

CSE 421 Winter 2025
Lecture 17:
Max Flow Running Time

Nathan Brunelle

<http://www.cs.uw.edu/421>

Flows

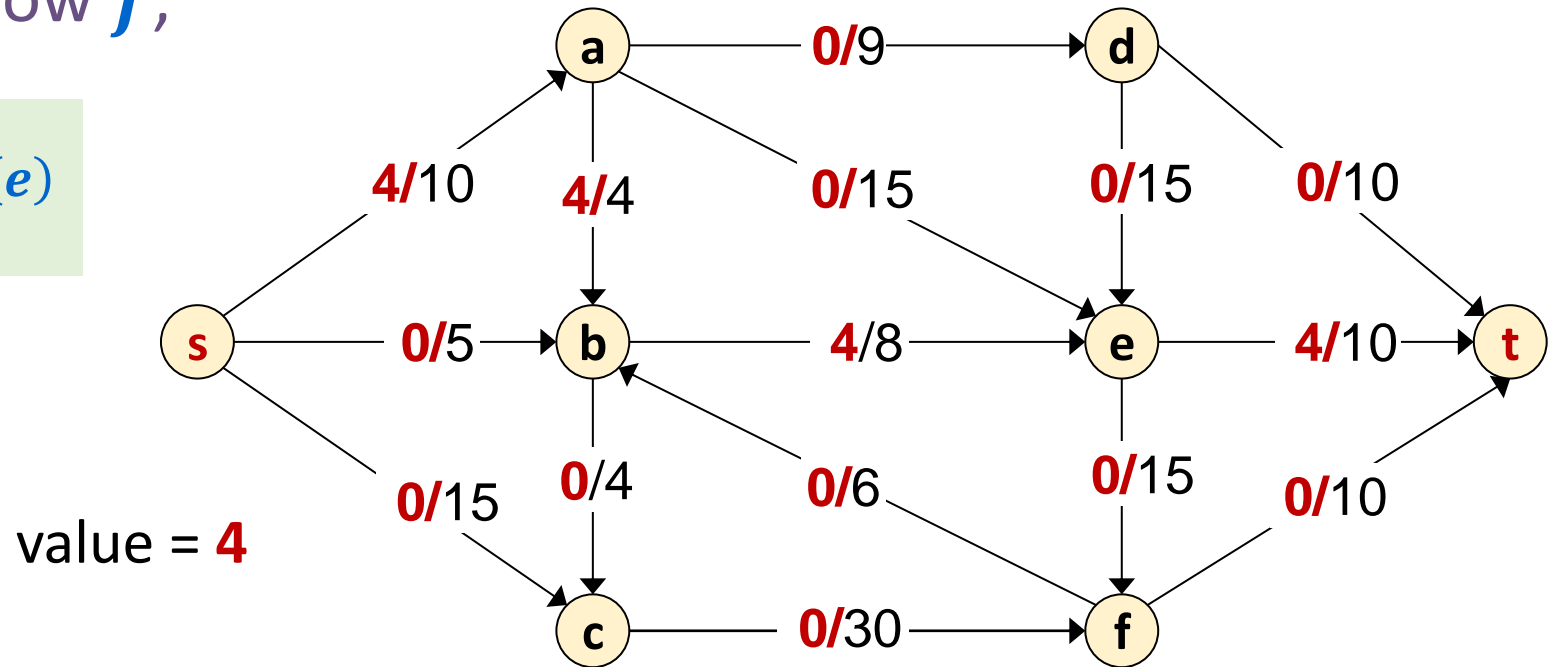
Defn: An **s-t flow** in a flow network is a function $f: E \rightarrow \mathbb{R}$ that satisfies:

- For each $e \in E$: $0 \leq f(e) \leq c(e)$ [capacity constraints]

- For each $v \in V - \{s, t\}$: $\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$ [flow conservation]

Defn: The **value** of flow f ,

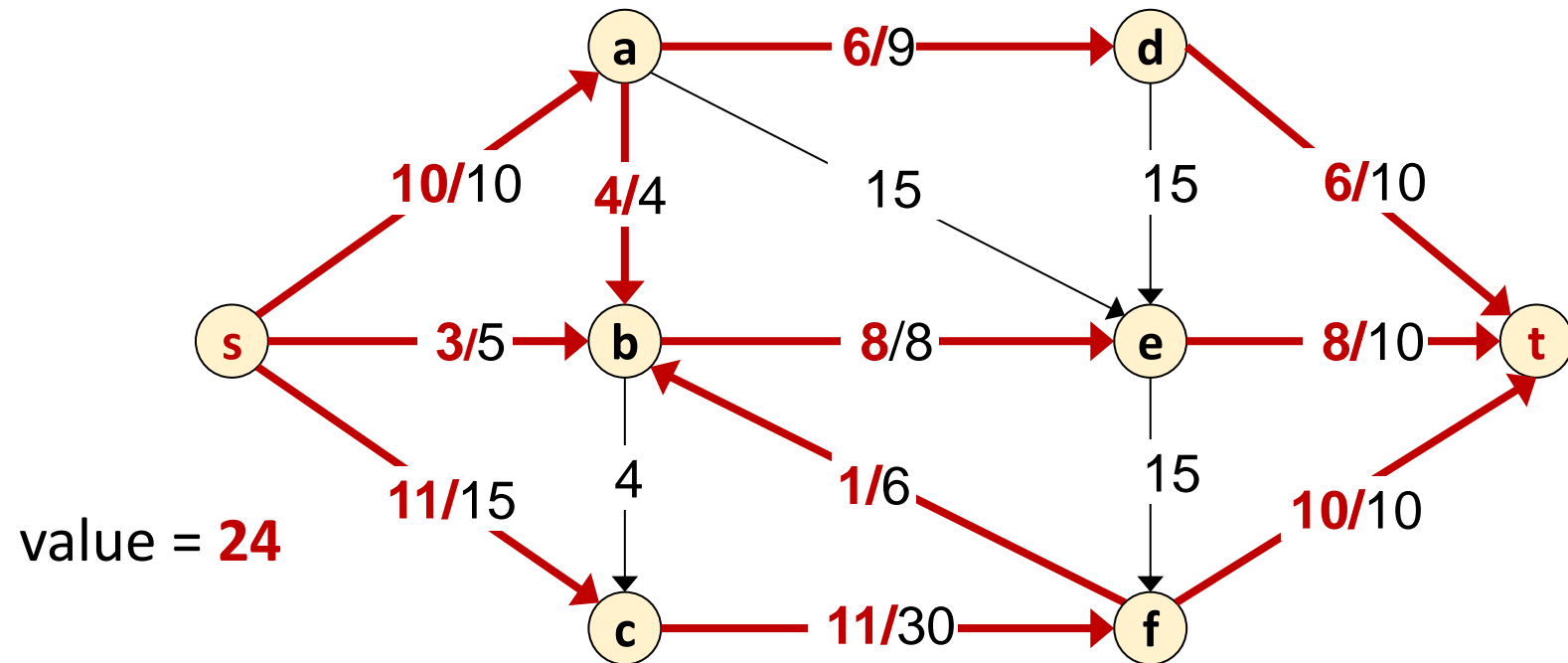
$$v(f) = \sum_{e \text{ out of } s} f(e)$$



