

CSE 421: Introduction to Algorithms

Network Flow

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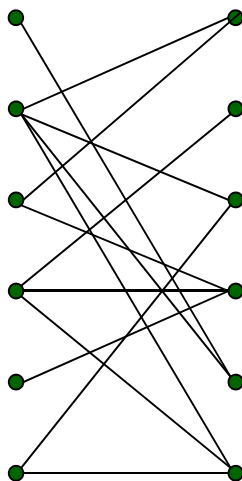
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Bipartite Matching

- **Given:** A bipartite graph $G=(V,E)$
 - $M \subseteq E$ is a matching in G iff no two edges in M share a vertex
- **Goal:** Find a matching M in G of maximum possible size

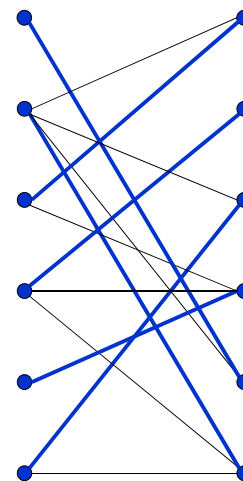
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Bipartite Matching



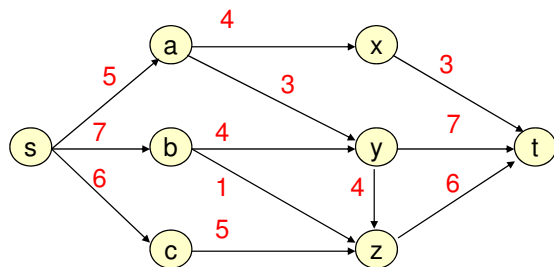
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Bipartite Matching



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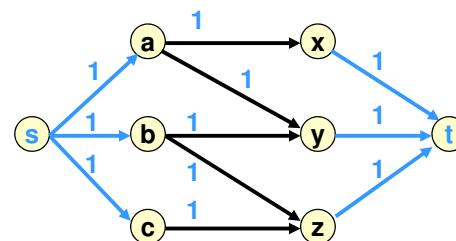
The Network Flow Problem



- How much stuff can flow from **s** to **t**?

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Bipartite matching as a special case of flow



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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** $f: E \rightarrow \mathbb{R}$ s.t., for all u, v :

- $0 \leq f(u, v) \leq c(u, v)$ [Capacity Constraint]
- if $u \neq s, t$, i.e. $f^{\text{out}}(u) = f^{\text{in}}(u)$ [Flow Conservation]

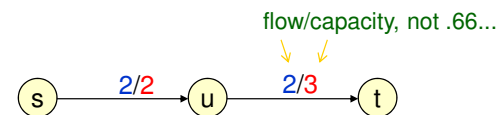
Maximizing total flow $v(f) = f^{\text{out}}(s)$

Notation:

$$f^{\text{in}}(v) = \sum_{e=(u,v) \in E} f(u, v) \quad f^{\text{out}}(v) = \sum_{e=(v,w) \in E} f(v, w)$$

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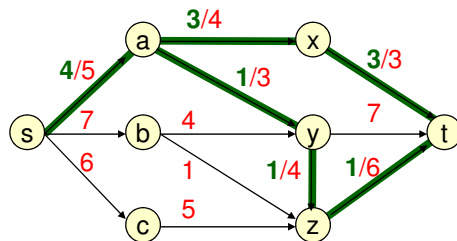
Example: A Flow Function



$$f^{\text{in}}(u) = f(s, u) = 2 = f(u, t) = f^{\text{out}}(u)$$

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Example: A Flow Function



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

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Max Flow via a Greedy Alg?

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

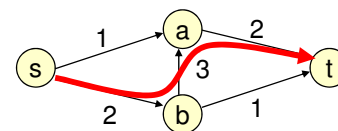
Pick such a path, p

Find c , the min capacity of any edge in p

Subtract c from all capacities on p

Delete edges of capacity 0

- This does **NOT** always find a max flow:



If pick $s \rightarrow b \rightarrow a \rightarrow t$ first, flow stuck at 2. But flow 3 possible.

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A Brief History of Flow

#	year	discoverer(s)	bound
1	1951	Dantzig	$O(n^2 m U)$
2	1955	Ford & Fulkerson	$O(nmU)$
3	1970	Dinitz	$O(nm^2)$
4	1970	Edmonds & Karp	$O(n^2 m)$
5	1972	Dinitz	$O(m^2 \log U)$
6	1973	Edmonds & Karp	$O(m^2 \log U)$
7	1974	Dinitz	$O(nm \log U)$
8	1974	Gabow	$O(nm \log U)$
9	1974	Karzanov	$O(n^3)$
10	1977	Cherkassky	$O(n^2 \sqrt{m})$
11	1980	Galil & Naamad	$O(nm \log^2 n)$
12	1983	Sleator & Tarjan	$O(nm \log n)$
13	1986	Goldberg & Tarjan	$O(nm \log(n^2/m))$
14	1987	Ahuja & Orlin	$O(nm + n^2 \log U)$
15	1987	Ahuja et al.	$O(nm \log(n \sqrt{\log U} / (m+2)))$
16	1989	Cheriyani & Hagerup	$E(nm + n^2 \log^2 n)$
17	1990	Cheriyani et al.	$O(n^3 / \log n)$
18	1990	Alon	$O(nm + n^{5/3} \log n)$
19	1992	King et al.	$O(nm + n^{2+\epsilon})$
20	1993	Phillips & Westbrook	$O(nm(\log_{m/n} n + \log^{2+\epsilon} n))$
21	1994	King et al.	$O(nm \log_{m/n} n)$
22	1997	Goldberg & Rao	$O(m^{3/2} \log(n^2/m) \log U)$
23	2012	Orlin & King et al.	$O(nm)$

