

Algorithms - Fall 2023

Applications  
of Flow



# Recap

- Sign up for oral grading  
+ HW of groups
- Sample final: I'll bring  
next week
- Final readings added to  
Canvas
- Review session:  
Wed. of finals week at  
either 2 or 3pm

# Max flow / Min Cut:

- FF: Residual graphs:  $\mathcal{O}(V+E) \cdot f^*$
- Edmonds - Karp:  $\mathcal{O}(E^2 \log E \log f^*)$
- BFS based:  $\mathcal{O}(VE^2)$

Technique	Direct	With dynamic trees	Source(s)
Blocking flow	$\mathcal{O}(V^2E)$	$\mathcal{O}(VE \log V)$	[Dinitz; Karzanov; Even and Itai; Sleator and Tarjan]
Network simplex	$\mathcal{O}(V^2E)$	$\mathcal{O}(VE \log V)$	[Dantzig; Goldfarb and Hao; Goldberg, Grigoriadis, and Tarjan]
Push-relabel (generic)	$\mathcal{O}(V^2E)$	—	[Goldberg and Tarjan]
Push-relabel (FIFO)	$\mathcal{O}(V^3)$	$\mathcal{O}(VE \log(V^2/E))$	[Goldberg and Tarjan]
Push-relabel (highest label)	$\mathcal{O}(V^2\sqrt{E})$	—	[Cheriyan and Maheshwari; Tunçel]
Push-relabel-add games	—	$\mathcal{O}(VE \log_{E/(V \log V)} V)$	[Cheriyan and Hagerup; King, Rao, and Tarjan]
Pseudoflow	$\mathcal{O}(V^2E)$	$\mathcal{O}(VE \log V)$	[Hochbaum]
Pseudoflow (highest label)	$\mathcal{O}(V^3)$	$\mathcal{O}(VE \log(V^2/E))$	[Hochbaum and Orlin]
Incremental BFS	$\mathcal{O}(V^2E)$	$\mathcal{O}(VE \log(V^2/E))$	[Goldberg, Held, Kaplan, Tarjan, and Werneck]
Compact networks	—	$\boxed{\mathcal{O}(VE)}$	[Orlin]

Figure 10.10. Several purely combinatorial maximum-flow algorithms and their running times.

Many use very different techniques

- linear programming
- complex data structures
- not residual graphs

# Topics in Ch. 11

A mess of different ideas!

① Matchings:

Identify a way to pair up items

Build  $G'$ : More pairs  $\Leftrightarrow$  larger flow

② Disjoint paths:

Modify  $G$ :

Find paths that avoid each other.

③ "Tuple" Selection

Build a graph: flow paths give selection

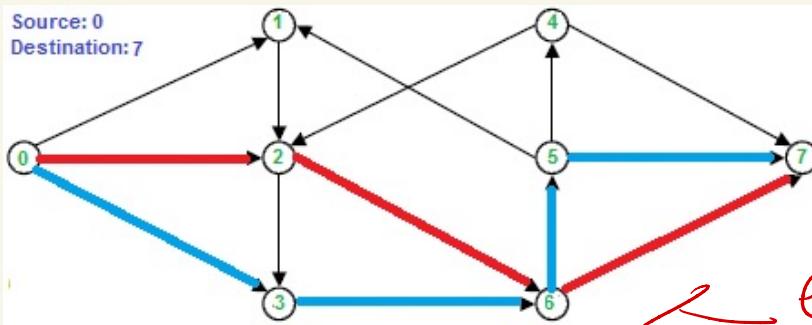
Magic:

## First problem

What if we want non-intersecting paths from  $s$  to  $t$ ?

Two variants:

- Edge disjoint: No 2 paths visit the same edge



edge disjoint  
but not vertex  
disjoint

- Vertex disjoint: no 2 paths visit the same vertex

Note: Different! And both model useful cases

Key: Flow will decompose into paths!

# Edge disjoint

Input: unweighted graph  $G = (V, E)$   
plus  $s, t \in V$ .

For each edge  $e = \overrightarrow{uv}$ :

set  $\text{capacity}(e) \leftarrow 1$  ]  $O(E)$

$\uparrow f \leftarrow \text{Call flow algorithm}$  ]  $O(VE)$

How to get paths?

