

Scaling Graph Clustering with Distributed Sketches

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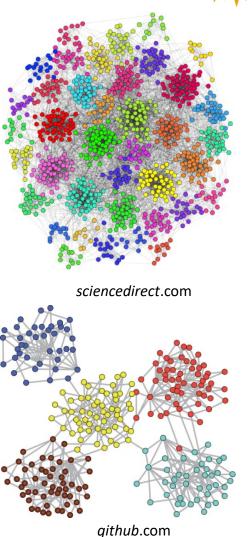
Graph Clustering

- Goal: Identify "like" vertices as groups in a network.
- Applications: Contact Tracing!
- Accepted Approach (Spectral):
 - Embed graph matrix in low dimensional space spanned by subset of its eigenvectors.
 - Feed the embedding to a clustering algorithm.

Challenges:

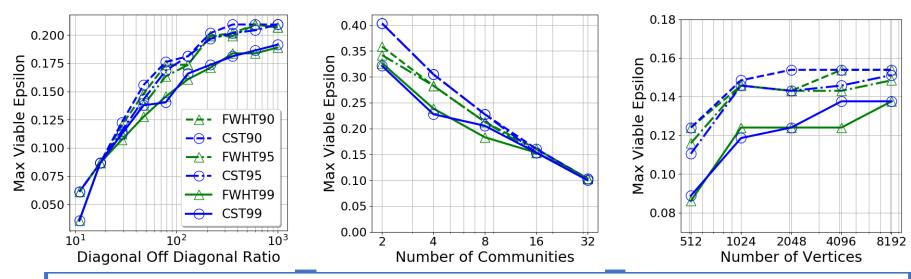
- Distributed: bulk synchronous communication.
- Streaming: spectral methods slow to update.





Proposed Solution:

Matrix Sketching: Overview and Results

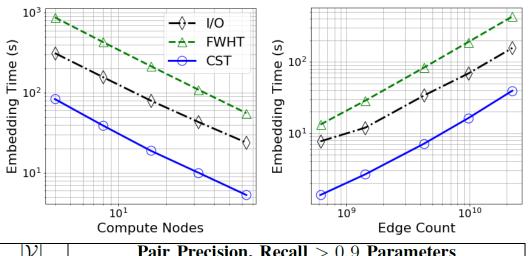


Y-axes: Required geometric fidelity of sketch. **X-axes:** (left) ratio of intra-group connectivity to inter-group connectivity; (center) number of true groups, fixed size; (right) size of graph, fixed number of groups.

- Randomly embed a graph matrix in a lower dimension, preserve important geometric structure.
 - Fast Walsh-Hadamard Transform (FWHT): theoretically sound, harder to implement.
 - CountSketch (CST): less rigorous, but faster and easily implemented in distributed setting.
- Use Stochastic Block Models (SBMs) in our numerical experiments.
 - SBMs function as surrogates for real world networks.
 - Randomly assign vertices to groups; high connectivity within groups, low between groups.

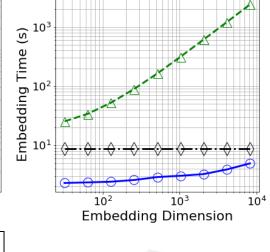


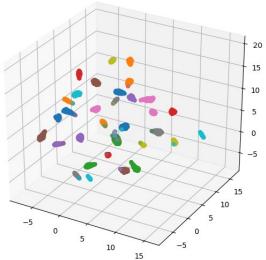
Results, Next Steps, Questions?



$ \mathcal{V} $	Pair Precision, Recall > 0.9 Parameters				
	# Partitions	ε	PP	PR	Accuracy
500	8	0.1	0.96983	0.97719	0.986
1000	11	0.1	0.95991	0.95301	0.976
5000	19	0.05	0.97103	0.97395	0.9878
20000	32	0.018	0.91305	0.90455	0.9588
50000	44	0.01	0.55959	0.12414	0.73773

- CST is faster than I/O, as accurate as FWHT!
- Next: Optimization of Clustering Algorithms
- See our paper: https://arxiv.org/abs/2007.12669







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