Maximum Flow Problem

Given: a flow network

Find: an s - t flow of maximum value



Residual Graphs and Augmenting Paths

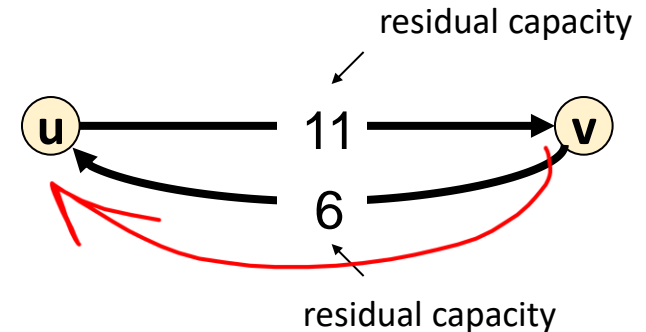
Residual edges of two kinds:

- Forward: $e = (u, v)$ with capacity $c_f(e) = c(e) - f(e)$
 - Amount of extra flow we can add along e
- Backward: $e^R = (v, u)$ with capacity $c_f(e) = f(e)$
 - Amount we can reduce/undo flow along e



Residual graph: $G_f = (V, E_f)$.

- Residual edges with residual capacity $c_f(e) > 0$.
- $E_f = \{e : f(e) < c(e)\} \cup \{e^R : f(e) > 0\}$.



Augmenting Path: Any s - t path P in G_f . Let $\text{bottleneck}(P) = \min_{e \in P} c_f(e)$.

Ford-Fulkerson idea: Repeat “find an augmenting path P and increase flow by $\text{bottleneck}(P)$ ” until none left.

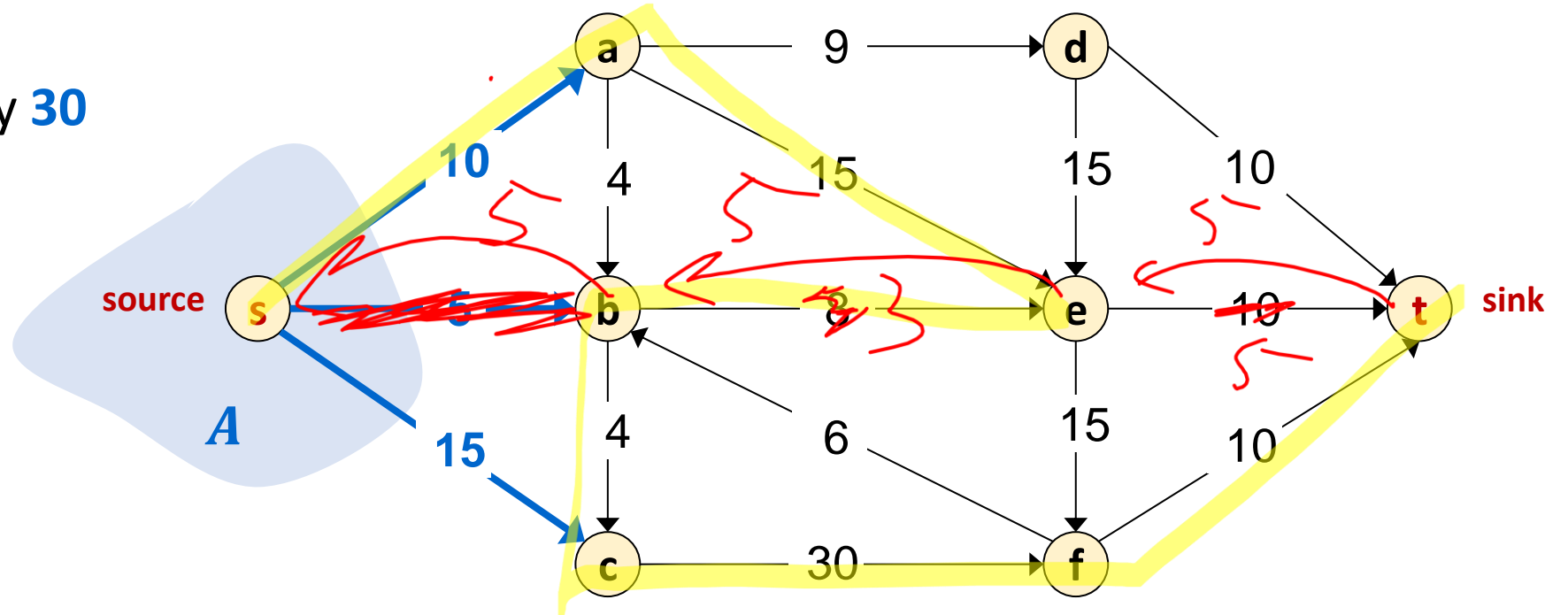
Cuts

Defn: An **s-t cut** is a partition (A, B) of V with $s \in A$ and $t \in B$.