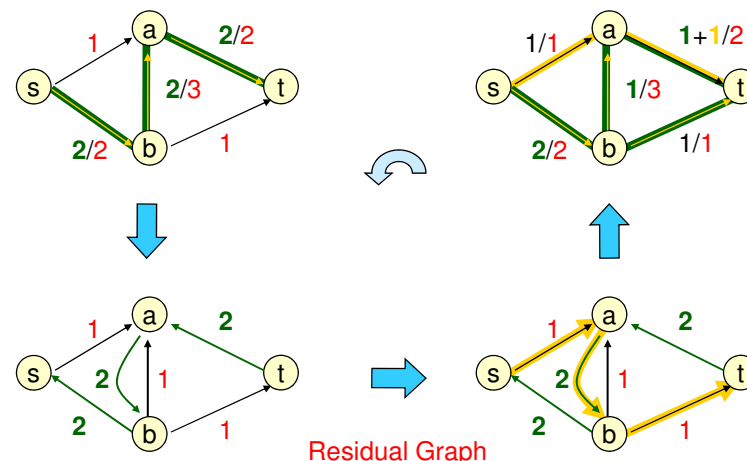
2012 Orlin & King et al. $O(nm)$

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

Source: Goldberg & Rao, FOCS '97

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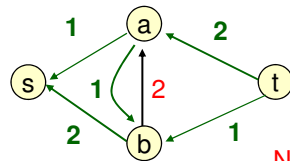
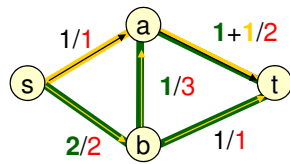
Greedy Revisited: Residual Graph & Augmenting Path



Residual Graph

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Greedy Revisited: An Augmenting Path

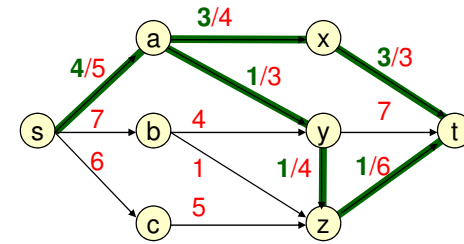


New Residual Graph

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Residual Capacity

- The *residual capacity* (w.r.t. f) of (u,v) is $c_f(u,v) = c(u,v) - f(u,v)$ if $f(u,v) \leq c(u,v)$ and $c_f(u,v) = f(v,u)$ if $f(v,u) > 0$



- e.g. $c_f(s,b)=7$; $c_f(a,x) = 1$; $c_f(x,a) = 3$

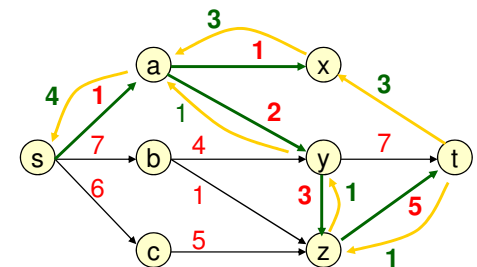
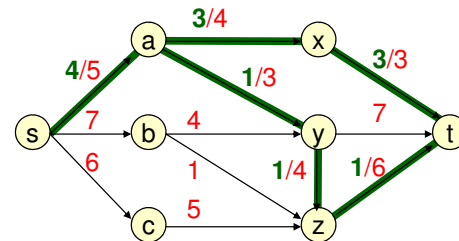
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Residual Graph & Augmenting Paths

- The *residual graph* (w.r.t. f) is the graph $G_f = (V, E_f)$, where $E_f = \{ (u,v) \mid c_f(u,v) > 0 \}$
 - Two kinds of edges
 - Forward edges
 - $f(u,v) < c(u,v)$ so $c_f(u,v) = c(u,v) - f(u,v) > 0$
 - Backward edges
 - $f(u,v) > 0$ so $c_f(v,u) \geq -f(u,v) = f(u,v) > 0$
 - An *augmenting path* (w.r.t. f) is a simple $s \rightarrow t$ path in G_f .

