

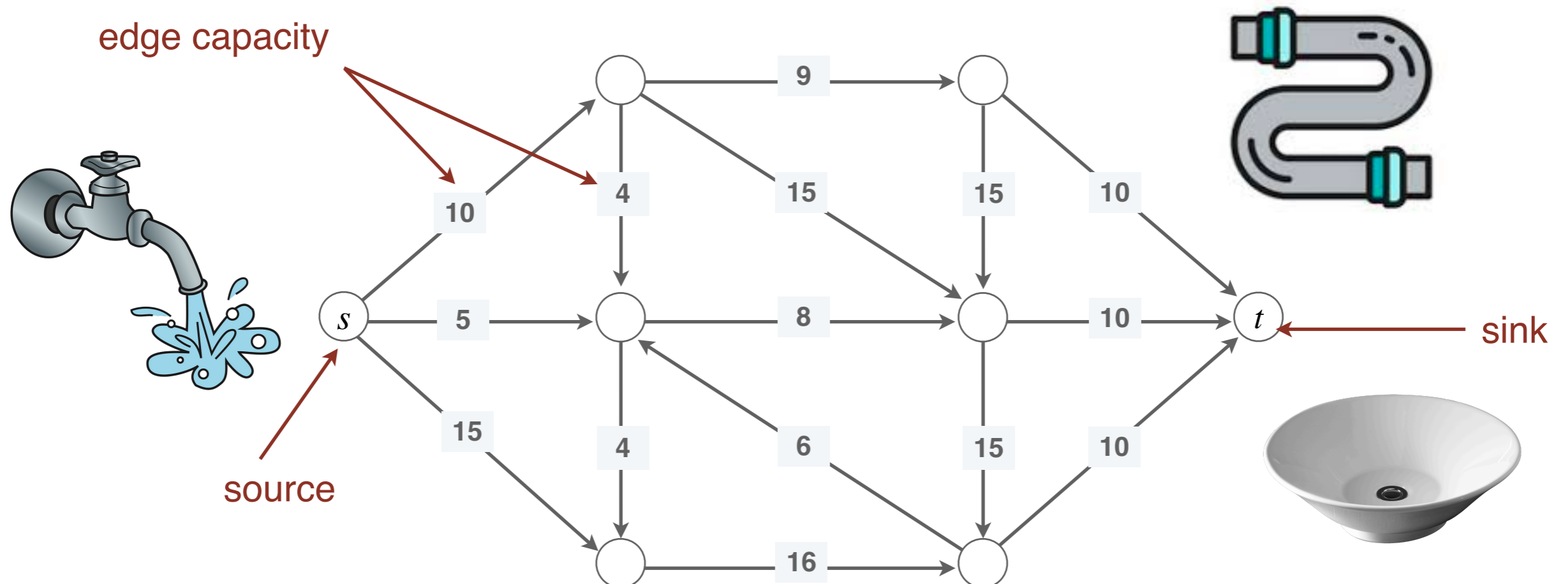
Network Flows

Admin

- TA evaluation form! <https://forms.gle/nZSPcwbaP3WCWxqEA>
 - Please fill out by next Friday
- Assignment 4 back to you ASAP (Friday?); Assignment 5 after
- Network flow practice problem canceled
 - Just do the practice midterm instead
- Practice midterm posted on GLOW, linked from website
 - *Draft* solutions out ASAP (Friday?)
 - Final will look very similar with better (?) problems, length
- Questions?

What's a Flow Network?

- A flow network is a directed graph $G = (V, E)$ with a
 - A **source** is a vertex s with in degree 0
 - A **sink** is a vertex t with out degree 0
 - Each edge $e \in E$ has **edge capacity** $c(e) > 0$



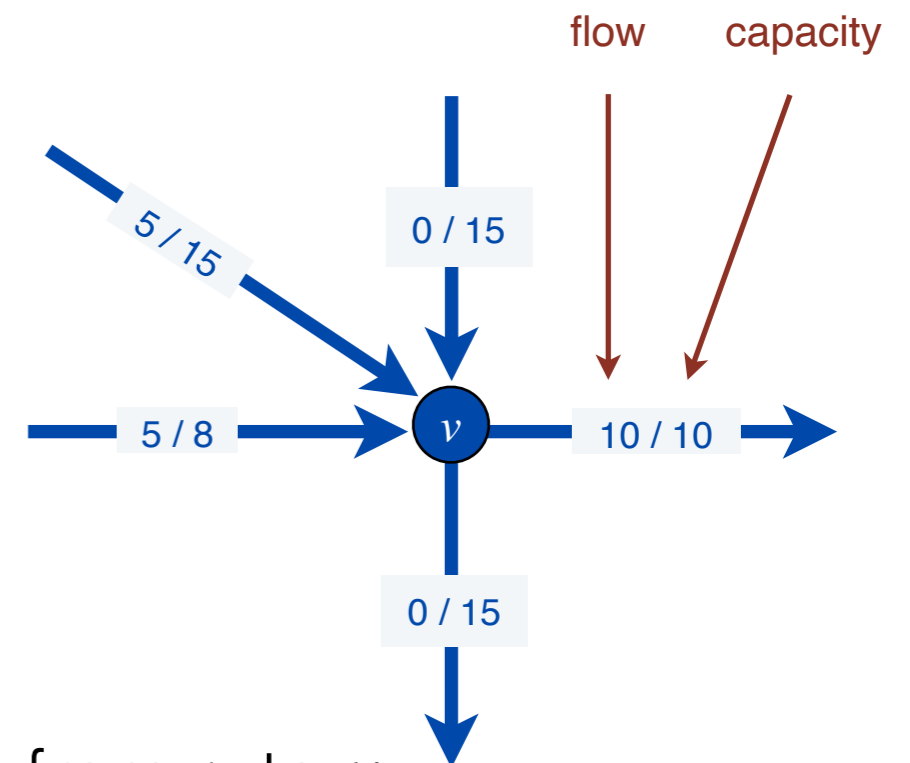
Assumptions

- Assume that each node v is on some s - t path, that is, $s \rightsquigarrow v \rightsquigarrow t$ exists, for any vertex $v \in V$
 - Implies G is connected and $m \geq n - 1$
- Assume **capacities are integers**
 - Will revisit this assumption and what happens if not
- Directed edge (u, v) written as $u \rightarrow v$
- For simplifying expositions, we will sometimes write $c(u \rightarrow v) = 0$ when $(u, v) \notin E$

What's a Flow?

- Given a flow network, an (s, t) -**flow** or just **flow** (if source s and sink t are clear from context) $f: E \rightarrow \mathbb{Z}^+$ satisfies the following two constraints:
- [Flow conservation]** $f_{in}(v) = f_{out}(v)$, for $v \neq s, t$ where

$$f_{in}(v) = \sum_u f(u \rightarrow v)$$
$$f_{out}(v) = \sum_w f(v \rightarrow w)$$

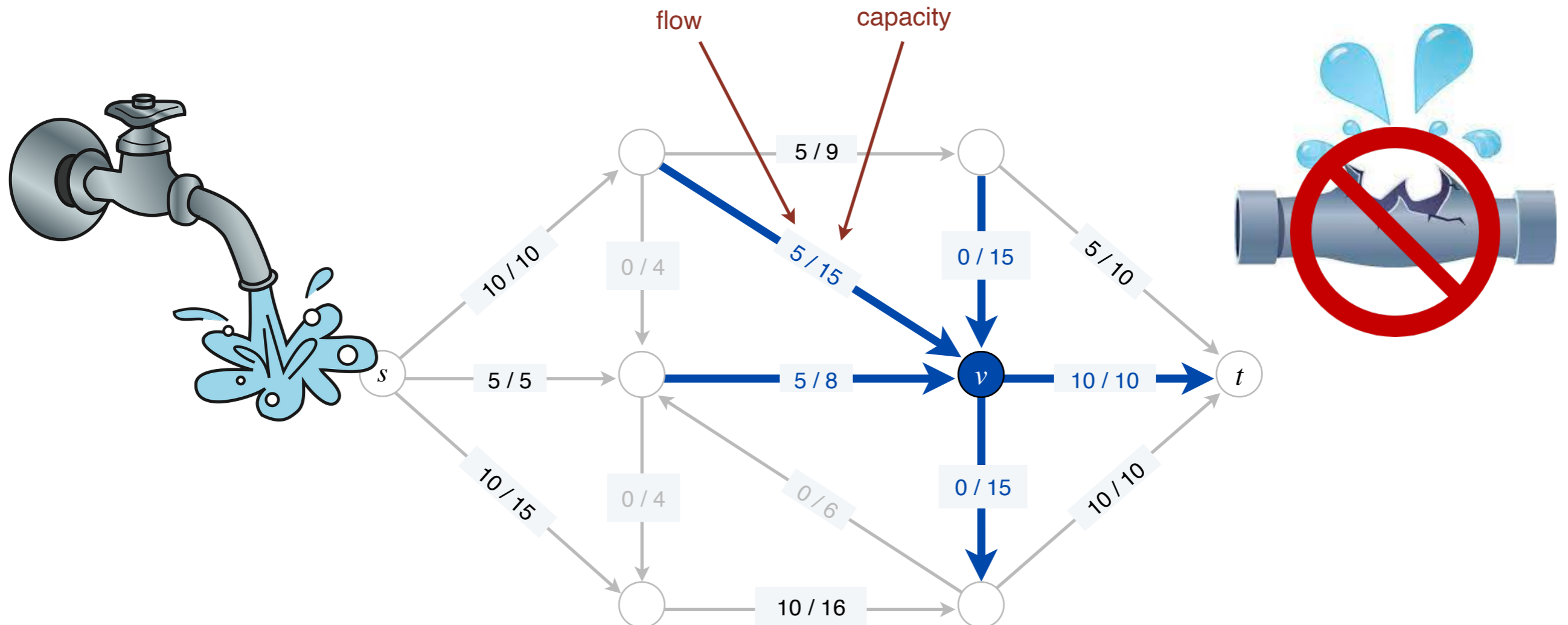


- To simplify, $f(u \rightarrow v) = 0$ if there is no edge from u to v

Feasible Flow

- And second, a feasible flow must satisfy the capacity constraints of the network, that is,

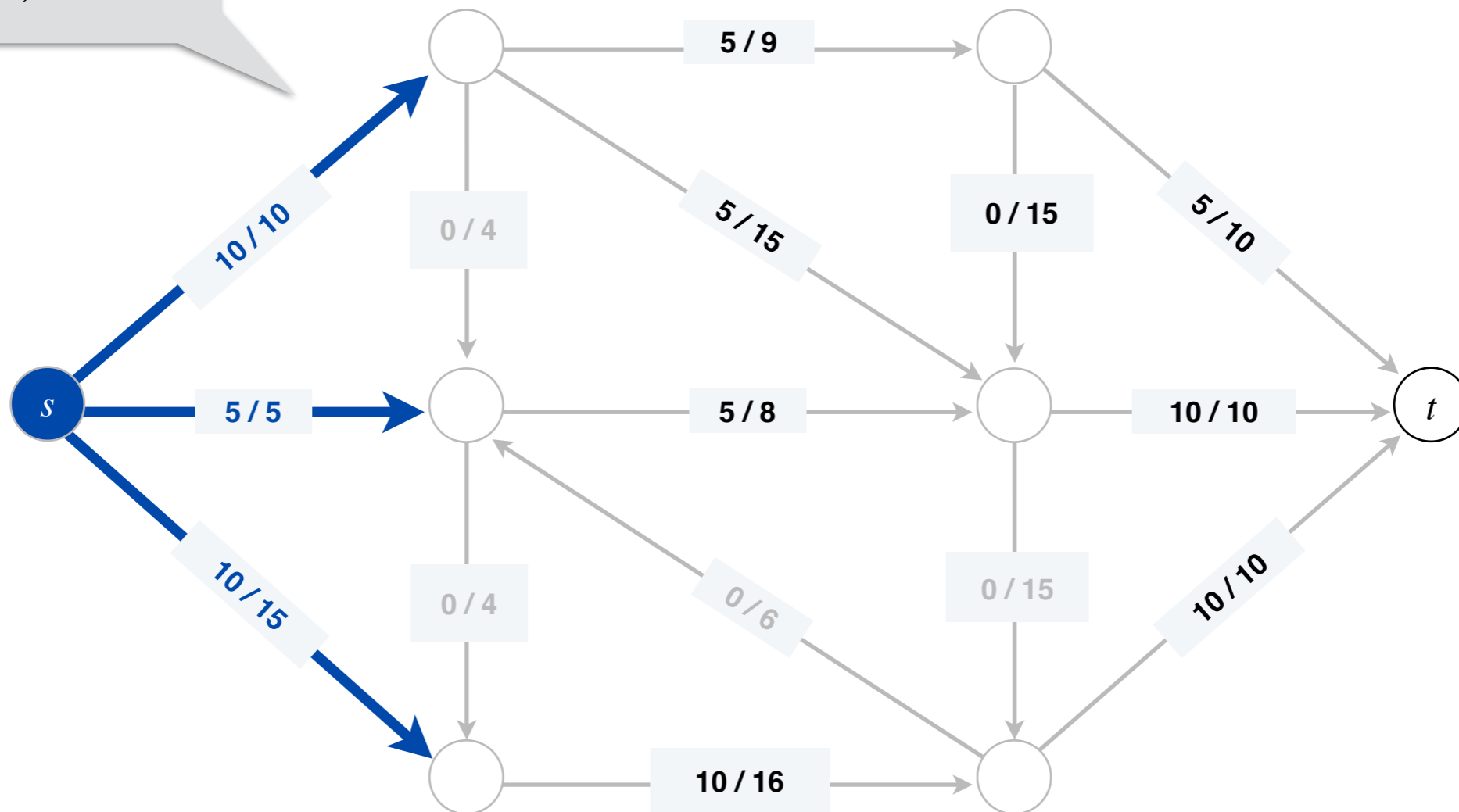
[Capacity constraint] for each $e \in E$, $0 \leq f(e) \leq c(e)$



Value of a Flow

- **Definition.** The **value** of a flow f , written $v(f)$, is $f_{out}(s)$.

What is $v(f)$ here?

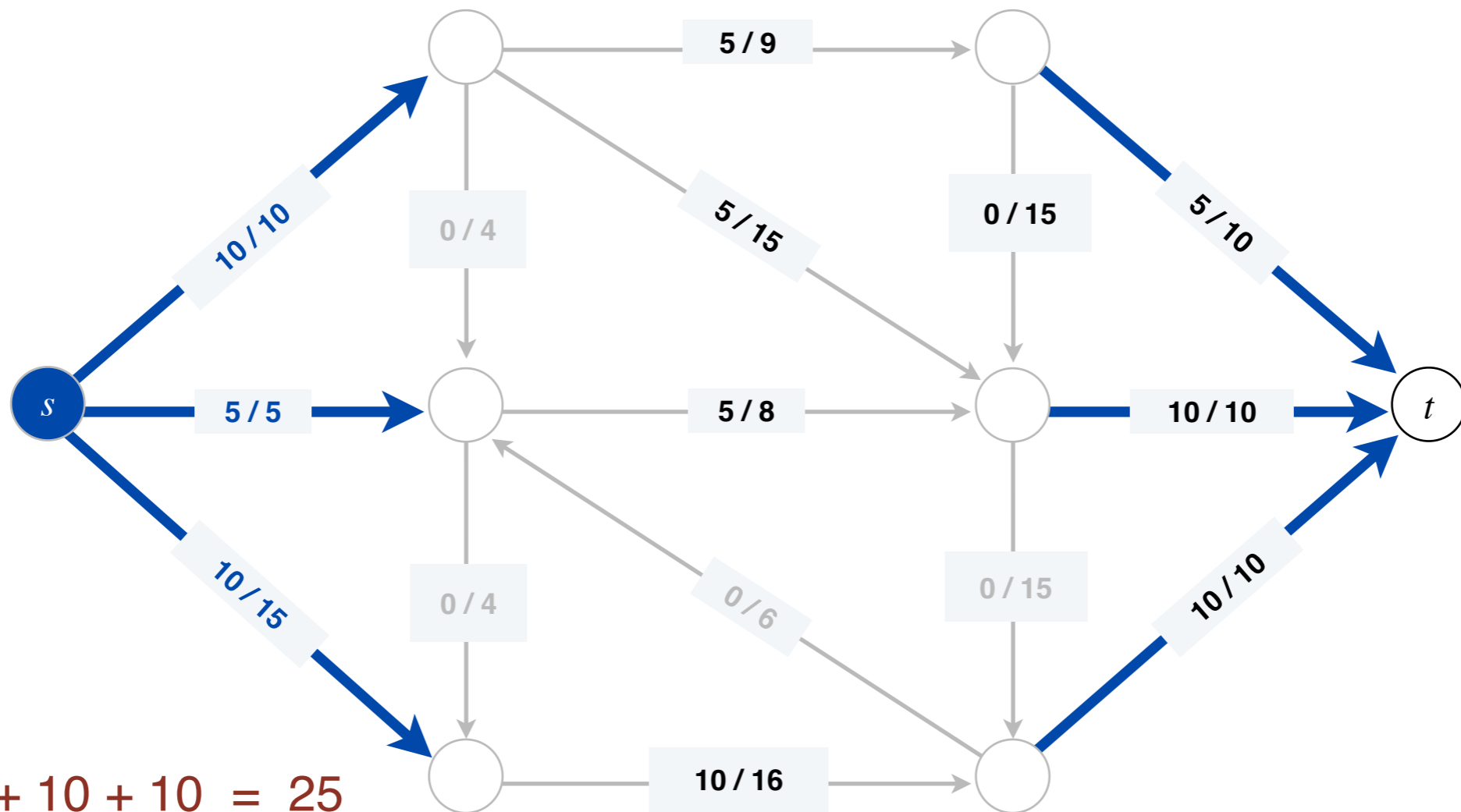


$$v(f) = 5 + 10 + 10 = 25$$

Value of a Flow

- **Definition.** The **value** of a flow f , written $v(f)$, is $f_{out}(s)$.
- **Lemma.** $f_{out}(s) = f_{in}(t)$

Intuitively, why do you think this is true?



