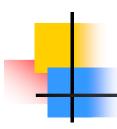
# CSE 521: Design & Analysis of Algorithms I

#### **Network Flow**

Paul Beame



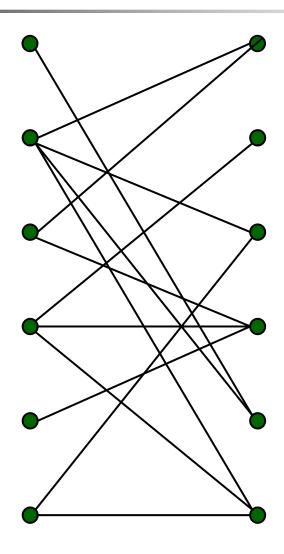
### **Bipartite Matching**

- Given: A bipartite graph G=(V,E)
  - M⊆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

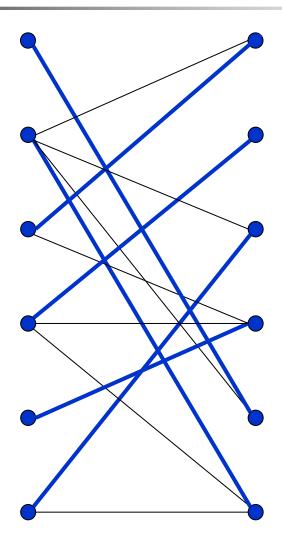


## **Bipartite Matching**



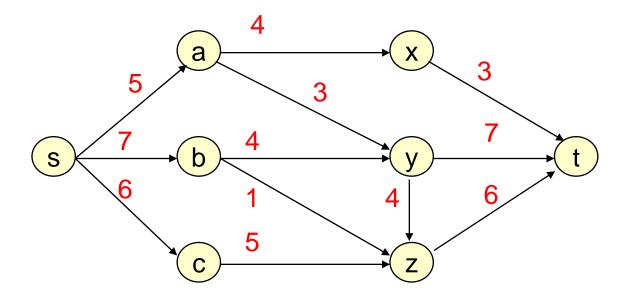


## **Bipartite Matching**





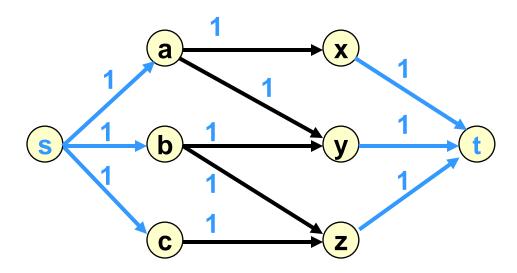
#### The Network Flow Problem



How much stuff can flow from s to t?



## Bipartite matching as a special case of flow



#### **Net Flow: Formal Definition**

#### Given:

A digraph G = (V, E)

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

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

#### Find:

A *flow function*  $f: E \rightarrow R$  s.t., for all u,v:

$$0 \le f(u,v) \le c(u,v)$$

[Capacity Constraint]

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

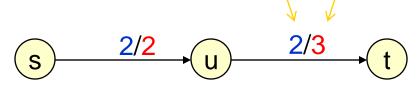
#### Notation:

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



## **Example: A Flow Function**

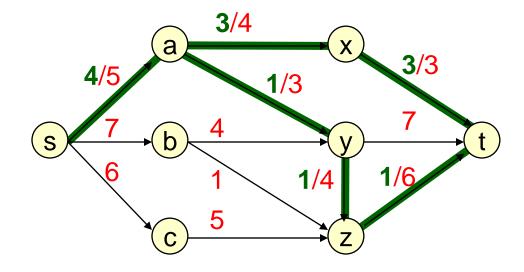
flow/capacity, not .66...



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

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### **Example: A Flow Function**

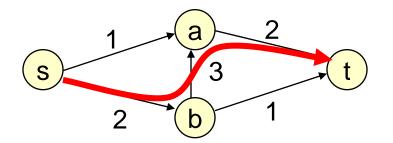


- Not shown: f(u,v) if = 0
- Note: max flow ≥ 4 since f is a flow function, with v(f) = 4

## Max Flow via a Greedy Alg?

While there is an s → 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.

