

# Lecture 19: Max-Flow II

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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  $s \rightarrow t$  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

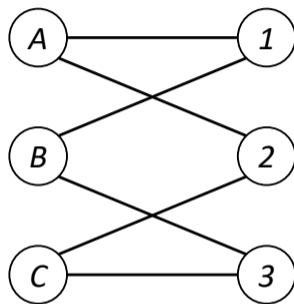
# Setup

## 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)



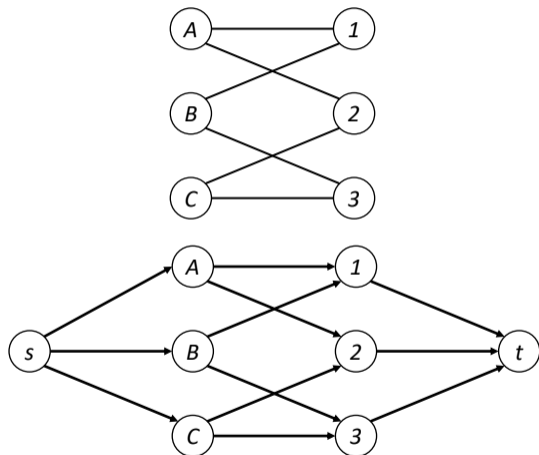
**Bipartite Maximum Matching:** Given bipartite graph  $G = (V, E)$ , find matching  $M$  maximizing  $|M|$

- ▶ 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**

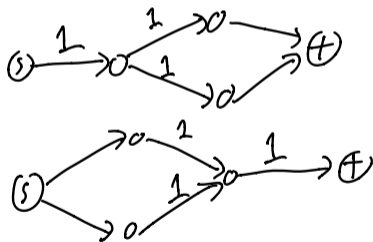
Run FF to get flow **f**  
Return  $\mathbf{M} = \{\mathbf{e} \in \mathbf{L} \times \mathbf{R} : \mathbf{f}(\mathbf{e}) > \mathbf{0}\}$



# Correctness

**Claim:**  $M$  is a matching

**Proof:** capacities in  $\{0, 1\} \implies f(e) \in \{0, 1\}$   
for all  $e$  (integrality)



**Claim:**  $M$  is maximum matching

**Proof:** Suppose larger matching  $M'$   
Can send  $|M'|$  flow using  $M'$ !

- ▶  $f'(s, u) = 1$  if  $u$  matched in  $M'$ , otherwise 0
- ▶  $f'(v, t) = 1$  if  $v$  matched in  $M'$ , otherwise 0
- ▶  $f'(u, v) = 1$  if  $\{u, v\} \in M'$ , otherwise 0
- ▶  $|f'| = |M'| > |M| = |f|$
- ▶ Contradiction

# Running Time

Running Time:

- ▶  $O(n + m)$  to make new graph
- ▶  $|f| = |M| \leq n/2$  iterations of FF

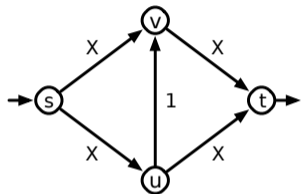
$\Rightarrow O(n(m + n)) = O(mn)$  time (assuming  $m \geq \Omega(n)$ )

# Edmonds-Karp



## Intuition

Bad example for Ford-Fulkerson:



If Ford-Fulkerson chooses bad augmenting paths, super slow!

Obvious idea: Choose better paths!

A bad example for the Ford-Fulkerson algorithm.

Obvious path to pick:

$$\arg \max_{\text{augmenting paths } P} \min_{e \in P} c_f(e).$$

- ▶ “Widest” path: push as much flow as possible each iteration

## Edmonds-Karp #1

Edmonds-Karp #1: Ford-Fulkerson, always choose “widest” path.

- ▶ Correct, since FF. Running time?

### Lemma

*In any graph with max  $s - t$  flow  $F$ , there exists a path from  $s$  to  $t$  with capacity at least  $F/m$*

### Proof.

Let  $X = \{e \in E : c(e) < F/m\}$ .