Value of a Flow

- **Lemma.** $f_{out}(s) = f_{in}(t)$

- **Proof.** Let $f(E) = \sum_{e \in E} f(e)$

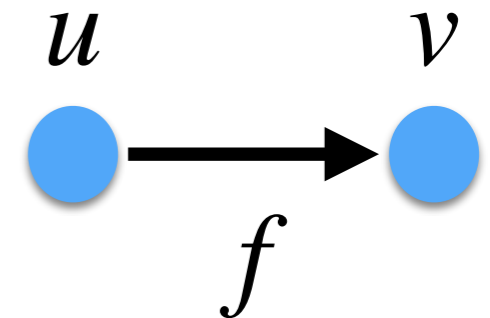
- Then, $\sum_{v \in V} f_{in}(v) = f(E) = \sum_{v \in V} f_{out}(v)$

- For every $v \neq s, t$ flow conservation implies $f_{in}(v) = f_{out}(v)$

- Thus all terms cancel out on both sides except

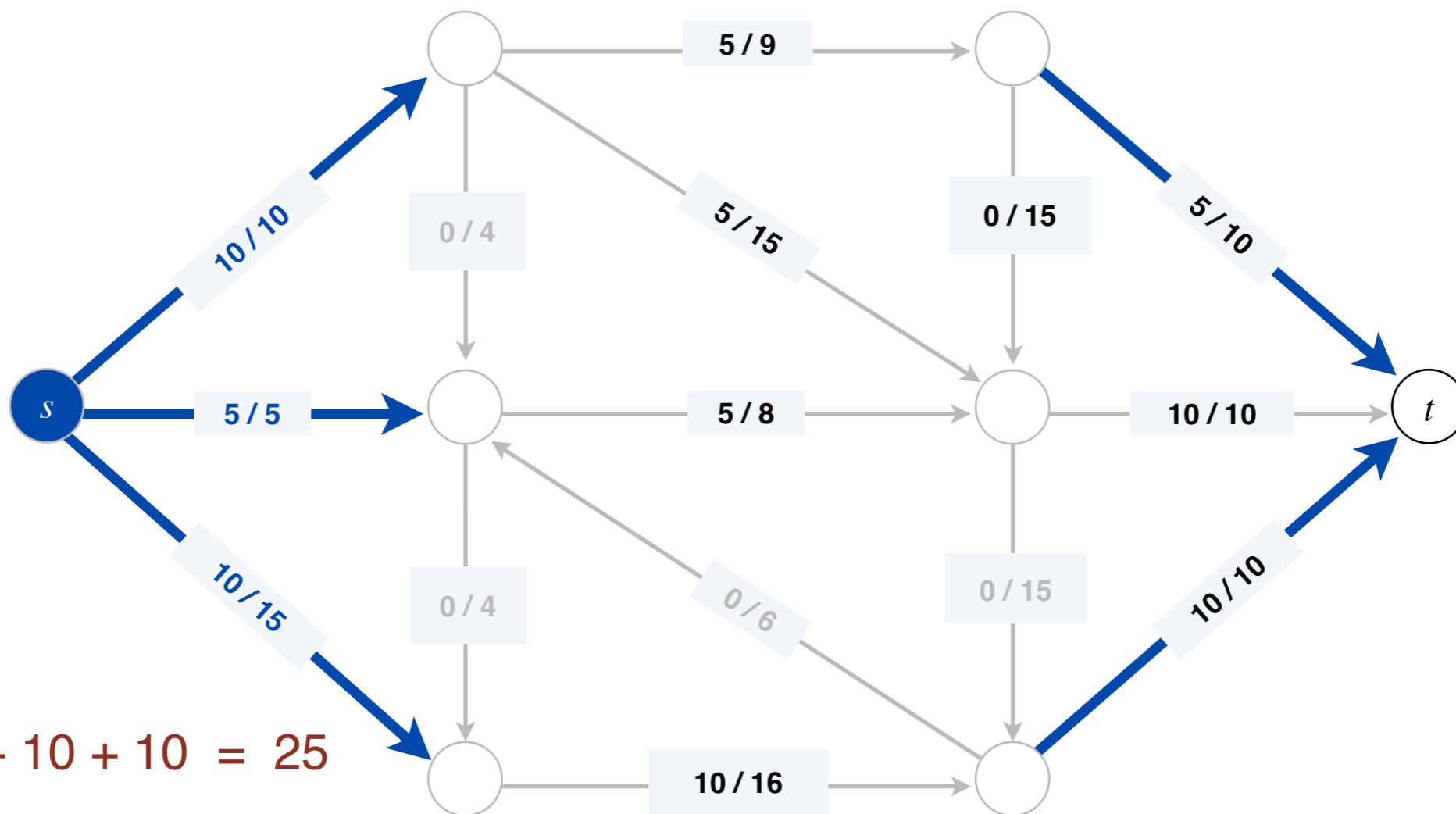
$$f_{in}(s) + f_{in}(t) = f_{out}(s) + f_{out}(t)$$

- But $f_{in}(s) = f_{out}(t) = 0$ ■



Value of a Flow

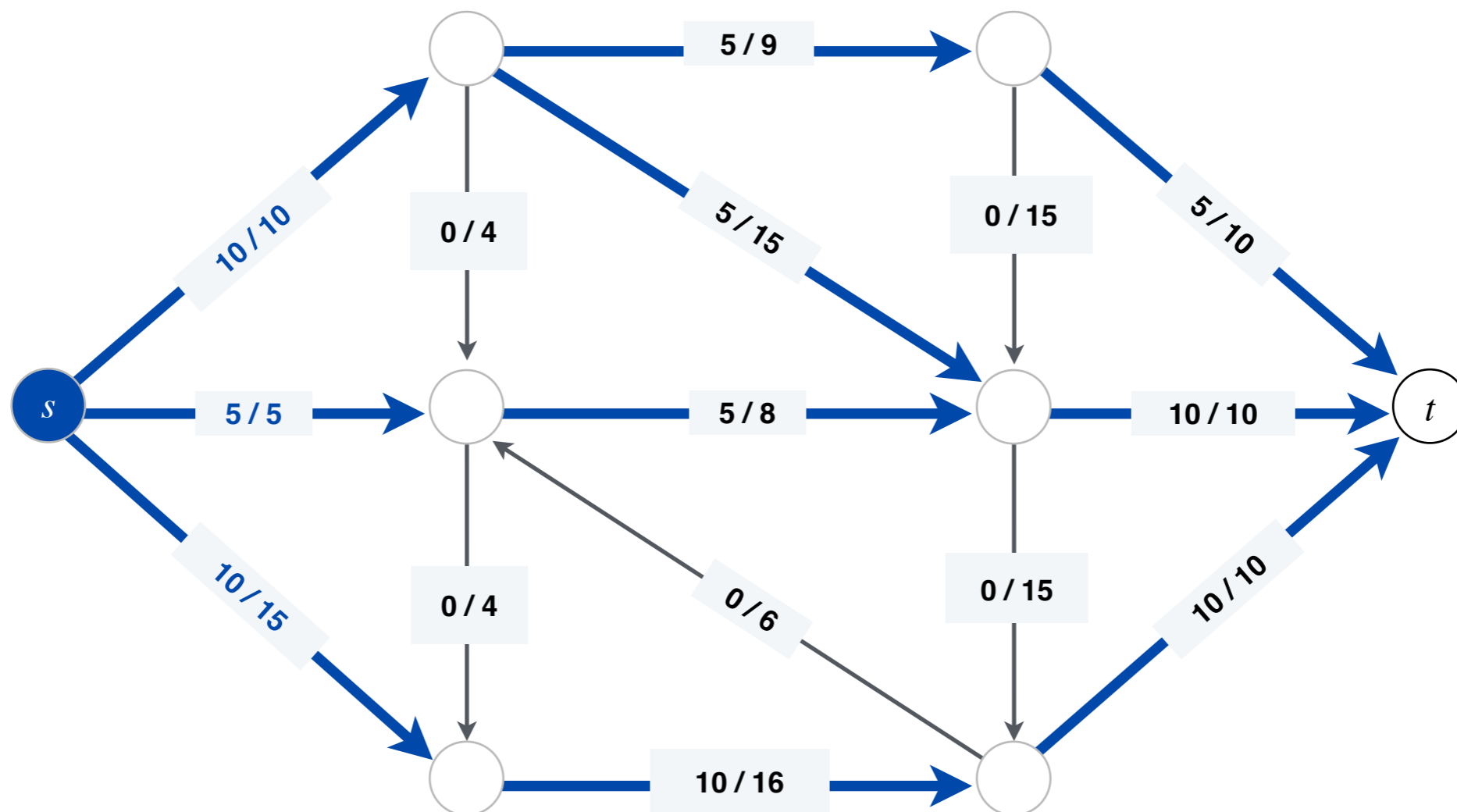
- **Lemma.** $f_{out}(s) = f_{in}(t)$
- **Corollary.** $v(f) = f_{in}(t)$.



value = 5 + 10 + 10 = 25

Max-Flow Problem

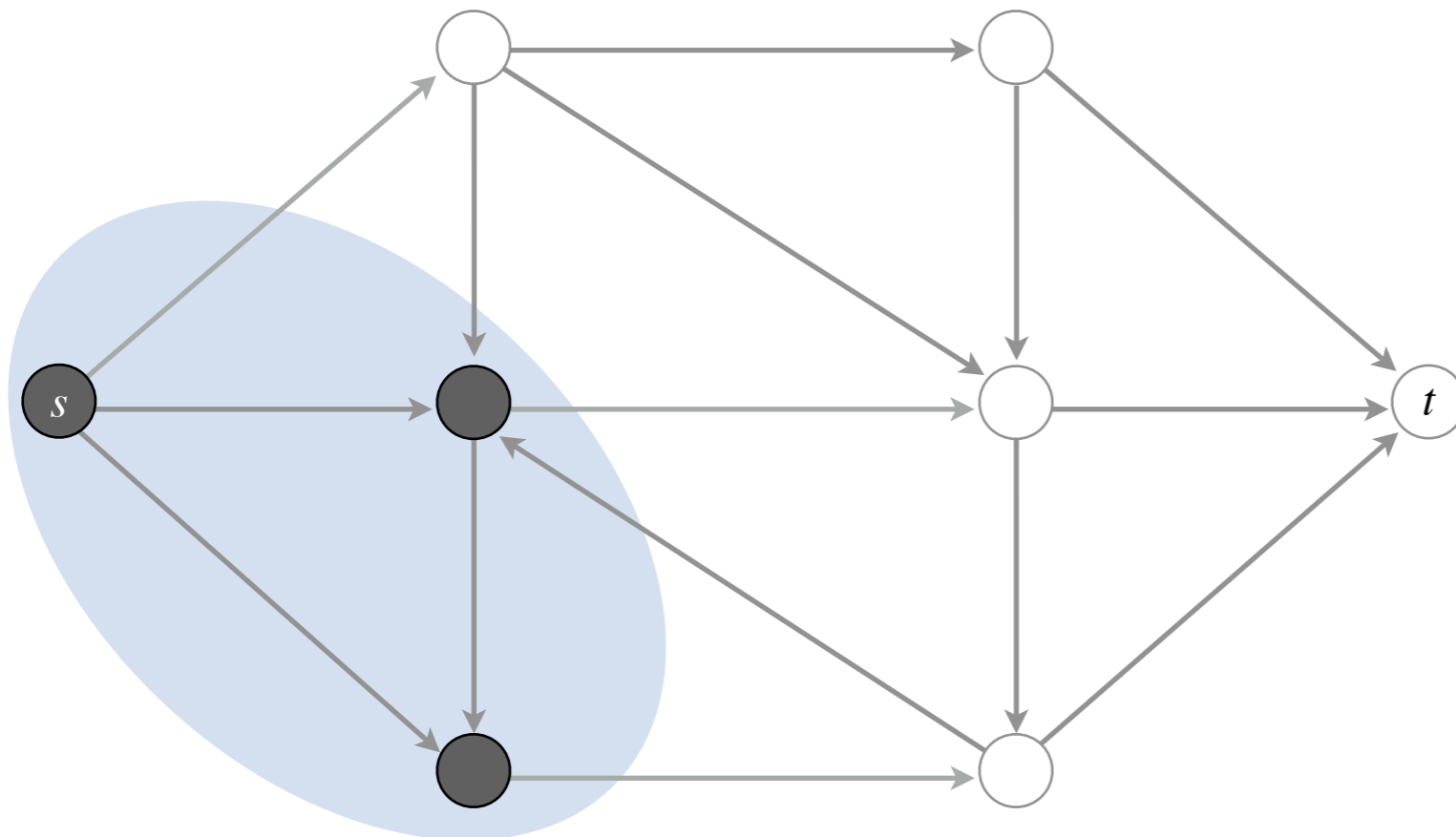
- **Problem.** Given an s - t flow network, find a feasible s - t flow of **maximum** value.



Minimum Cut Problem

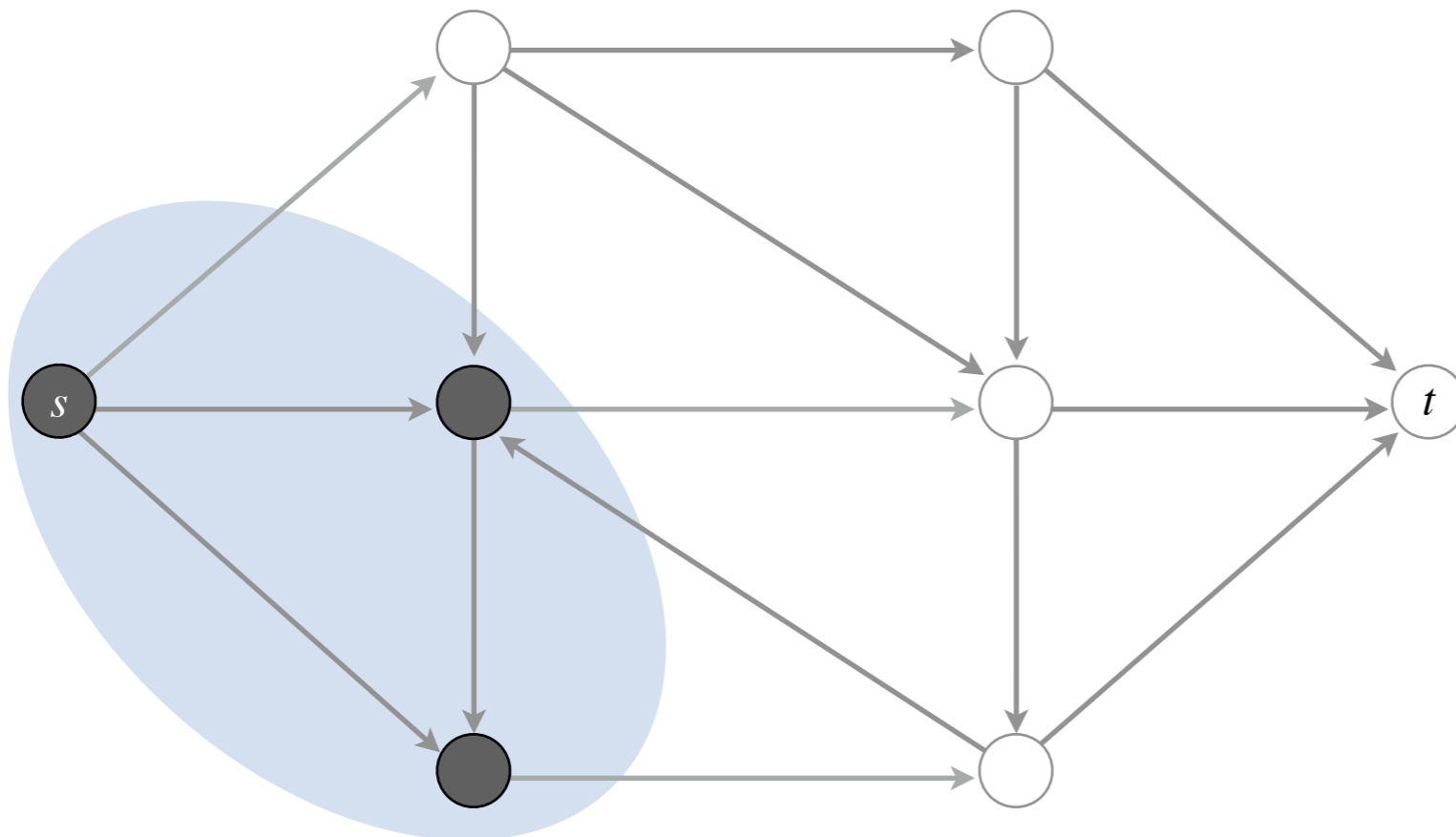
Cuts are Back!

- Cuts in graphs played a lead role when we were designing algorithms for MSTs
- What is the definition of a cut?



Cuts in Flow Networks

- **Recall.** A cut (S, T) in a graph is a partition of vertices such that $S \cup T = V$, $S \cap T = \emptyset$ and S, T are non-empty.
- **Definition.** An (s, t) -cut is a cut (S, T) s.t. $s \in S$ and $t \in T$.



Cut Capacity

- **Recall.** A cut (S, T) in a graph is a partition of vertices such that $S \cup T = V$, $S \cap T = \emptyset$ and S, T are non-empty.
- **Definition.** An (s, t) -cut is a cut (S, T) s.t. $s \in S$ and $t \in T$.
- **Capacity** of a (s, t) -cut (S, T) is the sum of the capacities of edges **leaving** S :

$$c(S, T) = \sum_{v \in S, w \in T} c(v \rightarrow w)$$

Quick Quiz

Question. What is the capacity of the s - t cut given by grey and white nodes?