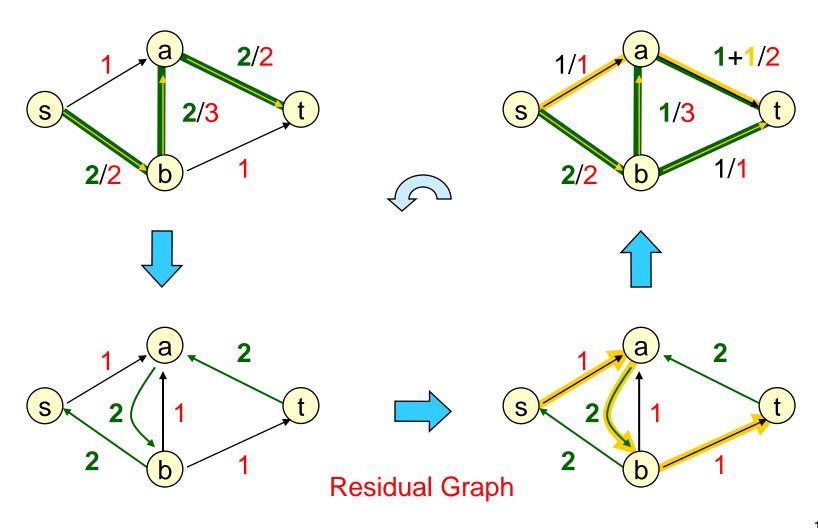
## **A Brief History of Flow**

#	year	discoverer(s)	bound
1	1951	Dantzig	$O(n^2mU)$
2	1955	Ford & Fulkerson	O(nmU)
3	1970	Dinitz	$O(nm^2)$
		Edmonds & Karp	
4	1970	Dinitz	$O(n^2m)$
5	1972	Edmonds & Karp	$O(m^2 \log U)$
		Dinitz	
6	1973	Dinitz	$O(nm \log U)$
		Gabow	
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	Cheriyan & Hagerup	$E(nm + n^2 \log^2 n)$
15	1990	Cheriyan 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	$O(m^{3/2}\log(n^2/m)\log U)$
L		A 100 Miles (100 Miles	$O(n^{2/3}m\log(n^2/m)\log U)$

n = # of verticesm= # of edgesU = Max capacity

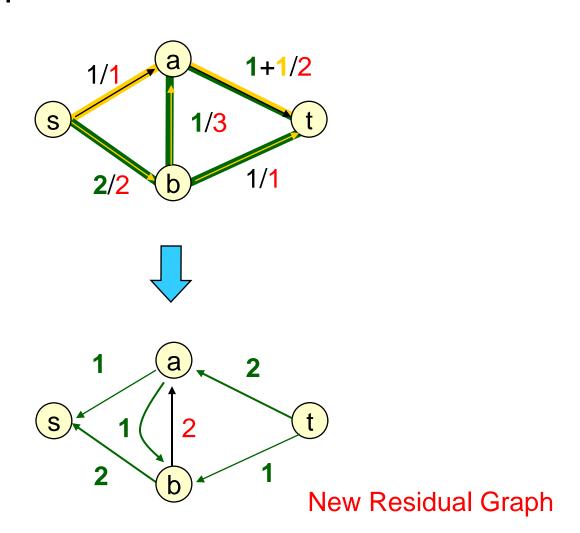
Source: Goldberg & Rao, FOCS '97

## **Greed Revisited: Residual Graph & Augmenting Path**



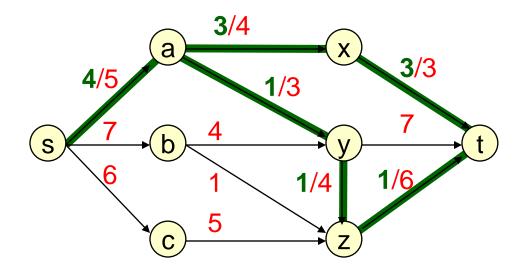


### **Greed Revisited: An Augmenting Path**



## **Residual Capacity**

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



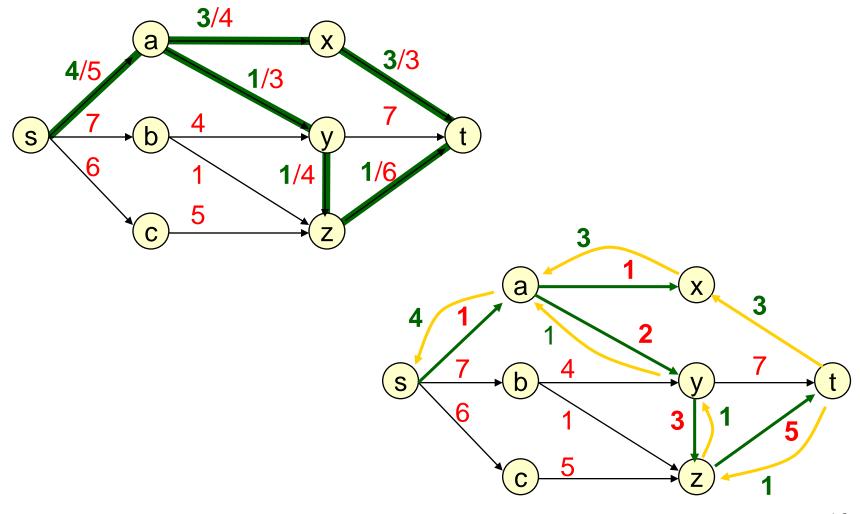
• e.g.  $c_f(s,b)=7$ ;  $c_f(a,x)=1$ ;  $c_f(x,a)=3$ 

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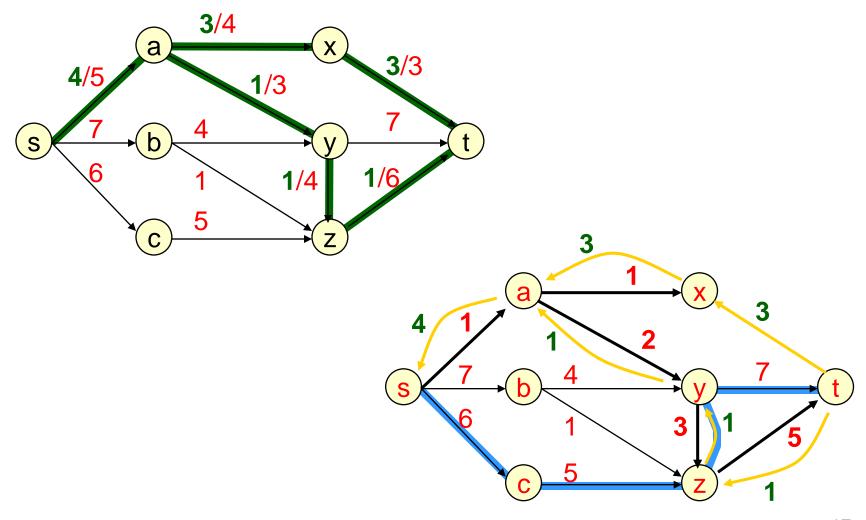
## Residual Graph & Augmenting Paths

