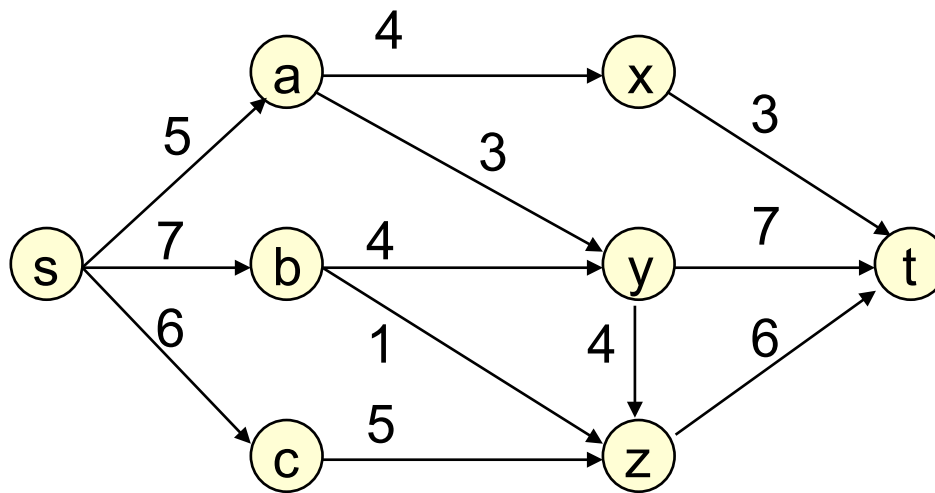

CSE 421

Introduction to Algorithms

Summer 2011

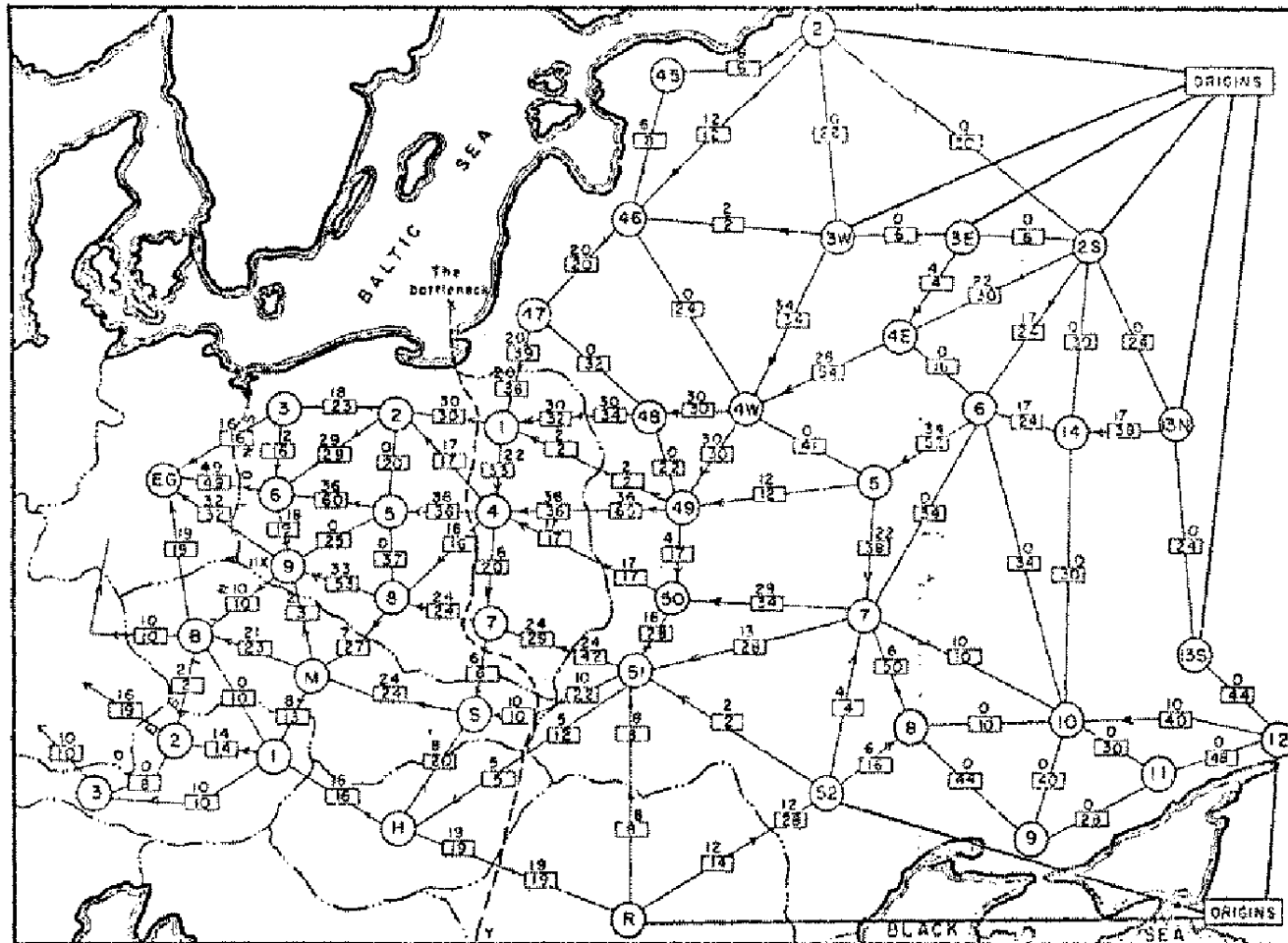
The Network Flow Problem

The Network Flow Problem



How much stuff can flow from s to t?

Soviet Rail Network, 1955



Reference: *On the history of the transportation and maximum flow problems.*
Alexander Schrijver in *Math Programming*, 91: 3, 2002.

Net Flow: Formal Definition

Given:

A digraph $G = (V, E)$

Two vertices s, t in V
(**source & sink**)

A **capacity** $c(u, v) \geq 0$
for each $(u, v) \in E$
(and $c(u, v) = 0$ for all non-edges (u, v))

Find:

A **flow function** $f: V \times V \rightarrow \mathbb{R}$ s.t.,
for all u, v :

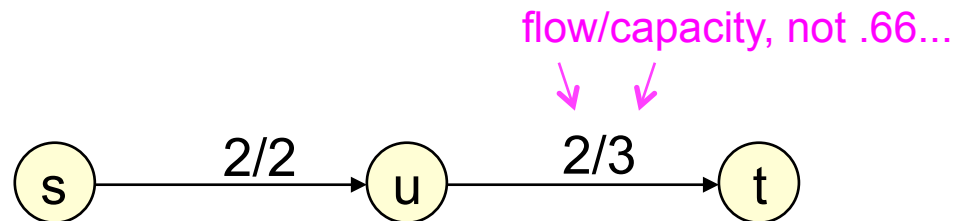
- $f(u, v) \leq c(u, v)$ [Capacity Constraint]
- $f(u, v) = -f(v, u)$ [Skew Symmetry]
- if $u \neq s, t$, $f(u, V) = 0$ [Flow Conservation]

Maximizing total flow $|f| = f(s, V)$

Notation:

$$f(X, Y) = \sum_{x \in X} \sum_{y \in Y} f(x, y)$$

Example: A Flow Function



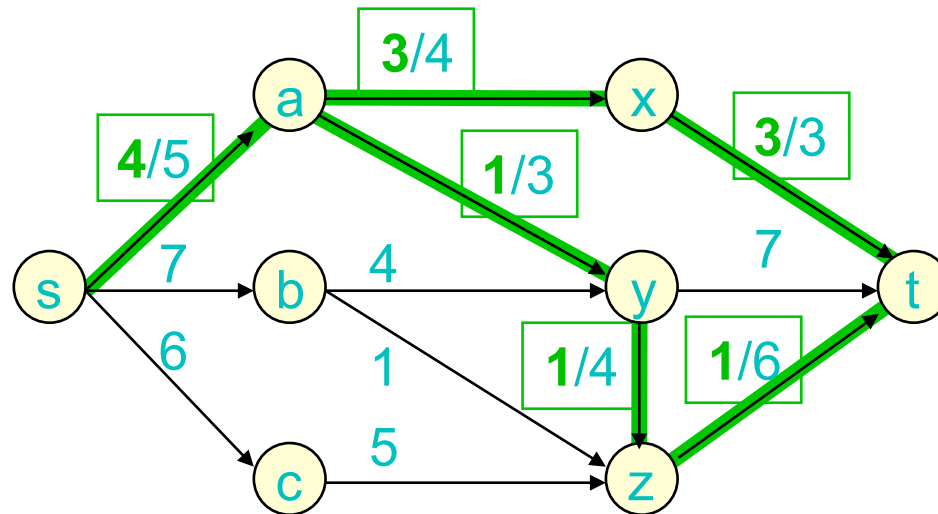
$$f(s,u) = f(u,t) = 2$$

$$f(u,s) = f(t,u) = -2 \quad (\text{Why?})$$

$$f(s,t) = -f(t,s) = 0 \quad (\text{In every flow function for this } G. \text{ Why?})$$

$$f(u,V) = \sum_{v \in V} f(u,v) = f(u,s) + f(u,t) = -2 + 2 = 0$$

Example: A Flow Function



- Not shown: $f(u,v)$ if ≤ 0
- Note: $\max \text{ flow} \geq 4$ since f is a flow function, with $|f| = 4$

Max Flow via a Greedy Alg?