The **capacity** of cut (A, B) is

$$c(A, B) = \sum_{e \text{ out of } A} c(e)$$

capacity **30**

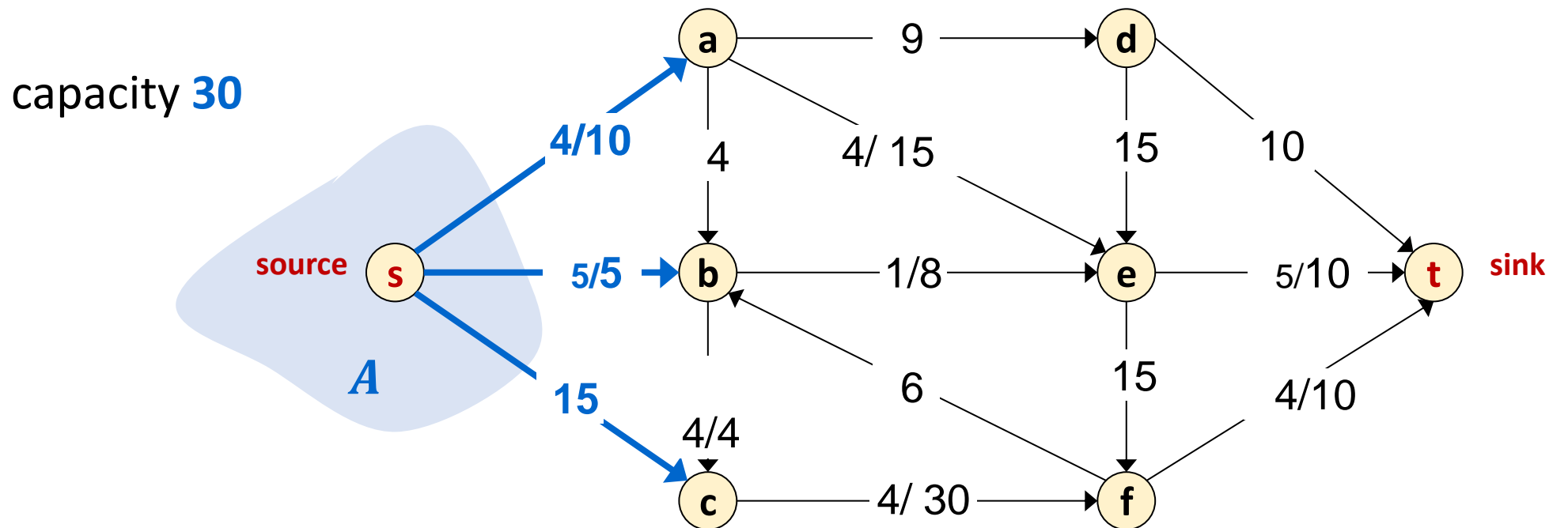


Cuts

Defn: An **s-t cut** is a partition (A, B) of V with $s \in A$ and $t \in B$.

The **capacity** of cut (A, B) is

$$c(A, B) = \sum_{e \text{ out of } A} c(e)$$



Minimum Cut Problem

Defn: An **s-t cut** is a partition (A, B) of V with $s \in A$ and $t \in B$.

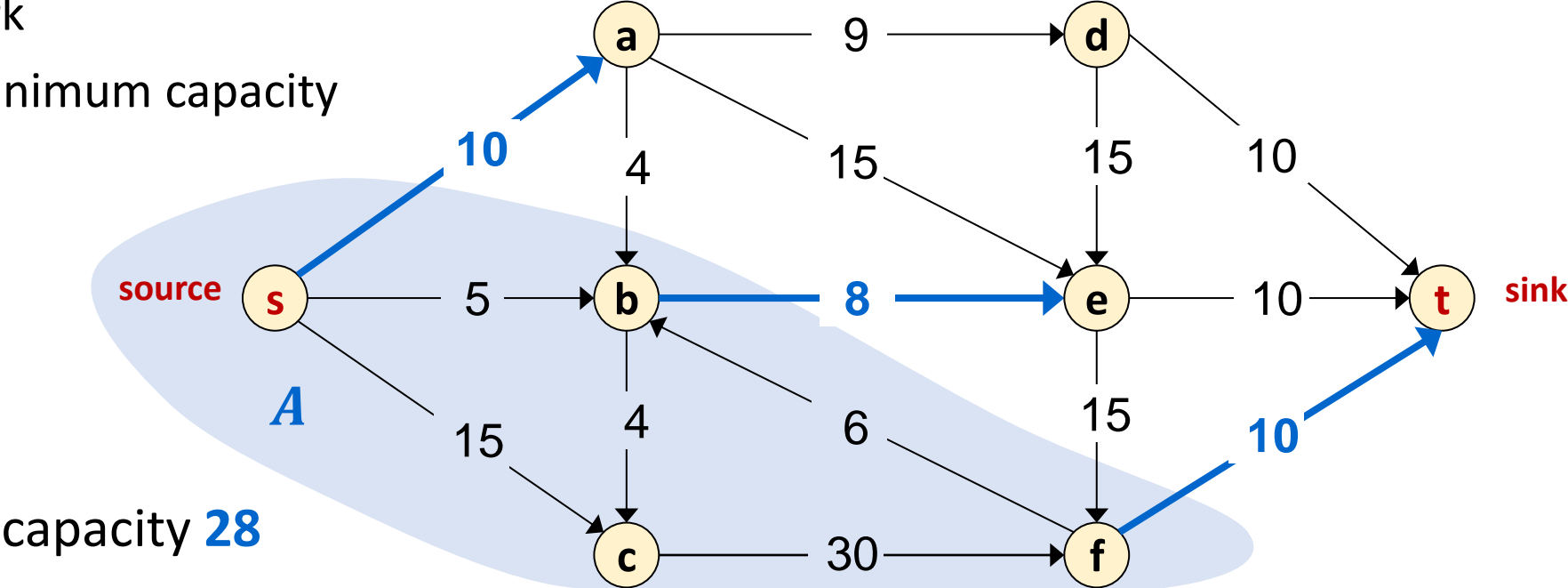
The **capacity** of cut (A, B) is

$$c(A, B) = \sum_{e \text{ out of } A} c(e)$$

Minimum s-t cut problem:

Given: a flow network

Find: an **s-t** cut of minimum capacity



Flows and Cuts

Let f be any s - t flow and (A, B) be any s - t cut:

Flow Value Lemma: The net value of the flow sent across (A, B) equals $v(f)$.

Intuition: All flow coming from s must eventually reach t , and so must cross that cut

Weak Duality: The value of the flow is at most the capacity of the cut;
i.e., $v(f) \leq c(A, B)$.

Intuition: Since all flow must cross any cut, any cut's capacity is an upper bound on the flow

Corollary: If $v(f) = c(A, B)$ then f is a maximum flow and (A, B) is a minimum cut.

Intuition: If we find a cut whose capacity matches the flow, we can't push more flow through that cut because it's already at capacity. We additionally can't find a smaller cut, since that flow was achievable.

Max-Flow Min-Cut Theorem

Augmenting Path Theorem: Flow f is a max flow \Leftrightarrow there are no augmenting paths wrt f

Max-Flow Min-Cut Theorem: The value of the max flow equals the value of the min cut.

[Elias-Feinstein-Shannon 1956, Ford-Fulkerson 1956]

“Maxflow = Mincut”

Proof: We prove both together by showing that all of these are equivalent:

(i) There is a cut (A, B) such that $v(f) = c(A, B)$.

(ii) Flow f is a max flow.

(iii) There is no augmenting path w.r.t. f .

(i) \Rightarrow (ii): Comes from weak duality lemma.

(ii) \Rightarrow (iii): (by contradiction)

If there is an augmenting path w.r.t. flow f then we can improve f . Therefore f is not a max flow.