- The residual graph (w.r.t. f) is the graph G<sub>f</sub> = (V,E<sub>f</sub>), where E<sub>f</sub> = { (u,v) | c<sub>f</sub>(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  $c_f(v,u) \ge -f(v,u)=f(u,v)>0$
- An augmenting path (w.r.t. f) is a simple
   s → t path in G<sub>f</sub>.

### **A Residual Network**



## **An Augmenting Path**

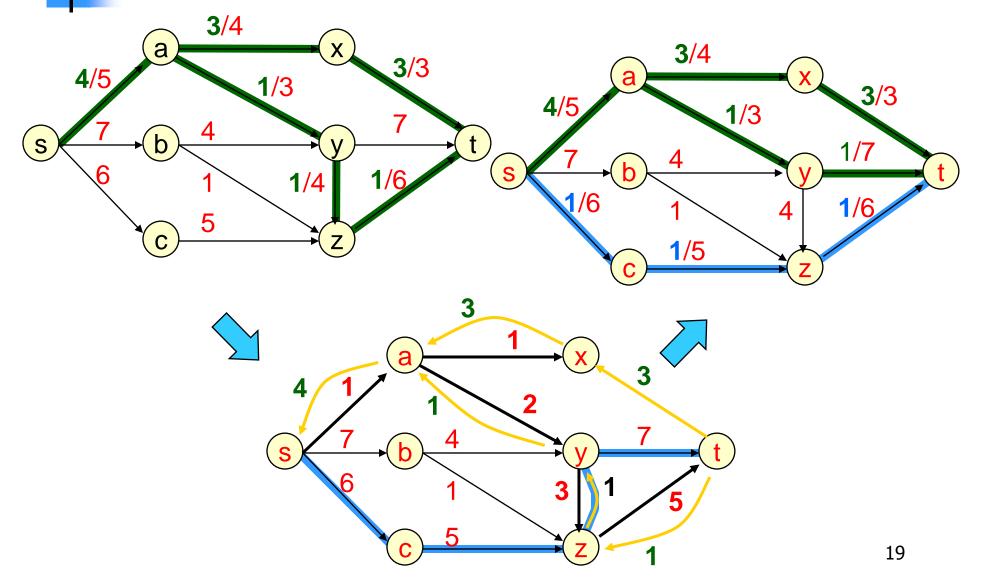


## **Augmenting A Flow**

```
augment(f,P)
    c_P \leftarrow \min_{(u,v) \in P} c_f(u,v) "bottleneck(P)"
    for each e∈ 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<sub>P</sub>
        endif
    endfor
    return(f)
```



## **Augmenting A Flow**





#### **Claim 7.1**

If G<sub>f</sub> has an augmenting path P, then the function f'=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  $f'(u,v)=f(u,v)+c_{P} \leq f(u,v)+c_{f}(u,v)$  = f(u,v)+c(u,v)-f(u,v) = c(u,v)

- If (u,v) is a backward edge then f and f' differ on flow along (v,u) instead of (u,v) f'(v,u)=f(v,u)-c<sub>P</sub> ≥ f(v,u)-c<sub>f</sub>(u,v) = f(v,u)-f(v,u)=0
- Other conditions like flow conservation still met



#### **Ford-Fulkerson Method**

Start with f=0 for every edge
While G<sub>f</sub> has an augmenting path,
augment

#### Questions:

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



## 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<sub>P</sub>>v(f) for f'=augment(f,P)
  - Since edges of residual capacity 0 do not appear in the residual graph
- Let  $C = \sum_{(s,u) \in E} c(s,u)$ 
  - ν(f)≤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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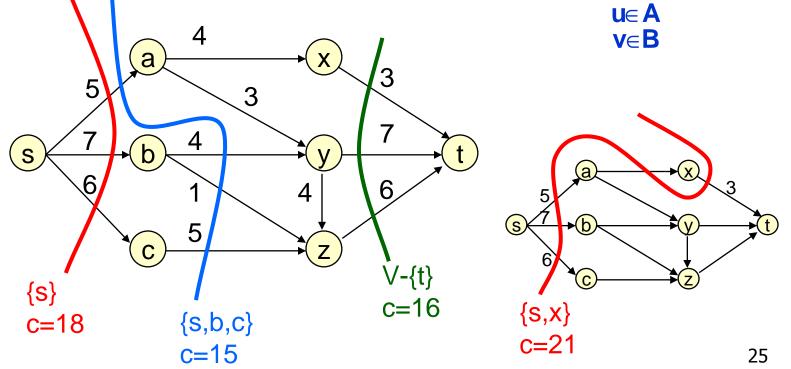
### Running Time of Ford-Fulkerson