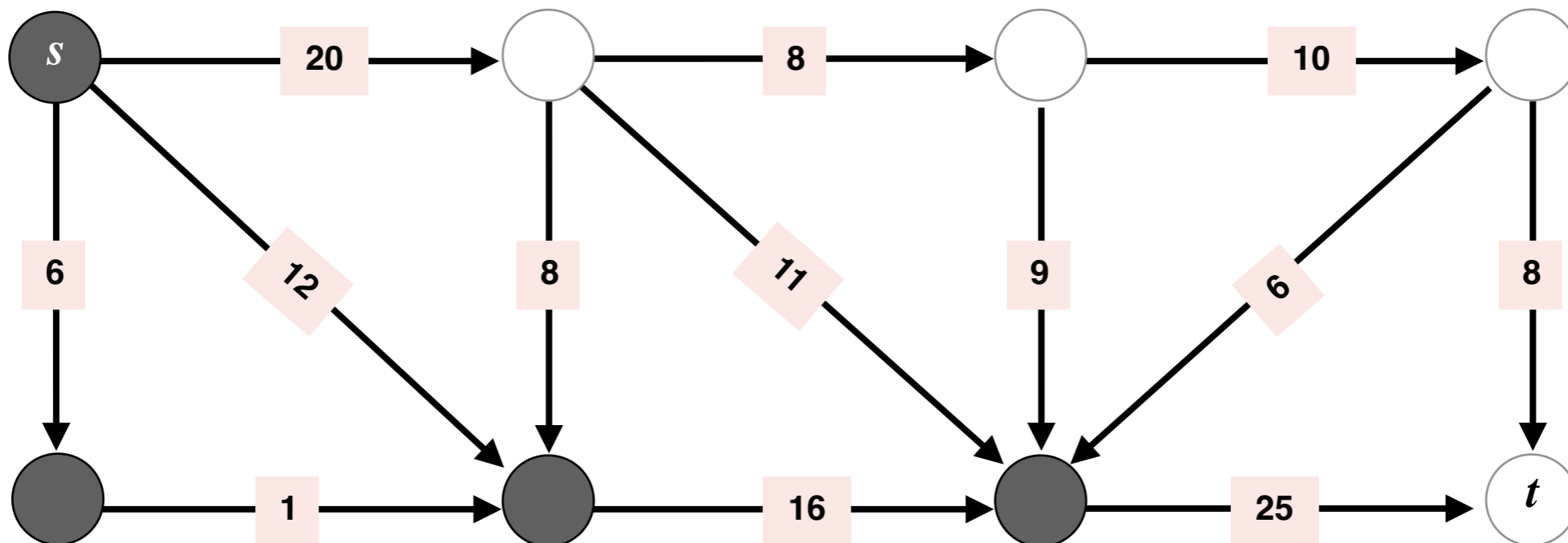
A. 11 (20 + 25 - 8 - 11 - 9 - 6)

B. 34 (8 + 11 + 9 + 6)

C. 45 (20 + 25)

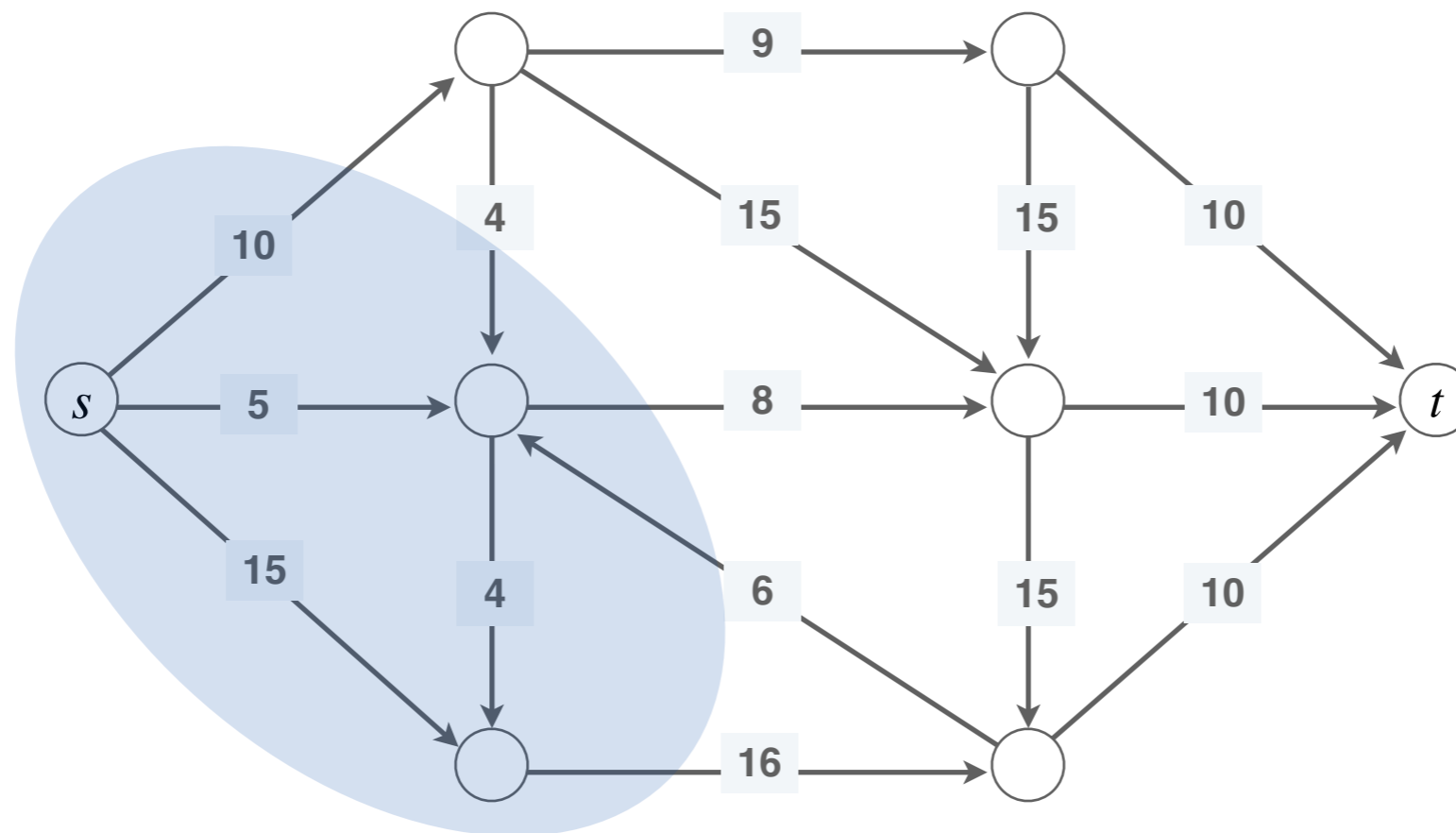
D. 79 (20 + 25 + 8 + 11 + 9 + 6)

$$c(S, T) = \sum_{v \in S, w \in T} c(v \rightarrow w)$$



Min Cut Problem

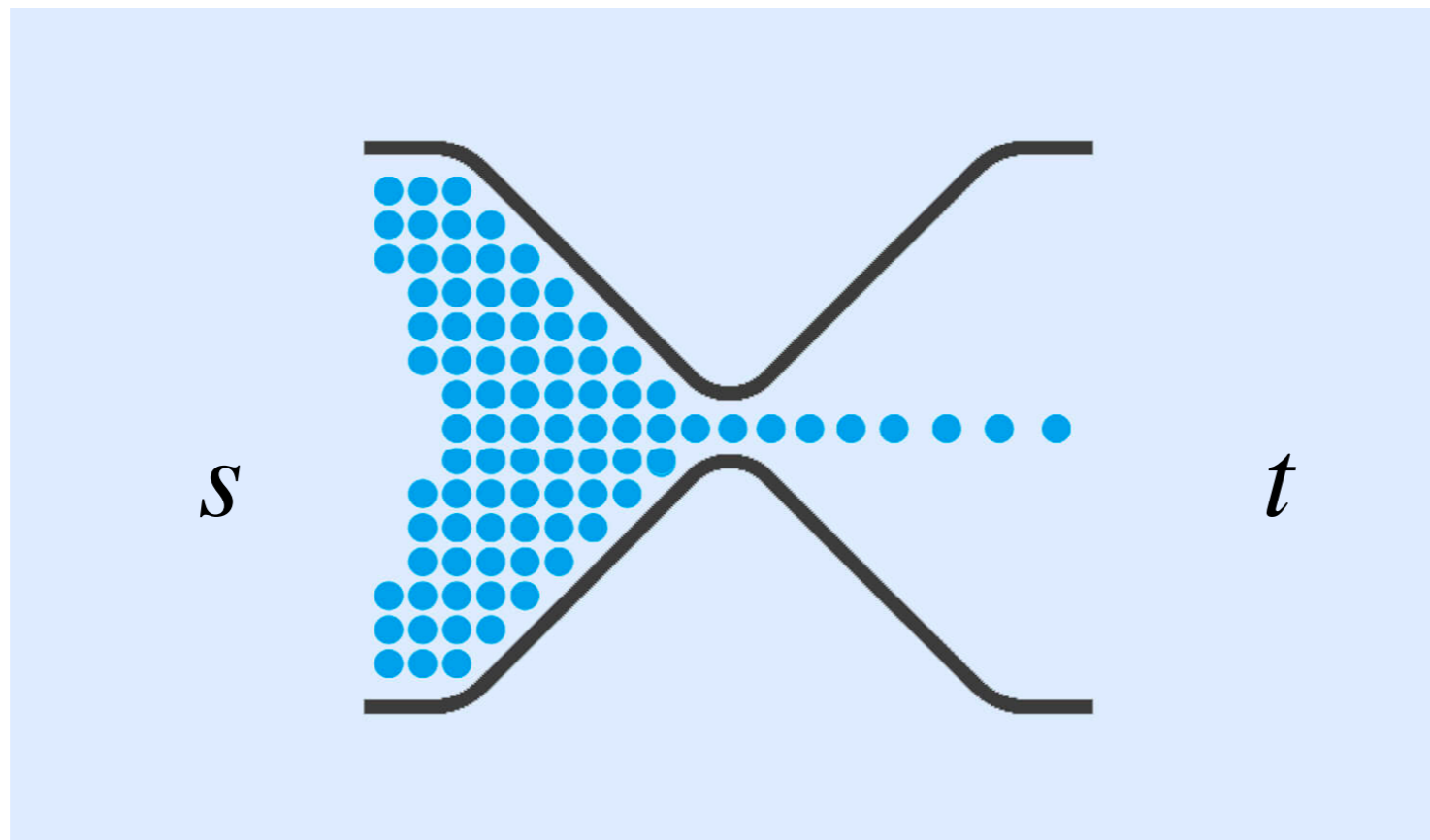
- **Problem.** Given an s - t flow network, find an s - t cut of **minimum** capacity.



Relationship between Flows and Cuts

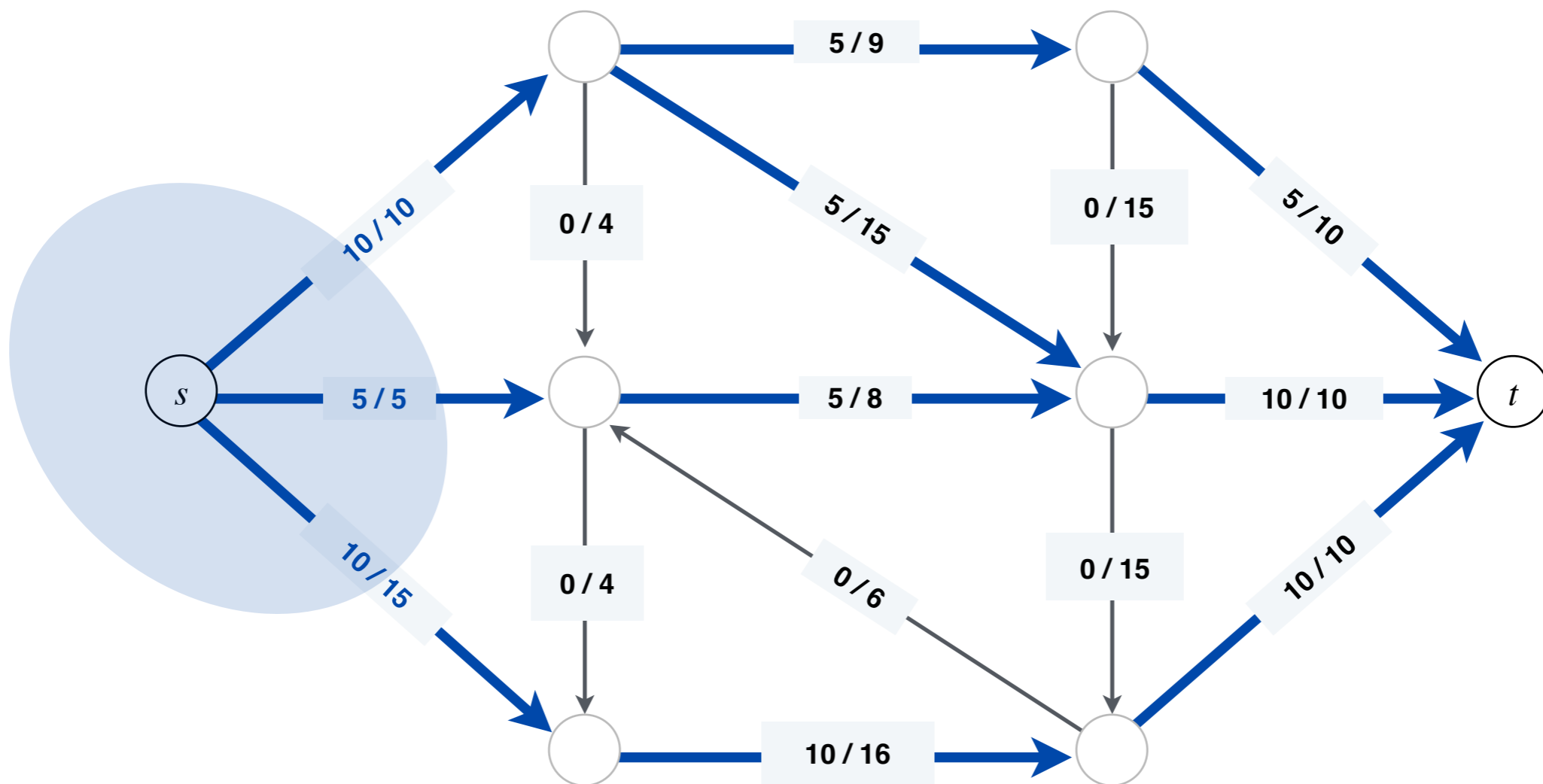
Flows and Cuts

- Cuts represent "**bottlenecks**" in a flow network
- For any cut, our flow needs to "get out" of that cut on its route from s to t
- Let us formalize this intuition



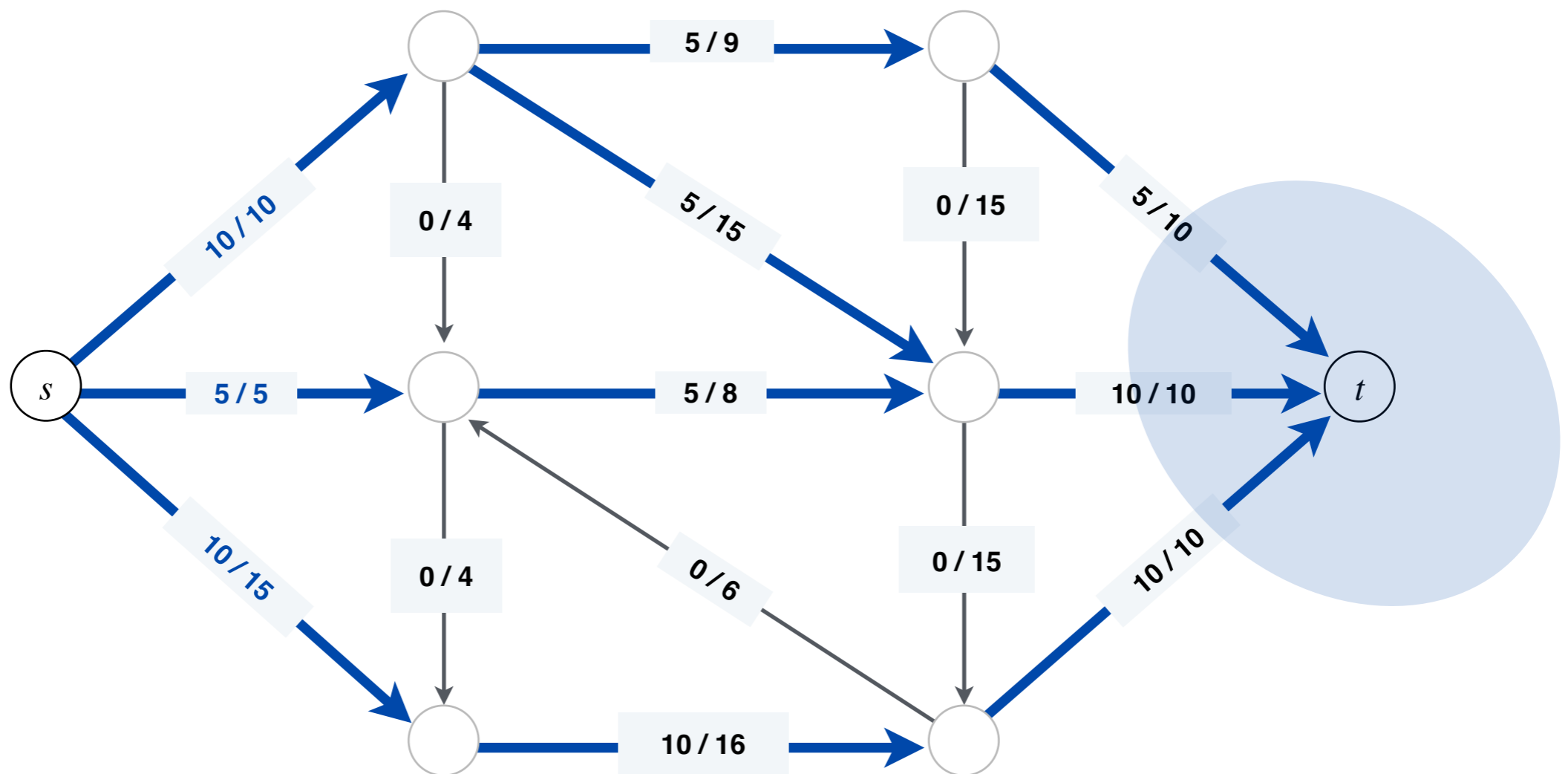
Flows and Cuts

- **Claim.** Let f be **any** s - t flow and (S, T) be **any** s - t cut then $v(f) \leq c(S, T)$
- There are two s - t cuts for which this is easy to see, which ones?