(iii) \Rightarrow (i): We will use the residual graph to identify a cut whose capacity matches the flow

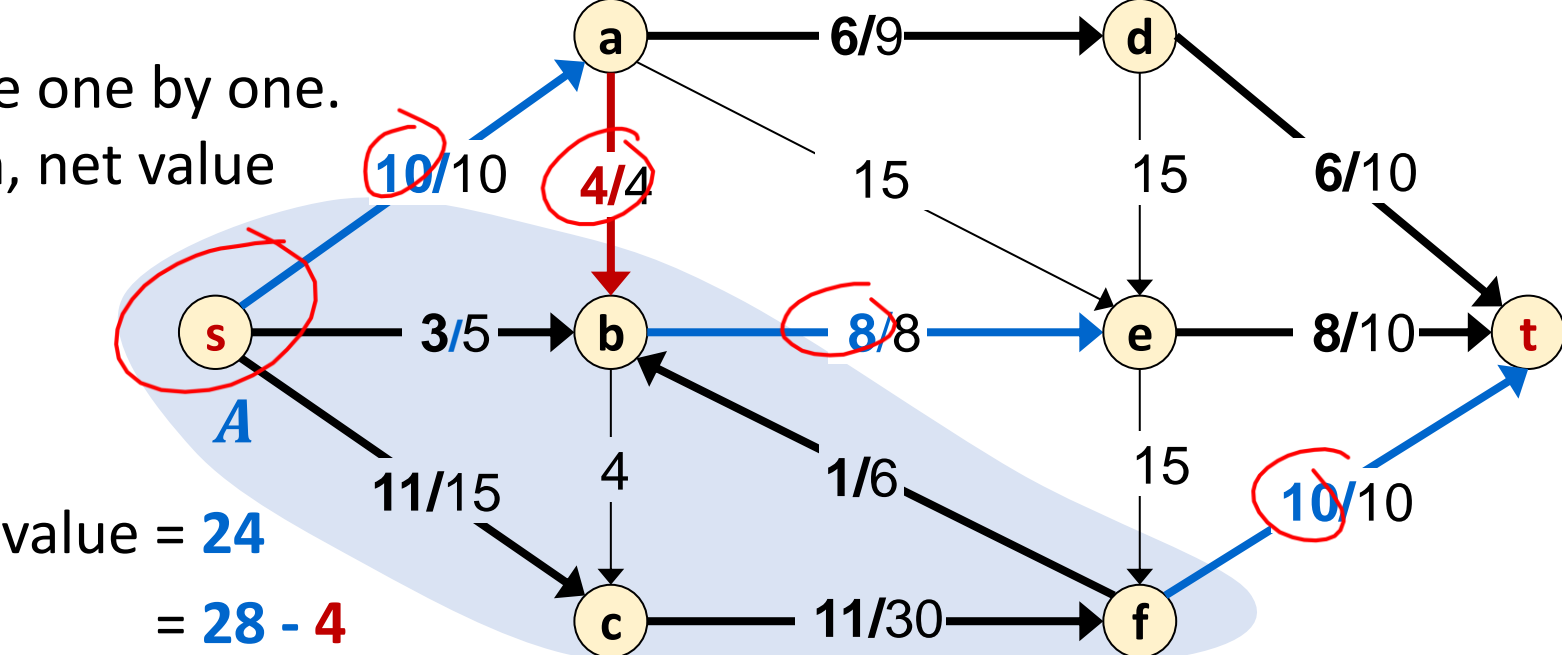
Flow Value Lemma – Idea

Flow Value Lemma: Let f be any s - t flow and (A, B) be any s - t cut. The net value of the flow sent across the cut equals $v(f)$:

$$\sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e) = v(f)$$

Why is it true?

- Add vertices to s side one by one.
- By flow conservation, net value doesn't change



Flow Value Lemma – Proof

Flow Value Lemma: Let f be any s - t flow and (A, B) be any s - t cut. The net value of the flow sent across the cut equals $v(f)$:

$$\sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e) = v(f)$$

Proof:

$$\begin{aligned}
 v(f) &= \sum_{e \text{ out of } s} f(e) \\
 &= \sum_{e \text{ out of } s} f(e) - \sum_{e \text{ into } s} f(e) + \sum_{v \in A - \{s\}} \left[\sum_{e \text{ out of } v} f(e) - \sum_{e \text{ into } v} f(e) \right] \\
 &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e)
 \end{aligned}$$

= 0. No edges into s since it is a source

Contributions from internal edges of A cancel.

= 0 by flow conservation.

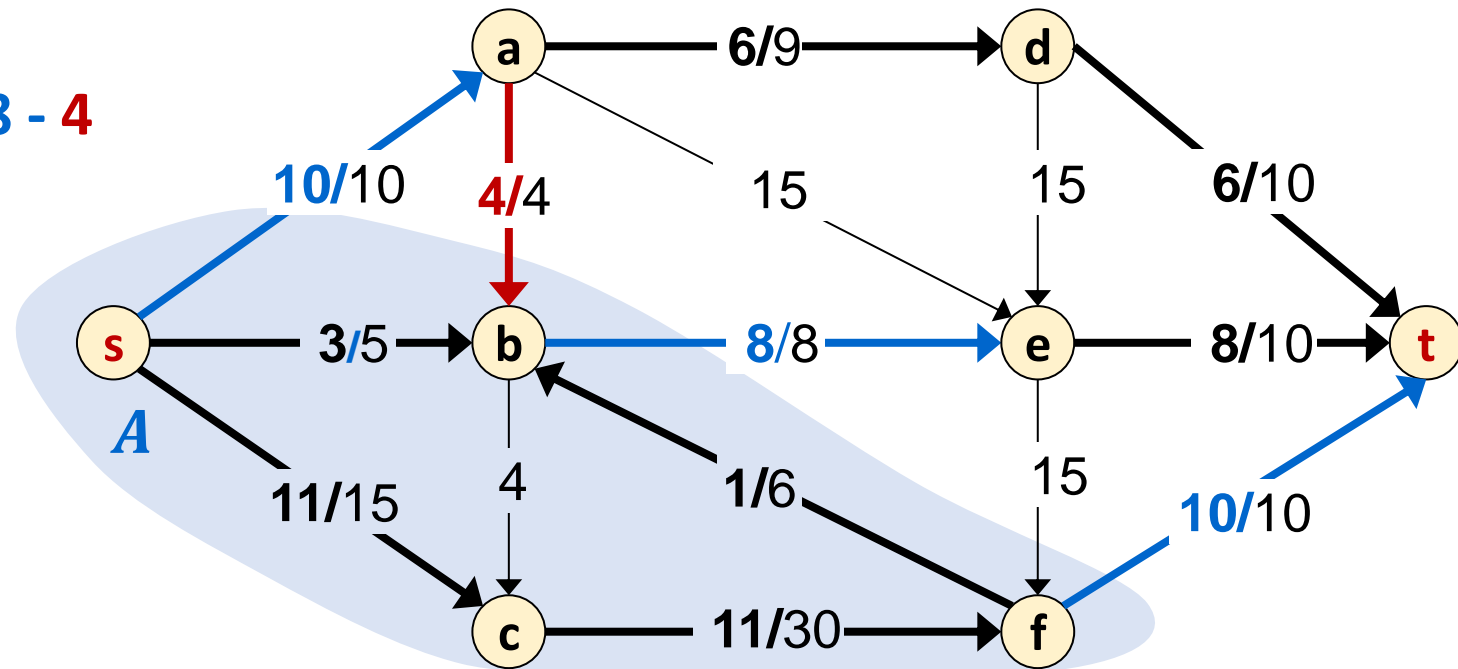
Weak Duality - Idea

(i) \Rightarrow (ii)

Weak Duality: Let f be any s - t flow and (A, B) be any s - t cut. The value of the flow is at most the capacity of the cut; i.e., $v(f) \leq c(A, B)$:

Value of flow = $24 = 28 - 4$

Capacity of cut = 28



Weak Duality - Proof

(i) \Rightarrow (ii)

Weak Duality: Let f be any s - t flow and (A, B) be any s - t cut. The value of the flow is at most the capacity of the cut; i.e., $v(f) \leq c(A, B)$.

