

Elements of Spectral Graph Theory II

A Toolkit for Fast Graph Algorithms

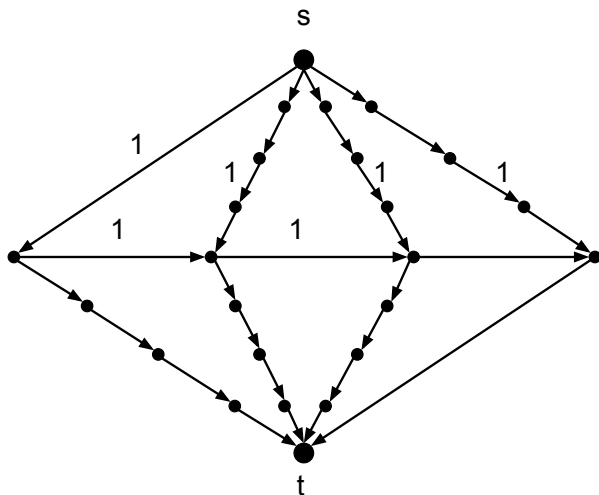
Aaron Schild

UC Berkeley EECS

March 27th, 2018

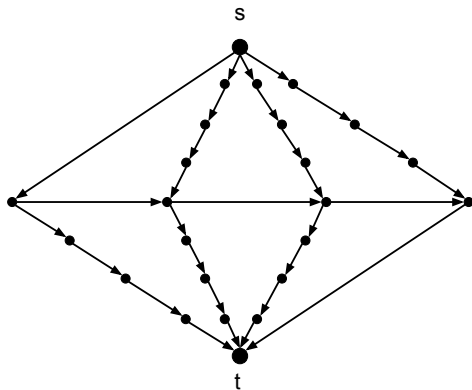
Motivation: Maximum Flow

Push as much flow from s to t without violating edge capacities.



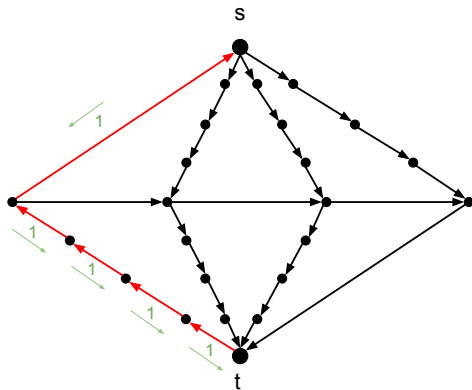
Classic technique: Ford-Fulkerson

- F : maximum flow value
- Push flow along paths one at a time
- $O(m)$ time to find each path, F paths
- $O(mF)$ total work