If no  $s \rightarrow t$  path in  $G \setminus X$ , then  $X$  an (edge) cut. Let  $S$  = nodes reachable from  $s$  in  $G \setminus X$ .

$$\text{cap}(S, \bar{S}) \leq \text{cap}(X) = \sum_{e \in X} c(e) < m \cdot (F/m) = F$$

$\implies \min(s, t) \text{ cut} \leq \text{cap}(S, \bar{S}) < F$ . Contradiction.

$\implies \exists s \rightarrow t$  path  $P$  in  $G \setminus X$ : every edge of  $P$  has capacity at least  $F/m$  □

Does this implies at most  $m$  iterations?

## EK 1 Running Time

### Theorem

If  $F$  is the value of the maximum flow and all capacities are integers, # iterations of EK1 is at most  $O(m \log F)$

How much flow remains to be sent after iteration  $i$ ?

- ▶  $i = 0$ :  $F$
- ▶  $i = 1$ : Sent at least  $F/m$ , so at most  $F - F/m = F(1 - 1/m)$  remaining
- ▶  $i = 2$ : Sent at least  $R/m$  if  $R$  was remaining after iteration 1, so at most  $R - R/m = R(1 - 1/m) \leq F(1 - 1/m)^2$  remaining

By induction: after iteration  $i$ , at most  $F(1 - 1/m)^i$  flow remaining to be sent.

Super useful inequality:  $1 + x \leq e^x$  for all  $x \in \mathbb{R}$

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

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

But all capacities integers, so must be finished!

## Finishing EK1

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

- ▶ Total time  $O(m \log n \cdot m \log F) = O(m^2 \log n \log F)$
- ▶ Polynomial time!

Question: can we get running time independent of  $F$ ?

- ▶ *Strongly* polynomial-time algorithm.

## Edmonds-Karp #2

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

- ▶ Ignore capacities, just find augmenting path with fewest hops!
- ▶ Easy to compute with BFS in  $O(m + n)$  time.

Main question: how many iterations?

### Theorem

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

## Proof (sketch) of EK2

Idea: prove that distance from  $\mathbf{s}$  to  $\mathbf{t}$  (unweighted) goes up by at least one every  $\leq \mathbf{m}$  iterations.

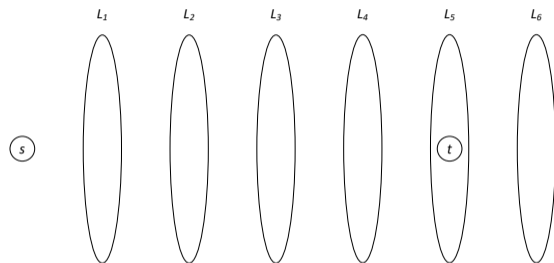
- ▶ Distance initially  $\geq \mathbf{1} \implies$  distance  $> \mathbf{n}$  after at most  $\mathbf{mn}$  iterations
- ▶ Only distance larger than  $\mathbf{n}$  is  $\infty$ : no  $\mathbf{s} \rightarrow \mathbf{t}$  path

$\implies$  Terminates after at most  $\mathbf{mn}$  iterations.

## Proof (sketch) of EK2 (continued)

Suppose  $s \rightarrow t$  distance is  $d$ .

“Lay out” residual graph in levels by BFS (distance from  $s$ )



Edge types:

- ▶ Forward edges: **1** level
- ▶ Edges inside level
- ▶ Backwards edges

What happens when we choose a *shortest* augmenting path? Only uses forward edges!

- ▶ At least **1** forward edge gets removed, replaced with backwards edge.
- ▶ No backwards edges turned forward

So after  $m$  iterations (same layout): no path using only forward edges  $\implies$  distance larger than  $d$ !

## Finishing EK2

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

Total time:  $O(mn(m + n)) = O(m^2n)$ . 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:  $\mathbf{O}(mn)$ . Orlin [2013] & King, Rao, Tarjan [1994]
  - ▶ Weakly Polynomial:  $\tilde{\mathbf{O}}(m^{\frac{3}{2}-\frac{1}{328}} \log \mathbf{U})$  (where  $\mathbf{U}$  is maximum capacity). Gao, Liu, Peng [2021]

Many other variants of flows, some of which are just  $\mathbf{s} - \mathbf{t}$  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!