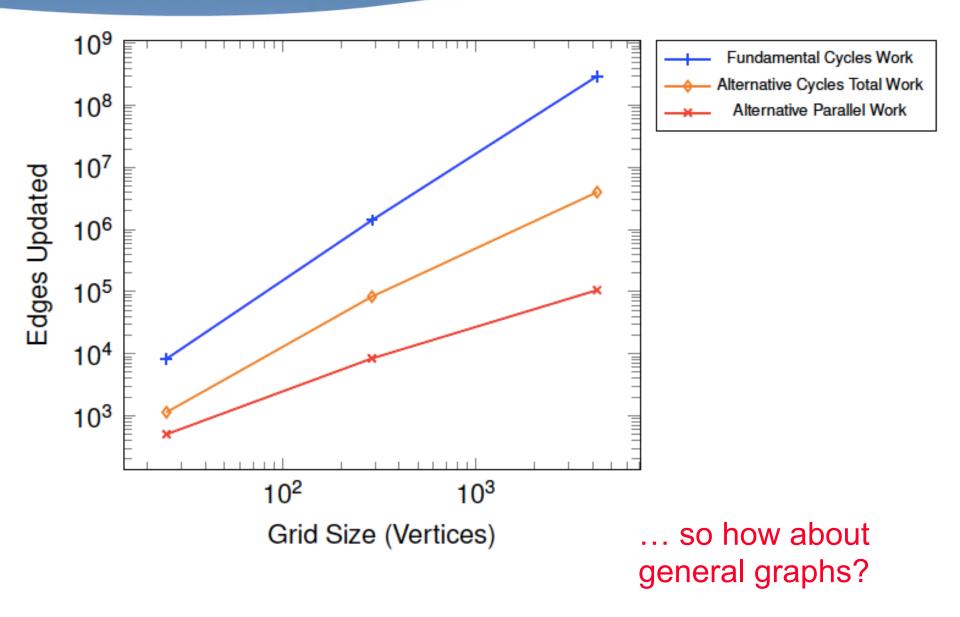


Example: non-fundamental cycles on a square grid

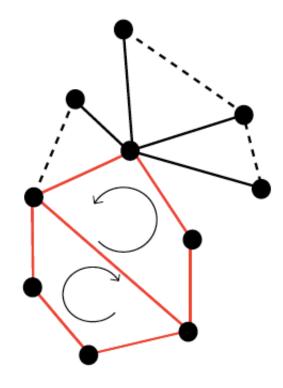
 Edge-disjoint cycles can be updated in parallel.



Work and span for square grid example

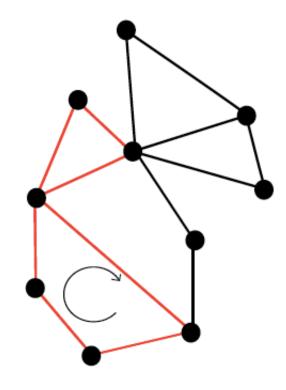


Tree shortcut cycles



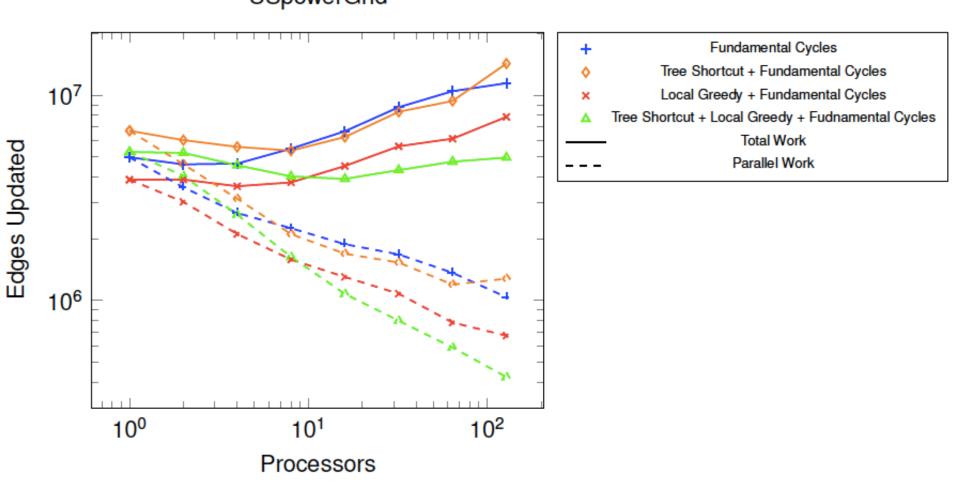
- Select an off-tree edge.
- Search for a shortcut between endpoints of this edge.
 - Only search edges closer to root of tree.
 - Truncate search to control cost.
- Replace tree cycle with shortcut cycle.
 - Cycle space dimension remains the same.

Local greedy cycles



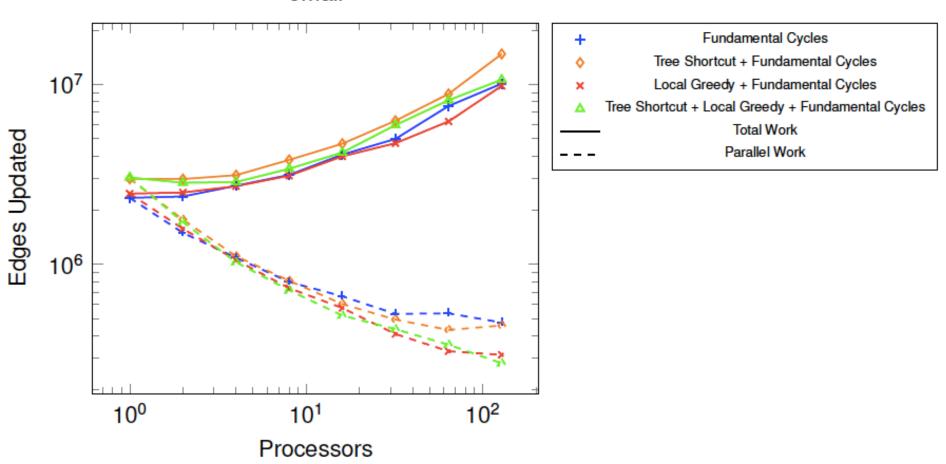
- Select an unmarked edge.
- Find smallest cycle containing this edge.
 - Truncate search to control cost.
 - If found mark all the used edges.
- This set is not guaranteed to span the cycle space.

Comparing cycle sets: work and parallel work



USpowerGrid

Comparing cycle sets: work and parallel work



email

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