Proof:

$$\begin{aligned} v(f) &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e) \\ &\leq \sum_{e \text{ out of } A} f(e) && \text{since } f(e) \geq 0 \\ &\leq \sum_{e \text{ out of } A} c(e) && \text{since } f(e) \leq c(e) \\ &= c(A, B) \end{aligned}$$



Proof of Max-Flow Min-Cut Theorem

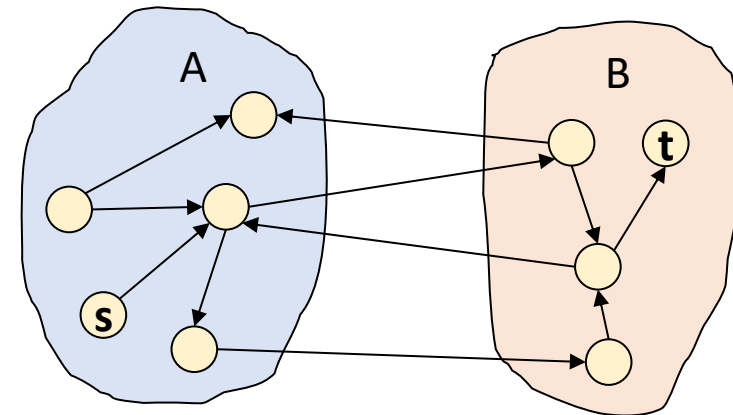
(iii) \Rightarrow (i):

Claim: If there is no augmenting path w.r.t. f , there is a cut (A, B) s.t. $v(f) = c(A, B)$.

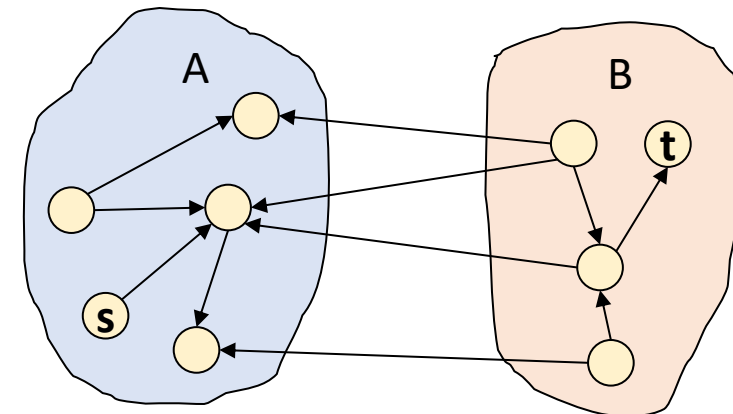
Proof of Claim: Let f be a flow with no augmenting paths.

Let A be the set of vertices reachable from s in residual graph G_f .

- By definition of A , $s \in A$.
- Since no augmenting path (s - t path in G_f), $t \notin A$.



original network



residual graph

Proof: Identifying the Min Cut

(iii) \Rightarrow (i):

Claim: If there is no augmenting path w.r.t. f , there is a cut (A, B) s.t. $v(f) = c(A, B)$.

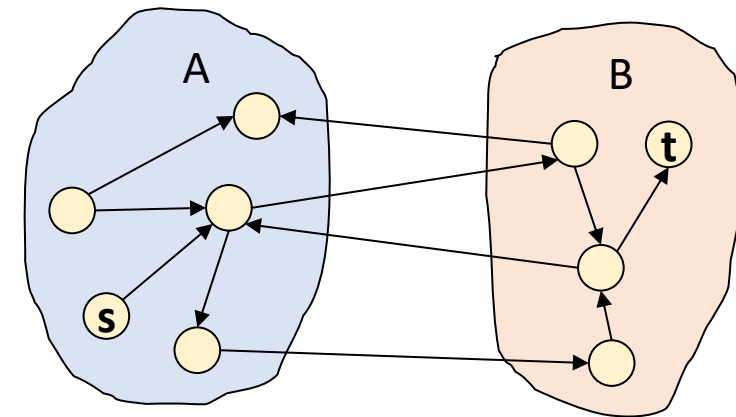
Proof of Claim: Let f be a flow with no augmenting paths.

Let A be the set of vertices reachable from s in residual graph G_f .

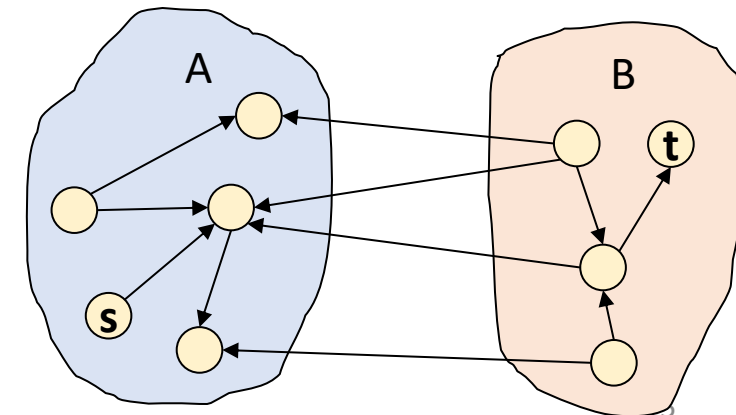
- By definition of A , $s \in A$.
- Since no augmenting path (s - t path in G_f), $t \notin A$.

Then

$$v(f) = \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e) \quad (\text{by Flow-Value Lemma})$$



original network



residual graph

Identifying the Min Cut: No Inflow

(iii) \Rightarrow (i):

Claim: If there is no augmenting path w.r.t. f , there is a cut (A, B) s.t. $v(f) = c(A, B)$.

Proof of Claim: Let f be a flow with no augmenting paths.

Let A be the set of vertices reachable from s in residual graph G_f .

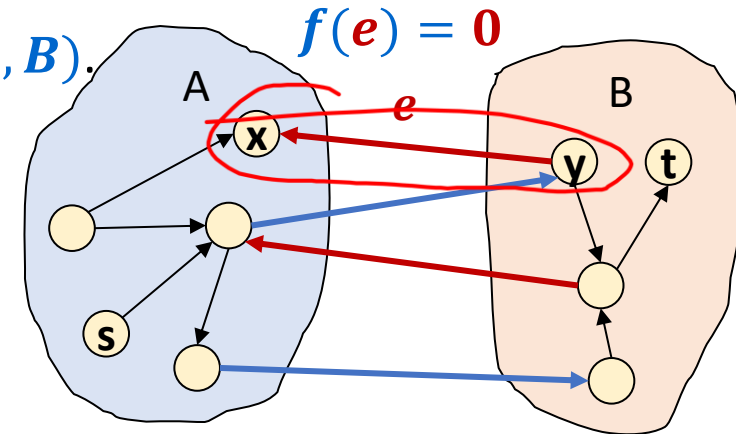
- By definition of A , $s \in A$.
- Since no augmenting path (s - t path in G_f), $t \notin A$.