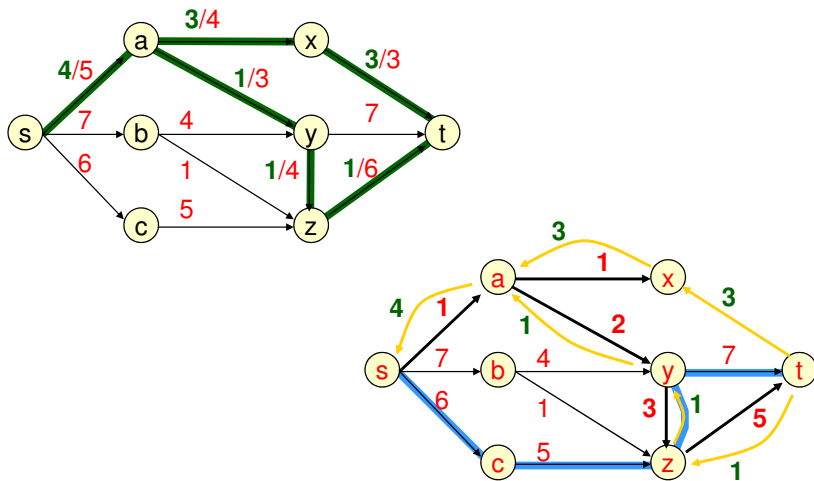
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A Residual Network



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An Augmenting Path



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Augmenting A Flow

augment(f, P)

$c_P \leftarrow \min_{(u,v) \in P} c_f(u,v)$ "bottleneck(P)"

for each $e \in P$

if e is a forward edge then

increase $f(e)$ by c_P

else (e is a backward edge)

decrease $f(e)$ by c_P

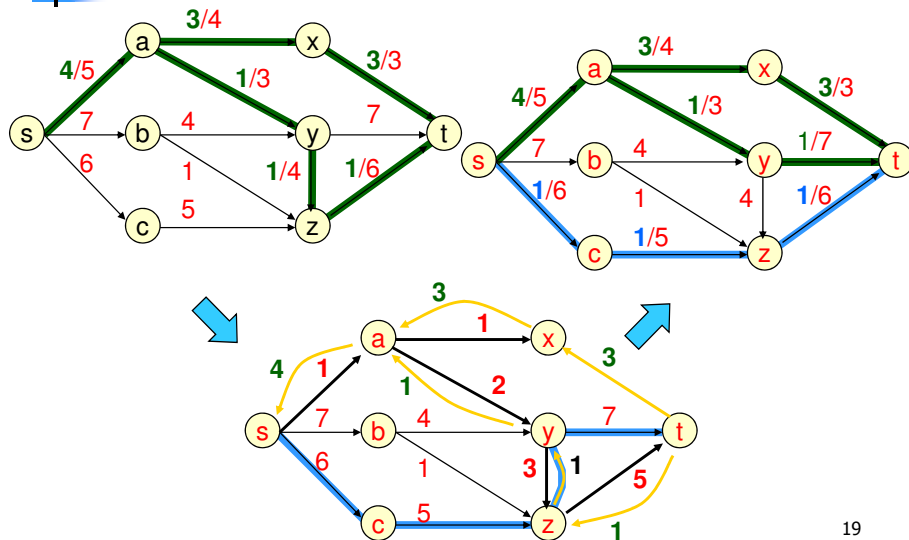
endif

endfor

return(f)

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Augmenting A Flow



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Claim 7.1

If G_f has an augmenting path P , then the function $f' = \text{augment}(f, P)$ is a legal flow.

Proof:

- f' and f differ only on the edges of P so only need to consider such edges (u,v)

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Proof of Claim 7.1

- If (u,v) is a forward edge then

$$\begin{aligned} f'(u,v) &= f(u,v) + c_p \leq f(u,v) + c_f(u,v) \\ &= f(u,v) + c(u,v) - f(u,v) \\ &= c(u,v) \end{aligned}$$
- If (u,v) is a backward edge then f and f' differ on flow along (v,u) instead of (u,v)

$$\begin{aligned} f'(v,u) &= f(v,u) - c_p \geq f(v,u) - c_f(u,v) \\ &= f(v,u) - f(v,u) = 0 \end{aligned}$$
- Other conditions like flow conservation still met

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Ford-Fulkerson Method

Start with $f=0$ for every edge

While G_f has an augmenting path, augment

- Questions:
 - Does it halt?
 - Does it find a maximum flow?
 - How fast?