Flows and Cuts

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- There are two s - t cuts for which this is easy to see, which ones?



Flows and Cuts

- To prove this for any cut, we first relate the flow value in a network to the net flow leaving a cut
- **Lemma.** For any feasible (s, t) -flow f on $G = (V, E)$ and any (s, t) -cut, $v(f) = f_{out}(S) - f_{in}(S)$, where

- $f_{out}(S) = \sum_{v \in S, w \in T} f(v \rightarrow w)$ (sum of flow 'leaving' S)

- $f_{in}(S) = \sum_{v \in S, w \in T} f(w \rightarrow v)$ (sum of flow 'entering' S)

- Note: $f_{out}(S) = f_{in}(T)$ and $f_{in}(S) = f_{out}(T)$

Flows and Cuts

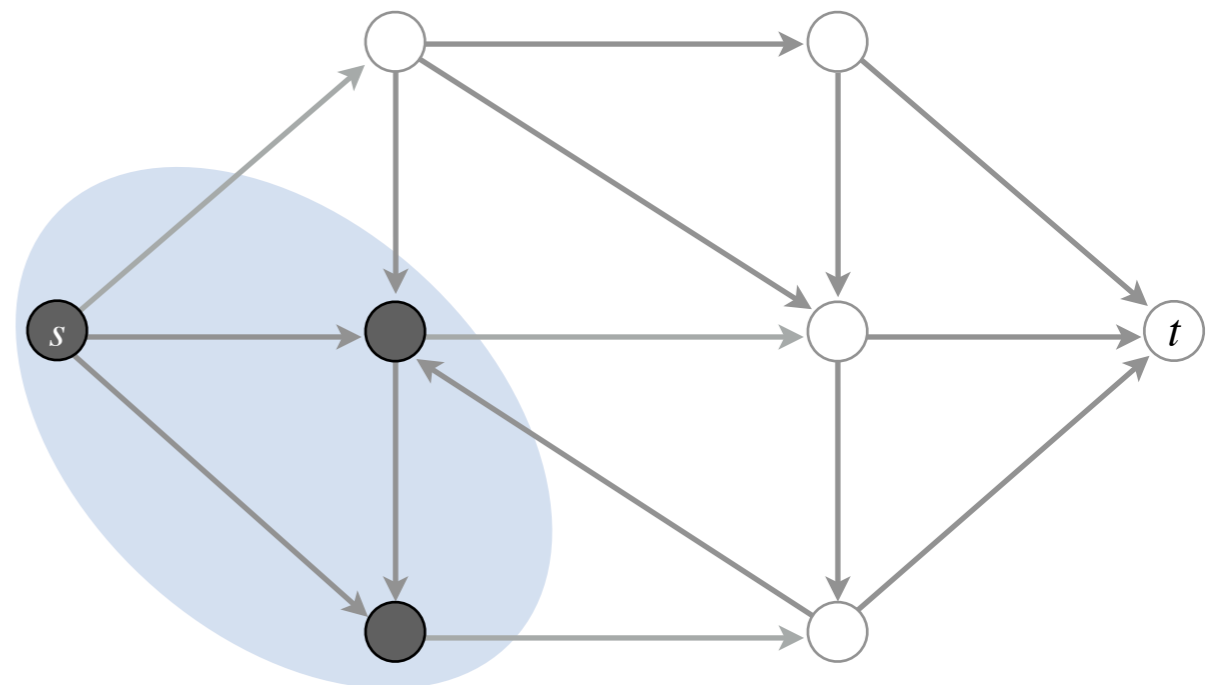
Proof. $f_{out}(S) - f_{in}(S)$

$$= \sum_{v \in S, w \in T} f(v \rightarrow w) - \sum_{v \in S, u \in T} f(u \rightarrow v) \quad [\text{by definition}]$$

Adding zero terms

$$= \left[\sum_{v, w \in S} f(v \rightarrow w) - \sum_{v, u \in S} f(u \rightarrow v) \right] + \sum_{v \in S, w \in T} f(v \rightarrow w) - \sum_{v \in S, u \in T} f(u \rightarrow v)$$

These are the same sum:
they sum the flow of all edges
with both vertices in S



Flows and Cuts

Proof. $f_{out}(S) - f_{in}(S)$

$$= \left[\sum_{v,w \in S} f(v \rightarrow w) - \sum_{v,u \in S} f(u \rightarrow v) \right] + \sum_{v \in S, w \in T} f(v \rightarrow w) - \sum_{v \in S, u \in T} f(u \rightarrow v)$$

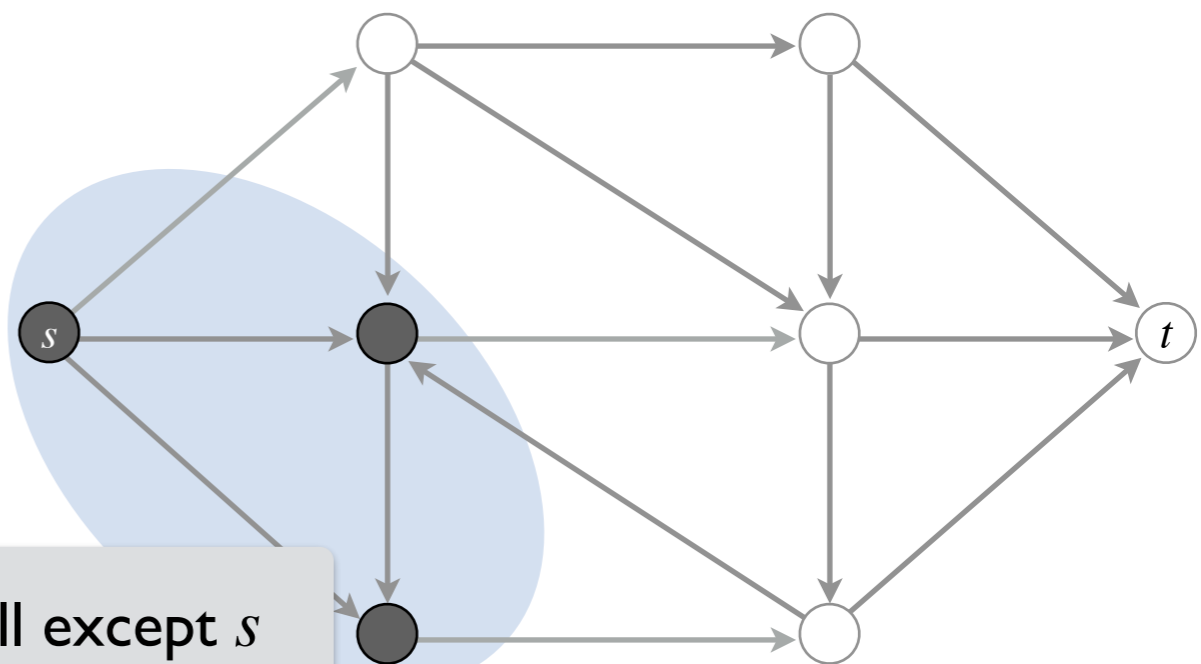
$$= \sum_{v,w \in S} f(v \rightarrow w) + \sum_{v \in S, w \in T} f(v \rightarrow w) - \sum_{v,u \in S} f(u \rightarrow v) - \sum_{v \in S, u \in T} f(u \rightarrow v)$$

$$= \sum_{v \in S} \left(\sum_w f(v \rightarrow w) - \sum_u f(u \rightarrow v) \right)$$

$$= \sum_{v \in S} f_{out}(v) - f_{in}(v)$$

$$= f_{out}(s) = v(f) \quad \blacksquare$$

Rearranging terms



Cancels out for all except s

Flows and Cuts

- We use this result to prove that the value of a flow cannot exceed the capacity of any cut in the network
- **Claim.** Let f be any s - t flow and (S, T) be any s - t cut then $v(f) \leq c(S, T)$
- **Proof.** $v(f) = f_{out}(S) - f_{in}(S)$

$$\leq f_{out}(S) = \sum_{v \in S, w \in T} f(v \rightarrow w)$$

$$\leq \sum_{v \in S, w \in T} c(v, w) = c(S, T)$$

When is $v(f) = c(S, T)$?

$f_{in}(S) = 0, f_{out}(S) = c(S, T)$

Max-Flow & Min-Cut

- Suppose the c_{\min} is the capacity of the minimum cut in a network
- What can we say about the feasible flow we can send through it
 - cannot be more than c_{\min}
- In fact, whenever we find any s - t flow f and any s - t cut (S, T) such that, $v(f) = c(S, T)$ we can conclude that:
 - f is the maximum flow, and,
 - (S, T) is the minimum cut
- The question now is, given any flow network with min cut c_{\min} , is it always possible to route a feasible s - t flow f with $v(f) = c_{\min}$

Max-Flow Min-Cut Theorem

- A beautiful, powerful relationship between these two problems is given by the following theorem
- **Theorem.** Given any flow network G , there exists a feasible (s, t) -flow f and a (s, t) -cut (S, T) such that,

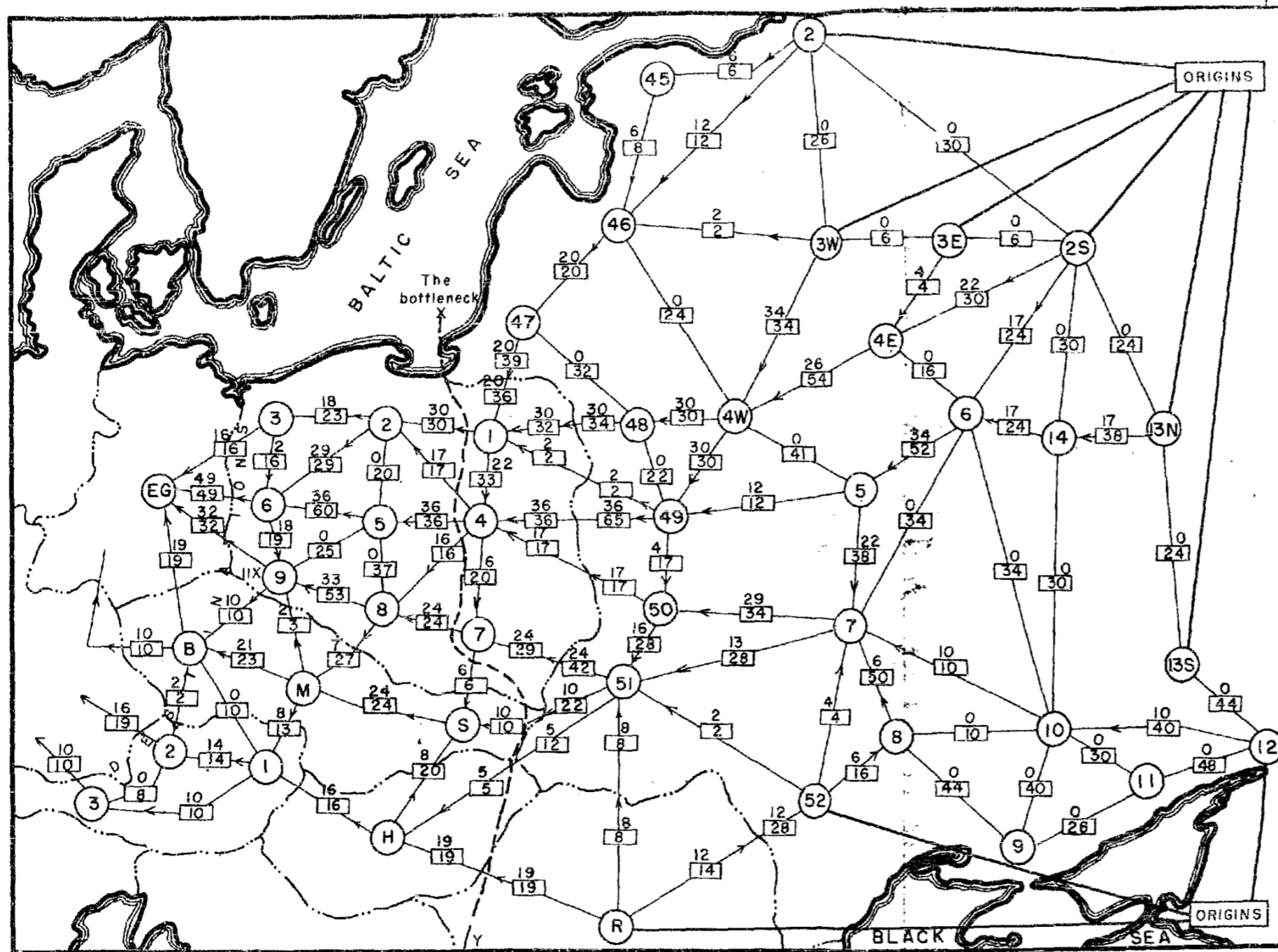
$$v(f) = c(S, T)$$

- Informally, in a flow network, the max-flow = min-cut
- This will guide our algorithm design for finding max flow
- (Will prove this theorem by construction in a bit—our algorithm will prove the theorem! (like with Gale-Shapley))

Network Flow History

- In 1950s, US military researchers Harris and Ross wrote a classified report about the rail network linking Soviet Union and Eastern Europe
 - Vertices were the geographic regions
 - Edges were railway links between the regions
 - Edge weights were the rate at which material could be shipped from one region to next
- Ross and Harris determined:
 - Maximum amount of stuff that could be moved from Russia to Europe (**max flow**)
 - Cheapest way to disrupt the network by removing rail links (**min cut**)

Network Flow History



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Fig. 7 — Traffic pattern: entire network available

Legend:

— — — International boundary

⊙ Railway operating division

← $\frac{9}{12}$ → Capacity: 12 each way per day. Required flow of 9 per day toward destinations (in direction of arrow) with equivalent number of returning trains in opposite direction

All capacities in $\frac{\text{trains}}{\sqrt{1000}\text{'s of tons}}$ each way per day

Origins: Divisions 2, 3W, 3E, 25, 13N, 13S, 12, 52 (USSR), and Roumania

Destinations: Divisions 3, 6, 9 (Poland); B (Czechoslovakia); and 2, 3 (Austria)

Alternative destinations: Germany or East Germany

Note IIX at Division 9, Poland

Towards a Max-Flow Algorithm

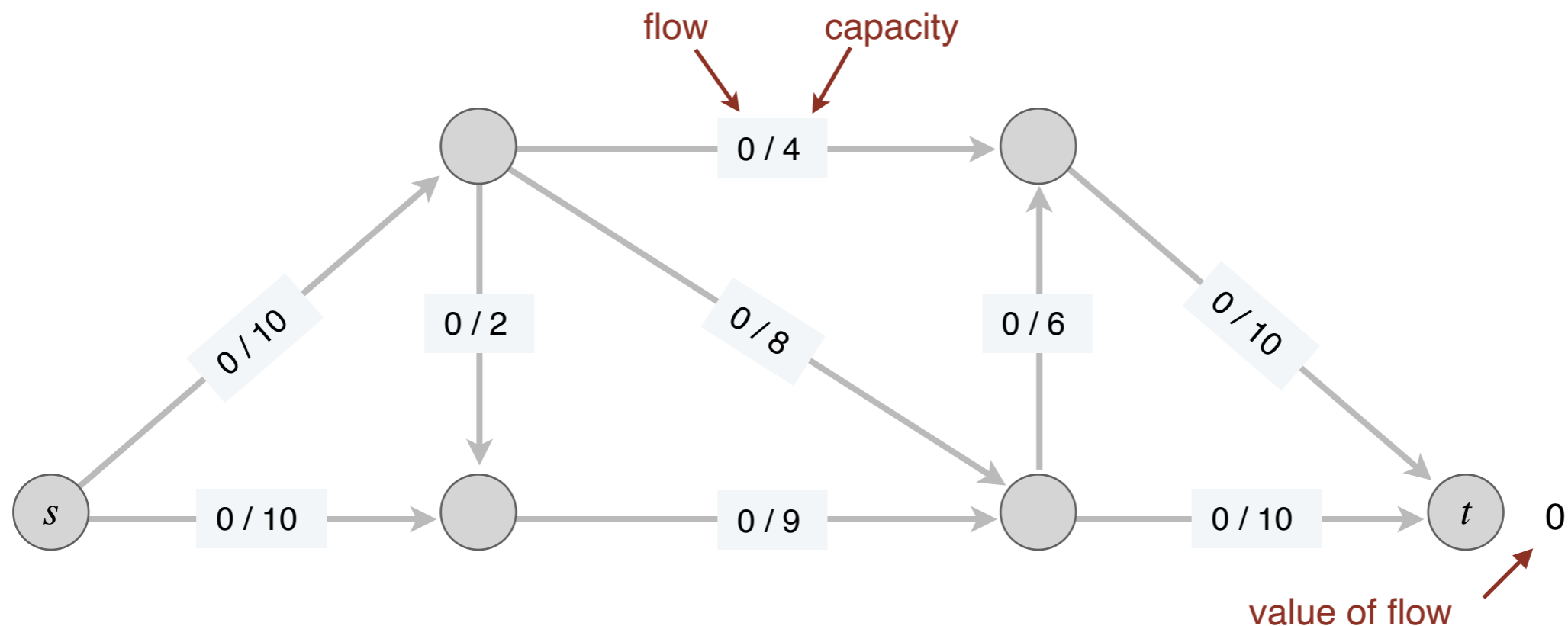
- Today: we will prove the max-flow min-cut theorem
constructively
- We will design a max-flow algorithm and show that there is a s - t cut s.t. value of flow computed by algorithm = capacity of cut
- Let's start with a greedy approach
 - Push as much flow as possible down a s - t path
 - This won't actually work
 - But gives us a sense of what we need to keep track off to improve upon it

Towards a Max-Flow Algorithm

- Greedy strategy:
 - Start with $f(e) = 0$ for each edge
 - Find an $s \rightsquigarrow t$ path P where each edge has $f(e) < c(e)$
 - “Augment” flow (as much as possible) along path P
 - Repeat until you get stuck
- Let’s take an example