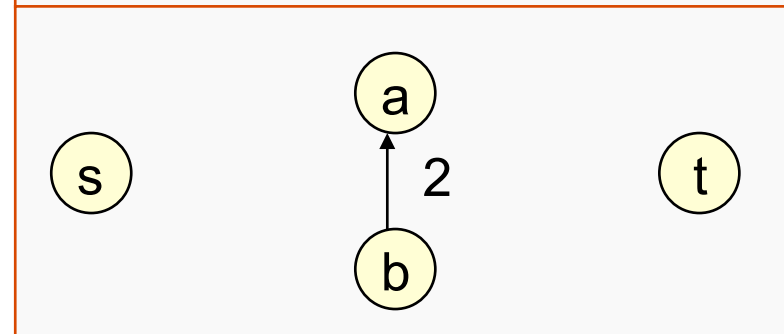
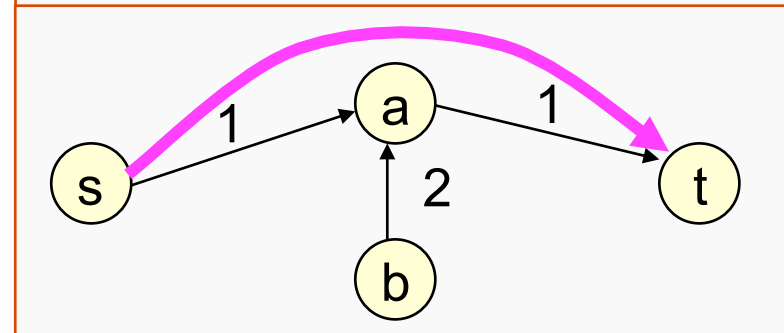
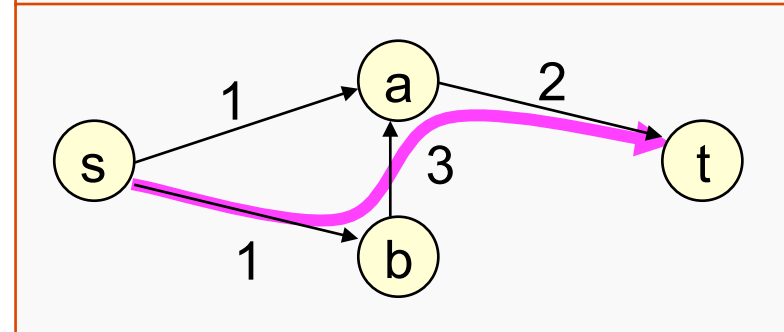
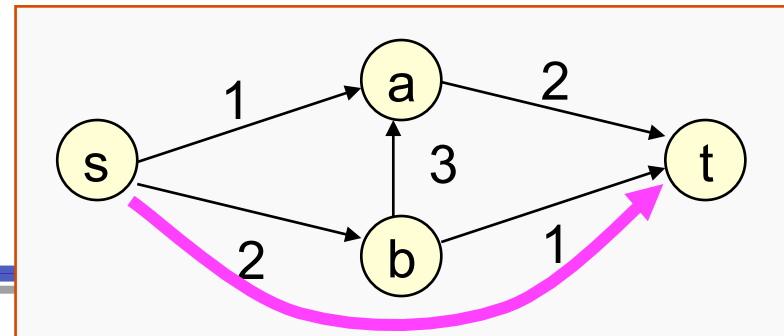
While there is an $s \rightarrow t$ path in G

Pick such a path, p

Find c_p , the min capacity of any edge in p

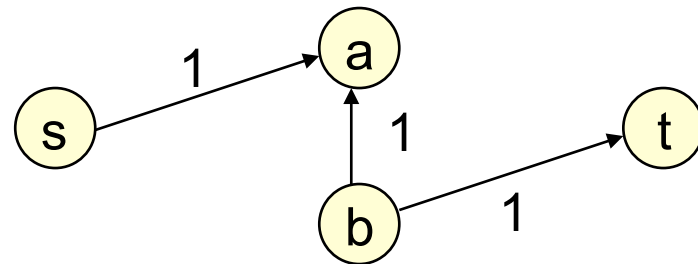
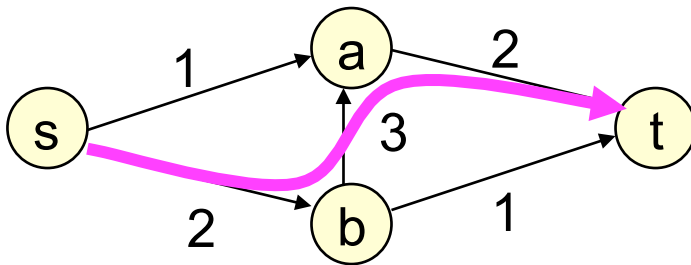
Subtract c_p from all capacities on p

Delete edges of capacity 0



Max Flow via a Greedy Alg?

This does **NOT** always find a max flow:
If you pick $s \rightarrow b \rightarrow a \rightarrow t$ first,



Flow stuck at 2. But flow 3 possible.

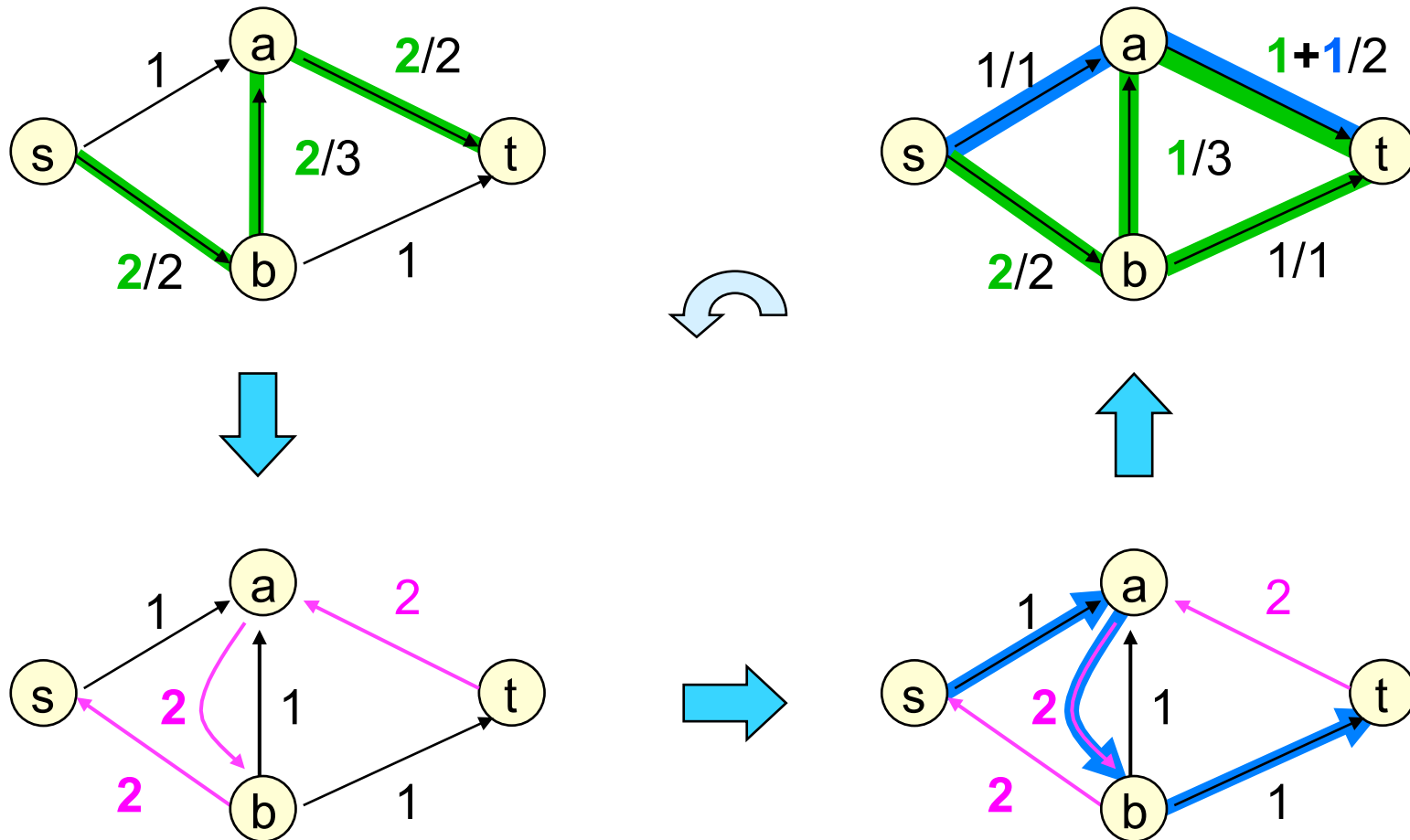
A Brief History of Flow

#	Year	Discoverer(s)	Bound
1	1951	Dantzig	$O(n^2mU)$
2	1955	Ford & Fulkerson	$O(nmU)$
3	1970	Dinitz; Edmonds & Karp	$O(nm^2)$
4	1970	Dinitz	$O(n^2m)$
5	1972	Edmonds & Karp; Dinitz	$O(m^2 \log U)$
6	1973	Dinitz; Gabow	$O(nm \log U)$
7	1974	Karzanov	$O(n^3)$
8	1977	Cherkassky	$O(n^2 \sqrt{m})$
9	1980	Galil & Naamad	$O(nm \log^2 n)$
10	1983	Sleator & Tarjan	$O(nm \log n)$
11	1986	Goldberg & Tarjan	$O(nm \log(n^2/m))$
12	1987	Ahuja & Orlin	$O(nm + n^2 \log U)$
13	1987	Ahuja et al.	$O(nm \log(n \sqrt{\log U})/(m+2))$
14	1989	Cheriyani & Hagerup	$E(nm + n^2 \log^2 n)$
15	1990	Cheriyani et al.	$O(n^3/\log n)$
16	1990	Alon	$O(nm + n^{8/3} \log n)$
17	1992	King et al.	$O(nm + n^{2+\epsilon})$
18	1993	Phillips & Westbrook	$O(nm(\log_{m/n} n + \log^{2+\epsilon} n))$
19	1994	King et al.	$O(nm(\log_{m/(n \log n)} n))$
20	1997	Goldberg & Rao	$O(m^{3/2} \log(n^2/m) \log U) ; O(n^{2/3} m \log(n^2/m) \log U)$
...

n = # of vertices
 m = # of edges
 U = Max capacity

Source: Goldberg & Rao, FOCS '97

Greed Revisited



Residual Capacity

The *residual capacity* (w.r.t. f) of (u,v) is
 $c_f(u,v) = c(u,v) - f(u,v)$

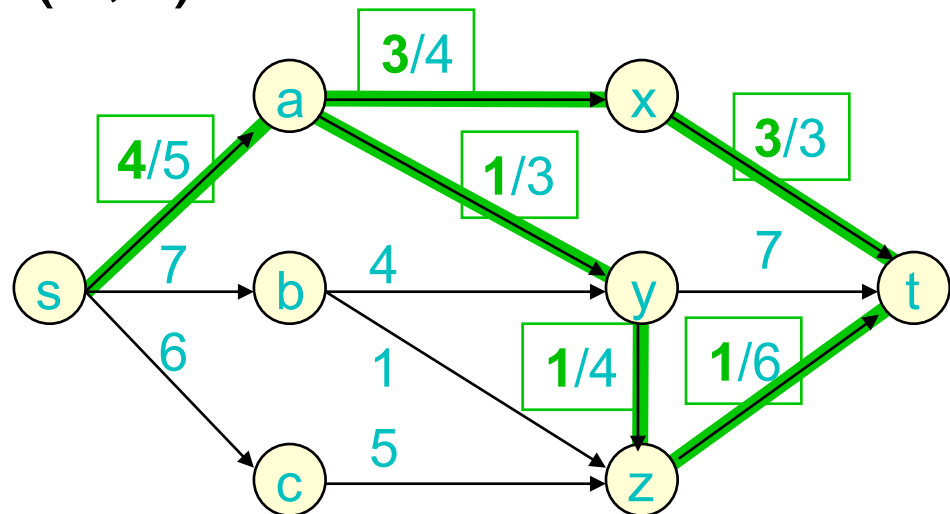
E.g.:

$$c_f(s,b) = 7;$$

$$c_f(a,x) = 1;$$

$$c_f(x,a) = 3;$$

$$c_f(x,t) = 0 \text{ (a saturated edge)}$$



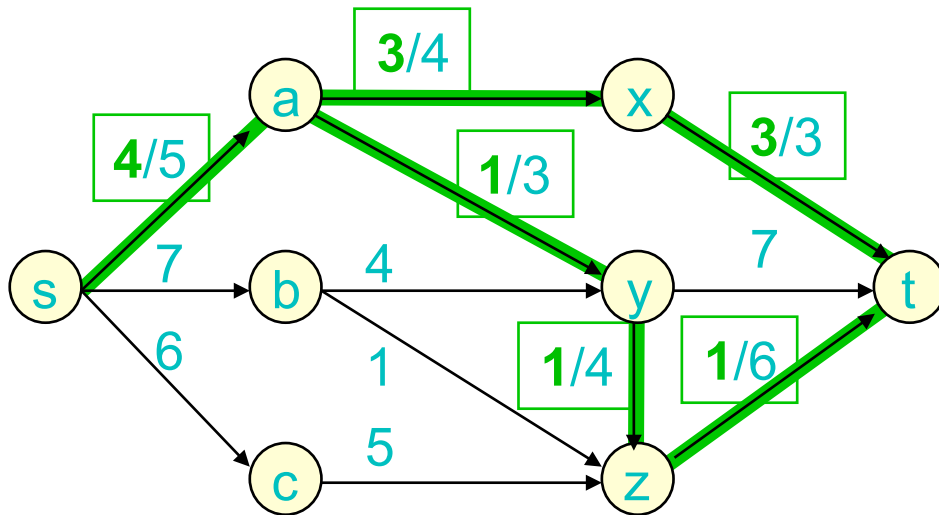
Residual Networks & Augmenting Paths

The *residual network* (w.r.t. f) is the graph $G_f = (V, E_f)$, where

$$E_f = \{ (u, v) \mid c_f(u, v) > 0 \}$$

An *augmenting path* (w.r.t. f) is a simple $s \rightarrow t$ path in G_f .

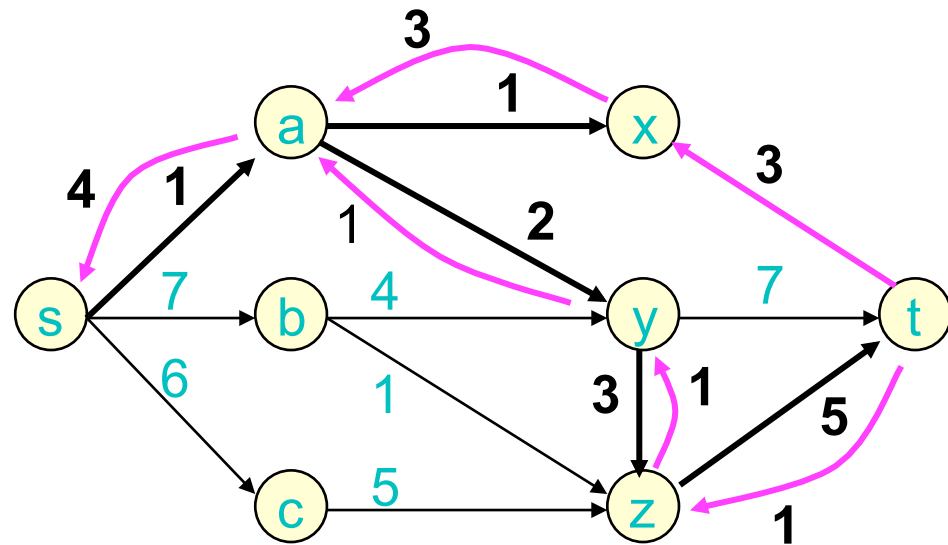
A Residual Network



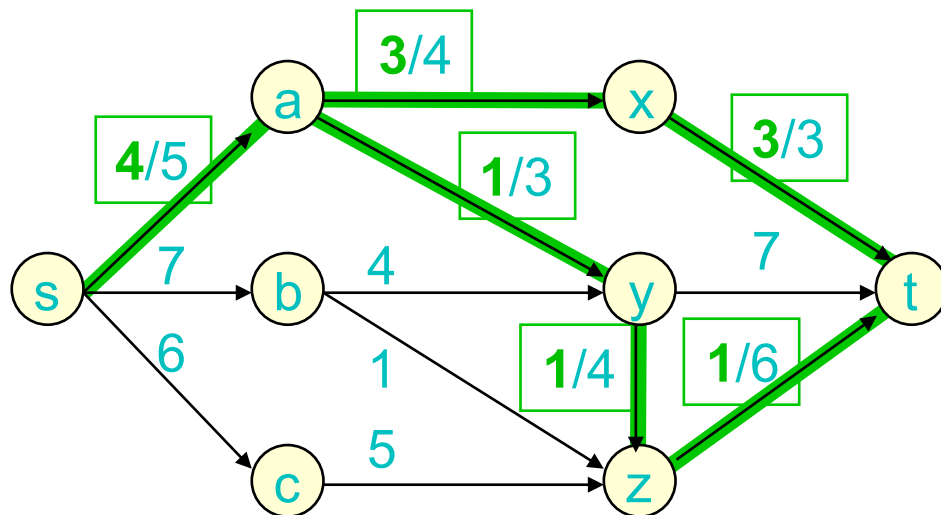
residual network: the graph

$G_f = (V, E_f)$, where

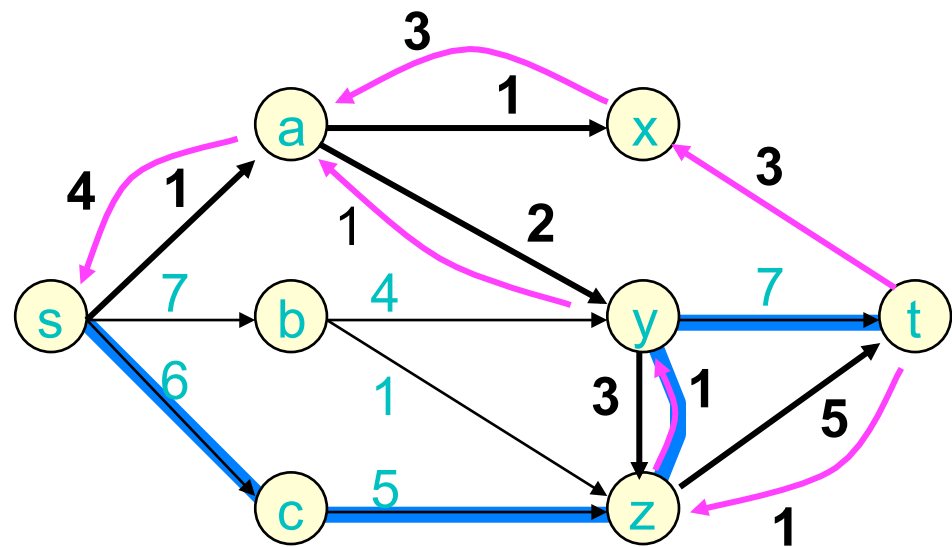
$E_f = \{ (u,v) \mid c_f(u,v) > 0 \}$



An Augmenting Path



augmenting path:
a simple $s \rightarrow t$ path in G_f .

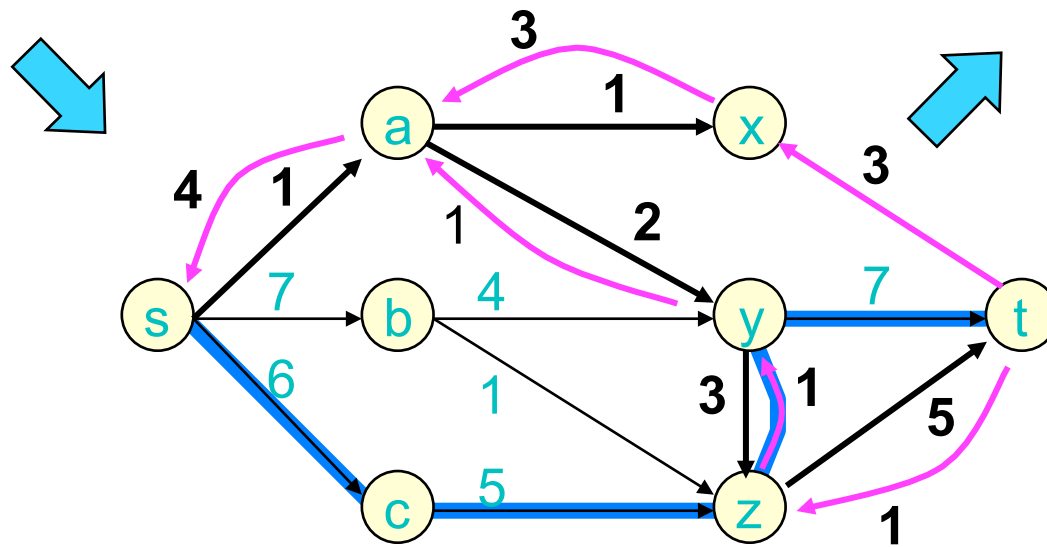
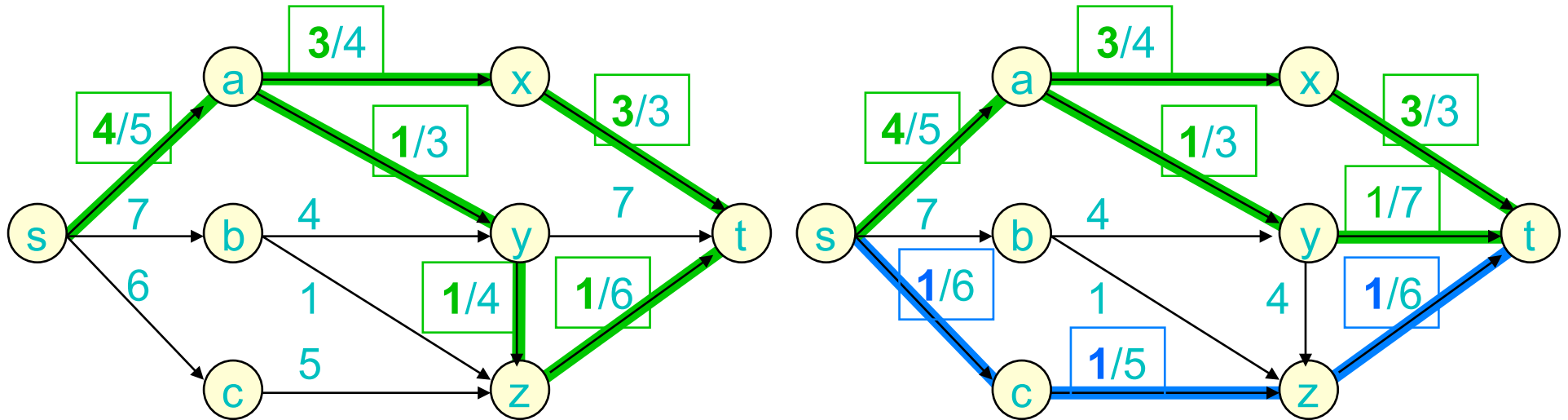


Lemma 1

If f admits an augmenting path p , then f is not maximal.