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Observations about Ford-Fulkerson Algorithm

- At every stage the capacities and flow values are always integers (if they start that way)
- The flow value $v(f') = v(f) + c_p > v(f)$ for $f' = \text{augment}(f, P)$
 - Since edges of residual capacity 0 do not appear in the residual graph
- Let $C = \sum_{(s,u) \in E} c(s,u)$
 - $v(f) \leq C$
 - **F-F** does at most C rounds of augmentation since flows are integers and increase by at least 1 per step

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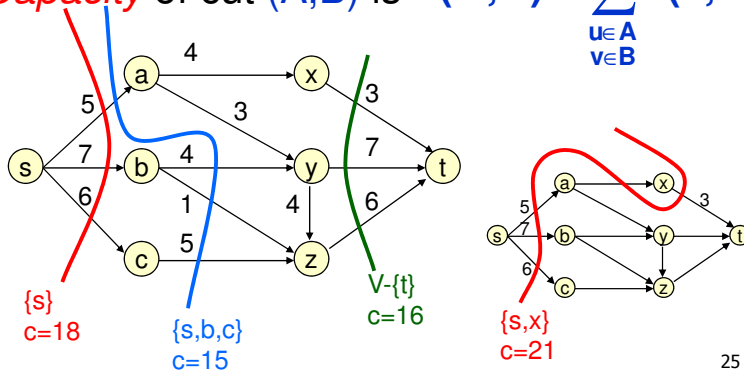
Running Time of Ford-Fulkerson

- For $f=0$, $G_f=G$
- Finding an augmenting path in G_f is graph search $O(n+m) = O(m)$ time
- Augmenting and updating G_f is $O(n)$ time
- Total $O(mC)$ time
- Does it find a maximum flow?
 - Need to show that for every flow f that isn't maximum G_f contains an s - t -path

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Cuts

- A partition (A, B) of V is an s - t -cut if
 - $s \in A, t \in B$
- Capacity** of cut (A, B) is $c(A, B) = \sum_{\substack{u \in A \\ v \in B}} c(u, v)$



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Convenient Definition

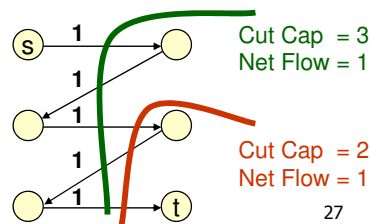
- $f^{\text{out}}(A) = \sum_{v \in A, w \notin A} f(v, w)$
- $f^{\text{in}}(A) = \sum_{v \in A, u \notin A} f(u, v)$

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Claims 7.6 and 7.8

- For any flow f and any cut (A, B) ,
 - the net flow across the cut equals the total flow, i.e., $v(f) = f^{\text{out}}(A) - f^{\text{in}}(A)$, and
 - the net flow across the cut cannot exceed the capacity of the cut, i.e. $f^{\text{out}}(A) - f^{\text{in}}(A) \leq c(A, B)$

- Corollary :**
Max flow \leq Min cut



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Proof of Claim 7.6

- Consider a set A with $s \in A, t \notin A$
- $f^{\text{out}}(A) - f^{\text{in}}(A) = \sum_{v \in A, w \notin A} f(v, w) - \sum_{v \in A, u \notin A} f(u, v)$
- We can add flow values for edges with both endpoints in A to **both** sums and they would cancel out so
- $$\begin{aligned} f^{\text{out}}(A) - f^{\text{in}}(A) &= \sum_{v \in A, w \in V} f(v, w) - \sum_{v \in A, u \in V} f(u, v) \\ &= \sum_{v \in A} (\sum_{w \in V} f(v, w) - \sum_{u \in V} f(u, v)) \\ &= \sum_{v \in A} f^{\text{out}}(v) - f^{\text{in}}(v) \\ &= f^{\text{out}}(s) - f^{\text{in}}(s) \end{aligned}$$
- since all other vertices have $f^{\text{out}}(v) = f^{\text{in}}(v)$
- $v(f) = f^{\text{out}}(s)$ and $f^{\text{in}}(s) = 0$

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Proof of Claim 7.8

$$\begin{aligned}
 v(f) &= f^{\text{out}}(A) - f^{\text{in}}(A) \\
 &\leq f^{\text{out}}(A) \\
 &= \sum_{v \in A, w \notin A} f(v, w) \\
 &\leq \sum_{v \in A, w \notin A} c(v, w) \\
 &\leq \sum_{v \in A, w \in B} c(v, w) \\
 &= c(A, B)
 \end{aligned}$$

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Max Flow / Min Cut Theorem

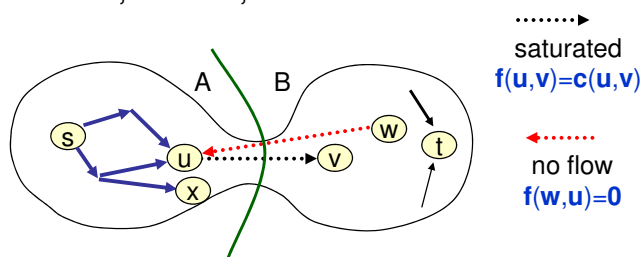
Claim 7.9 For any flow f , if G_f has no augmenting path then there is some s - t -cut (A, B) such that $v(f) = c(A, B)$ (proof on next slide)