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

## Cuts

- A partition (A,B) of V is an *s-t-cut* if
  - S∈ A, t∈ B

• Capacity of cut (A,B) is  $c(A,B) = \sum c(u,v)$ 



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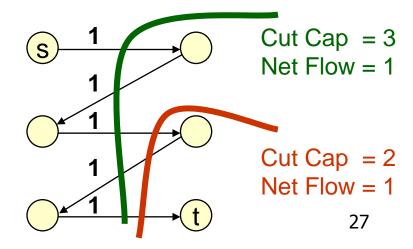
#### **Convenient Definition**

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

• 
$$f^{in}(A) = \sum_{v \in A, u \notin A} f(u,v)$$

#### **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) = fout(A)-fin(A), and
  - the net flow across the cut cannot exceed the capacity of the cut,
     i.e. fout(A)-fin(A) ≤ c(A,B)
- Corollary : Max flow ≤ Min cut



#### **Proof of Claim 7.6**

- Consider a set A with s∈ A, t∉ A
- $f^{out}(A)-f^{in}(A) = \sum_{v \in A, w \notin A} f(v,w) \sum_{v \in A, u \notin A} f(u,v)$
- We can add flow values for edges with both endpoints in A to both sums and they would cancel out so

since all other vertices have fout(v)=fin(v)

• 
$$v(f) = f^{out}(s)$$
 and  $f^{in}(s)=0$ 

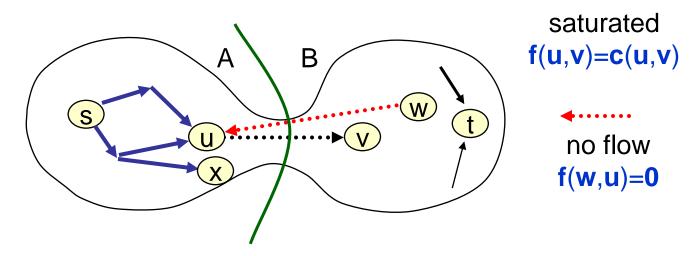
#### **Proof of Claim 7.8**

#### 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') ≤ 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') ≤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

#### **Claim 7.9**

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



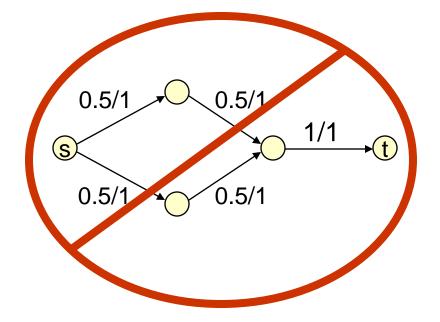
This is true for **every** edge crossing the cut, i.e.

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



#### If all capacities are integers

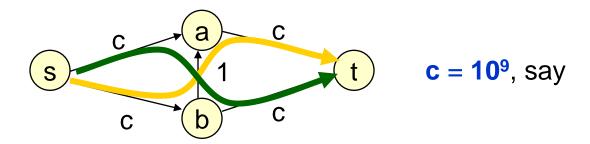
- The max flow has an integer value
- Ford-Fulkerson method finds a max flow in which f(u,v) is an integer for all edges (u,v)





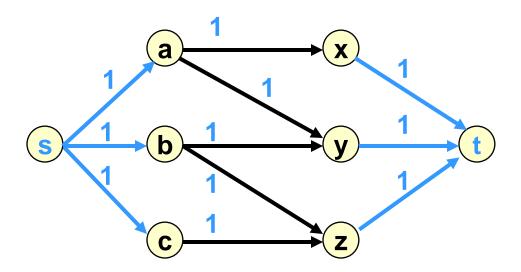
#### **Corollaries & Facts**

- 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



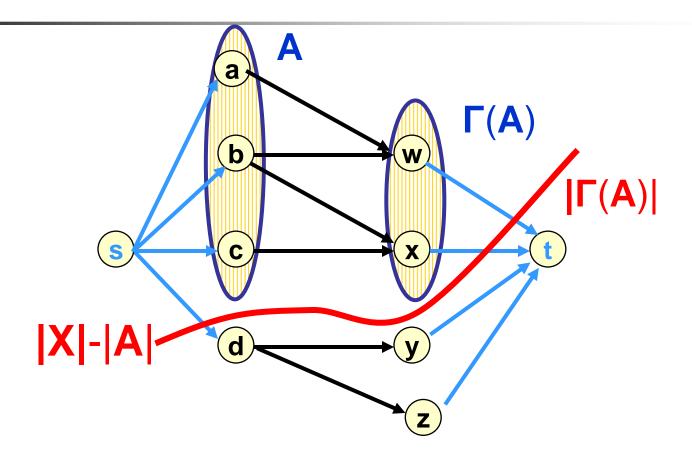
## Consequences of Ford-Fulkerson: Hall's Theorem

Def: For a graph G=(V,E) and A⊆V let the neighborhood of A be

Hall's Theorem: A graph G=(V,E) with bipartition V=X∪Y where |X|=|Y| has a perfect matching if and only if for every set A⊆X we have |Γ(A)|≥|A|.