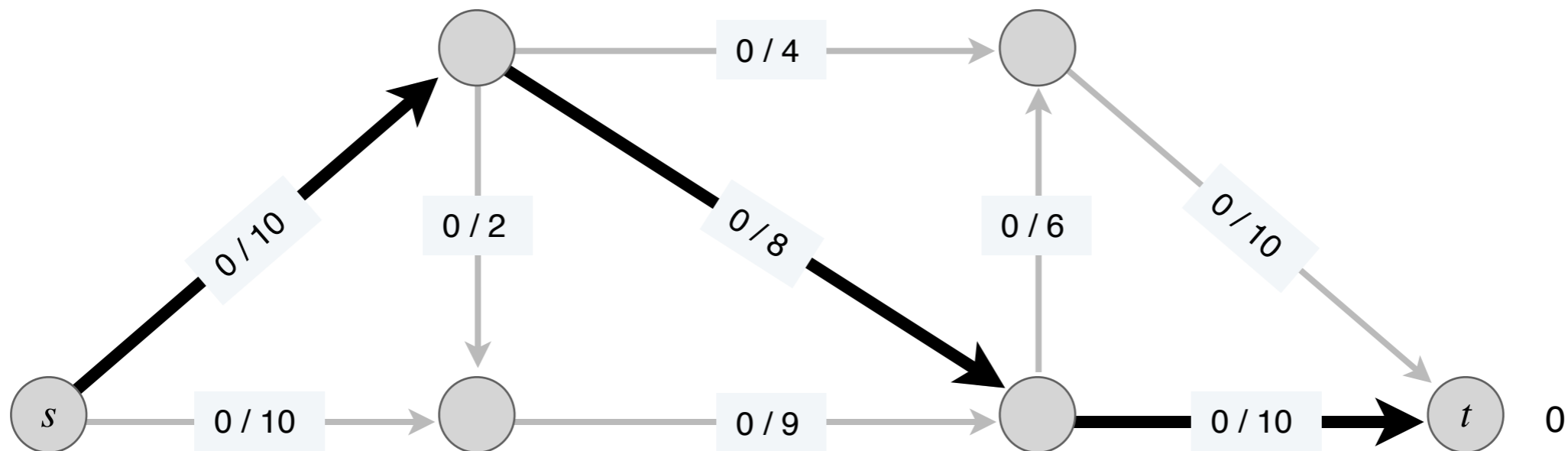
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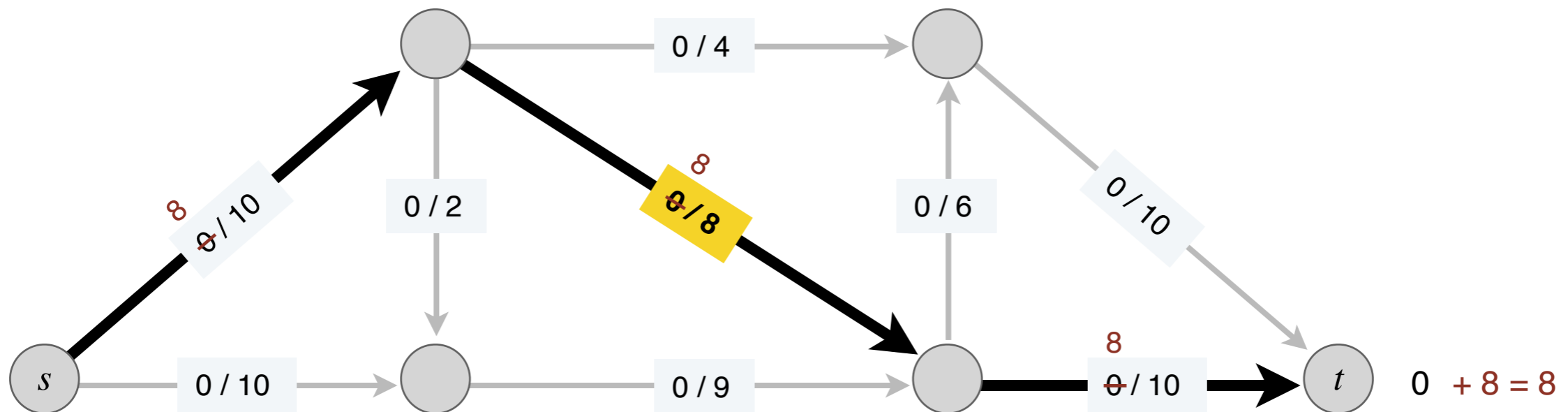
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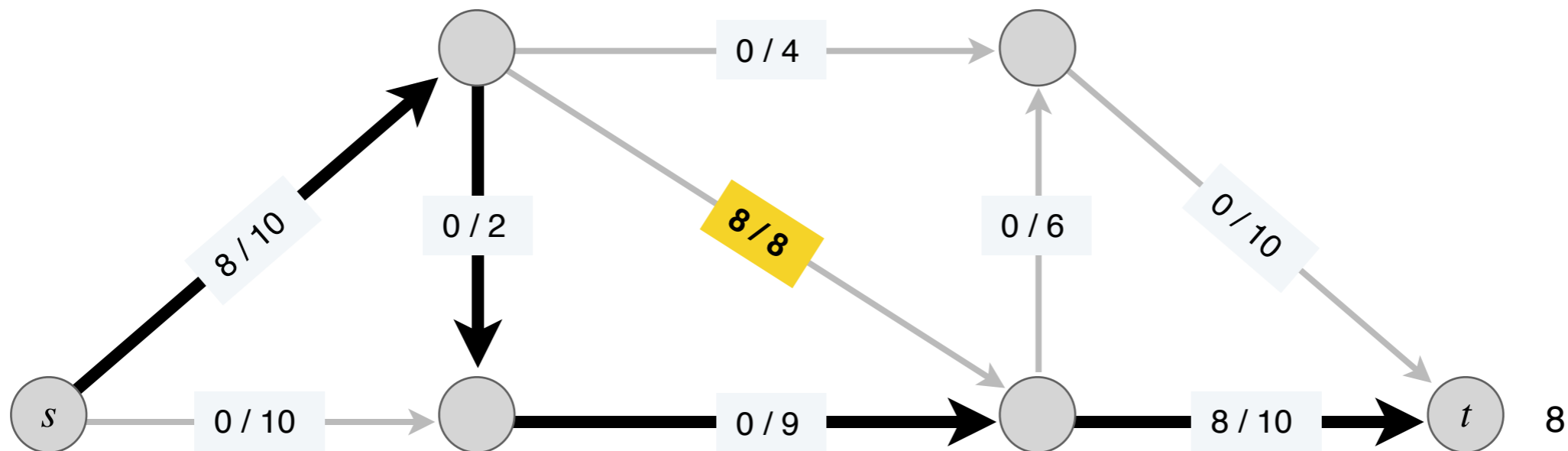
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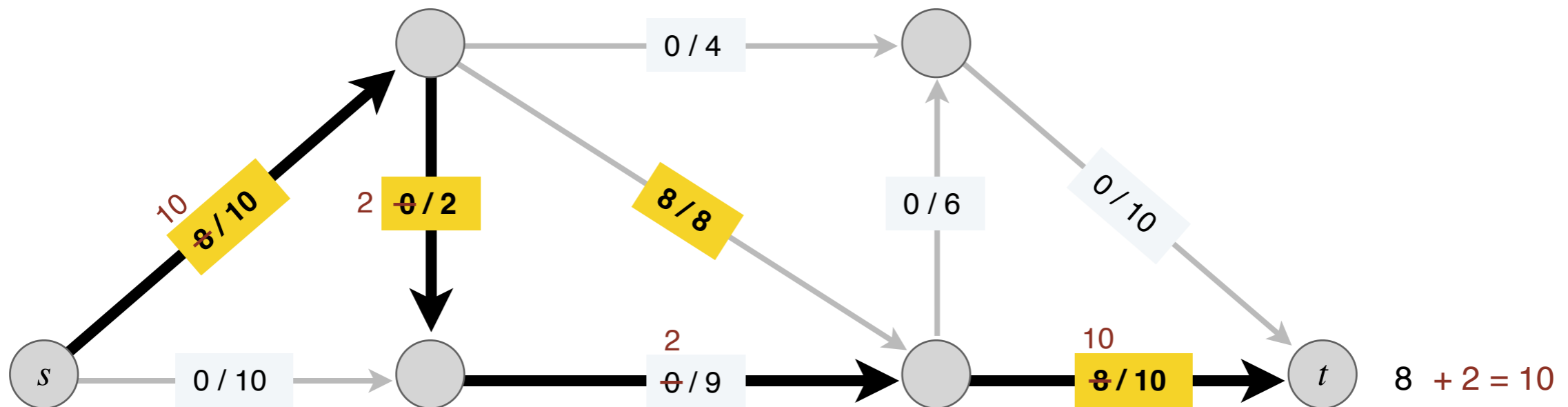
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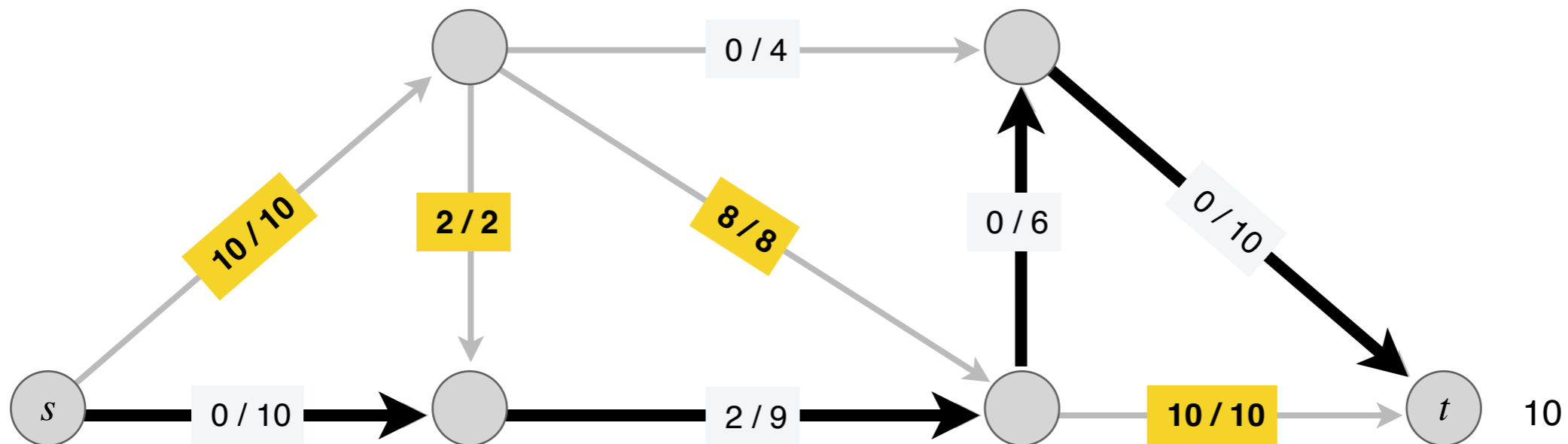
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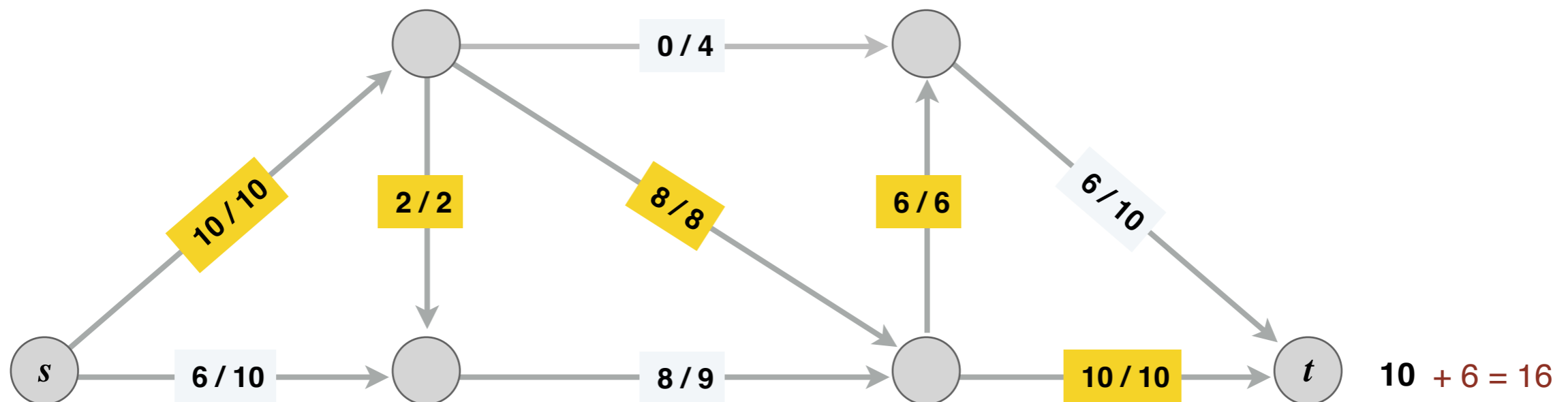


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Is this the best we can do?