- We know by **Claims 7.6 & 7.8** that any flow f' satisfies $v(f') \leq c(A, B)$ and we know that F-F runs for finite time until it finds a flow f satisfying conditions of **Claim 7.9**
 - Therefore by 7.9 for any flow f' , $v(f') \leq v(f)$
- Corollary (1)** F-F computes a maximum flow in G
- (2)** For any graph G , the value $v(f)$ of a maximum flow = minimum capacity $c(A, B)$ of any s - t -cut in G

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Claim 7.9

Let $A = \{u \mid \exists \text{ an path in } G_f \text{ from } s \text{ to } u\}$
 $B = V - A$; $s \in A, t \in B$



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

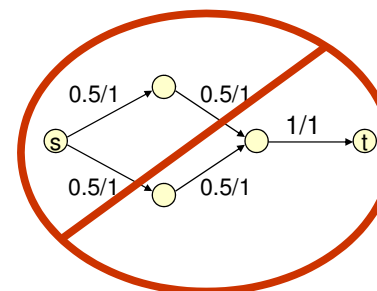
$$f^{\text{out}}(A) = \sum_{\substack{u \in A \\ v \in B}} f(u, v) = \sum_{\substack{u \in A \\ v \in B}} c(u, v) = c(A, B) \text{ and } f^{\text{in}}(A) = 0 \text{ so } v(f) = f^{\text{out}}(A) - f^{\text{in}}(A) = c(A, B)$$

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Flow Integrality Theorem

If all capacities are integers

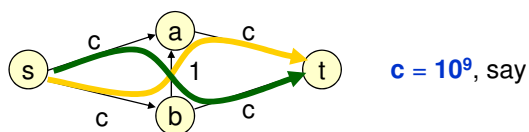
- The max flow has an integer value
- Ford-Fulkerson method finds a max flow in which $f(u, v)$ is an integer for all edges (u, v)



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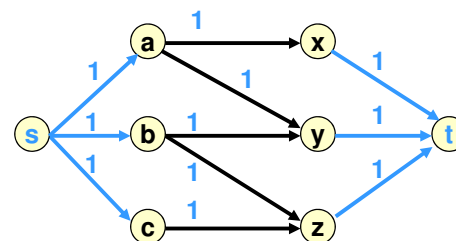
Corollaries & Facts

- 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:



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Bipartite matching as a special case of flow



Integer flows implies each flow is just a subset of the edges

Therefore flow corresponds to a matching

$O(mC) = O(nm)$ running time

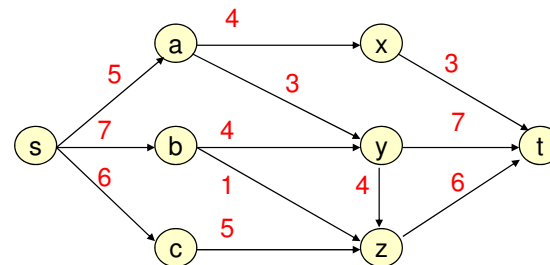
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Capacity-scaling algorithm

- General idea:
 - Choose augmenting paths P with 'large' capacity c_P
 - Can augment flows along a path P by any amount $\Delta \leq c_P$
 - Ford-Fulkerson still works
 - Get a flow that is maximum for the high-order bits first and then add more bits later

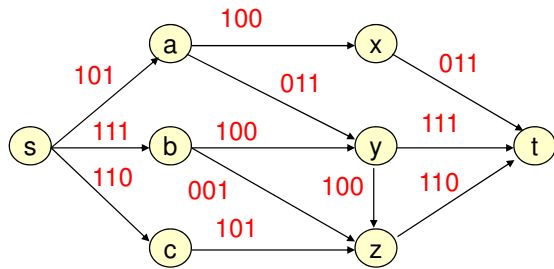
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Capacity Scaling



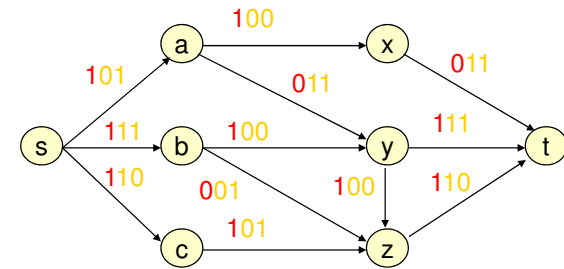
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Capacity Scaling



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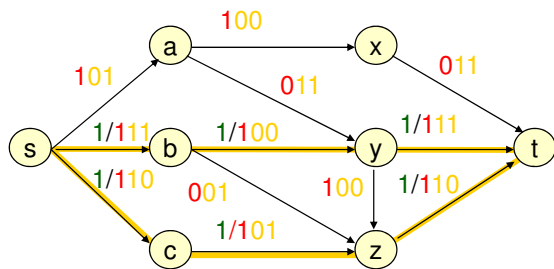
Capacity Scaling Bit 1



Capacity on each edge is at most **1**
(either **0** or **1** times $\Delta=4$)