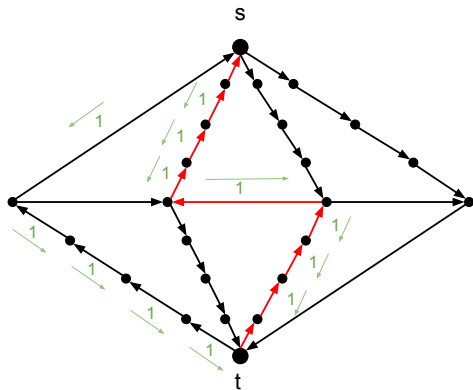
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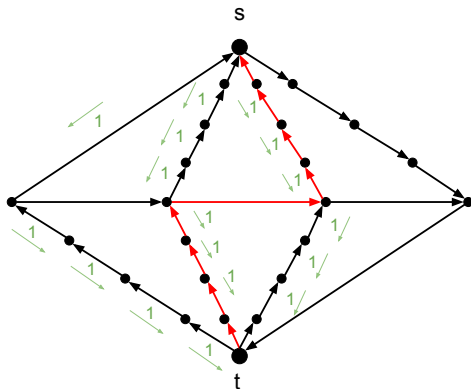
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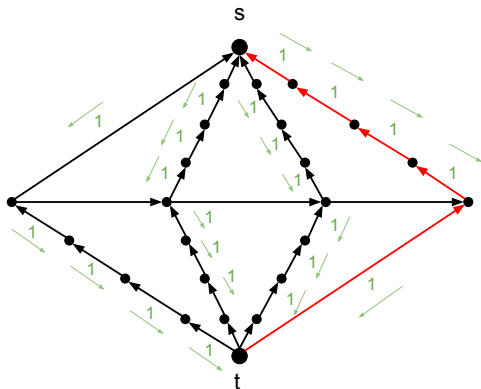
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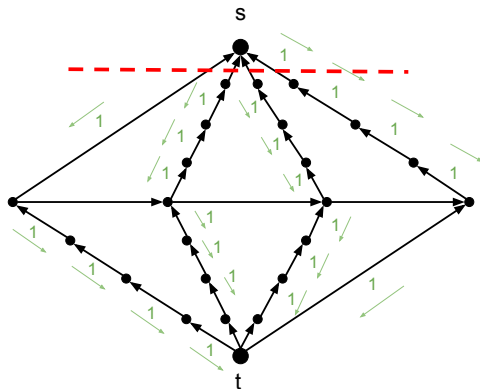
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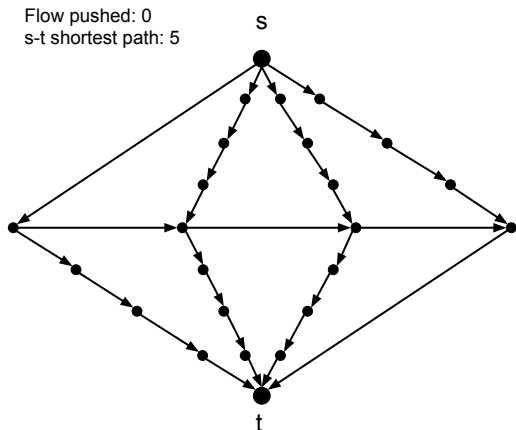
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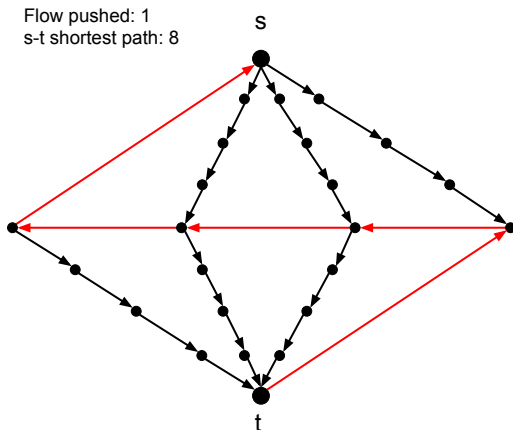
$O(m^{3/2})$ time: Blocking Flow

- Push flow along maximal collection of $s - t$ shortest paths
- $s - t$ distance increases after each augmentation