Initialize empty list for paths

while  $(\text{value}(f) > 0)$

find  $s \rightarrow v$  edge w/  $f(s \rightarrow v) > 1$

add  $(s \rightarrow v)$  to paths[i]

set  $f(s \rightarrow v) \leftarrow 0$

while ( $v \neq t$ )

find edge  $v \rightarrow u$

with  $f(v \rightarrow u) = 1$

add  $(v \rightarrow u)$  to paths[i]

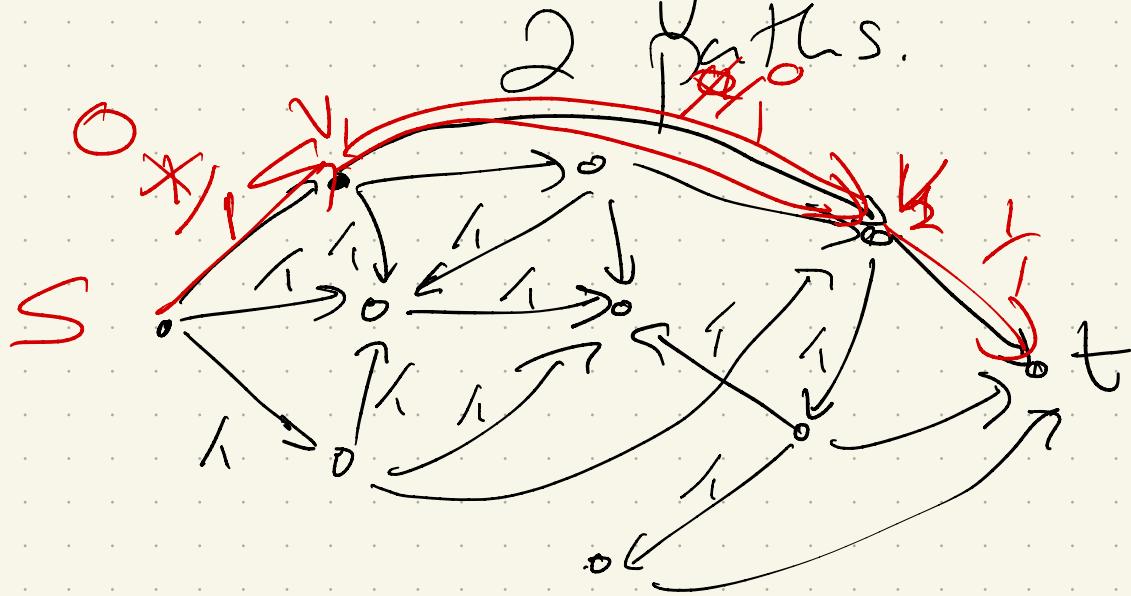
set  $f(v \rightarrow u) \leftarrow 0$

$v \leftarrow u$

i++

Since all flows are integral,  
+ capacity of every edge  
is  $\underline{= 1}$

$\Rightarrow$  no edge will be in  
2 paths.



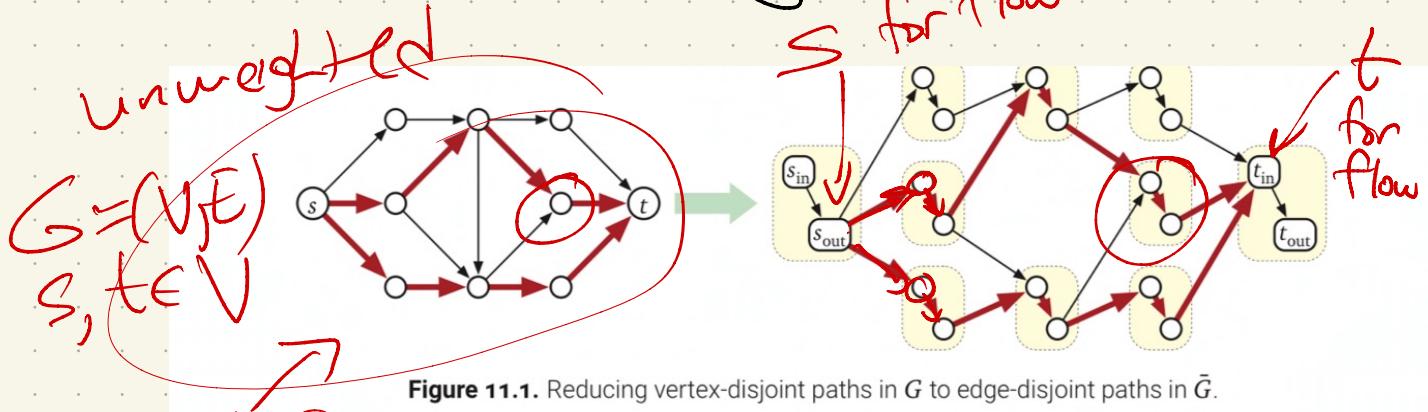
paths: 1:  $s \rightarrow v_1, v_1 \rightarrow v_2, v_2 \rightarrow t$   
2:

correctness: Any set of  $k$   
disjoint paths

$\Rightarrow$  flow of value  $k$   
any flow of value  $k \Rightarrow$  set of  $k$   
paths

Vertex disjoint: ← find paths which don't share a vertex

Build a new graph  $\tilde{G}$ :



$V \rightarrow V(\tilde{G}) = 2$  vertices for every  $v \in G$   
 $E \rightarrow E(\tilde{G}) =$  for any  $u \rightarrow v \in E$ , add  $u_{out} \rightarrow v_{in}$   
 $V + E \rightarrow E(\tilde{G}) =$  add  $v_{in} \rightarrow v_{out}$  for every  $v \in G$

Add capacity = 1 to each edge.

Any flow path that enters  $v_{in}$  will exit  $v_{out}$ , so

$$f(v_{in} \rightarrow v_{out}) = 1 = C(v_{in} \rightarrow v_{out})$$

Result:

Another (his) variant

Suppose edges are unlimited,  
but vertices have capacities.

Ex: Internet routing:

packets queue up at  
routers/switches

So:  $G = (V, E)$  +  $c(v)$  give  
capacities on vertices

How to do flow?

Build  $\bar{G}$ : (with only edge  
capacities)

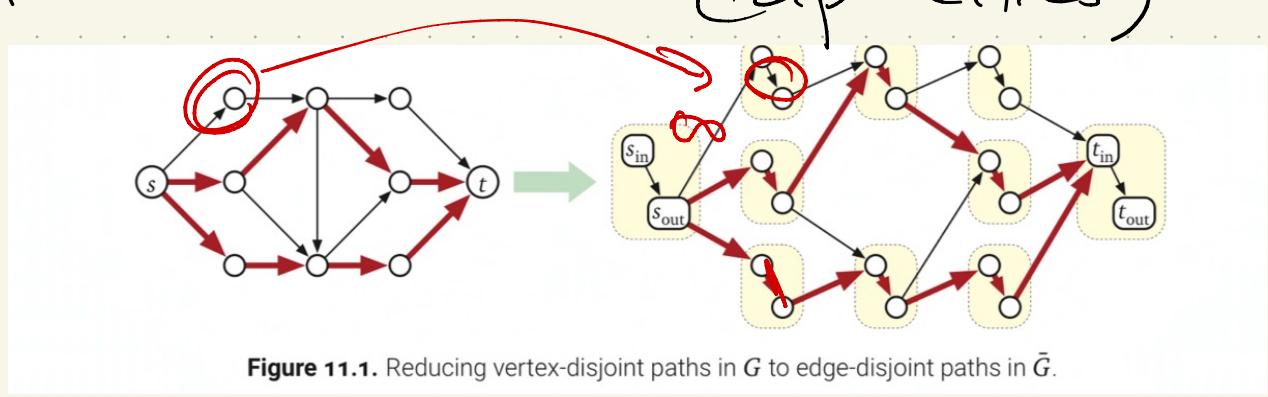


Figure 11.1. Reducing vertex-disjoint paths in  $G$  to edge-disjoint paths in  $\bar{G}$ .

# Reductions: Correctness

Solution in  $G$

$\Rightarrow$  max flow in  $\tilde{G}$

max flow in  $\tilde{G}$

$\Rightarrow$  ~~max flow in~~  $G$   
Solution

So: Compute in  $\tilde{G}$

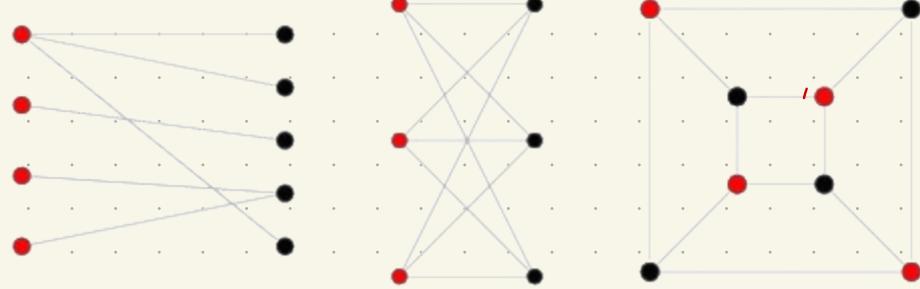
Runtime:  $O(V(\tilde{G}) \cdot E(\tilde{G}))$