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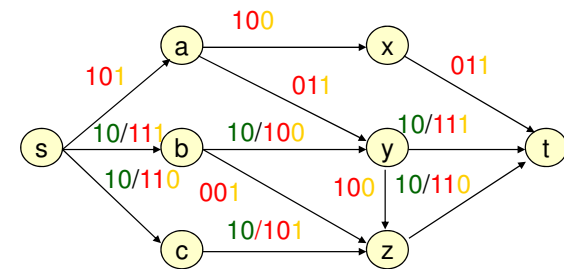
Capacity Scaling Bit 1



$O(nm)$ time

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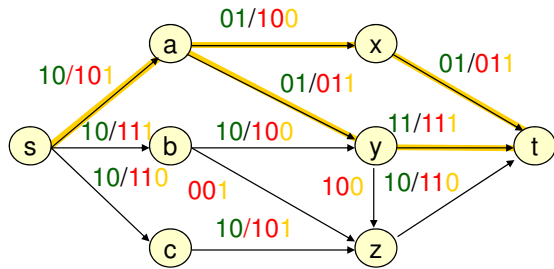
Capacity Scaling Bit 2



Residual capacity across min cut is at most **m**
(either **0** or **1** times $\Delta=2$)

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Capacity Scaling Bit 2

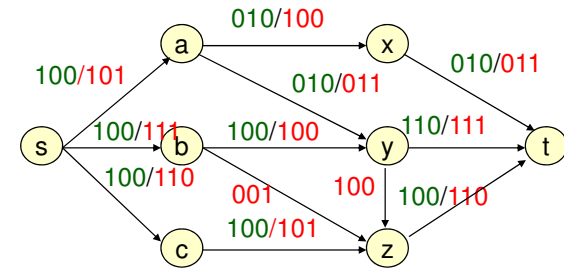


Residual capacity across min cut is at most m

$\Rightarrow \leq m$ augmentations

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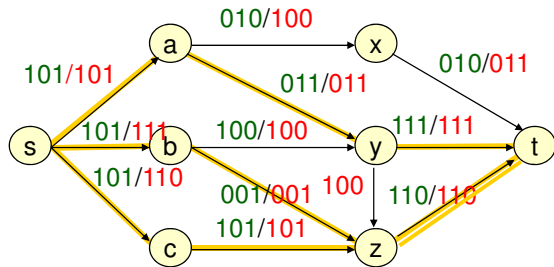
Capacity Scaling Bit 3



Residual capacity across min cut is at most m
(either 0 or 1 times $\Delta=1$)

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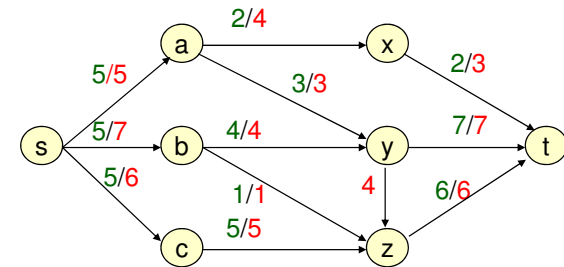
Capacity Scaling Bit 3



After $\leq m$ augmentations

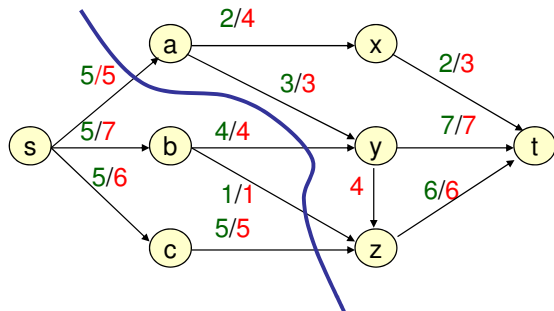
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Capacity Scaling Final



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Capacity Scaling Min Cut



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Total time for capacity scaling

- $\log_2 U$ rounds where U is largest capacity
- At most m augmentations per round
 - Let c_i be the capacities used in the i^{th} round and f_i be the maxflow found in the i^{th} round
 - For any edge (u,v) , $c_{i+1}(u,v) \leq 2c_i(u,v) + 1$
 - $i+1^{\text{st}}$ round starts with flow $f = 2 f_i$
 - Let (A,B) be a min cut from the i^{th} round
 - $v(f_i) = c_i(A,B)$ so $v(f) = 2c_i(A,B)$
 - $v(f_{i+1}) \leq c_{i+1}(A,B) \leq 2c_i(A,B) + m = v(f) + m$
- $O(m)$ time per augmentation
- Total time $O(m^2 \log U)$

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Edmonds-Karp Algorithm

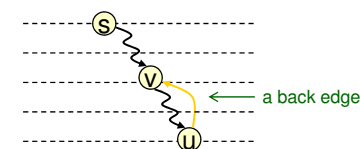
- Use a **shortest** augmenting path (via Breadth First Search in residual graph)
- Time: $O(n m^2)$