Then

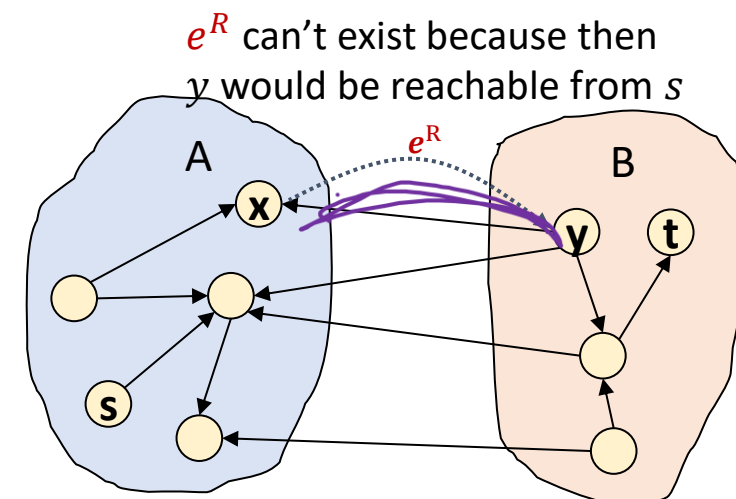
$$v(f) = \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e)$$

$$= \sum_{e \text{ out of } A} f(e)$$

(By contradiction: If an edge going into A had flow then the backward edge would be in the residual graph, so the edge should not cross the cut)



original network



residual graph

Identifying the Min Cut: Saturated Outflow

(iii) \Rightarrow (i):

Claim: If there is no augmenting path w.r.t. f , there is a cut (A, B) s.t. $v(f) = c(A, B)$.

Proof of Claim: Let f be a flow with no augmenting paths.

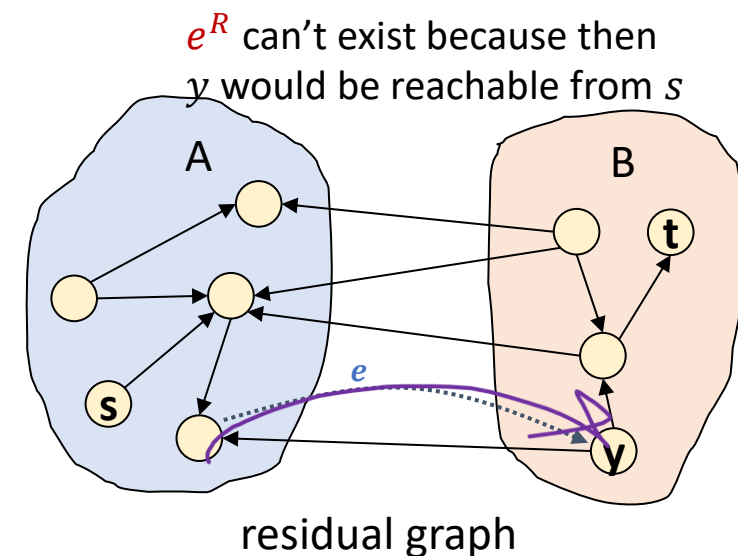
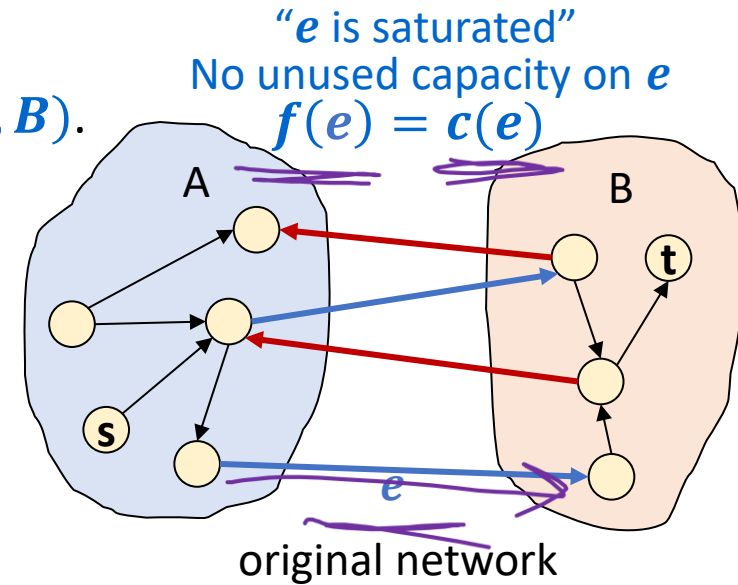
Let A be the set of vertices reachable from s in residual graph G_f .

- By definition of A , $s \in A$.
- Since no augmenting path (s - t path in G_f), $t \notin A$.

Then

$$\begin{aligned}
 v(f) &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e) \\
 &= \sum_{e \text{ out of } A} c(e) \\
 &= \sum_{e \text{ out of } A} c(e)
 \end{aligned}$$

(By contradiction: If an edge going out of A had unused capacity then the forward edge would be in the residual graph, so the edge should not cross the cut)



Identifying the Min Cut: Conclusion

(iii) \Rightarrow (i):

Claim: If there is no augmenting path w.r.t. f , there is a cut (A, B) s.t. $v(f) = c(A, B)$.

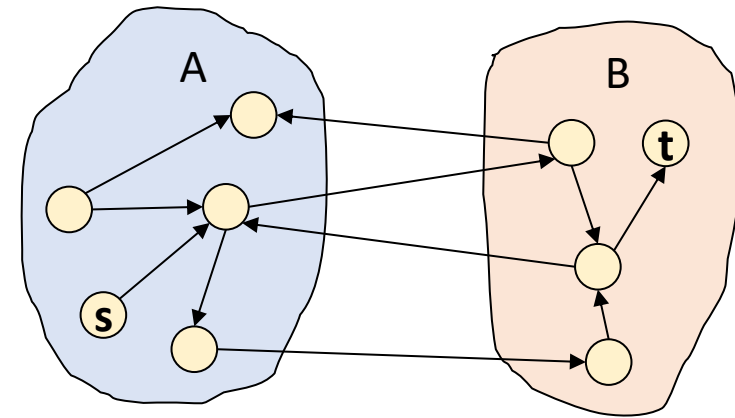
Proof of Claim: Let f be a flow with no augmenting paths.

Let A be the set of vertices reachable from s in residual graph G_f .