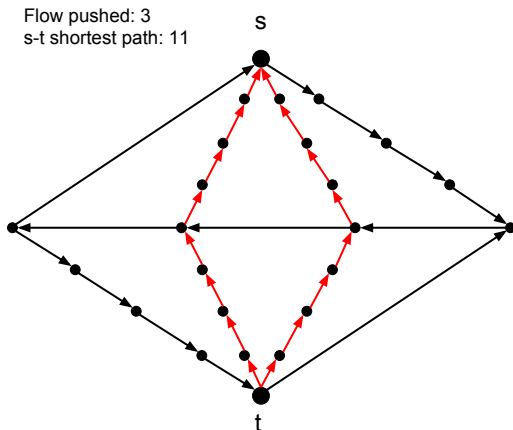
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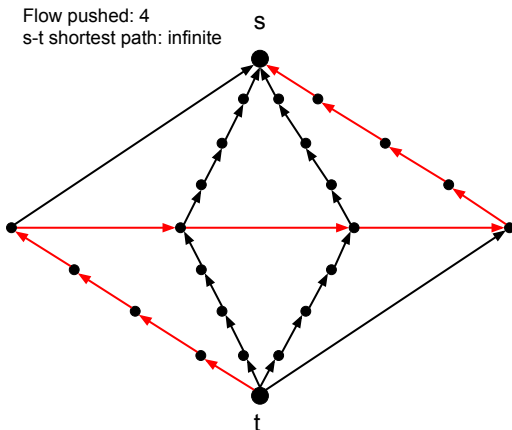
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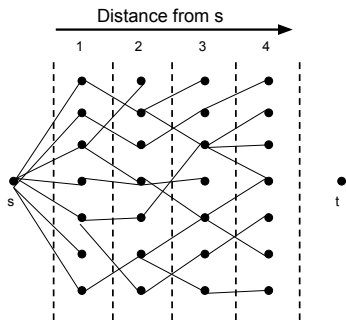
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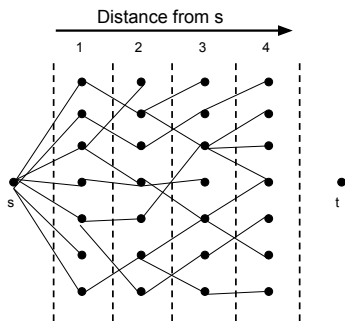
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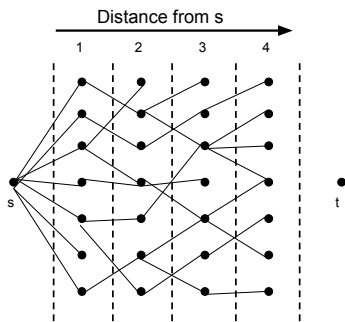
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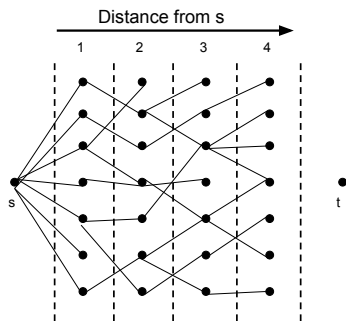
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- $s - t$ distance increases after each augmentation
- **If distance $> D$, there must be a cut with size at most m/D**
- Only m/D more paths needed at this point
- Runtime: $O(m(D + m/D))$ for any $D > 0$, minimized with $D = m^{1/2} \rightarrow O(m^{3/2})$ time

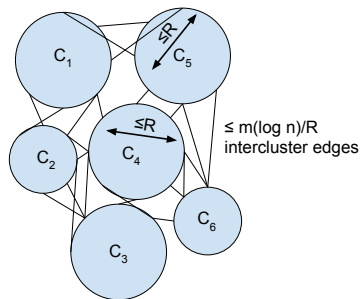


Graph Partitioning and Region Growing

Theorem ([LR99])

For any value $R > 0$, can partition a graph into clusters C_1, C_2, \dots, C_k with two properties:

- each cluster has diameter at most R
- the number of edges between clusters is at most $\tilde{O}(m/R)$

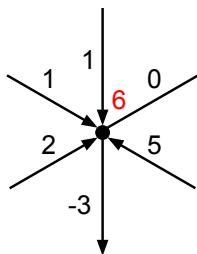


Maximum Flow as Norm Minimization

- suppose graph has unit capacities, m edges, and n vertices

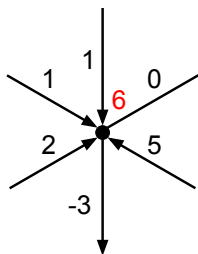
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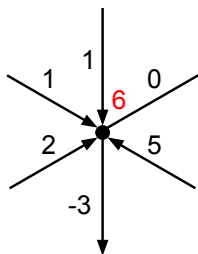
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Maximum flow problem: find maximum value of α for which there is a vector $f \in \mathbb{R}^m$ with $Mf = \alpha\chi_{st}$ with $\|f\|_\infty \leq 1$.

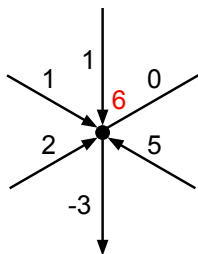


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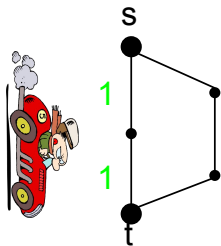
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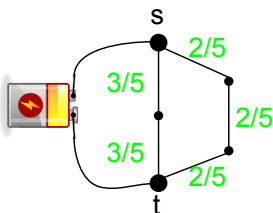
Equivalently up to scaling: $\min_f \|f\|_\infty$ subject to $Mf = \chi_{st}$