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BFS/Shortest Path Lemmas

Distance from s in G_f 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



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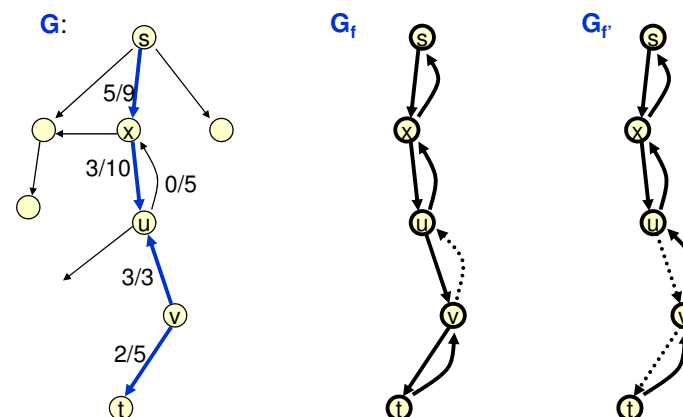
Key Lemma

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 along P only deletes forward edges, or adds back edges that go to previous vertices along P

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Augmentation vs BFS



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Theorem

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

Proof:

Call (u,v) **critical** for augmenting path P if it's closest to s having min residual capacity

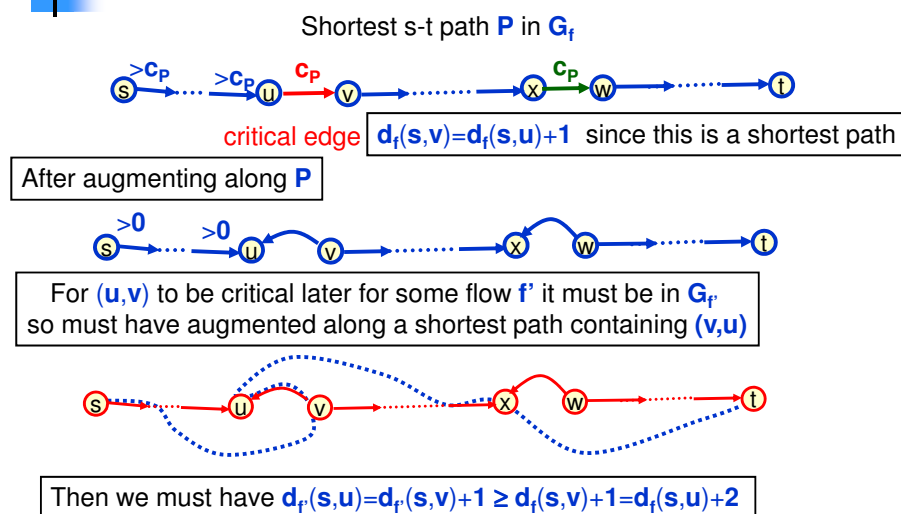
It will disappear from G_f after augmenting along P

In order for (u,v) to be critical again the (u,v) edge must re-appear in G_f but that will only happen when the distance to u has increased by 2 (next slide)

It won't be critical again until farther from s so each edge critical at most $n/2$ times

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Critical Edges in G_f



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Corollary

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

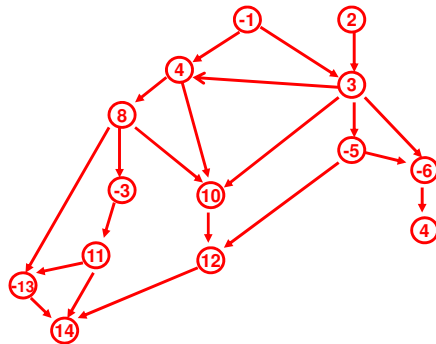
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Project Selection a.k.a. The Strip Mining Problem

- Given**
 - a directed acyclic graph $G=(V,E)$ representing precedence constraints on tasks (a task points to its predecessors)
 - a profit value $p(v)$ associated with each task $v \in V$ (may be positive or negative)
- Find**
 - a set $A \subseteq V$ of tasks that is closed under predecessors, i.e. if $(u,v) \in E$ and $u \in A$ then $v \in A$, that maximizes $\text{Profit}(A) = \sum_{v \in A} p(v)$

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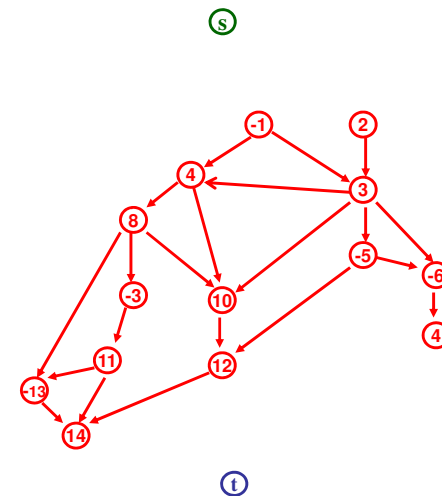
Project Selection Graph