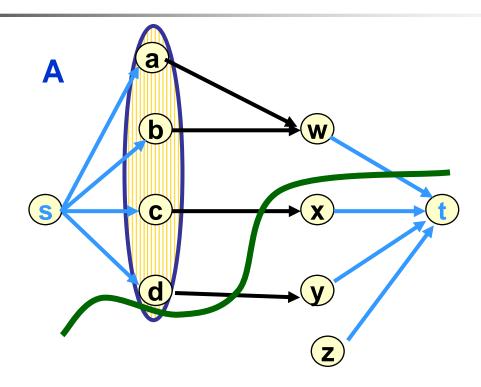
#### Hall's Theorem Proof: Part 1



Min cut  $\leq |X|-|A|+|\Gamma(A)|<|X|$  so matching is not perfect



### Hall's Theorem Proof: Part 2

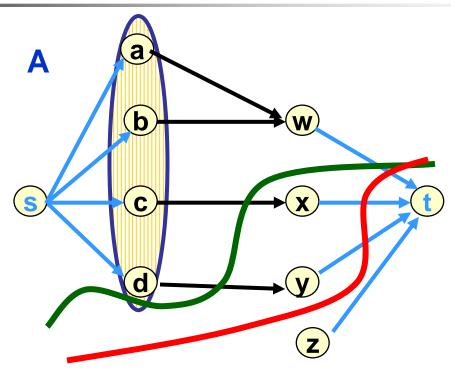


If no perfect matching there is some **st**-cut (**A**',**B**) with cut value < |X|

Let A=part of A' in X

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#### Hall's Theorem Proof: Part 2



- Modify A' to include all vertices in Γ(A)
- Value of mincut cannot go up so still < |X|</li>
- # of edges from s cut is |X|-|A|
- # of edges to t cut is ≥ |\(\Gamma(A)\)|
- Total is  $\geq |X|-|A|+|\Gamma(A)|$  but <|X| so  $|\Gamma(A)|<|A|$



# Consequences of Ford-Fulkerson: Edge-disjoint paths

- Given a directed graph G=(V,E) and vertices s and t find a maximum set of edge-disjoint (simple) paths from s to t.
- Lemma: Any 0-1 flow f in the network flow graph on G=(V,E), s, t with all capacities=1 contains a set of ν(f) edgedisjoint simple paths from s to t.
- Proof: Follow the edges of positive flow through the graph. Flow conservation



# Edge-disjoint paths in directed graphs

- Since a collection of k edge-disjoint paths is always a legal 0-1 flow of value k, by Maxflow=Mincut we have:
- Directed Menger's Theorem: The maximum number of edge-disjoint paths from s to t is equal to the minimum number of edges that need to be cut to separate s from t.
- We can use Ford-Fulkerson to find the flow in O(mn) time (since capacities are all 1) and then prune the flow as on previous slide in O(m+n) extra time.



- Given an undirected graph G=(V,E) and nodes s and t, we can modify G to create a directed graph G'=(V,E') where we include both (u,v) and (v,u) in E' whenever {u,v} is in E.
- Observation: If in a flow f we have f(v,u)=f(u,v) >0 then we can set f(v,u)=f(u,v)=0 and keep the same flow value
- Algorithm: Run F-F on G' in O(mn) time and then remove flow cycles in O(m+n) time.
- Undirected Menger's Theorem: The maximum number of edge-disjoint paths from s to t is equal to the minimum number of edges that need to be cut to separate s from t.



# Polynomial running time?

- F-F uses O(mC) time where C is the total capacity leaving s.
  - If all capacities are at most some upper bound U then C < nU but each input capacity takes only log<sub>2</sub>U bits to represent so the running time is not polynomial in the input size
- We will give two modifications that yield polynomial-time algorithms for general network flow problems

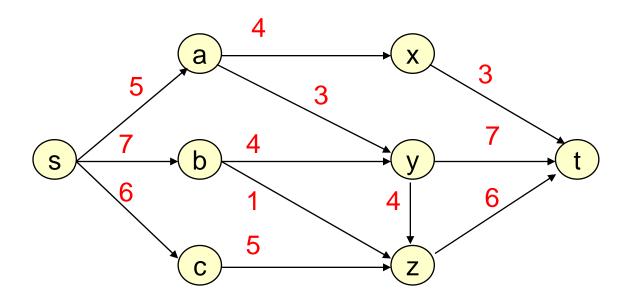


# **Capacity-scaling algorithm**

- General idea:
  - Choose augmenting paths P with 'large' capacity c<sub>P</sub>
  - Can augment flows along a path P by any amount b≤c<sub>P</sub>
    - Ford-Fulkerson still works
  - Get a flow that is maximum for the highorder bits first and then add more bits later

