Lecture 20: Max-Flow II

Michael Dinitz

November 7, 2024 601.433/633 Introduction to Algorithms

Introduction

Last time:

- ► Max-Flow = Min-Cut
- Can compute max flow and min cut using Ford-Fulkerson: while residual graph has an s → t path, push flow along it.
 - Corollary: if all capacities integers, max-flow is integral
 - If max-flow has value F, time O(F(m + n)) (if all capacities integers)
 - Exponential time!

Today:

- Important setting where FF is enough: max bipartite matching
- Two ways of making FF faster: Edmonds-Karp

Max Bipartite Matching

Setup

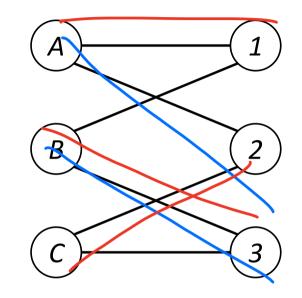
Definition

A graph G = (V, E) is *bipartite* if V can be partitioned into two parts L, R such that every edge in E has one endpoint in L and one endpoint in R.

Definition

A matching is a subset $M \subseteq E$ such that $e \cap e' = \emptyset$ for all $e, e' \in M$ with $e \neq e'$ (no two edges share an endpoint)

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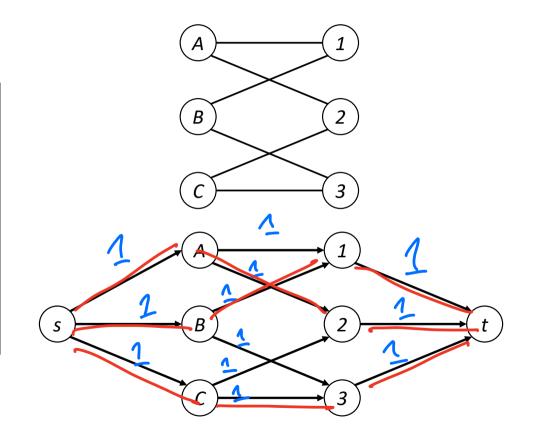
Bipartite Maximum Matching: Given bipartite graph G = (V, E), find matching M maximizing |M|

Extremely important problem, doesn't seem to have much to do with flow!

Algorithm

Give all edges capacity 1 Direct all edges from *L* to *R* Add source *s* and sink *t* Add edges of capacity 1 from *s* to *L* Add edges of capacity 1 from *R* to *t*

Run FF to get flow fReturn $M = \{e \in L \times R : f(e) > 0\}$





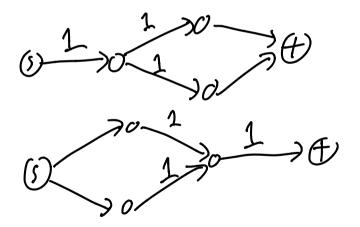
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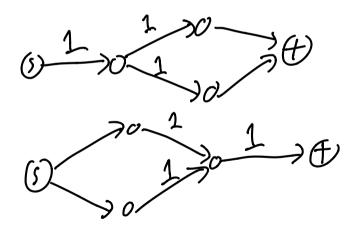
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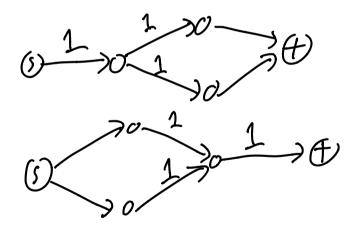
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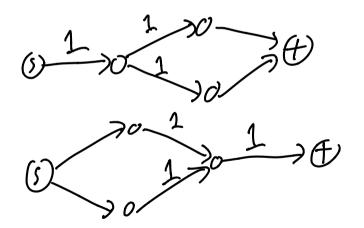
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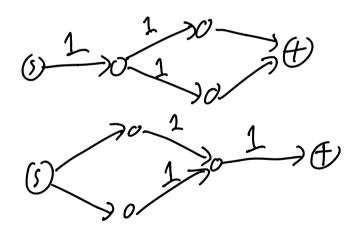
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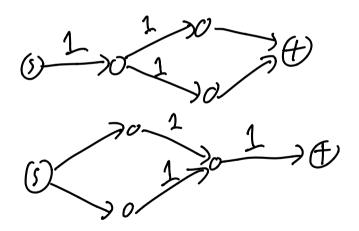
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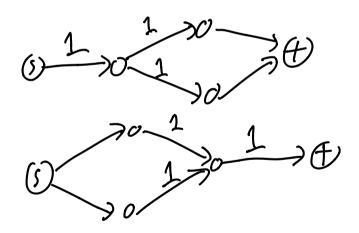
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- |f'| = |M'| > |M| = |f|
- Contradiction

Running Time

Running Time:

- O(n + m) to make new graph
- $|f| = |M| \le n/2$ iterations of FF
- $\implies O(n(m+n)) = O(mn)$ time (assuming $m \ge \Omega(n)$)

Exensions

. . .

Many extensions:

- Max-weight bipartite matching
- Min-cost perfect matching
- Matchings in general graphs

Exensions

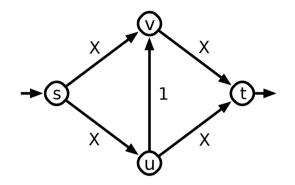
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Still active area of study!

- Michael Dinitz, Sungjin Im, Thomas Lavastida, Benjamin Moseley, Sergei Vassilvitskii. Faster Matchings via Learned Duals. NeurIPS 2021.
- Michael Dinitz, George Li, Quanquan Liu, Felix Zhou. Differentially Private Matchings. Submitted (Monday), hopefully on arXiv soon.