Each task points to its predecessor tasks

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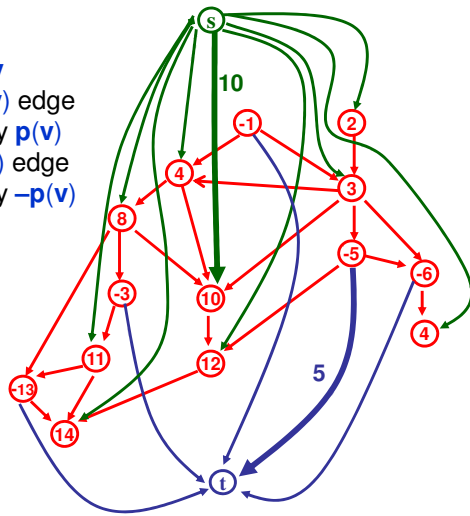
Extended Graph



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Extended Graph G'

For each vertex v
 If $p(v) \geq 0$ add (s, v) edge
 with capacity $p(v)$
 If $p(v) < 0$ add (v, t) edge
 with capacity $-p(v)$



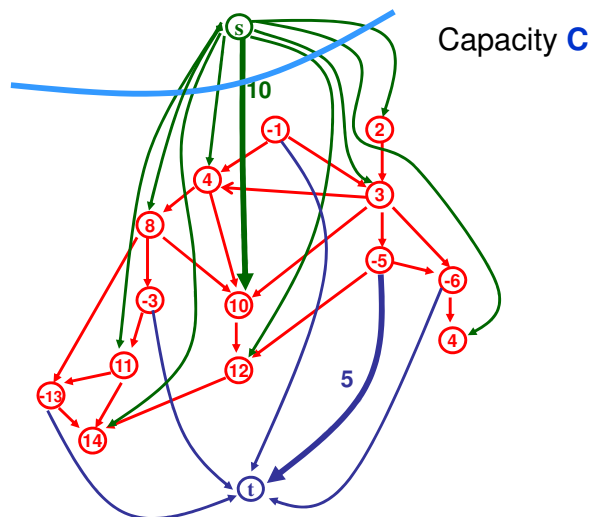
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Extended Graph G'

- Want to arrange capacities on edges of G so that for minimum s - t -cut (S, T) in G' , the set $A = S - \{s\}$
 - satisfies precedence constraints
 - has maximum possible profit in G
- Cut capacity with $S = \{s\}$ is just $C = \sum_{v: p(v) \geq 0} p(v)$
 - $\text{Profit}(A) \leq C$ for any set A
- To satisfy precedence constraints don't want any original edges of G going forward across the minimum cut
 - That would correspond to a task in $A = S - \{s\}$ that had a predecessor not in $A = S - \{s\}$
- Set capacity of each of the edges of G to $C+1$
 - The minimum cut has size at most C

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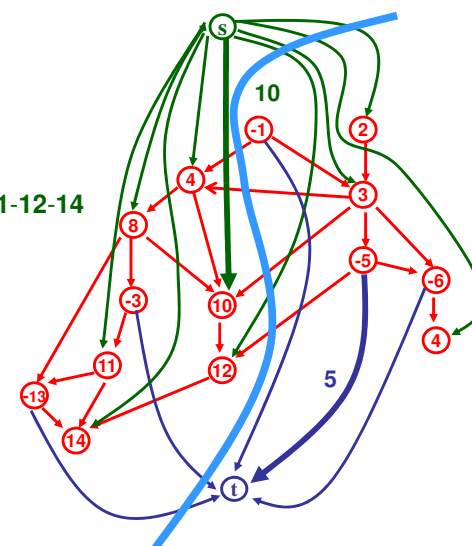
Extended Graph G'



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Extended Graph G'

Cut value
 $= 13 + 3 + 2 + 3 + 4$
 $= 13 + 3$
 $+ C - 4 - 8 - 10 - 11 - 12 - 14$



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Project Selection

- **Claim** Any s - t -cut (S, T) in G' such that $A = S - \{s\}$ satisfies precedence constraints has capacity

$$c(S, T) = C - \sum_{v \in A} p(v) = C - \text{Profit}(A)$$
- **Corollary** A minimum cut (S, T) in G' yields an optimal solution $A = S - \{s\}$ to the profit selection problem
- **Algorithm** Compute maximum flow f in G' , find the set S of nodes reachable from s in G' , and return $S - \{s\}$

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Proof of Claim

- $A = S - \{s\}$ satisfies precedence constraints
 - No edge of G crosses forward out of A since those edges have capacity $C+1$
 - Only forward edges cut are of the form (v, t) for $v \in A$ or (s, v) for $v \notin A$
 - The (v, t) edges for $v \in A$ contribute

$$\sum_{v \in A: p(v) < 0} -p(v) = - \sum_{v \in A: p(v) < 0} p(v)$$
 - The (s, v) edges for $v \notin A$ contribute

$$\sum_{v \notin A: p(v) \geq 0} p(v) = C - \sum_{v \in A: p(v) \geq 0} p(v)$$
 - Therefore the total capacity of the cut is

$$c(S, T) = C - \sum_{v \in A} p(v) = C - \text{Profit}(A)$$

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