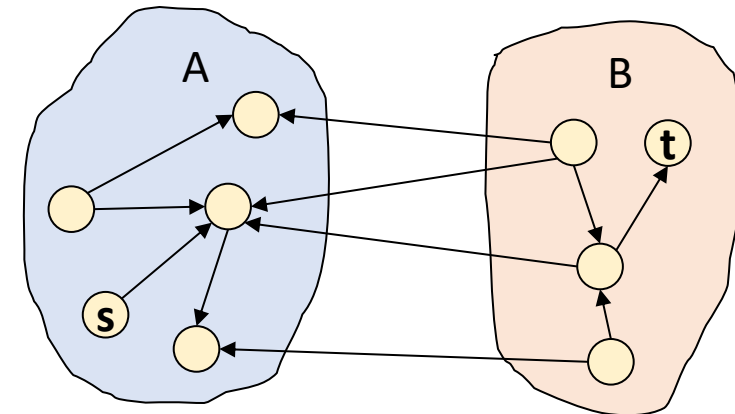
- By definition of A , $s \in A$.
- Since no augmenting path (s - t path in G_f), $t \notin A$.

Then

$$\begin{aligned} v(f) &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e) \\ &= \sum_{e \text{ out of } A} c(e) = c(A, B) \quad (\text{by Definition}) \end{aligned}$$



original network



residual graph

Fork Fulkerson Algorithm

```
FordFulkerson(G, s, t, c){  
  for each  $e \in E$ {  
    set  $f(e) = 0$   
  }  
  calculate residual graph  $G_f$   
  while  $G_f$  has an  $s - t$  path  $P$ {  
    augment( $f, c, P$ )  
    update  $G_f$   
  }  
  return  $f$ 
```

```
augment( $f, c, P$ ){  
   $b = \text{bottleneck}(P)$   
  for each  $e \in P$ {  
     $f(e) += b$   
     $f(e^R) -= b$   
  }  
  return  $f$ 
```

MaxFlow/MinCut & Ford-Fulkerson Algorithm

Augmenting Path Theorem: Flow f is a max flow \Leftrightarrow there are no augmenting paths wrt f

Max-Flow Min-Cut Theorem: The value of the max flow equals the value of the min cut.

[Elias-Feinstein-Shannon 1956, Ford-Fulkerson 1956]

“MaxFlow = MinCut”

Flow Integrality Theorem: If all capacities are integers then there is a maximum flow with all-integer flow values.

Ford-Fulkerson Algorithm: $O(m)$ per iteration. With integer capacities each at most C need at most **MaxFlow** $< nC$ iterations for a total of $O(mnC)$ time.

$$\begin{aligned} m &= |E| \\ n &= |V| \end{aligned}$$

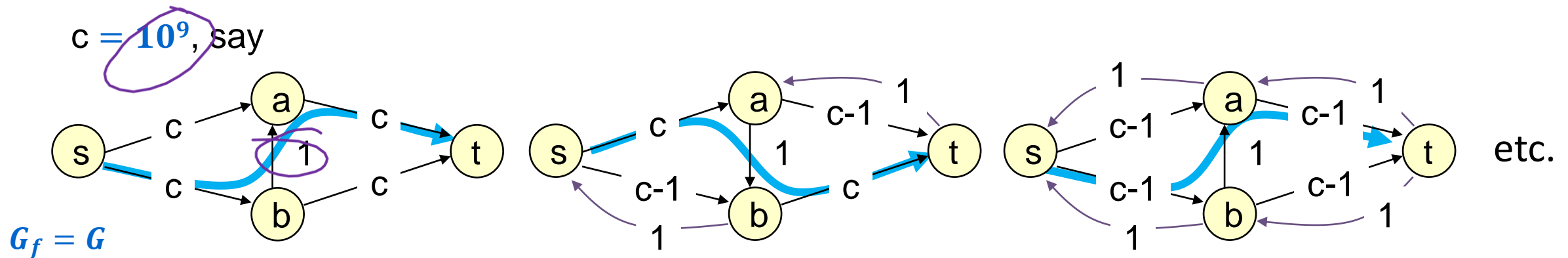
Ford-Fulkerson Efficiency

Worst case runtime $O(mnC)$ with integer capacities $\leq C$.

- $O(m)$ time per iteration.
- At most nC iterations.
- This is “pseudo-polynomial” running time.

$(\log(10^9))$

- May take exponential time, even with integer capacities:



Polynomial-Time Variant of Ford-Fulkerson

Use care when selecting augmenting paths.

- Some choices lead to exponential algorithms.
- Clever choices lead to polynomial algorithms.
- If capacities are irrational, algorithm not guaranteed to terminate!

Goal: Choose augmenting paths so that:

- Can find augmenting paths efficiently.
- Few iterations.

Choose augmenting paths with fewest number of edges. [Edmonds-Karp 1972 , Dinitz 1970]

- Just run BFS to find an augmenting path!

Edmonds-Karp Algorithm (Ford-Fulkerson with BFS)

Use Breadth First Search as the search algorithm to find an $s-t$ path in G_f .

- Using any **shortest** augmenting path

Theorem: Ford-Fulkerson using BFS terminates in $O(m^2n)$ time. [Edmonds-Karp, Dinitz]

“One of the most obvious ways to implement Ford-Fulkerson is always polynomial time”

Why might this be good intuitively?

- Longer augmenting paths involve more edges so may be more likely to hit a low residual capacity one which would limit the amount of flow improvement.

The proof uses a completely different idea...

Edmonds-Karp Algorithm (Ford-Fulkerson with BFS)

Analysis Focus:

For any edge e that could be in the residual graph G_f , (either an edge in G or its reverse) count # of iterations that e is the **first bottleneck edge** on the augmenting path chosen by the algorithm.

Claim: This can't happen in more than $n/2$ iterations.

Proof: Write $e = (u, v)$.

Show that each time it happens, the distance from s to u in the residual graph G_f is at least **2** more than it was the last time.

This would be enough since the distance is $< n$ (or infinite and hence u isn't reachable) so this can happen at most $n/2$ times.