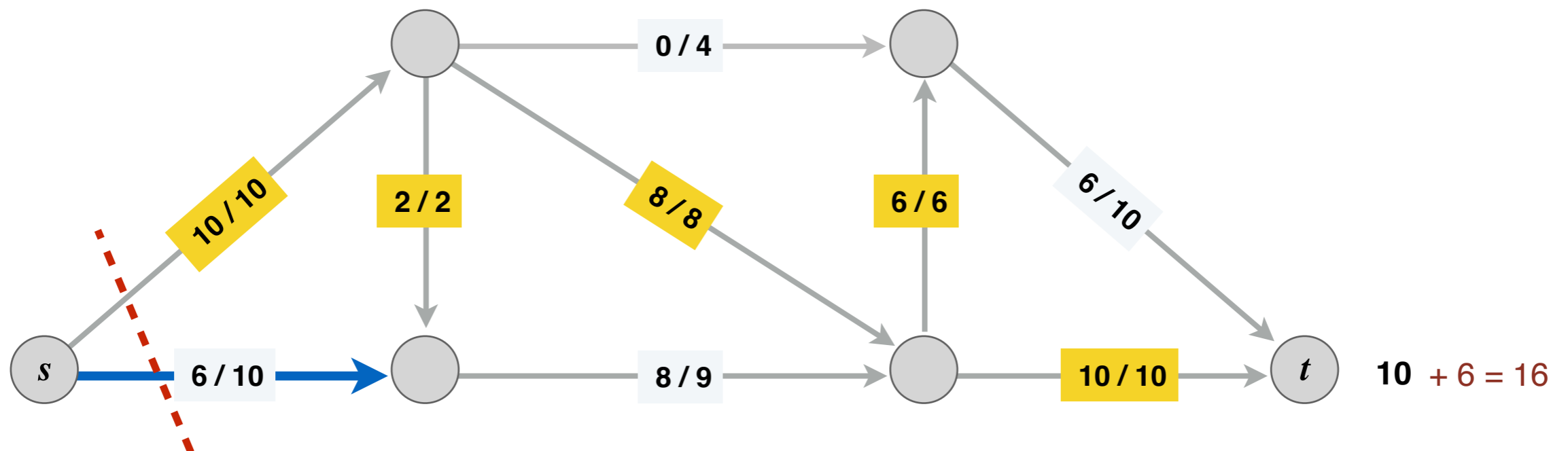
ending flow value = 16



Towards a Max-Flow Algorithm

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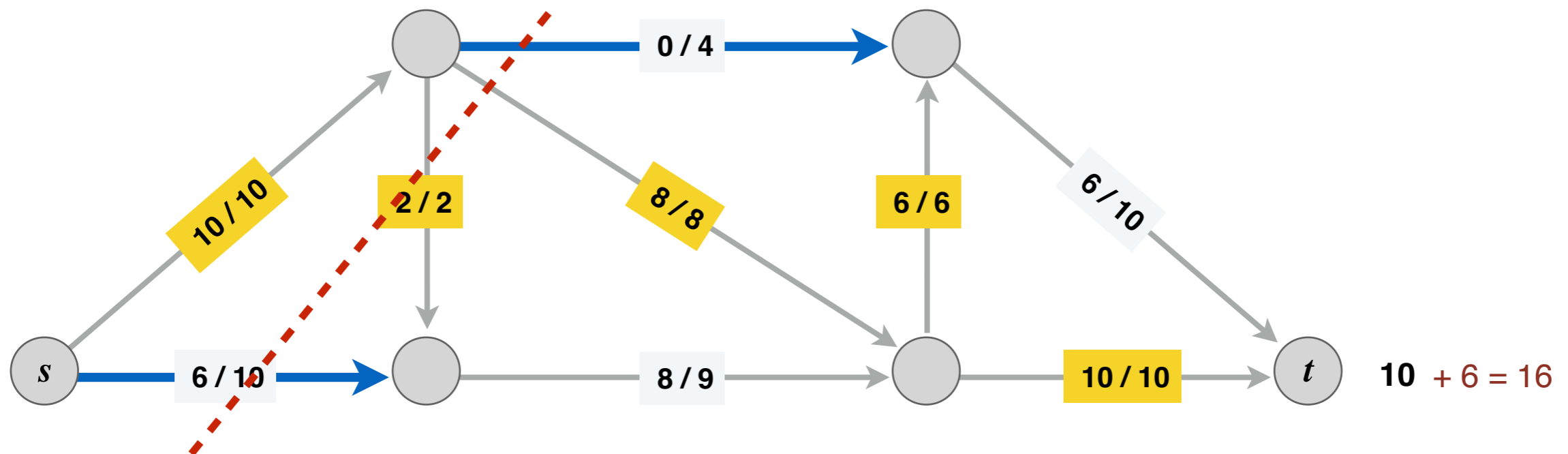
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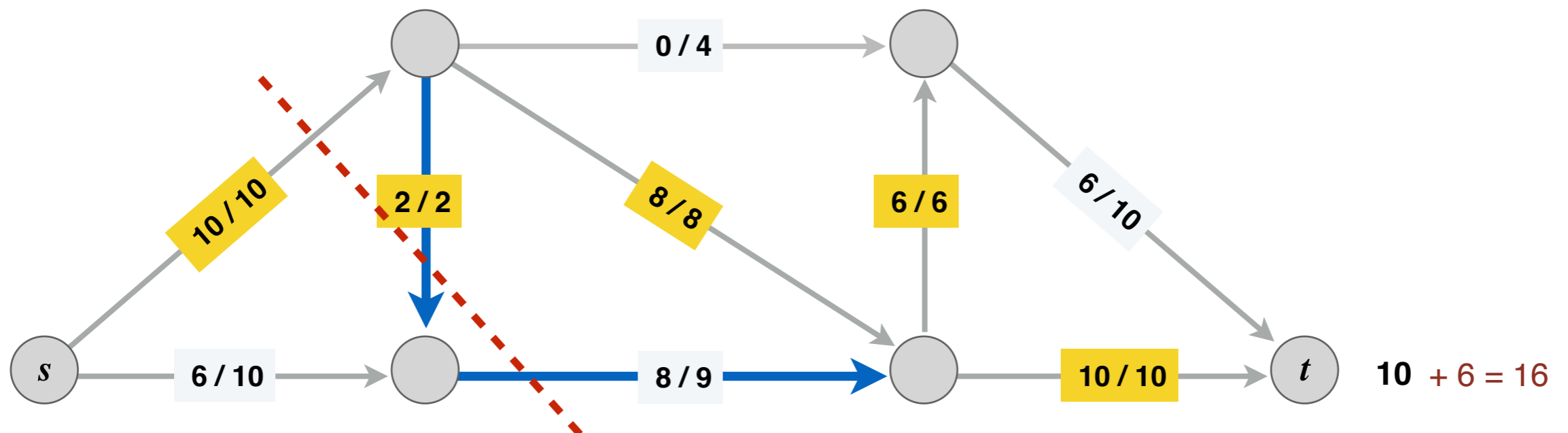
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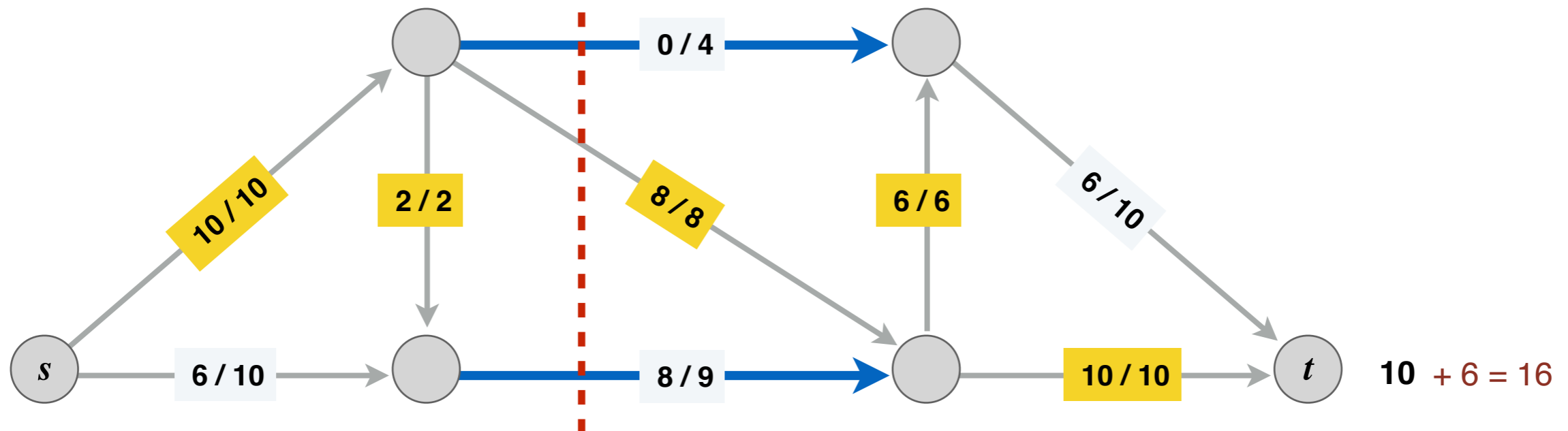
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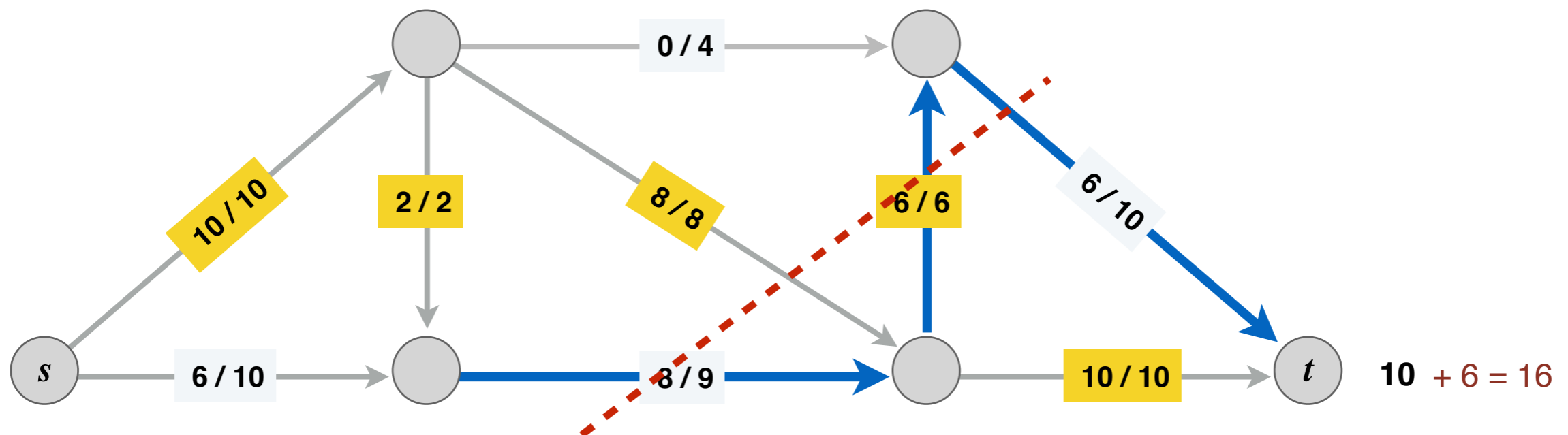
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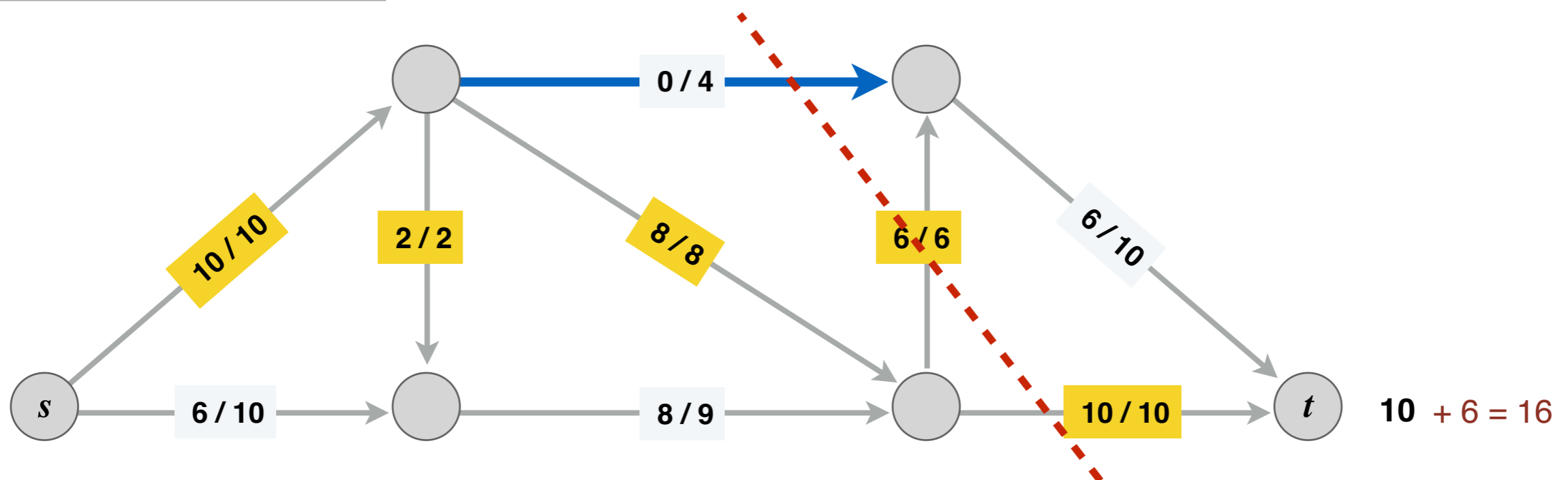
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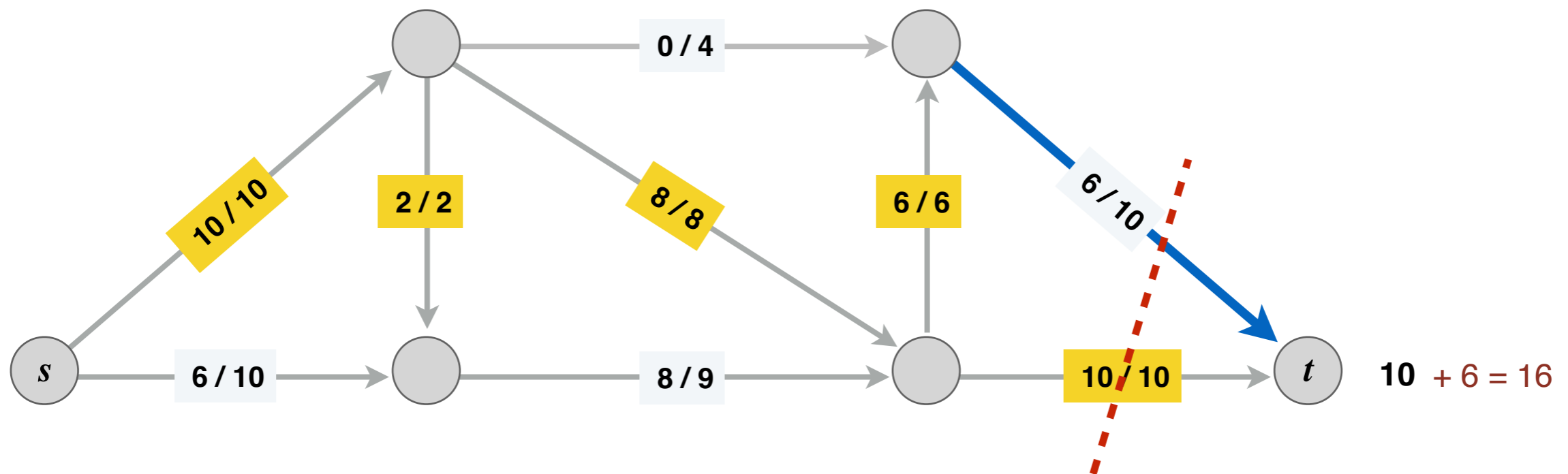
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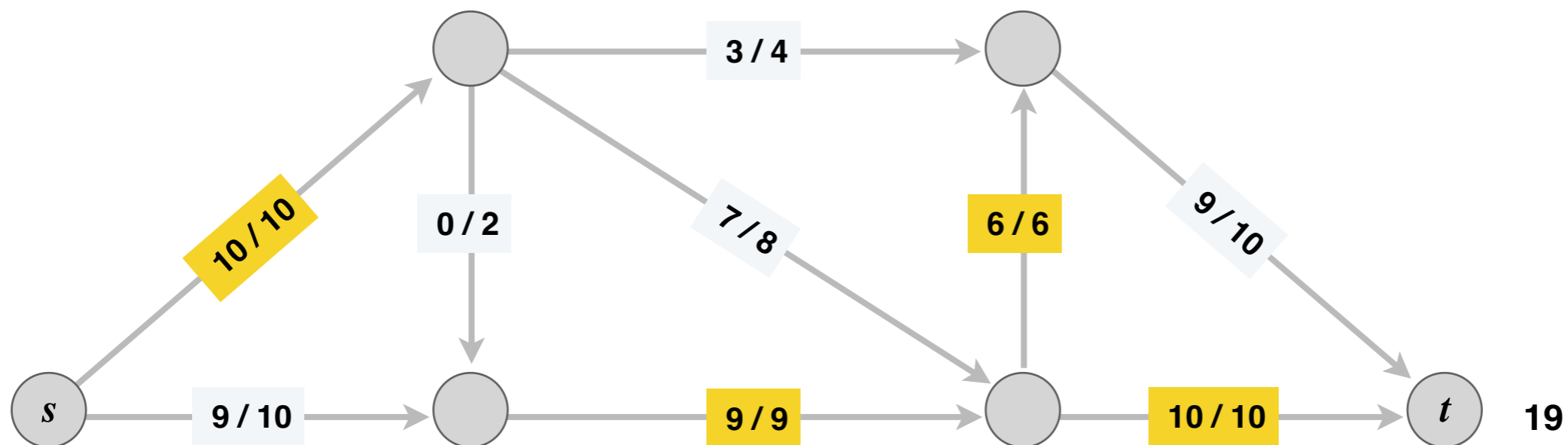
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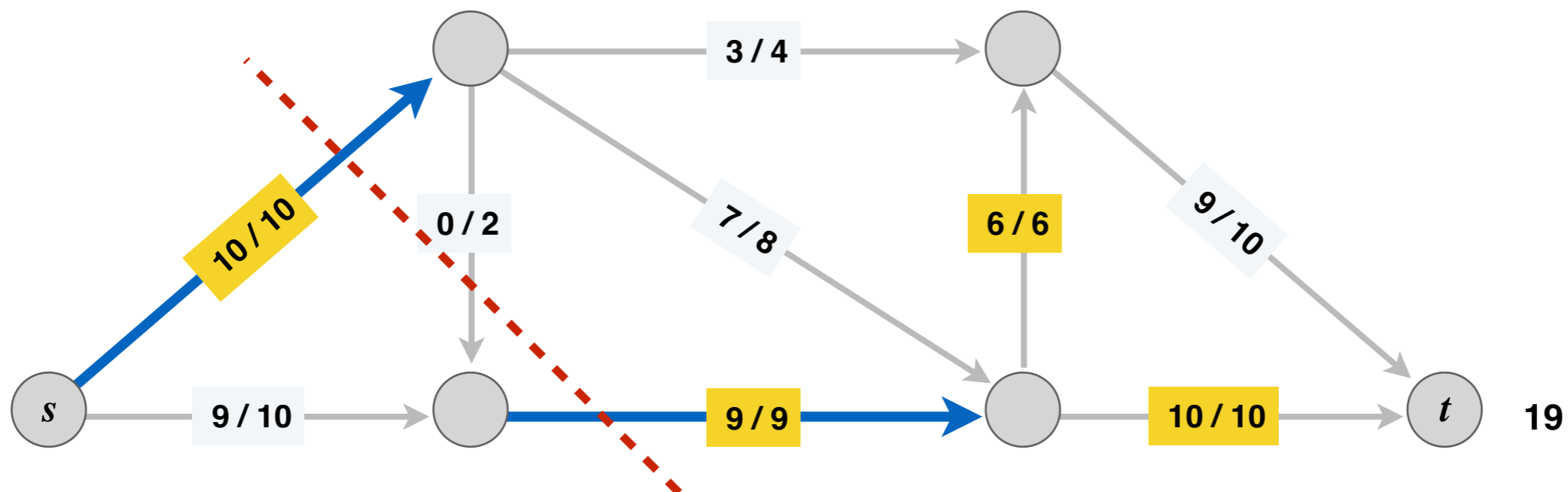
max-flow value = 19



Towards a Max-Flow Algorithm

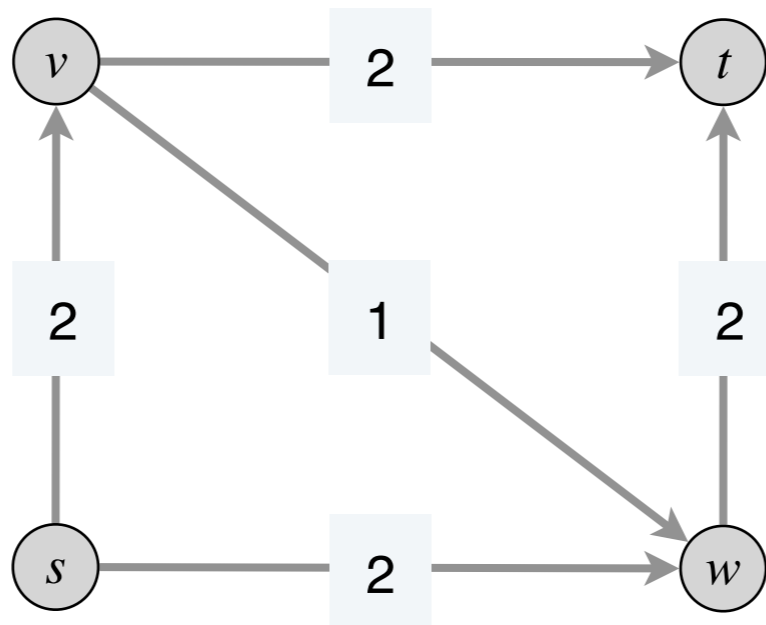
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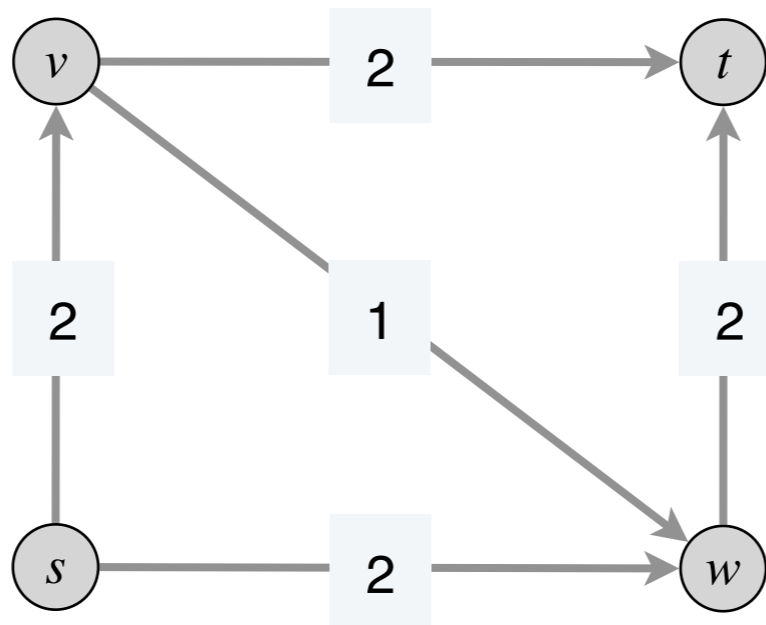
Why Greedy Fails