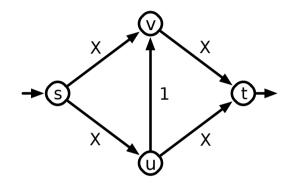
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A bad example for the Ford-Fulkerson algorithm.

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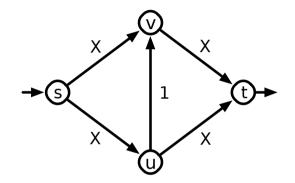


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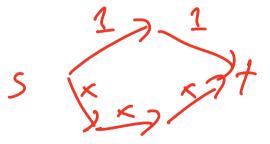


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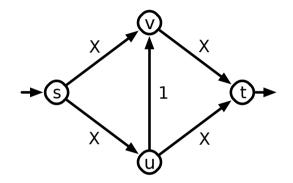
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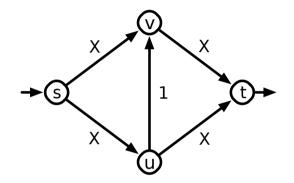
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"Widest" path: push as much flow as possible each iteration

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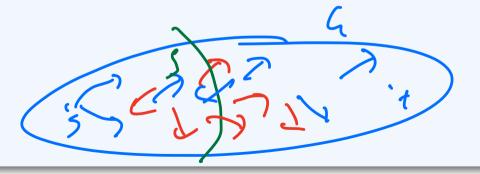
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Does this implies at most m iterations?

Michael Dinitz

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 \implies If $i > m \ln F$, amount remaining to be sent at most

$$F(1-1/m)^{i} < F(1-1/m)^{m \ln F} \le F(e^{-1/m})^{m \ln F} = F \cdot e^{-\ln F} = 1$$

But all capacities integers, so must be finished!

Modified version of Dijkstra: find widest path in $O(m \log n)$ time

- Total time $O(m \log n \cdot m \log F) = O(m^2 \log n \log F)$
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Question: can we get running time independent of **F**?

Strongly polynomial-time algorithm.

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Again use Ford-Fulkerson, but pick *shortest* augmenting path (unweighted)

- Ignore capacities, just find augmenting path with fewest hops!
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Theorem

EK2 has at most O(mn) iterations, so at most $O(m^2n)$ running time (if $m \ge n$)

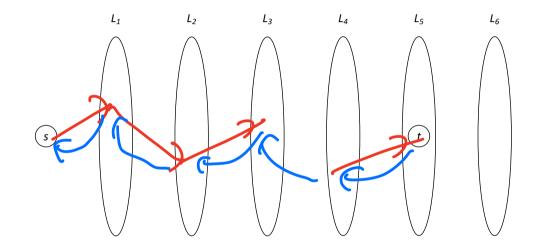
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- Distance initially $\geq 1 \implies$ distance > *n* after at most *mn* iterations
- Only distance larger than n is ∞ : no $s \rightarrow t$ path
- → Terminates after at most *mn* iterations.

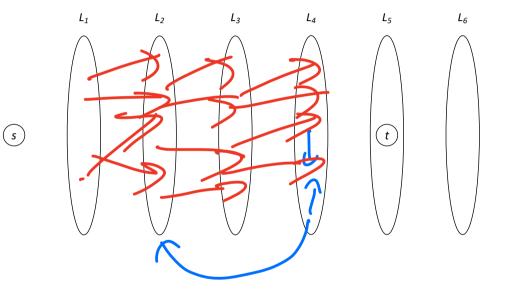
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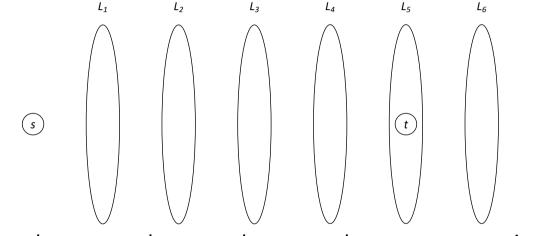


Edge types:

- ► Forward edges: 1 level
- Edges inside level
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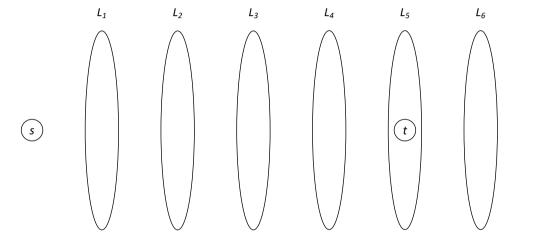
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So after m iterations (same layout): no path using only forward edges \implies distance larger than d!



So at most mn iterations. Each iteration unweighted shortest path: BFS, time O(m + n)



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Total time: $O(mn(m + n)) = O(m^2 n)$. Independent of F!

Extensions

Many better algorithms for max-flow: *blocking flows* (Dinitz's algorithm (not me)), *push-relabel* algorithms, etc.

- CLRS has a few of these.
- State of the art:
 - Strongly polynomial: O(mn). Orlin [2013] & King, Rao, Tarjan [1994]
 - Weakly Polynomial: O(m^{1+o(1)} log U) (where U is maximum capacity). Chen, Kyng, Liu, Peng, Gutenberg and Sachdeva [2022]

Many other variants of flows, some of which are just s - t max flow in disguise!

Min-Cost Max-Flow: every edge also has a cost. Find minimum cost max-flow. Can be solved with just normal max flow!