Proof: “obvious” -- augment along p by c_p , the min residual capacity of p 's edges.

Augmenting A Flow



Lemma 1': Augmented Flows are Flows

If f is a flow & p an augmenting path of capacity c_p , then f' is also a valid flow, where

$$f'(u, v) = \begin{cases} f(u, v) + c_p, & \text{if } (u, v) \text{ in path } p \\ f(u, v) - c_p, & \text{if } (v, u) \text{ in path } p \\ f(u, v), & \text{otherwise} \end{cases}$$

Proof:

- a) Flow conservation – easy
- b) Skew symmetry – easy
- c) Capacity constraints – pretty easy

Lma 1': Augmented Flows are Flows

$$f'(u,v) = \begin{cases} f(u,v) + c_p, & \text{if } (u,v) \text{ in path } p \\ f(u,v) - c_p, & \text{if } (v,u) \text{ in path } p \\ f(u,v), & \text{otherwise} \end{cases}$$

f a flow & p an aug path of cap c_p , then f' also a valid flow.

Proof (Capacity constraints):

$(u,v), (v,u)$ not on path: no change

(u,v) on path:

$$f'(u,v) = f(u,v) + c_p$$

$$\leq f(u,v) + c_f(u,v)$$

$$= f(u,v) + c(u,v) - f(u,v)$$

$$= c(u,v)$$

$$f'(v,u) = f(v,u) - c_p$$

$$< f(v,u)$$

$$\leq c(v,u)$$

Residual Capacity:

$$0 < c_p \leq c_f(u,v) = c(u,v) - f(u,v)$$

Cap Constraints:

$$-c(v,u) \leq f(u,v) \leq c(u,v)$$

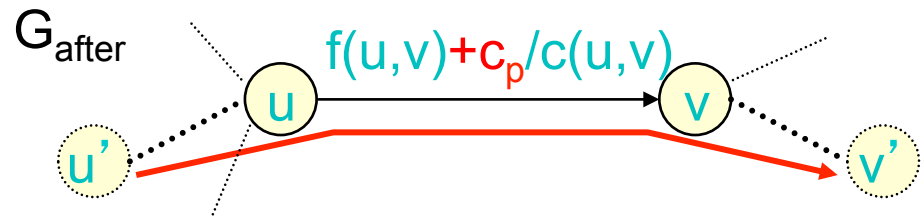
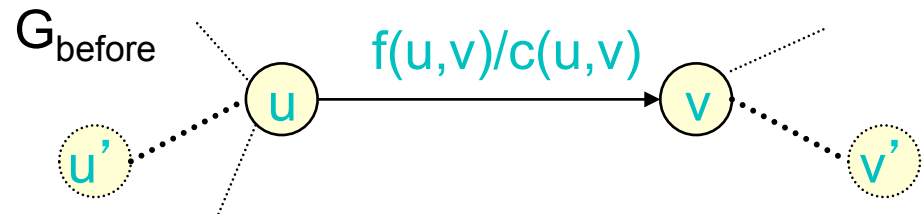
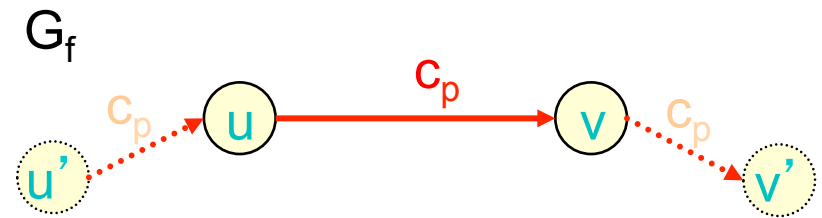
Lemma 1' Example – Case 1

Let (u,v) be any edge in augmenting path. Note

$$c_f(u,v) = c(u,v) - f(u,v) \geq c_p > 0$$

Case 1: $f(u,v) \geq 0$:

Add forward flow

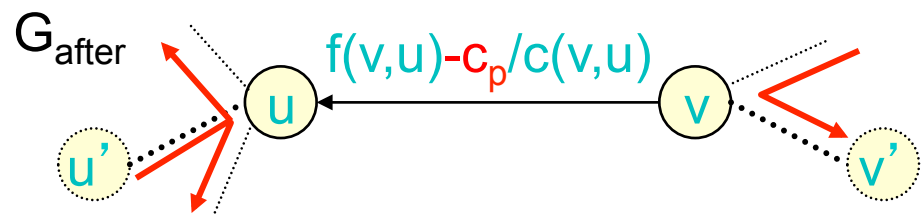
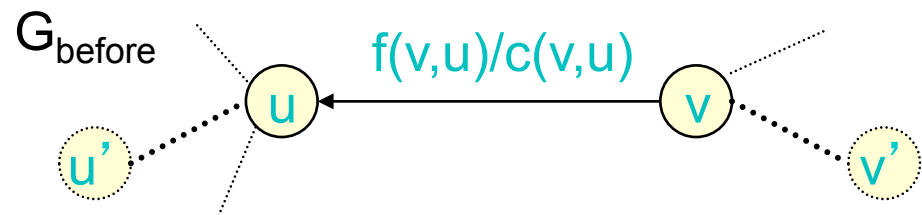
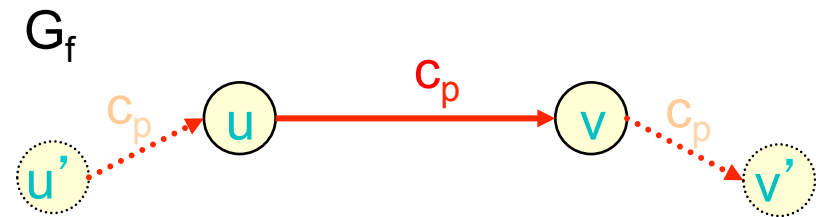


Lemma 1' Example – Case 2

Let (u,v) be any edge in augmenting path. Note $c_f(u,v) = c(u,v) - f(u,v) \geq c_p > 0$

Case 2: $f(u,v) \leq -c_p$:
 $f(v,u) = -f(u,v) \geq c_p$

Cancel/redirect reverse flow

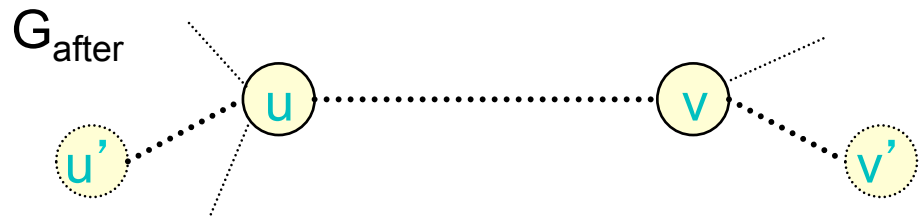
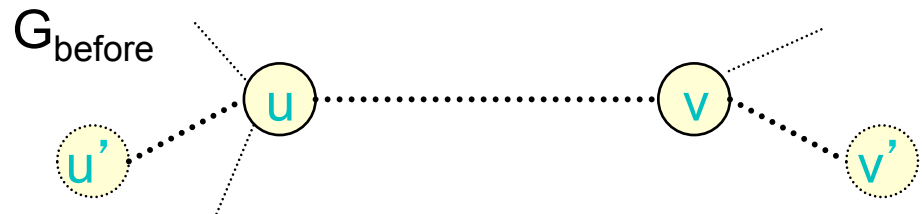
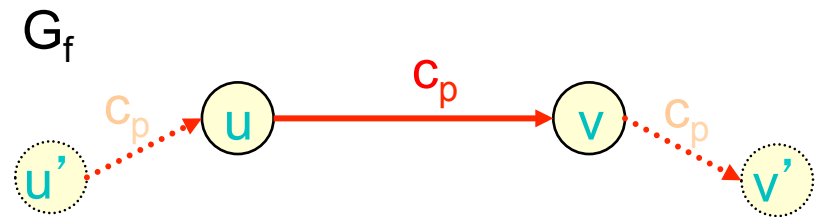


Lemma 1' Example – Case 3

Let (u,v) be any edge in augmenting path. Note $c_f(u,v) = c(u,v) - f(u,v) \geq c_p > 0$

Case 3: $-c_p < f(u,v) < 0$:

???



Lemma 1' Example – Case 3

Let (u,v) be any edge in augmenting path. Note $c_f(u,v) = c(u,v) - f(u,v) \geq c_p > 0$

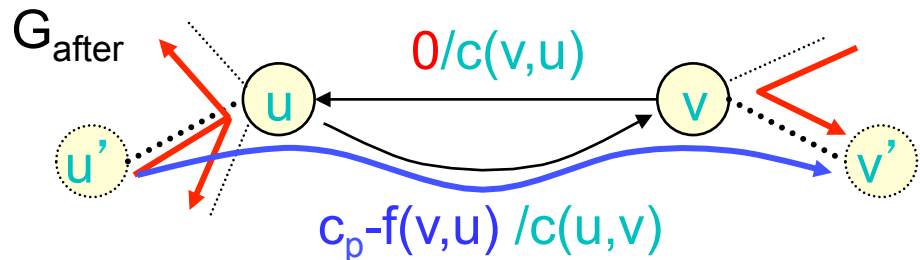
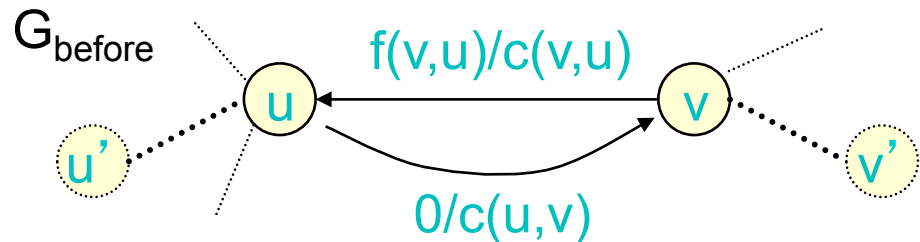
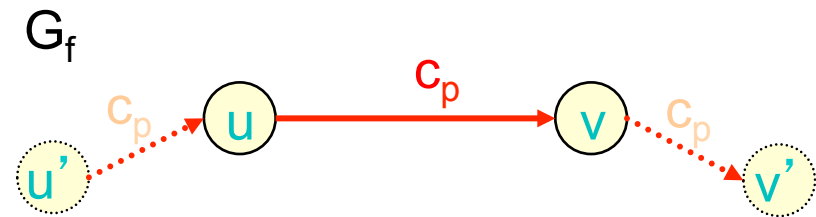
Case 3: $-c_p < f(u,v) < 0$
 $c_p > f(v,u) > 0$:

Both:

cancel/redirect
 reverse flow

and

add forward flow



Ford-Fulkerson Method

While G_f has an augmenting path,
augment

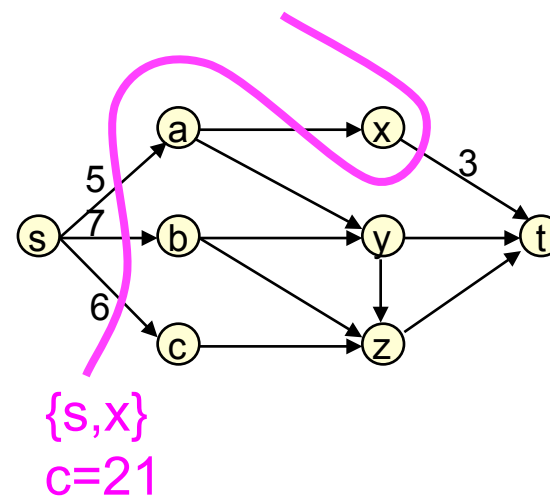
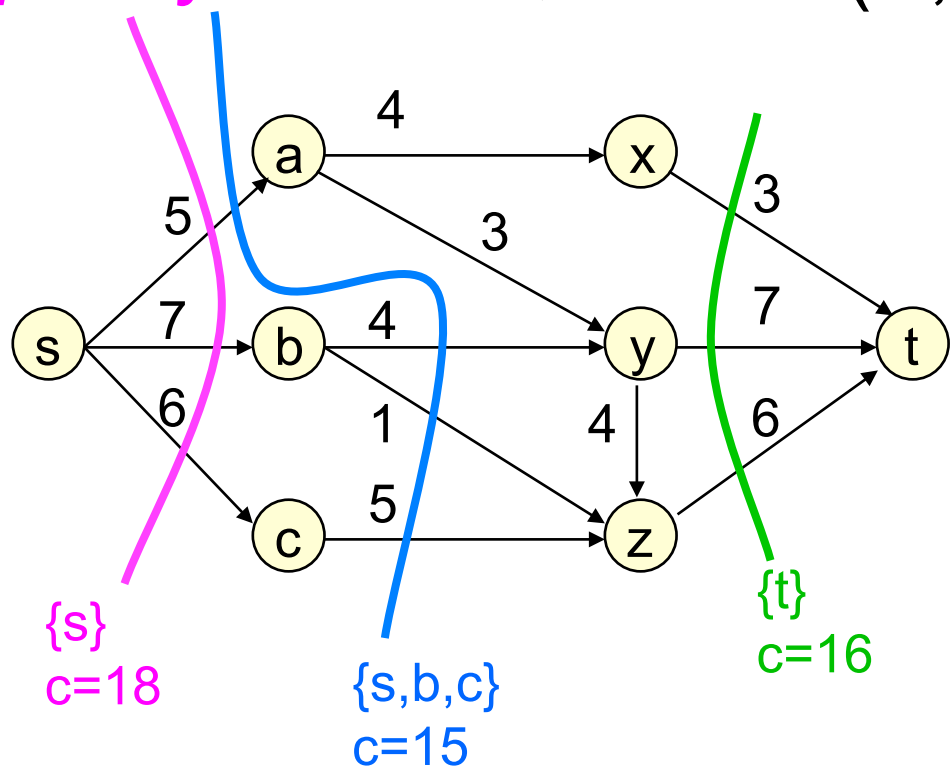
Questions:

- » Does it halt?
- » Does it find a maximum flow?
- » How fast?

Cuts

A partition S, T of V is a *cut* if $s \in S, t \in T$.

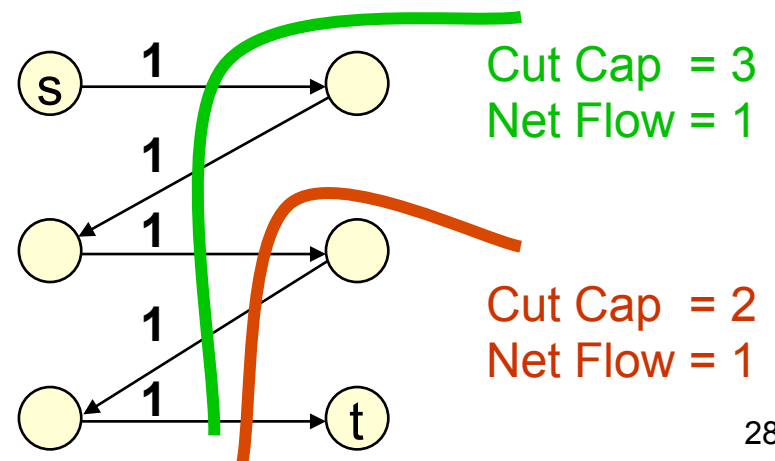
Capacity of cut S, T is $c(S, T) = \sum_{\substack{u \in S \\ v \in T}} c(u, v)$



Lemma 2

- For any flow f and any cut S, T ,
 - the net flow across the cut equals the total flow, i.e., $|f| = f(S, T)$, and
 - the net flow across the cut cannot exceed the capacity of the cut, i.e. $f(S, T) \leq c(S, T)$

- Corollary:
Max flow \leq Min cut



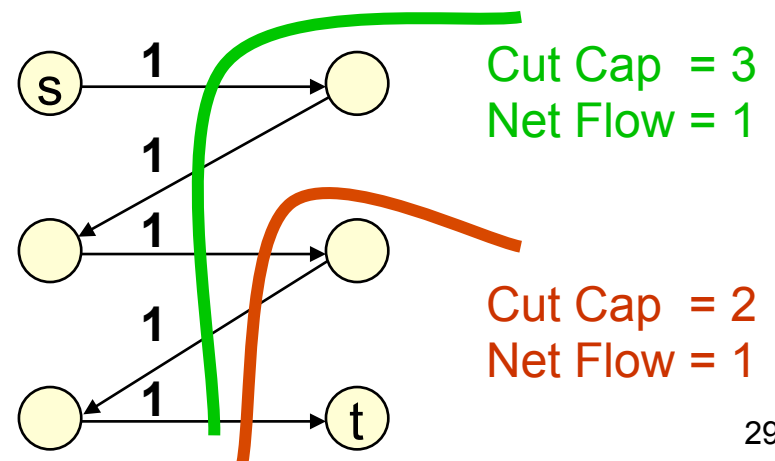
Lemma 2

For any flow f and any cut S, T ,
net flow across cut = total flow \leq cut capacity

Proof:

Track a flow unit. Starts at s , ends at t .
crosses cut an odd # of times; net = 1.

Last crossing uses a
forward edge totaled
in $C(S, T)$



Max Flow / Min Cut Theorem

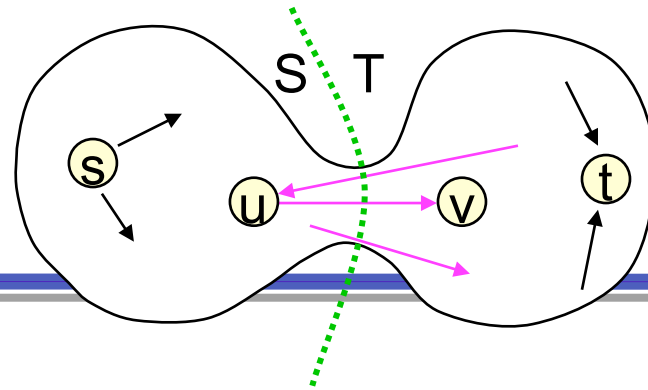
For any flow f , the following are equivalent

- (1) $|f| = c(S,T)$ for some cut S,T (a min cut)
- (2) f is a maximum flow
- (3) f admits no augmenting path

Proof:

- (1) \Rightarrow (2): corollary to lemma 2
- (2) \Rightarrow (3): contrapositive of lemma 1

(3) \Rightarrow (1)
 (no aug) \Rightarrow (cut)



Idea: where's bottleneck

$S = \{ u \mid \exists \text{ an augmenting path wrt } f \text{ from } s \text{ to } u \}$

$T = V - S; s \in S, t \in T$

For any (u,v) in $S \times T$, \exists an augmenting path from s to u , but **not** to v .

$\therefore (u,v)$ has 0 residual capacity:

$(u,v) \in E \Rightarrow$ saturated $f(u,v) = c(u,v)$

$(v,u) \in E \Rightarrow$ no flow $f(u,v) = 0 = -f(v,u)$

This is true for every edge crossing the cut, i.e.

$$|f| = f(S,T) = \sum_{u \in S} \sum_{v \in T} f(u,v) =$$

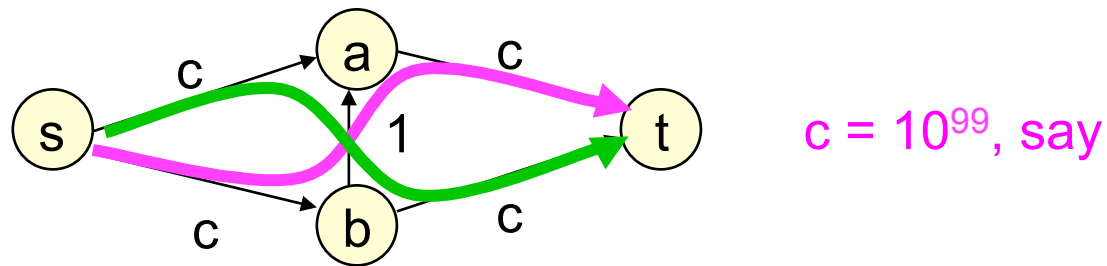
$$\sum_{u \in S, v \in T, (u,v) \in E} f(u,v) = \sum_{u \in S, v \in T, (u,v) \in E} c(u,v) = c(S,T)$$

Corollaries & Facts

If Ford-Fulkerson terminates, then it's found a max flow.

It will terminate if $c(e)$ integer or rational (but may not if they're irrational).