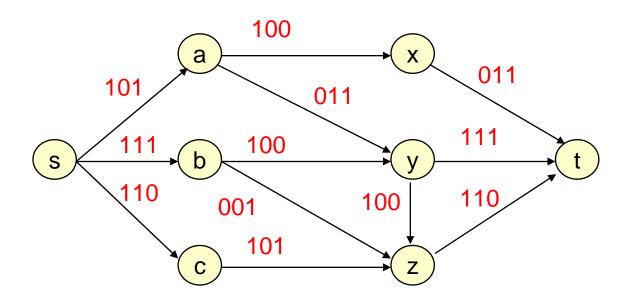


# **Capacity Scaling**

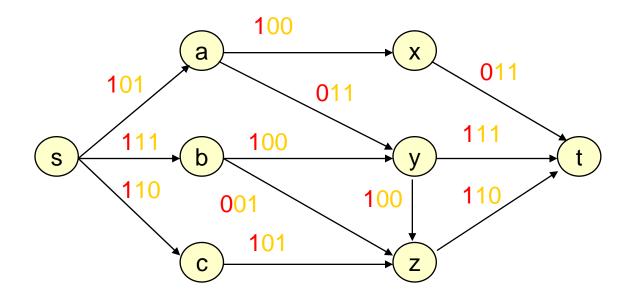




# **Capacity Scaling**

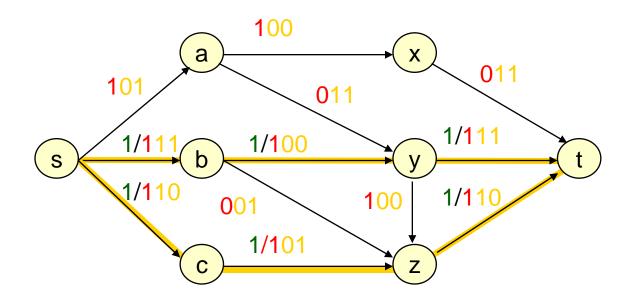






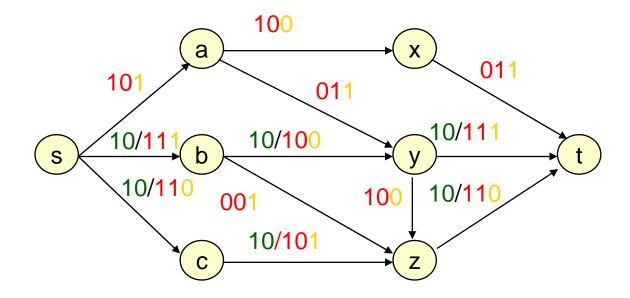
Capacity on each edge is at most 1 (either 0 or 1 times  $\Delta=4$ )





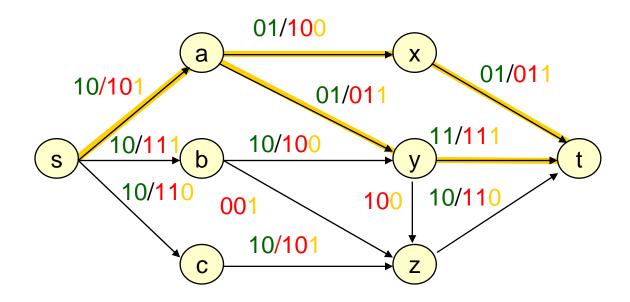
O(nm) time





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

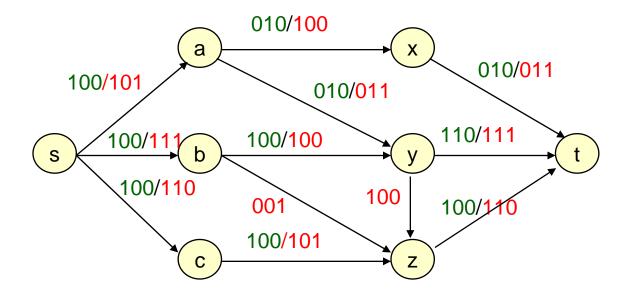




Residual capacity across min cut is at most m

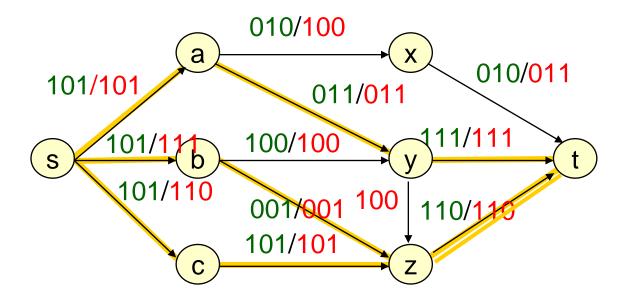
 $\Rightarrow \leq m$  augmentations





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

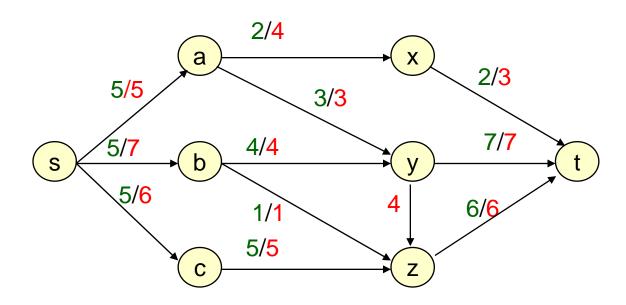




After ≤ m augmentations

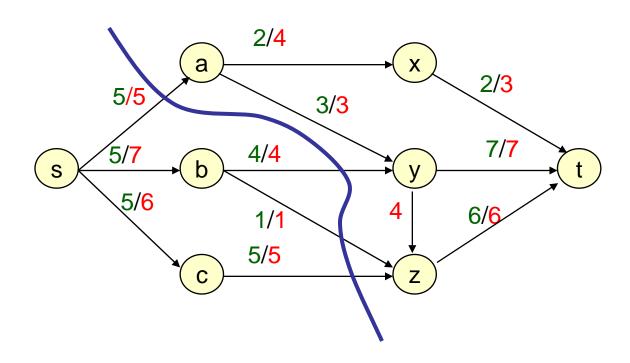


# **Capacity Scaling Final**





# **Capacity Scaling Min Cut**



# Total time for capacity scaling

- log<sub>2</sub> U rounds where U is largest capacity
- At most m augmentations per round
  - Let c<sub>i</sub> be the capacities used in the i<sup>th</sup> round and f<sub>i</sub> be the maxflow found in the i<sup>th</sup> round
    - For any edge (u,v),  $c_{i+1}(u,v) \le 2c_i(u,v)+1$
  - i+1<sup>st</sup> round starts with flow f = 2 f<sub>i</sub>
  - Let (A,B) be a min cut from the i<sup>th</sup> round
    - $v(f_i)=c_i(A,B)$  so  $v(f)=2c_i(A,B)$
  - $v(f_{i+1}) \le c_{i+1}(A,B) \le 2c_i(A,B) + m = v(f) + m$
- O(m) time per augmentation
- Total time O(m² log U)