=

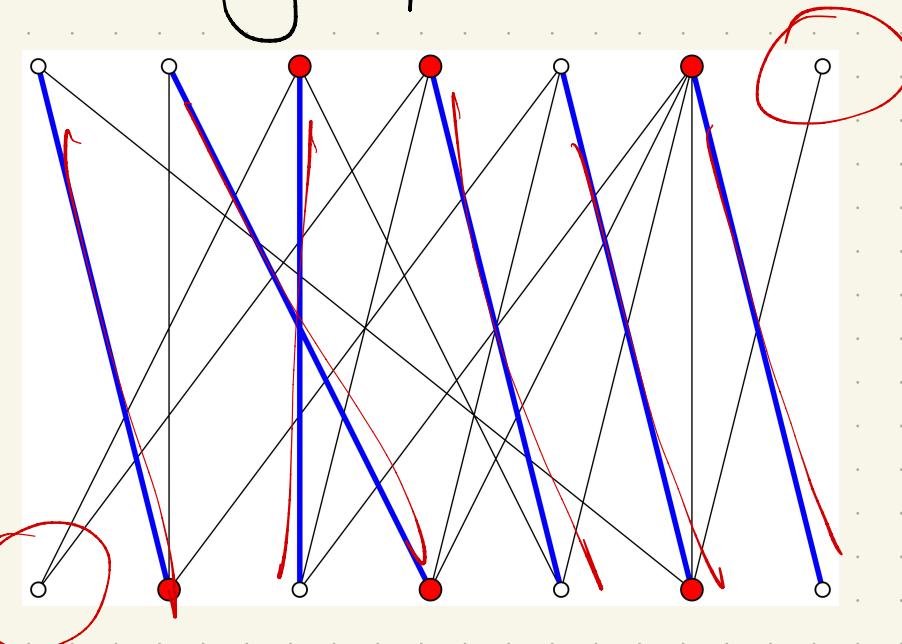
# Bipartite Graphs

Any graph where vertices can be divided into 2 sets  
(usually L & R)  
s.t. no edges exist inside  
L or R

Ex:



Maximum matching: find edges  
(no 2 edges per vertex)



Instead, use flows:

Convert  $G$  to  $\tilde{G}$ :

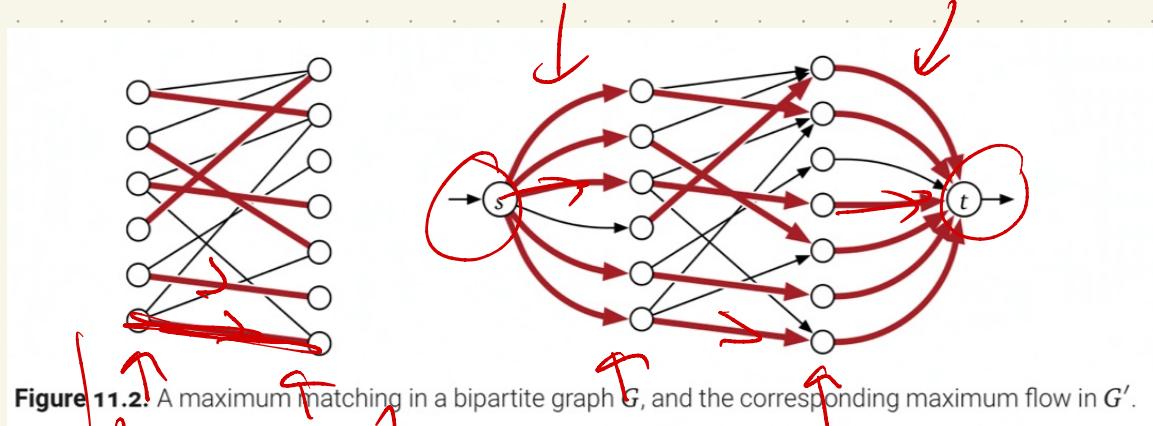


Figure 11.2: A maximum matching in a bipartite graph  $G$ , and the corresponding maximum flow in  $G'$ .