However, may take exponential time, even with integer capacities:



Edmonds-Karp Algorithm

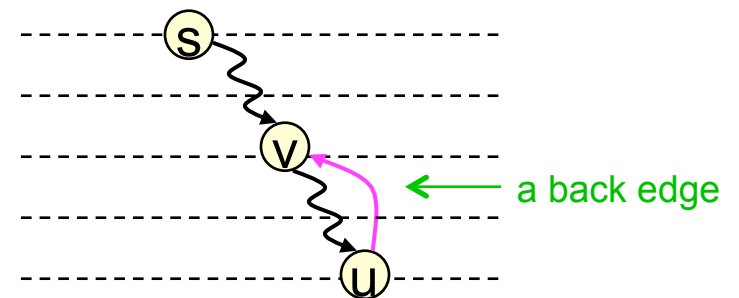
Use a **shortest** augmenting path
(via Breadth First Search in residual graph)

Time: $O(n m^2)$

BFS/Shortest Path Lemmas

Distance from s is never reduced by:

- **Deleting** an edge
proof: no new (hence no shorter) path created
- **Adding** an edge (u,v) , **provided** v is nearer than u
proof: BFS is unchanged, since v visited before (u,v) examined

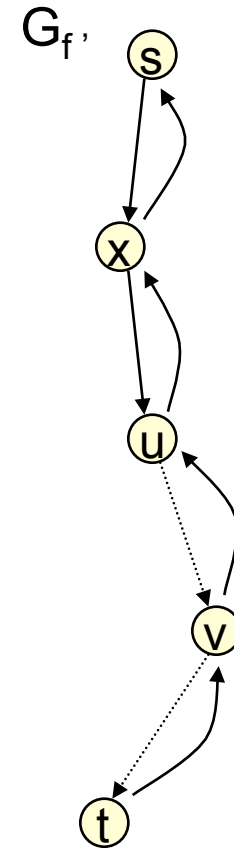
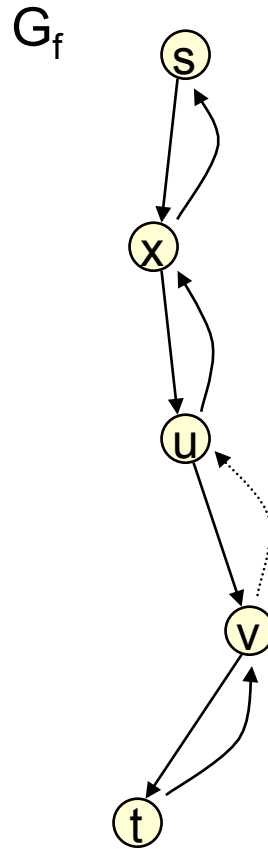
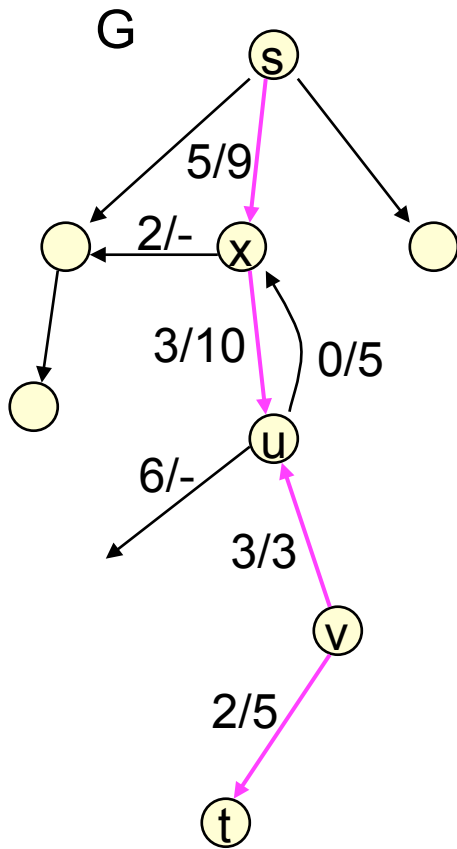


Lemma 3

Let f be a flow, G_f the residual graph, and p a shortest augmenting path. Then no vertex is closer to s after augmentation along p .

Proof: Augmentation only deletes edges, adds back edges

Augmentation vs BFS



Theorem 2

The Edmonds-Karp Algorithm performs $O(mn)$ flow augmentations

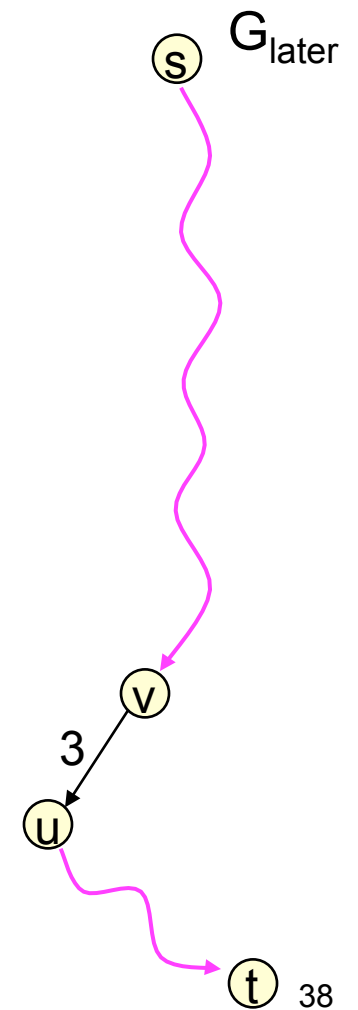
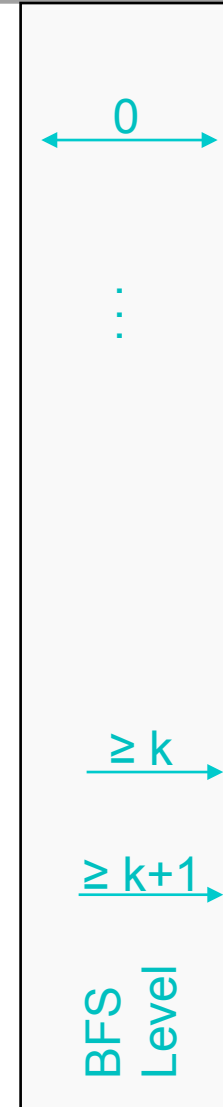
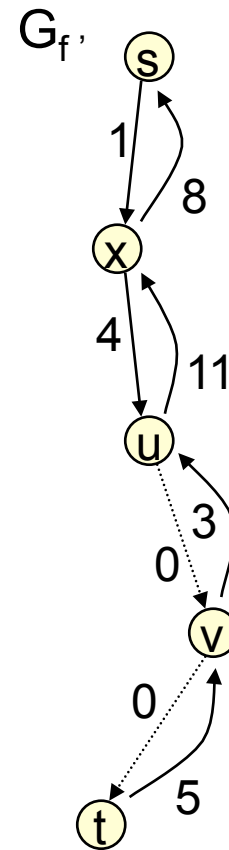
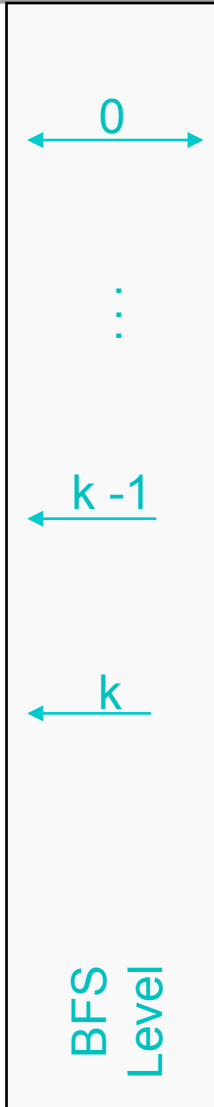
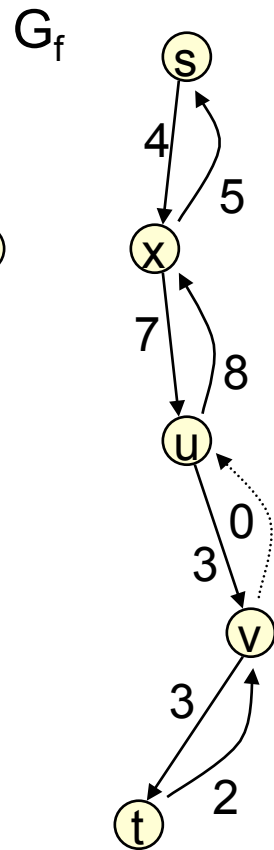
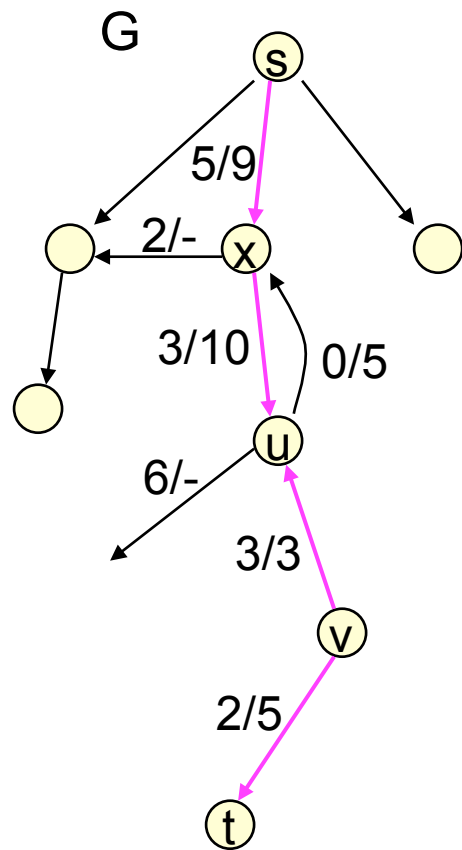
Proof:

$\{u, v\}$ is **critical** on augmenting path p if it's closest to s having min residual capacity.

Won't be critical again until farther from s .

So each edge critical at most n times.

Augmentation vs BFS Level



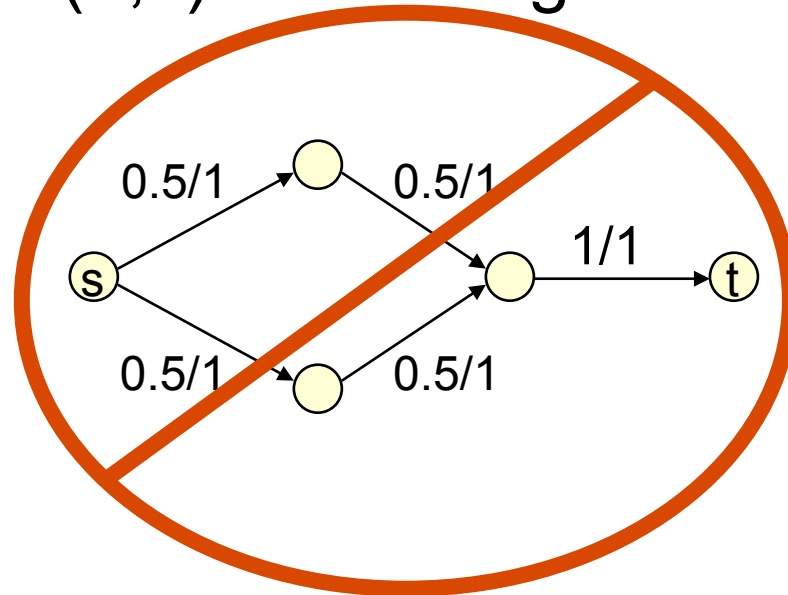
Corollary

Edmonds-Karp runs in $O(nm^2)$

Flow Integrality Theorem

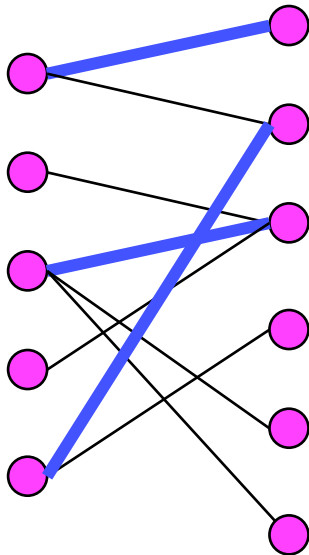
If all capacities are integers

- » The max flow has an integer value
- » Ford-Fulkerson method finds a max flow in which $f(u,v)$ is an integer for all edges (u,v)