# **Edmonds-Karp Algorithm**

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

■ Time: O(n m²)

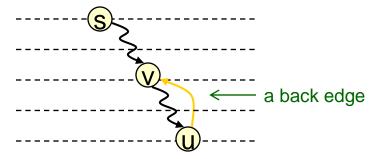


### **BFS/Shortest Path Lemmas**

### 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





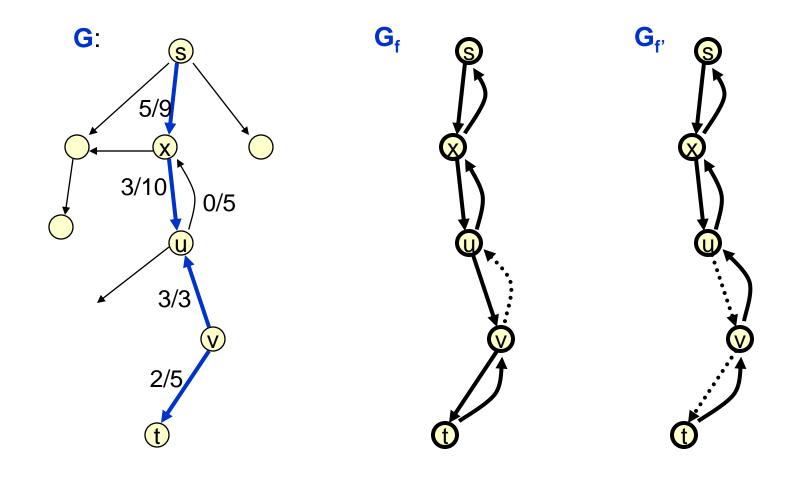
### **Key Lemma**

Let f be a flow, G<sub>f</sub> 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



# **Augmentation vs BFS**





#### **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<sub>f</sub> after augmenting along P

In order for (u,v) to be critical again the (u,v) edge must re-appear in G<sub>f</sub> 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 so each edge critical at most n/2 times



# Critical Edges in G<sub>f</sub>

Shortest s-t path P in G<sub>f</sub>

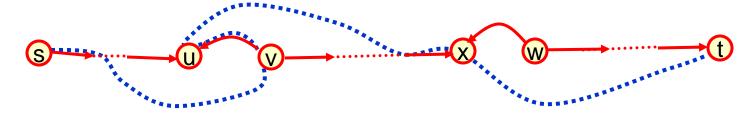


critical edge  $|\mathbf{d}_f(\mathbf{s}, \mathbf{v}) = \mathbf{d}_f(\mathbf{s}, \mathbf{u}) + 1$  since this is a shortest path

After augmenting along P



For  $(\mathbf{u}, \mathbf{v})$  to be critical later for some flow  $\mathbf{f}$  it must be in  $\mathbf{G}_{\mathbf{f}}$  so must have augmented along a shortest path containing  $(\mathbf{v}, \mathbf{u})$ 



Then we must have  $d_{f'}(s,u)=d_{f'}(s,v)+1 \ge d_f(s,v)+1=d_f(s,u)+2$ 



# Corollary

Edmonds-Karp runs in O(nm²) time



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

#### Given

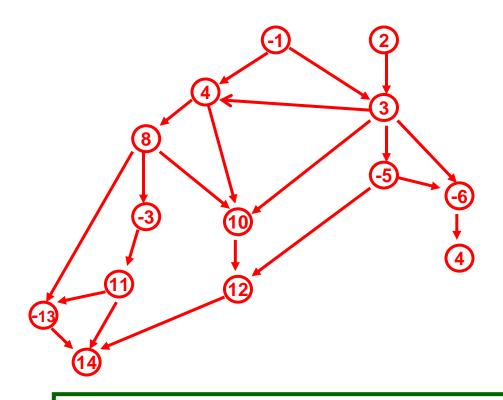
- a directed acyclic graph G=(V,E)
  representing precedence constraints on
  tasks (a task points to its predecessors)
- a profit value p(v) associated with each task v∈ V (may be positive or negative)

#### Find

a set A⊆V of tasks that is closed under predecessors, i.e. if (u,v)∈ E and u∈ A then
 v∈ A, that maximizes Profit(A)=∑<sub>v∈A</sub> p(v)