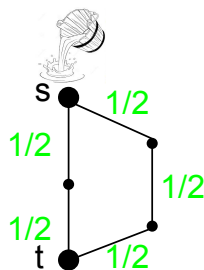
Other Norm-Minimizing Flows



(a) l_1 (shortest path)

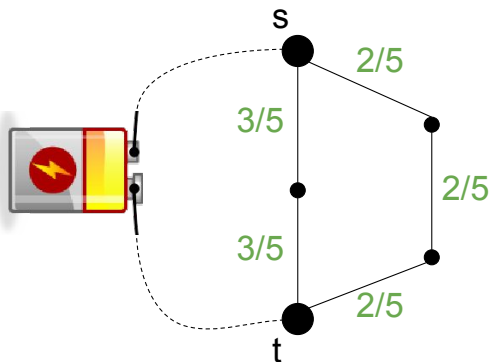


(b) l_2



(c) l_∞ (max flow)

Electrical Flows

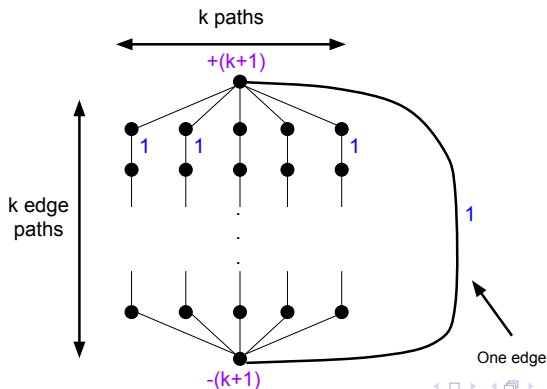


Can be computed in almost linear time by solving a Laplacian linear system! [ST003]

Finding maximum flows using electrical flows

Algorithm (similar to [CKM⁺10]):

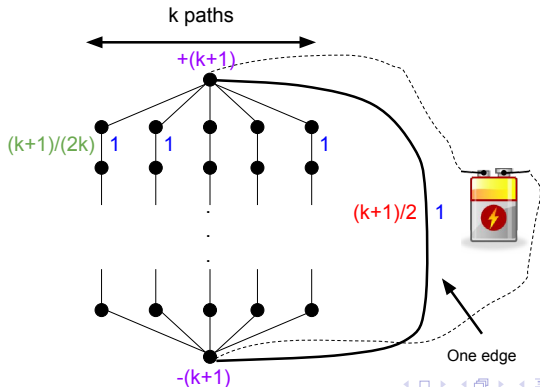
- Arbitrarily initialize resistances $\{\mathbf{r}_e\}_e$
- While there is some edge e with $\mathbf{f}_e > \mathbf{c}_e$
 - ▶ Let \mathbf{f} be the $s - t$ electrical flow with resistances \mathbf{r}
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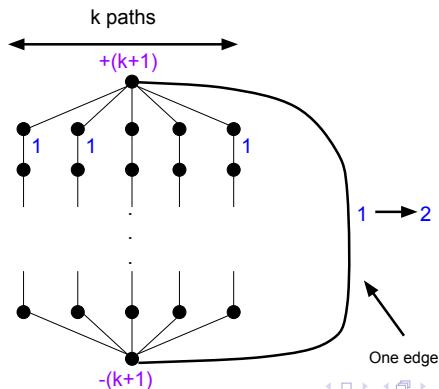
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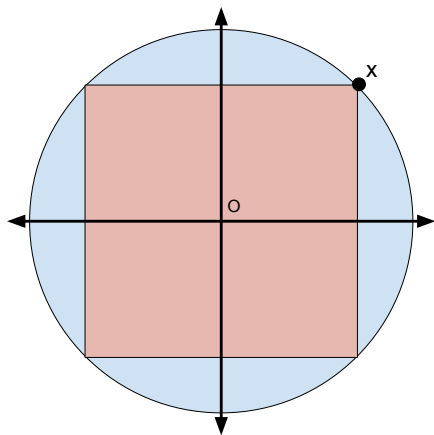
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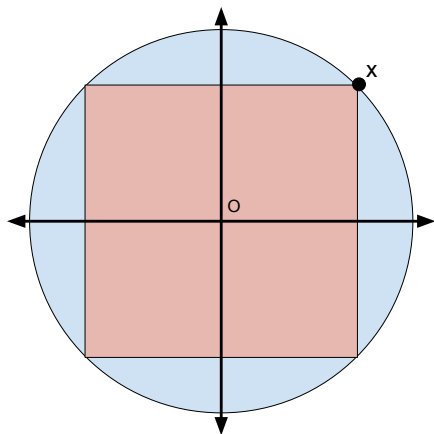
Geometry and its relationship to iteration count