A valid flow,
but unnecessary

Bipartite Maximum Matching



Bipartite Graphs:

- $G = (V, E)$
- $V = L \cup R$ ($L \cap R = \emptyset$)
- $E \subseteq L \times R$

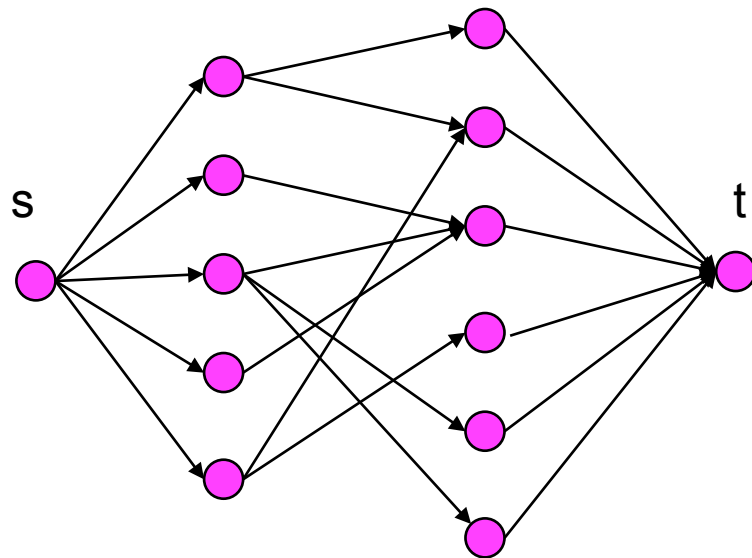
Matching:

- A set of edges $M \subseteq E$ such that no two edges touch a common vertex

Problem:

- Find a matching M of maximum size

Reducing Matching to Flow



Given bipartite G , build flow network N as follows:

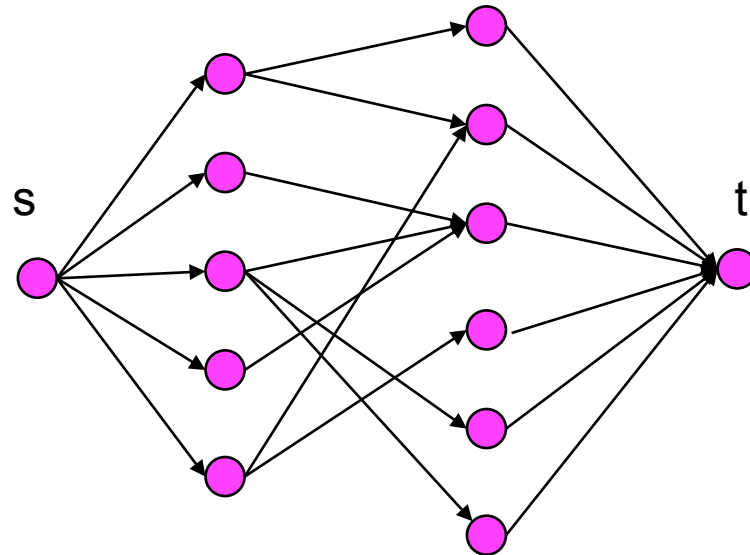
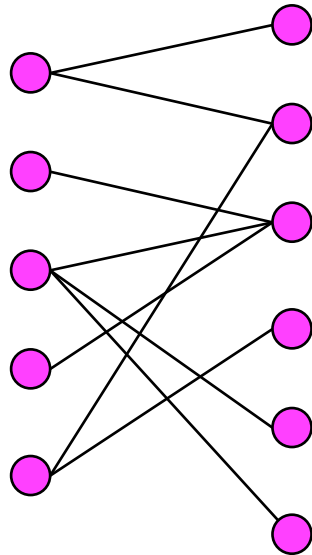
- Add source s , sink t
- Add edges $s \rightarrow L$
- Add edges $R \rightarrow t$
- All edge capacities 1

Theorem:

Max flow iff
max matching

Reducing Matching to Flow

Theorem: Max matching size = max flow value



$M \rightarrow f$? Easy – send flow only through M

$f \rightarrow M$? Flow integrality Thm, + cap constraints

Notes on Matching

- Max Flow Algorithm is probably overly general here
- But most direct matching algorithms use "augmenting path" type ideas similar to that in max flow – See text & homework
- Time $mn^{1/2}$ possible via Edmonds-Karp

7.12 Baseball Elimination

Some slides by Kevin Wayne

Baseball Elimination

Team i	Wins w_i	Losses l_i	To play g_i	Against = g_{ij}			
				Atl	Phi	NY	Mon
Atlanta	83	71	8	-	1	6	1
Philly	80	79	3	1	-	0	2
New York	78	78	6	6	0	-	0
Montreal	77	82	3	1	2	0	-

Which teams have a chance of finishing the season with most wins?

- » Montreal eliminated since it can finish with at most 80 wins, but Atlanta already has 83.
- » $w_i + g_i < w_j \Rightarrow$ team i eliminated.
- » Only reason sports writers appear to be aware of.
- » Sufficient, but not necessary!

Baseball Elimination

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Which teams have a chance of finishing the season with most wins?

- » Philly can win 83, but still eliminated . . .
- » If Atlanta loses a game, then some other team wins one.

Remark. Depends on *both* **how many** games already won and left to play, *and* on **whom** they're against.

Baseball Elimination

Baseball elimination problem.

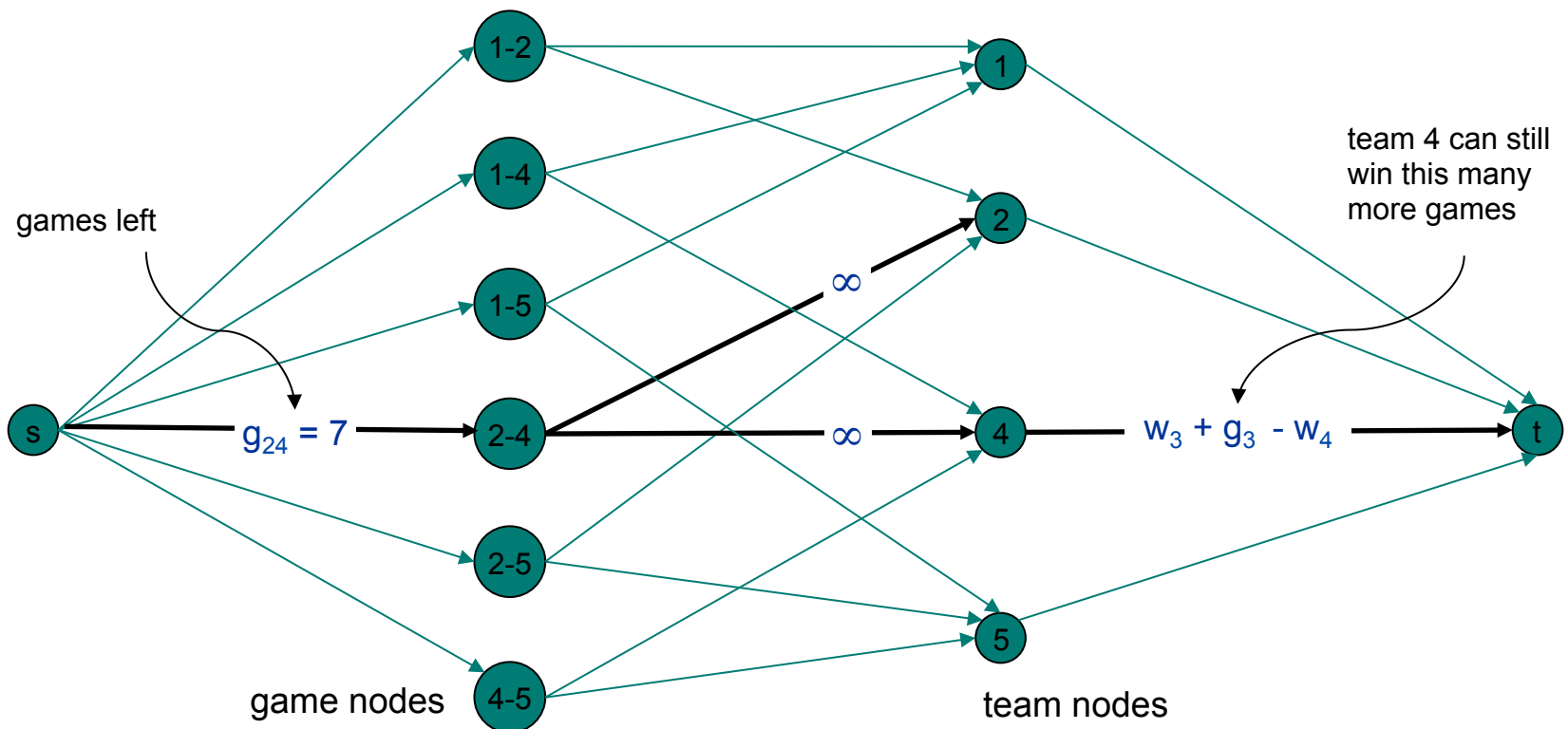
- » Set of teams S .
- » Distinguished team $s \in S$.
- » Team x has won w_x games already.
- » Teams x and y play each other g_{xy} additional times.
- » Is there any outcome of the remaining games in which team s finishes with the most (or tied for the most) wins?

Baseball Elimination: Max Flow Formulation

Can team 3 finish with most wins?

Assume team 3 wins all remaining games $\Rightarrow w_3 + g_3$ wins.

Divvy remaining games so that all teams have $\leq w_3 + g_3$ wins.