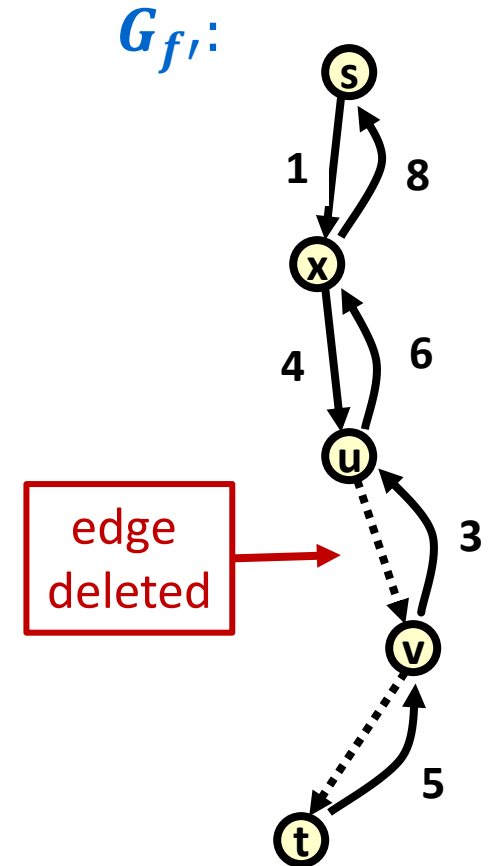
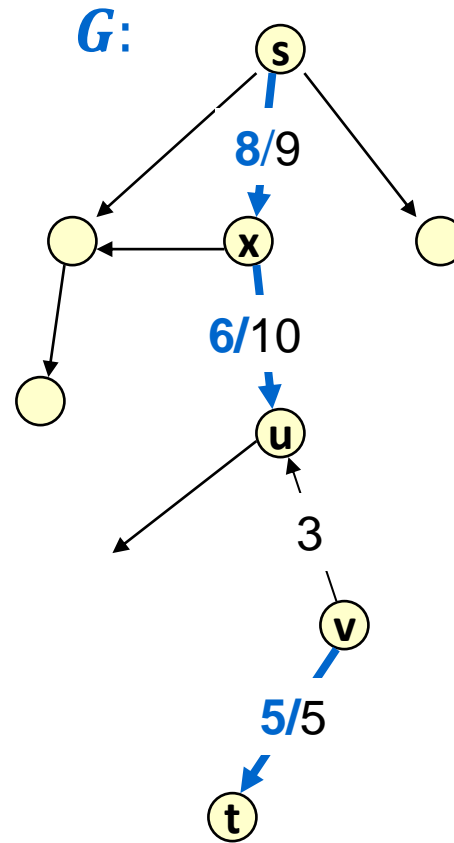
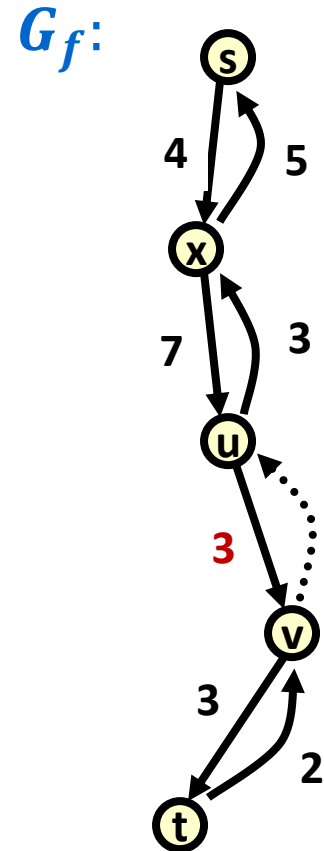
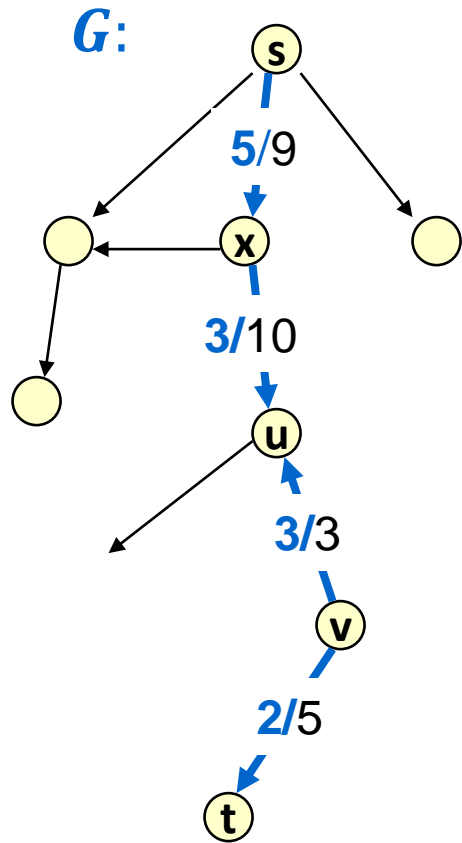
Distances in the Residual Graph

Key Lemma: Let f be a flow, G_f the residual graph, and P be a shortest augmenting path. No vertex is closer to s in the residual graph after augmenting along P .

Proof: Augmenting along P can only change the edges in G_f by either:

1. Deleting a forward edge
 - Deleting any edge can never reduce distances
2. Add a backward edge (v, u) that is the reverse of an edge (u, v) of P
 - Since P was a shortest path in G_f , the distance from s to v in G_f is already more than the distance from s to u . Using the new backward edge (v, u) to get to u would be an even longer path to u so it is never on a shortest path to any node in the new residual graph.

Augmentation vs BFS



First Bottleneck Edges in G_f

Shortest s - t path P in G_f

Write $c_P = \text{bottleneck}(P)$



$d_f(s, v) = d_f(s, u) + 1$ since P is a shortest path.

After augmenting along P , edge (u, v) disappears; but will have edge (v, u)



distance is ≥ 2
larger than before

For (u, v) to be a first bottleneck edge later, it must get added back to the residual graph by augmenting along a shortest path P' containing (v, u) in $G_{f'}$, for some flow f'

Since P' is shortest $d_{f'}(s, u) = d_{f'}(s, v) + 1 \geq d_f(s, v) + 1 = d_f(s, u) + 2$

The next time that (u, v) is first bottleneck edge is even later so distance is at least as large!

Edmonds-Karp Algorithm (Ford-Fulkerson with BFS)

Analysis Focus:

For any edge e that could be in the residual graph G_f , (either an edge in G or its reverse) count # of iterations that e is the **first bottleneck edge** on the augmenting path chosen by the algorithm.

Claim: This can't happen in more than $n/2$ iterations

Claim \Rightarrow Theorem:

Only $2m$ edges and $O(m)$ time per iteration so $O(m^2n)$ time overall.

History & State of the Art for MaxFlow Algorithms

#	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	$O(m^{3/2} \log(n^2/m) \log U)$ $O(n^{2/3} m \log(n^2/m) \log U)$

Source: Goldberg & Rao, FOCS '97

21	2013	Orlin	$O(mn)$
22	2014	Lee & Sidford	$m\sqrt{n} \log^{O(1)} n \log U$
23	2016	Madry	$m^{10/7} U^{1/7} \log^{O(1)} n$
24	2021	Gao, Liu, & Peng	$m^{3/2-1/328} \log^{O(1)} n \log U$
25	2022	van den Brand et al.	$m^{3/2-1/58} \log^{O(1)} n \log U$
26	2022	Chen et al.	$m^{1+o(1)} \log U$

Tables use U instead of C for the upper bound on capacities

Methods:

Augmenting Paths – increase flow to capacity

Preflow-Push – decrease flow to get flow conservation

Linear Programming – randomized, high probability of optimality