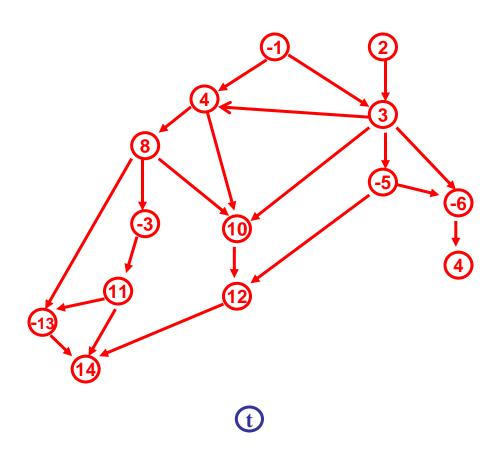
# **Project Selection Graph**



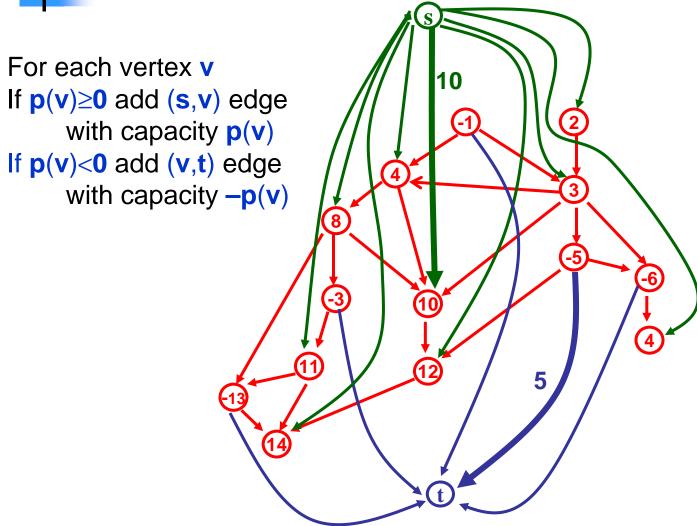
Each task points to its predecessor tasks





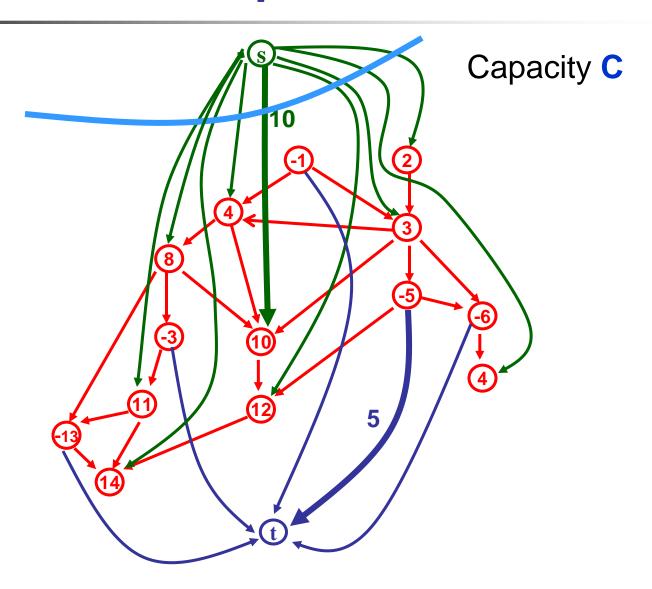




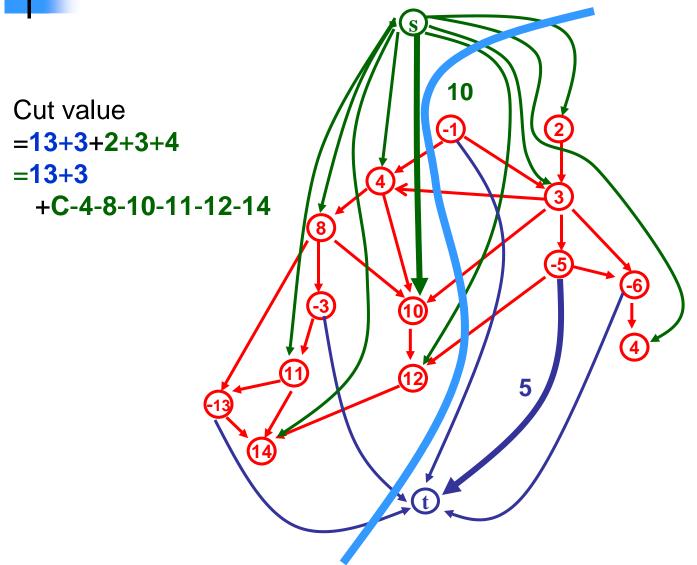


- Want to arrange capacities on edges of G so that for minimum s-t-cut (S,T) in G', the set A=S-{s}
  - satisfies precedence constraints
  - has maximum possible profit in G
- Cut capacity with  $S=\{s\}$  is just  $C=\sum_{v: p(v)\geq 0} p(v)$ 
  - Profit(A) ≤ C for any set A
- To satisfy precedence 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





# 





### **Project Selection**

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

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

- 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'<sub>f</sub> and return S-{s}

# **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∈ A or (s,v) for v∉ A
  - The (v,t) edges for v∈ A contribute

$$\sum_{\mathbf{v}\in A: p(\mathbf{v})<\mathbf{0}} -\mathbf{p}(\mathbf{v}) = -\sum_{\mathbf{v}\in A: p(\mathbf{v})<\mathbf{0}} \mathbf{p}(\mathbf{v})$$

The (s,v) edges for v∉ A contribute

$$\sum_{v \notin A: p(v) \ge 0} p(v) = C - \sum_{v \in A: p(v) \ge 0} p(v)$$

Therefore the total capacity of the cut is

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