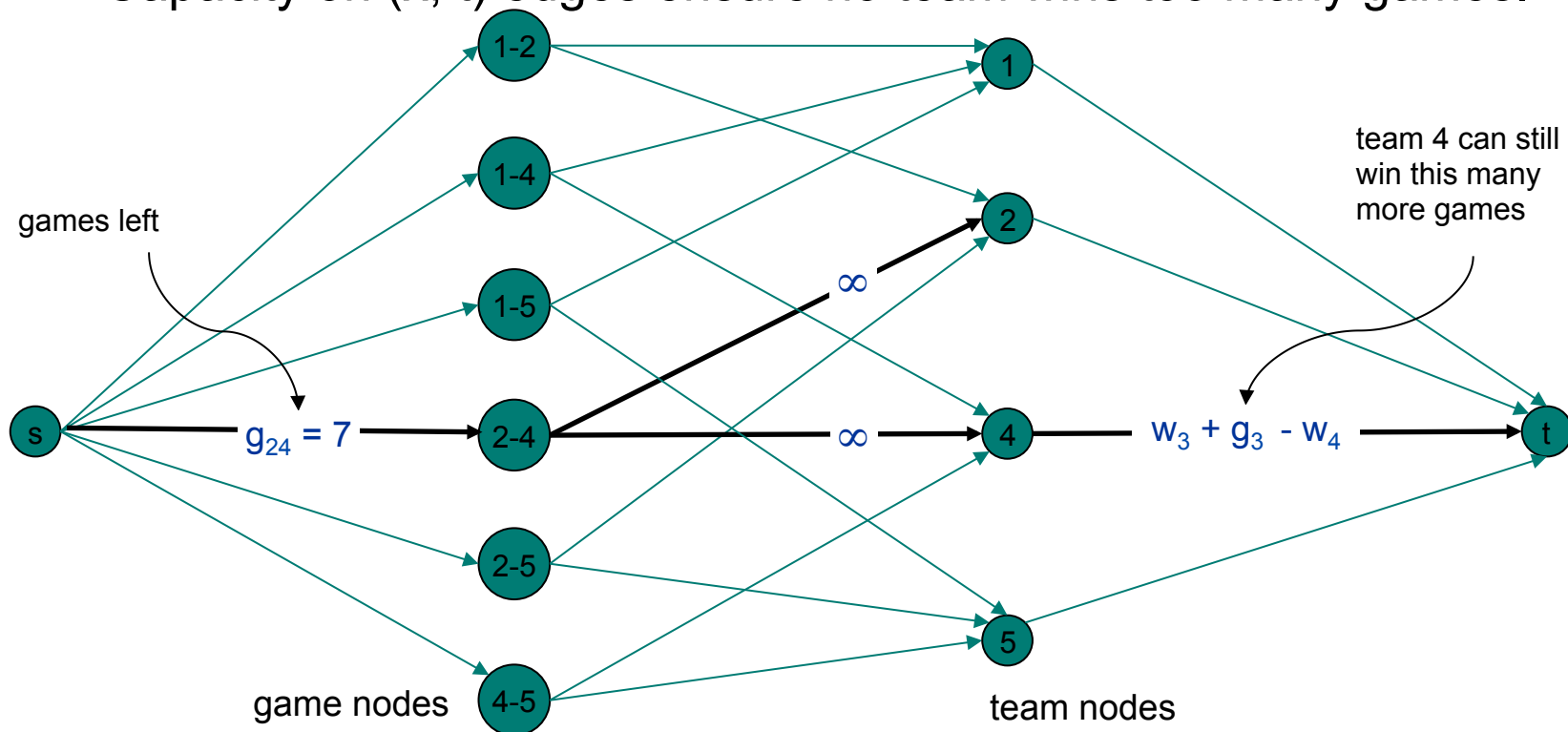


Baseball Elimination: Max Flow Formulation

Theorem. Team 3 is not eliminated iff max flow saturates all edges leaving source.

Integrality \Rightarrow each remaining x-y game added to # wins for x or y.

Capacity on (x, t) edges ensure no team wins too many games.



Baseball Elimination: Explanation for Sports Writers

Team i	Wins w_i	Losses l_i	To play g_i	Against = g_{ij}				
				NY	Bal	Bos	Tor	Det
NY	75	59	28	-	3	8	7	3
Baltimore	71	63	28	3	-	2	7	4
Boston	69	66	27	8	2	-	0	0
Toronto	63	72	27	7	7	0	-	-
Detroit	49	86	27	3	4	0	0	-

AL East: August 30, 1996

Which teams have a chance of finishing the season with most wins?

Detroit could finish season with $49 + 27 = 76$ wins.

Baseball Elimination: Explanation for Sports Writers

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AL East: August 30, 1996

Which teams could finish the season with most wins?

Detroit could finish season with $49 + 27 = 76$ wins.

Certificate of elimination. $R = \{NY, Bal, Bos, Tor\}$

Have already won $w(R) = 278$ games.

Must win at least $r(R) = 27$ more.

Average team in R wins at least $305/4 > 76$ games.

Baseball Elimination: Explanation for Sports Writers

*Certificate of
elimination*

$$T \subseteq S, \quad w(T) := \overbrace{\sum_{i \in T} w_i}^{\# \text{ wins}}, \quad g(T) := \overbrace{\sum_{\{x,y\} \subseteq T} g_{xy}}^{\# \text{ remaining games}},$$

LB on avg # games won

If $\frac{w(T) + g(T)}{|T|} > w_z + g_z$ then z **eliminated** (by subset T).



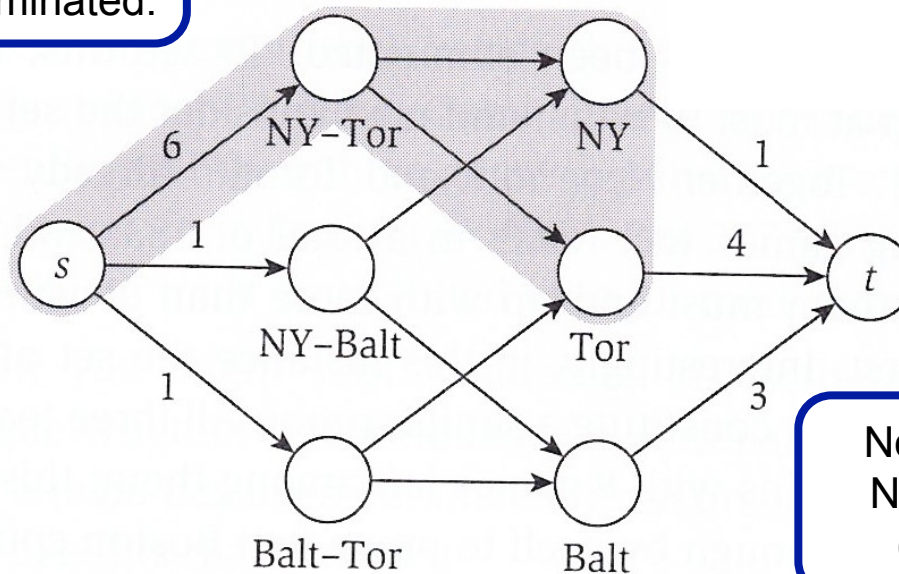
Theorem. [Hoffman-Rivlin 1967] Team z is eliminated iff there exists a subset T^* that eliminates z.

Proof idea. Let $T^* =$ teams on source side of min cut.

	w	l	g	NY	Balt	Tor	Bos
NY	90		11	-	1	6	4
Baltimore	88		6	1	-	1	4
Toronto	87		10	6	1	-	4
Boston	79		12	4	4	4	-

$$g^* = 1 + 6 + 1 = 8$$

$(90 + 87 + 6) / 2 > 91$,
so the set $T = \{NY, Tor\}$
proves Boston is eliminated.



Note: $T = \{NY, Tor, Balt\}$ is
NOT a certificate, since
 $(90 + 88 + 87 + 8) / 3 = 91$

Fig 7.21 Min cut \Rightarrow no flow of value g^* , so Boston eliminated.

Baseball Elimination: Explanation for Sports Writers

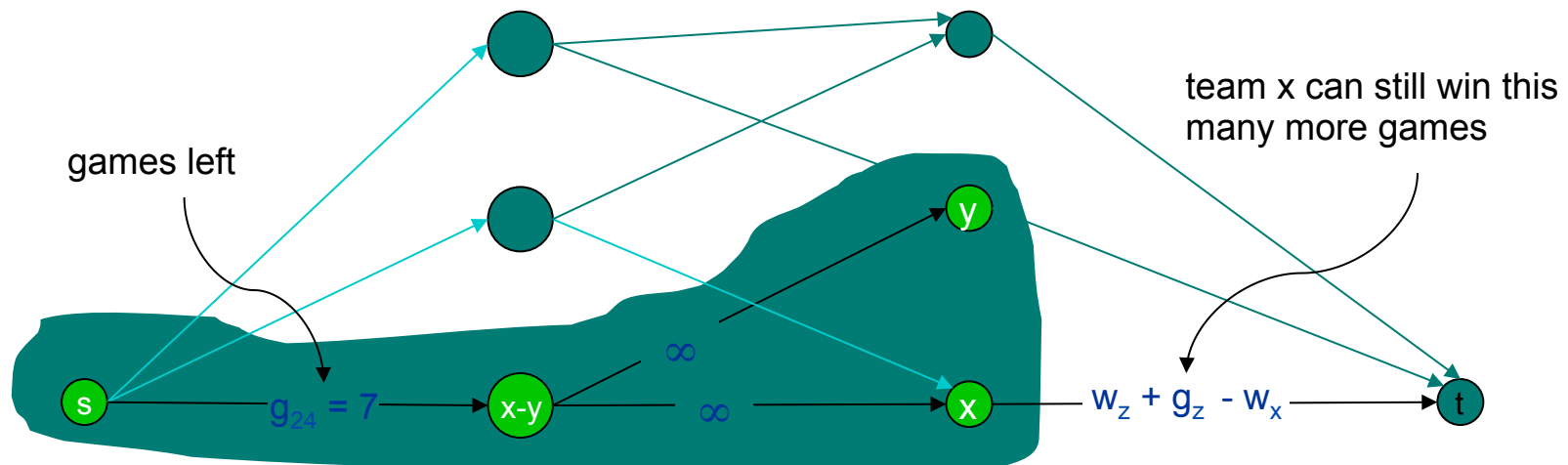
Pf of theorem.

Use max flow formulation, and consider min cut (A, B) .

Define T^* = team nodes on source side of min cut.

Observe $x-y \in A$ iff both $x \in T^*$ and $y \in T^*$.

infinite capacity edges ensure if $x-y \in A$ then $x \in A$ and $y \in A$
 if $x \in A$ and $y \in A$ but $x-y \notin T^*$, then adding $x-y$ to A decreases capacity of cut



Baseball Elimination: Explanation for Sports Writers

Pf of theorem.

Use max flow formulation, and consider min cut (A, B).

Define T^* = team nodes on source side of min cut.

Observe $x-y \in A$ iff both $x \in T^*$ and $y \in T^*$.

$$g(S - \{z\}) > \text{cap}(A, B)$$

$$\begin{aligned} &= \overbrace{g(S - \{z\}) - g(T^*)}^{\text{capacity of game edges leaving A}} + \overbrace{\sum_{x \in T^*} (w_z + g_z - w_x)}^{\text{capacity of team edges leaving A}} \\ &= g(S - \{z\}) - g(T^*) - w(T^*) + |T^*|(w_z + g_z) \end{aligned}$$

Rearranging:

$$w_z + g_z < \frac{w(T^*) + g(T^*)}{|T^*|}$$

Matching & Baseball: Key Points

- Can (sometimes) take problems that seemingly have *nothing* to do with flow & reduce them to a flow problem
- How? Build a clever network; map allocation of stuff in original problem (match edges; wins) to allocation of flow in network. Clever edge capacities constraint solution to mimic original problem in some way. Integrality useful.

Matching & Baseball: Key Points

- Furthermore, in the baseball example, min cut can be translated into a succinct *certificate* or *proof* of some property that is much more transparent than “see, I ran max-flow and it says flow must be less than g^* ”.
- These examples suggest why max flow is so important – it’s a very general tool used in many other algorithms.