$G$

$\tilde{G}$ : add 2 new vertices  $s$  &  $t$ ,  
all edges  $v \in G$  to  $\tilde{G}$

Edges for  $\tilde{G}$ : every edge in  
 $G$ , add directed version  $\rightarrow$   
 $\tilde{G}$

for all  $v \in L$ , add  $s \rightarrow v$

for all  $u \in R$ , add  $u \rightarrow t$

give every edge  $cap = 1$

Algorithm: Given  $G = (V, E)$   
with  $V = L \cup R$  (bipartite)

// build  $\tilde{G}$   
(new side)

$$\tilde{V} = V + 2$$

$$\tilde{E} = V + E$$

// run flow

$$O(\tilde{V}\tilde{E}) = O(V(V+E))$$

// get matching

track flow paths

↳ any  $L \rightarrow R$  edge w/  $\text{flow} = 1$   
is in matching

Runtime:

## Correctness:

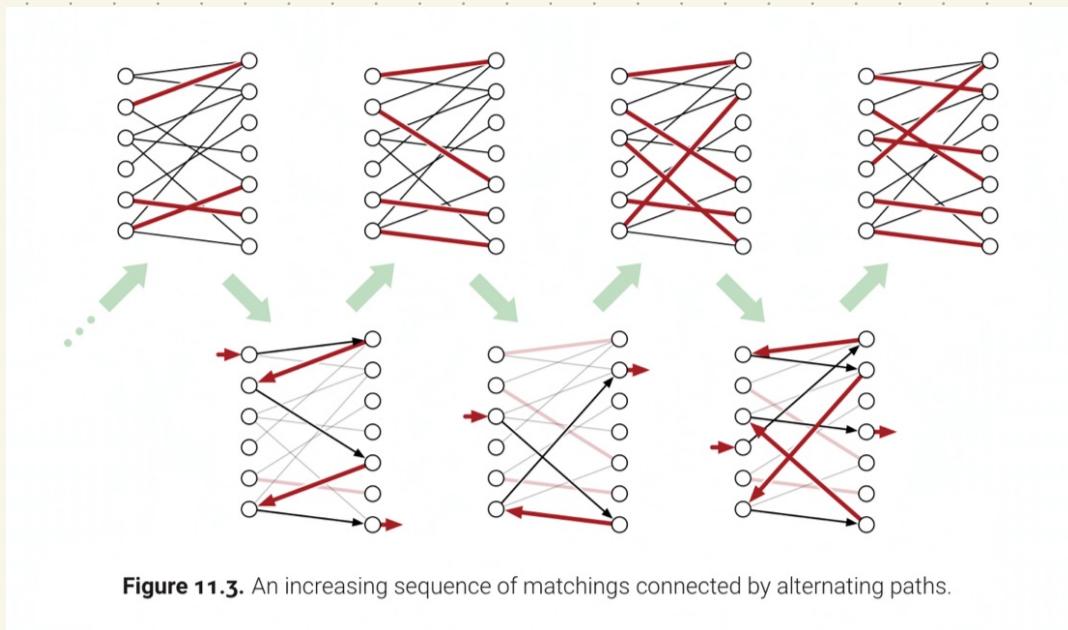
① Any matching in  $\tilde{G}$   
 $\Rightarrow$  flow in  $\tilde{G}$

② Any flow in  $\tilde{G} \Rightarrow$  matching  
in  $G$

therefore, max flow  $\Leftrightarrow$  max  
matching  
in  $G$

Aside: How??

(FF is somehow improving matching...)



## Crazier "word problem" examples

A company sells ~~to~~ products,  
~~& keeps~~ records on customers.

Goal: Design a survey to send  
to  $n$  customers, to  
get feed back.

- Each customer's survey  
shouldn't be too long,  
& should ask only about  
products they purchased
- Each product needs  
some # of reviews  
from different customers

- Input:
- $k$  products
  - $n$  customers
  - records of who bought what:  $a_{ij}$  for  $i \leq k, j \leq n$
  - For each customer,

$c_i$  is max # of products to ask them about

- for each product,  $p_i$  is minimum # of reviews needed

Can we design a survey?

Algorithm

# Runtime + correctness