- Can find an approximate max flow in $\tilde{O}(\sqrt{m}\text{poly}(1/\epsilon))$ iterations



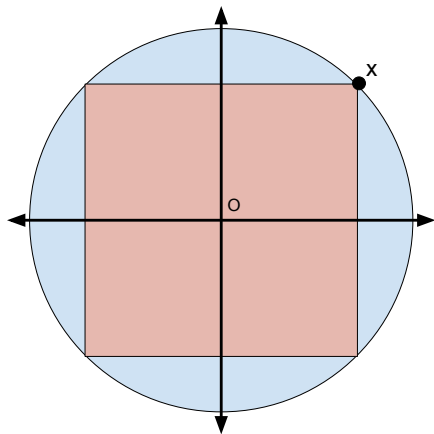
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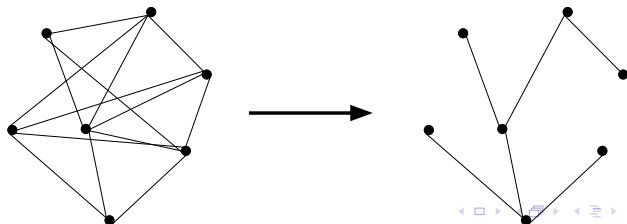
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- \sqrt{m} comes from gap between ℓ_2 and ℓ_∞ norm in \mathbb{R}^m
- $\tilde{O}(m)$ time per iteration $\rightarrow \tilde{O}(m^{3/2}\text{poly}(1/\epsilon))$ time



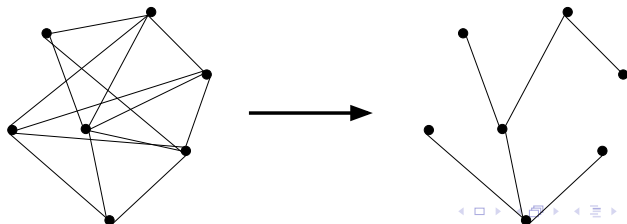
Better geometry with graph embeddings

- Embed graph into a simpler graph while preserving cuts/distances



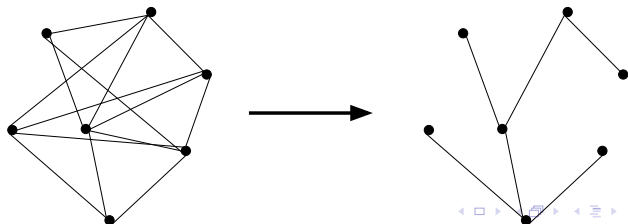
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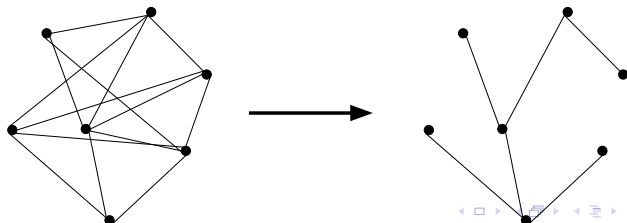
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- Unfortunately, can't do well with a single tree, but can do well with a distribution over trees!
- Related embedding technique can be used to find an $m^{o(1)}$ -approximate ℓ_∞ projection, which yields an $(1 + \epsilon)$ -approximate max flow in $O(m \text{polylog}(n)/\epsilon)$ time [KLOS14, She13, She17, Pen16]



Summary

- Graph partitioning (diameter v.s. cutsizes)
- Writing flow problems as norm minimization
- Graph embedding

Bibliography



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Electrical flows, laplacian systems, and faster approximation of maximum flow in undirected graphs.

[CoRR, abs/1010.2921, 2010.](#)



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An almost-linear-time algorithm for approximate max flow in undirected graphs, and its multicommodity generalizations.

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Multicommodity max-flow min-cut theorems and their use in