- **Problem:** greedy can never “undo” a bad flow decision
- Consider the following flow network



Why Greedy Fails

- **Problem:** greedy can never “undo” a bad flow decision
- Consider the following flow network
 - Unique max flow has $f(v \rightarrow w) = 0$
 - Greedy could choose $s \rightarrow v \rightarrow w \rightarrow t$ as first P



- **Takeaway:** Need a mechanism to “undo” bad flow decisions

Ford-Fulkerson Algorithm

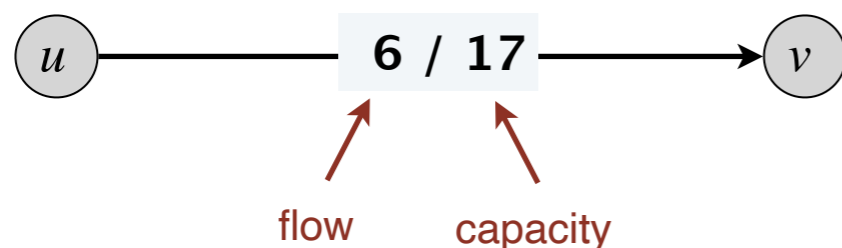
Ford Fulkerson: Idea

- Want to make “forward progress” while letting ourselves undo previous decisions if they’re getting in our way
- **Idea:** keep track of where we can push flow
 - Can push more flow along an edge with remaining capacity
 - Can also push flow “back” along an edge that already has flow down it
- Need a way to systematically track these decisions

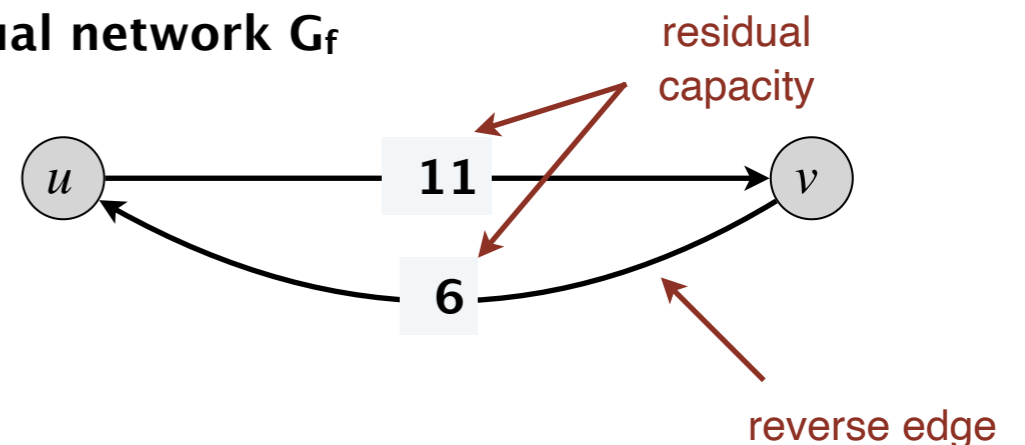
Residual Graph

- Given flow network $G = (V, E, c)$ and a feasible flow f on G , **the residual graph** $G_f = (V, E_f, c_f)$ is defined as:
 - Vertices in G_f same as G
 - (Forward edge)** For $e \in E$ with residual capacity $c(e) - f(e) > 0$, create $e \in E_f$ with capacity $c(e) - f(e)$
 - (Backward edge)** For $e \in E$ with $f(e) > 0$, create $e_{\text{reverse}} \in E_f$ with capacity $f(e)$

original flow network G



residual network G_f



Flow Algorithm Idea

- Now we have a residual graph that lets us make forward progress or push back existing flow
- We will look for $s \rightsquigarrow t$ paths in G_f rather than G
- Once we have a path, we will "augment" flow along it similar to greedy
 - find bottleneck capacity edge on the path and push that much flow through it in G_f
- When we translate this back to G , this means:
 - We increment existing flow on a forward edge
 - Or we decrement flow on a backward edge

Augmenting Path & Flow

- An **augmenting path** P is a **simple** $s \rightsquigarrow t$ path in the residual graph G_f
- The **bottleneck capacity** b of an augmenting path P is the minimum capacity of any edge in P .

The path P is in G_f

AUGMENT(f, P)

$b \leftarrow$ bottleneck capacity of augmenting path P .

FOREACH edge $e \in P$:

IF ($e \in E$, that is, e is forward edge)

Increase $f(e)$ in G by b

ELSE

Decrease $f(e)$ in G by b

RETURN f .

Updating flow in G

Ford-Fulkerson Algorithm

- Start with $f(e) = 0$ for each edge $e \in E$
- Find a simple $s \rightsquigarrow t$ path P in the residual network G_f
- Augment flow along path P by bottleneck capacity b
- Repeat until you get stuck

FORD-FULKERSON(G)

FOREACH edge $e \in E$: $f(e) \leftarrow 0$.

$G_f \leftarrow$ residual network of G with respect to flow f .

WHILE (there exists an $s \rightsquigarrow t$ path P in G_f)

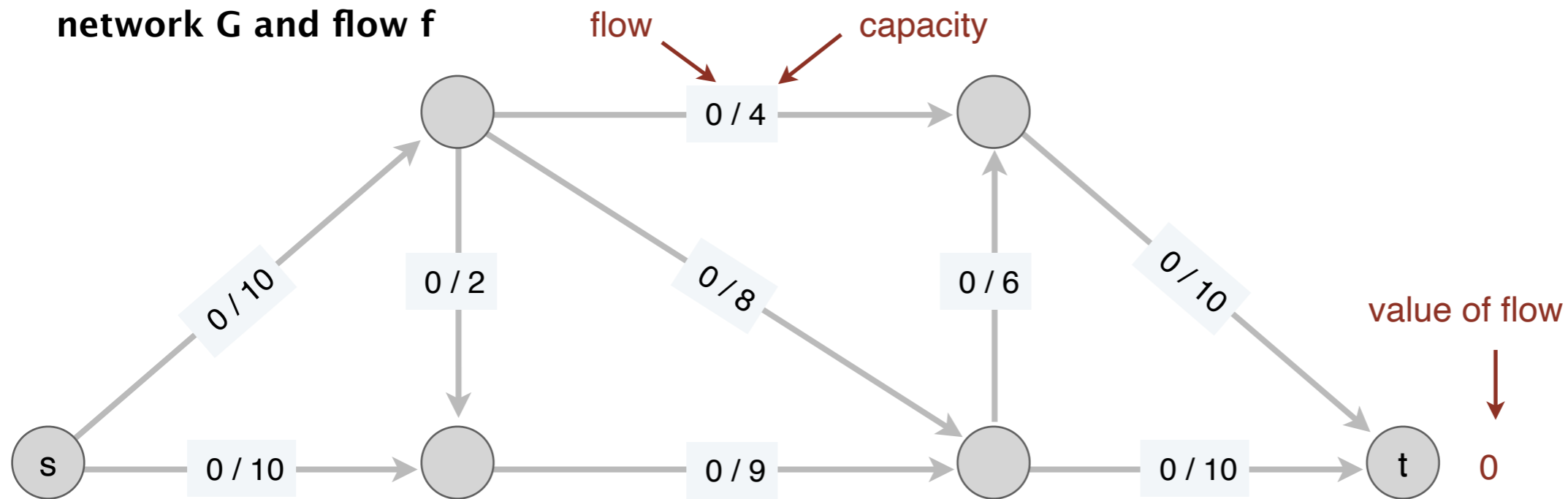
$f \leftarrow$ **AUGMENT**(f, P).

 Update G_f .

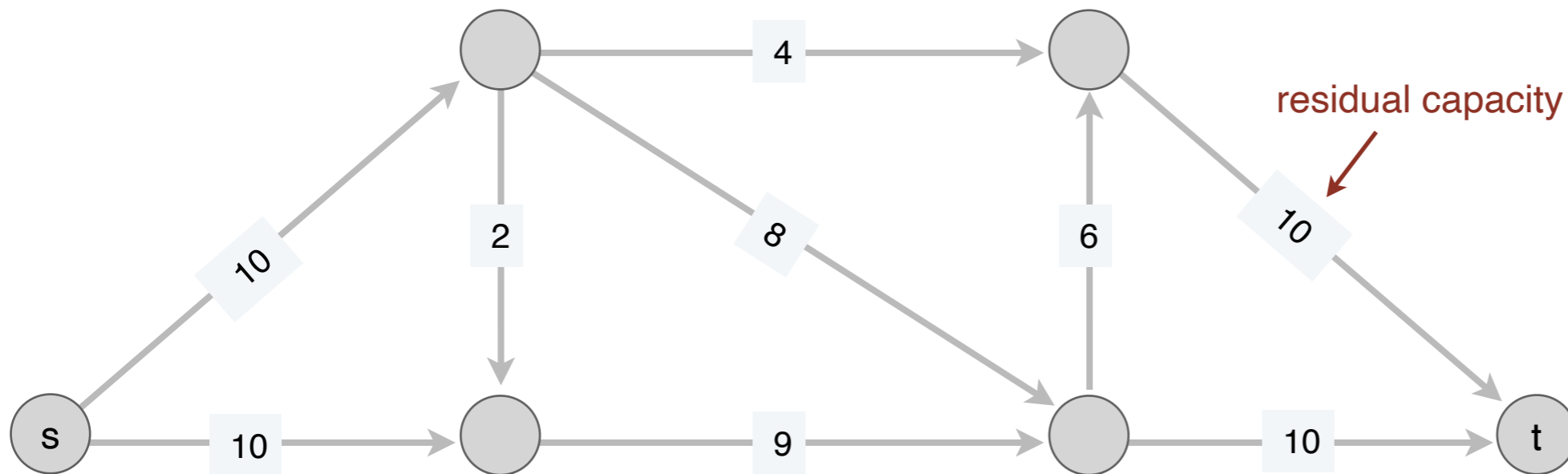
RETURN f .

