Reproducibility in Parallel Graph Algorithms

Samuel D. Pollard, Sudharshan Srinivasan, and Boyana Norris



SIAM CSE 2019, MS168: Reproducibility in Network Algorithms II February 26, 2019

Steps to (Compute a Graph Property Quickly)



- Find a package
- 2 Nail down dependencies
- 3 Reformat graph input file
- 4 Learn how to use CLI/API
- 5 Run gamut of experiments
- 6 Measure performance of experiments
- 7 Combine, analyze, and decide
- 8 Link with existing workflow





Steps to Quickly (Compute a Graph Property Quickly)

- 1 Find a package
- 2 Nail down dependencies
- **3** Reformat graph input file
- 4 Learn how to use CLI/API
- 5 Run gamut of experiments on my computer
- 6 Measure performance of such experiments
- 7 Combine, analyze, and decide
- 8 Link with existing workflow



Parallel Graph Algorithms

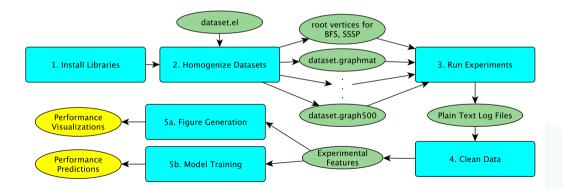


- ▶ Social network analysis [2]
 - Twitter has its own graph processing package (GraphJet), Google has its own language (Pregel)
- ▶ Bioinformatics [4]
 - Tend to have more complex analysis of smaller datasets
 - non-e.g. Needleman-Wunsch algorithm for global sequence alignment



Workflow







Before Reproducibility...



- Ensure we're actually measuring the right times!
- File read time was mistakenly used as performance measurement

- Abridged log file from GraphMat
 - finished file read. time: 2.65211
 - load graph: 5.91229 sec
 - initialize engine: 8.32081e-05 sec
 - run algorithm 1 (count degree): 0.0555639 sec
 - run algorithm 2 (compute PageRank): 0.149445 sec
 - print output: 0.0641179 sec
 - deinitialize engine: 0.00022006 sec



File Formats

UNIVERSITY OF

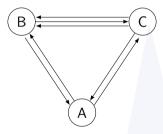


 Binary can be much faster (factor of at least 3) No parsing, can just store into 	this?	or this?	or this?
 array Less portable May be serialization of internal data structures D-indexed or 1-indexed? May not be interchangeable 	55555 44444	5 4	4 3
	44444 22222	4 2	3 1
	11111 66666	1 6	0 5

8/21

Differing Results

- PageRank stopping criterion
 - $||p_t p_{t-1}||_1$
 - $||p_t p_{t-1}||_{\infty}$
 - Stop when no weights change (machine ϵ)
- ► Triangle counting
 - Count both directions of triangle?
 - One, two, or three triangles?



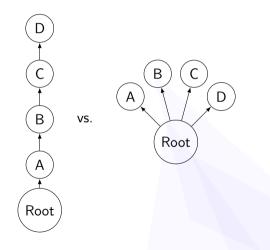




Starting (Root) Vertices



- Performance of BFS and SSSP depends on where you start
- ► More reachable vertices





Once We Have Reproducibility



► Running experiments is expensive

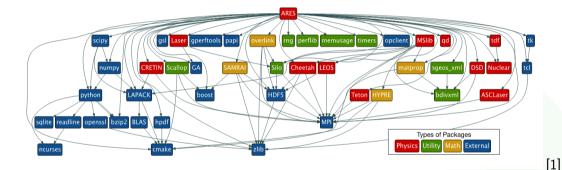
- Read in a 10+GB file just to do BFS once
- Batched is attractive but requires per-package modification

ST	TIME	NOD	ES
R	12:24:12		1
R	4:45:20		1
R	3:15:10		1
R	2:45:54		1
R	2:45:54		1
R	2:45:54		1
R	2:47:34		1
	R R R R R R R	R 12:24:12 R 4:45:20 R 3:15:10 R 2:45:54 R 2:45:54 R 2:45:54	R 12:24:12 R 4:45:20 R 3:15:10 R 2:45:54 R 2:45:54 R 2:45:54



How Reproducible is Reproducible Enough?





- ► We want better performance!
 - Computer upgrades, new package versions
 - All the lovely breaking changes that come along with them



Parameter Tuning

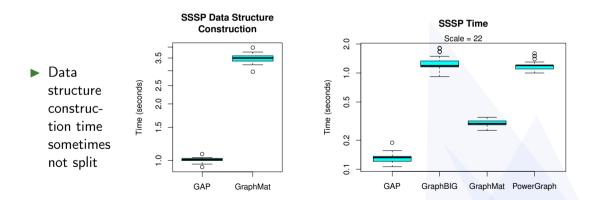


- ▶ BFS α and β (direction-optimizing). This is a combination of:
 - bottom-up: unvisited nodes searching for visited parents
 - top-down: visited nodes searching for unvisited children
- \blacktriangleright SSSP $\Delta\text{-stepping};$ order nodes using ranges of size Δ to distribute work



Early Results

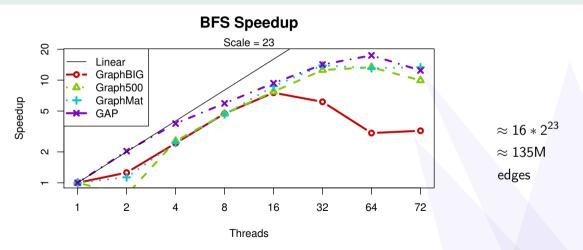






Early Results







Be Careful with Speedup



- ▶ What if I naïvely wrote serial implementation?
- ▶ Graph500 has Serial, OpenMP, and MPI implementations



Recommendation and Ranking



- $\blacktriangleright\,$ Pick the best graph package for your hardware and graph
 - 1 Compute features of a graph
 - This may be expensive
 - 2 Train a model based on these features
- ▶ Apply work in linear solver recommendations [5] to graph processing packages



Computing Features



17/21

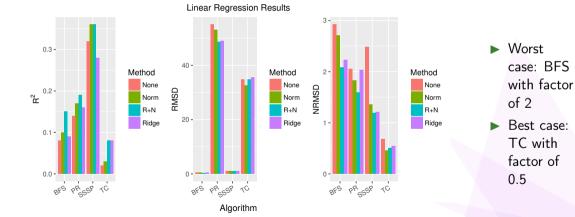
Figure: A scale free network has low diameter [3]

- We use 12 features (the ones computed on SNAP)
- e.g. # vertices, # edges, diameter, clustering coefficient

UNIVERSITY OF

Training Models





O UNIVERSITY OF OREGON

Classification



(b) Random Forest					
		good	bad		
тс	good	43	1		
	bad	0	118		
		good	bad		
BFS	good	110	0		
	bad	0	147		
		good	bad		
PR	good	75	0		
	bad	1	193		
		good	bad		
SSSP	good	51	11		
	bad	9	198		



Conclusion



- ▶ Manually selecting between dozens of implementations is infeasible
- > Automated installation, performance experiments can help reproducibility
- Computing graph features and performance models beforehand can facilitate package selection



Future Directions



- ► Leaderboard of performance results (like Graph500)
- Containerization
 - Singularity
 - Docker
 - Extreme-Scale Scientific Software Stack (E4S) software stack



References I



- Gamblin, T., LeGendre, M., Collette, M. R., Lee, G. L., Moody, A., de Supinski, B. R., and Futral, S. The spack package manager: bringing order to hpc software chaos. In <u>Proceedings of the International Conference for High Performance Computing, Networking, Storage and Analysis</u> (Nov. 2015), SC '15, pp. 40:1–40:12.
- Kang, U., Meeder, B., and Faloutsos, C.
 Spectral analysis for billion-scale graphs: Discoveries and implementation.
 In Advances in Knowledge Discovery and Data Mining (Berlin, 2011), vol. 6635 of Lecture Notes in Computer Science, Springer.
- [3] Lamberson, P. J.
 Scale-free network.
 Available at http://social-dynamics.org/scale-free-network/.



References II



- [4] Pavlopoulos, G. A., Secrier, M., Moschopoulos, C. N., Soldatos, T. G., Kossida, S., Aerts, J., Schneider, R., and Bagos, P. G.
 Using graph theory to analyze biological networks. In <u>BioData Mining</u> (Bethesda, MD, 2011), PubMed Central.
- [5] Sood, K., Norris, B., and Jessup, E. Comparative performance modeling of parallel preconditioned krylov methods. In IEEE 19th International Conference on High Performance Computing and Communications; 15th International Conference on Smart City; 3rd International Conference on Data Science and Systems (Dec.
 - 2017), HPCC/SmartCity/DSS, pp. 26–33.