Ford-Fulkerson Example

network G and flow f

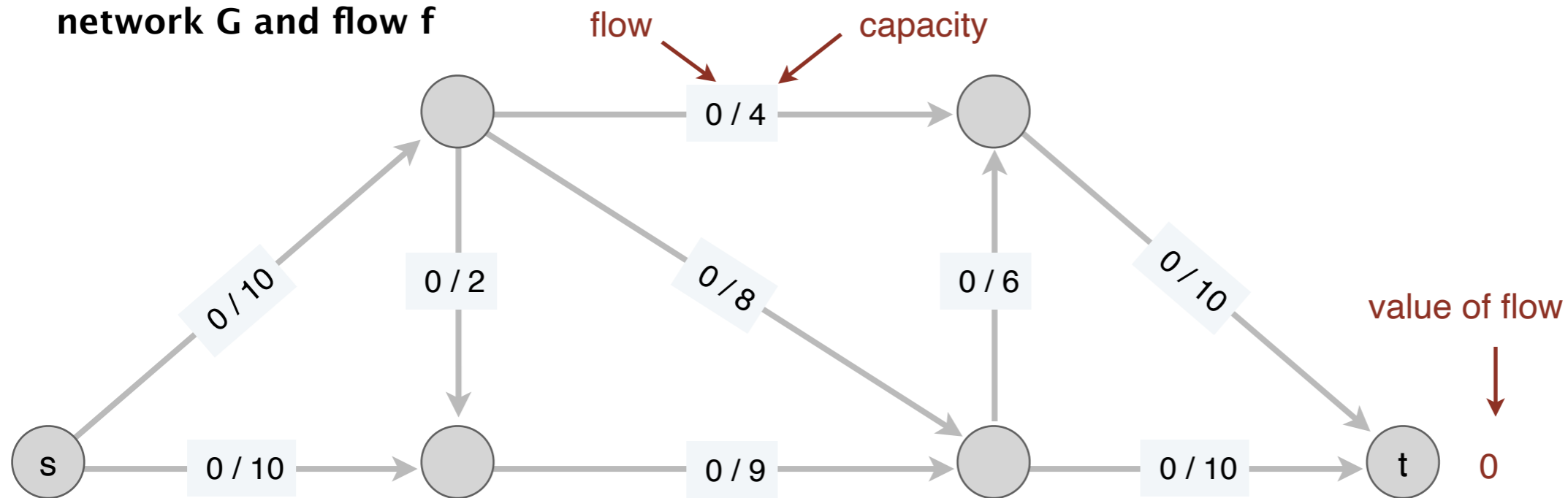


residual network G_f

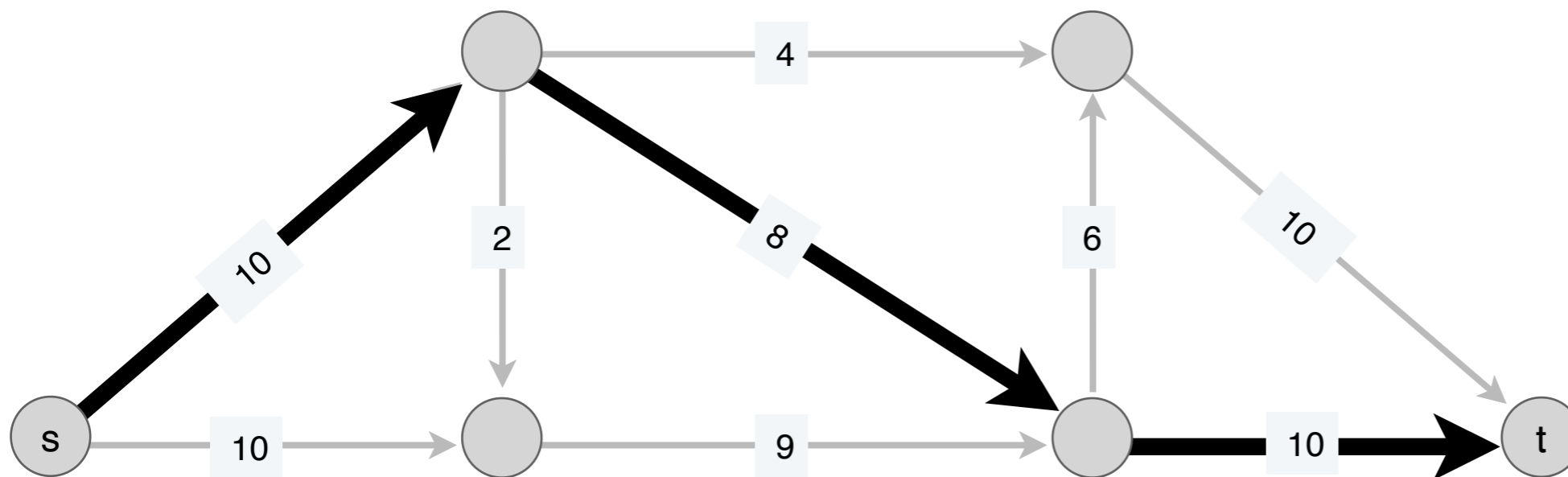


Ford-Fulkerson Example

network G and flow f

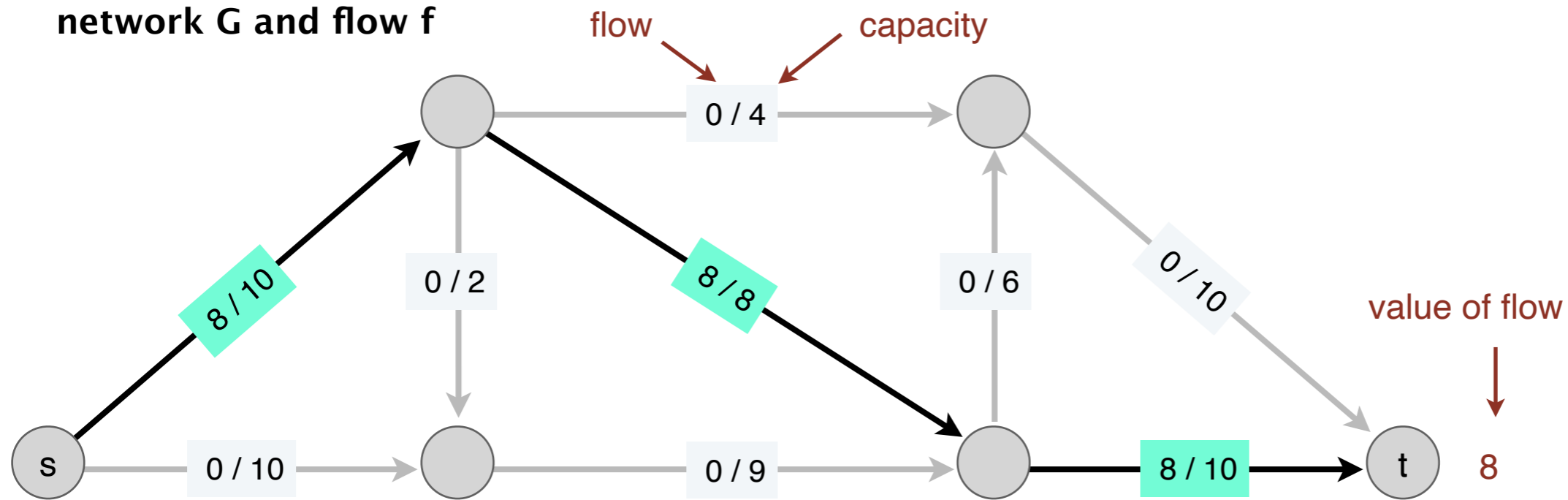


P in residual network G_f

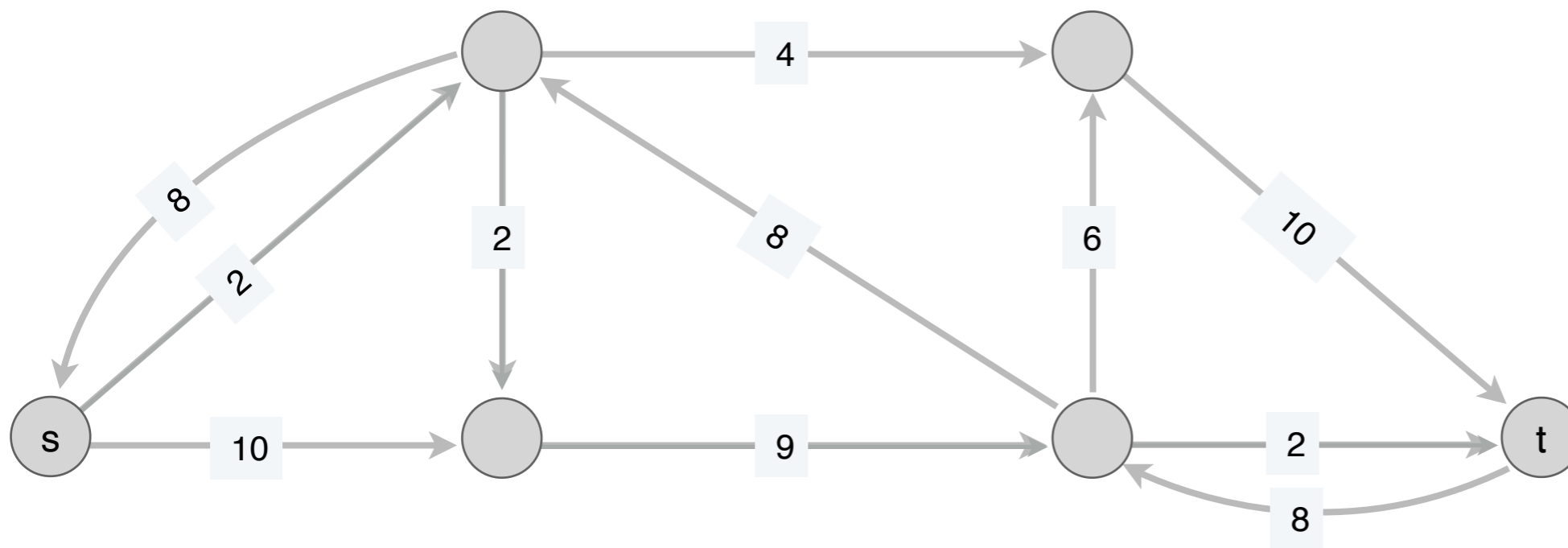


Ford-Fulkerson Example

network G and flow f

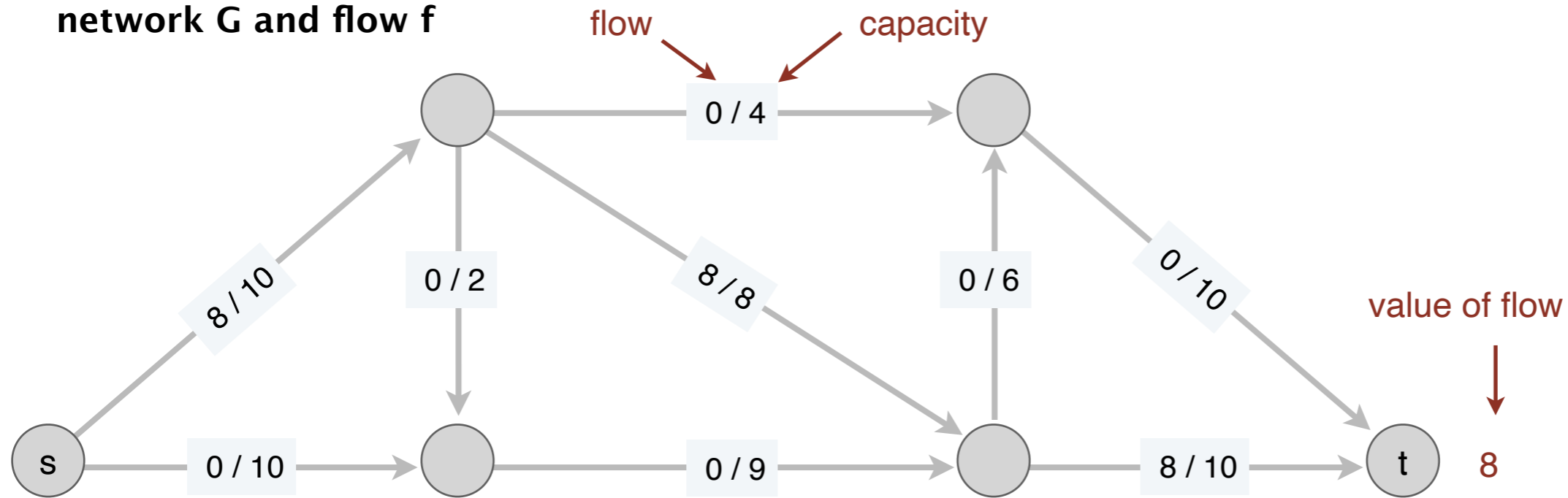


residual network G_f

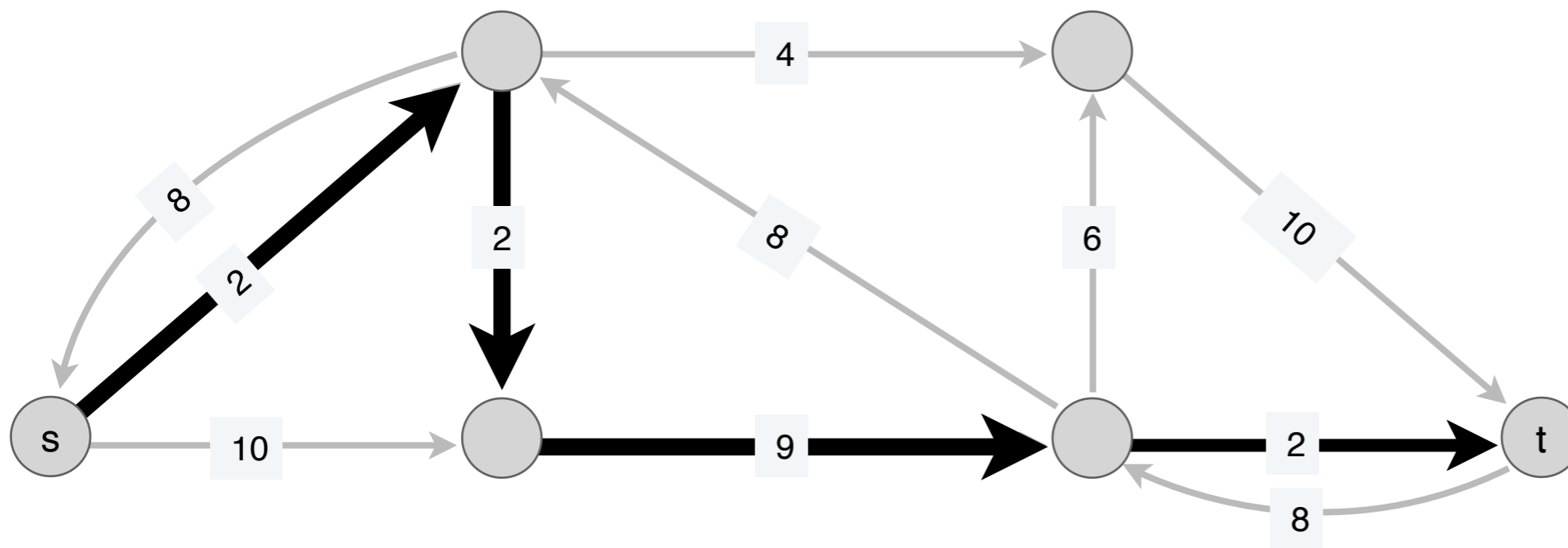


Ford-Fulkerson Example

network G and flow f

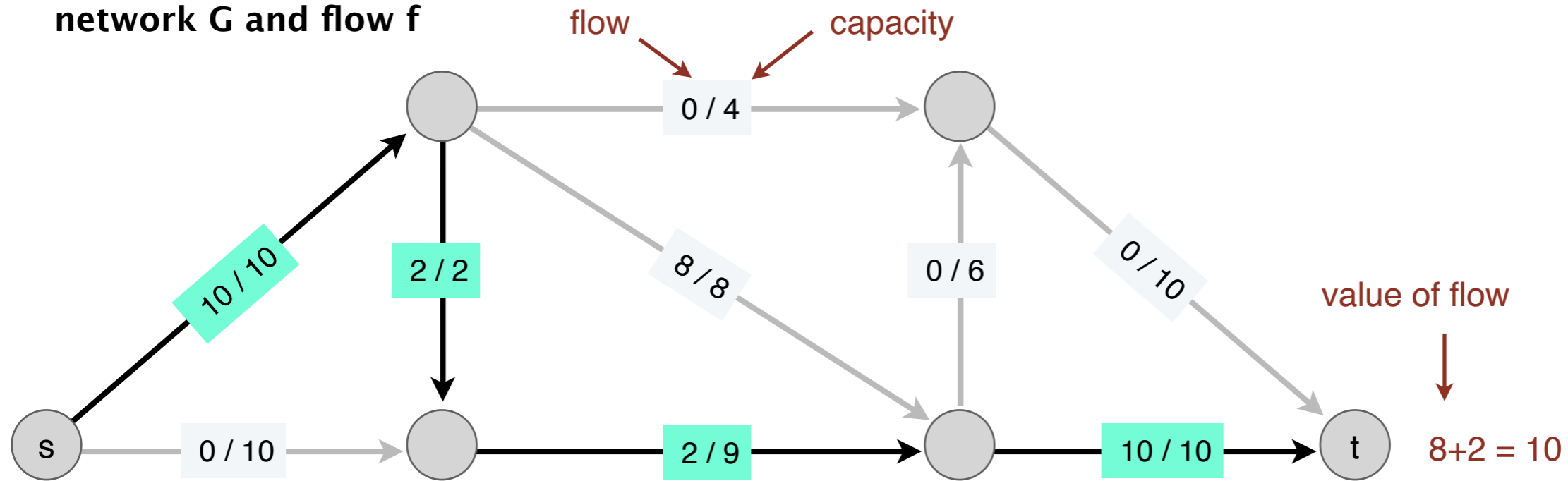


P in residual network G_f

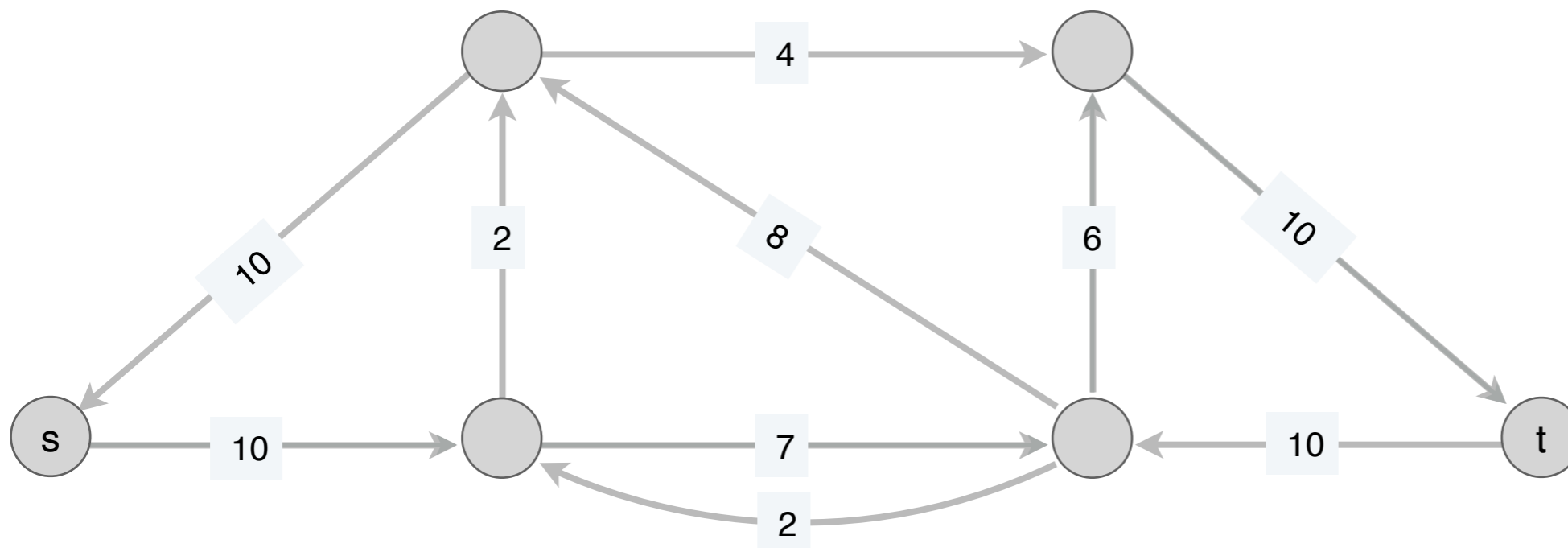


Ford-Fulkerson Example

network G and flow f

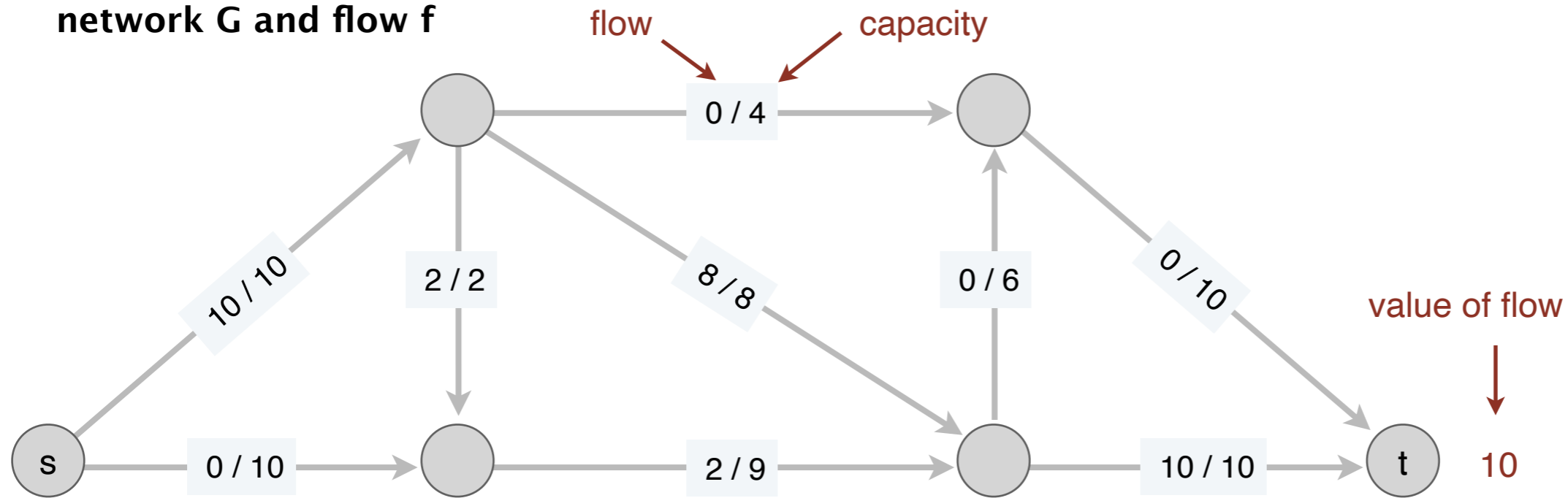


residual network G_f

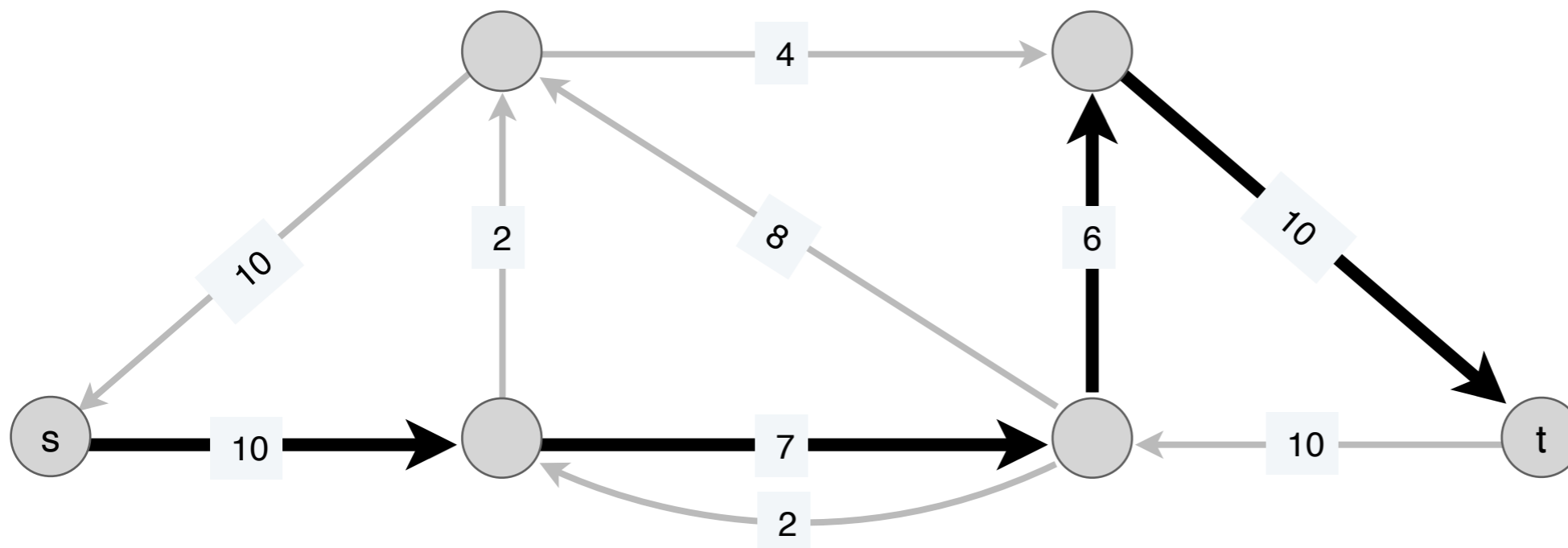


Ford-Fulkerson Example

network G and flow f

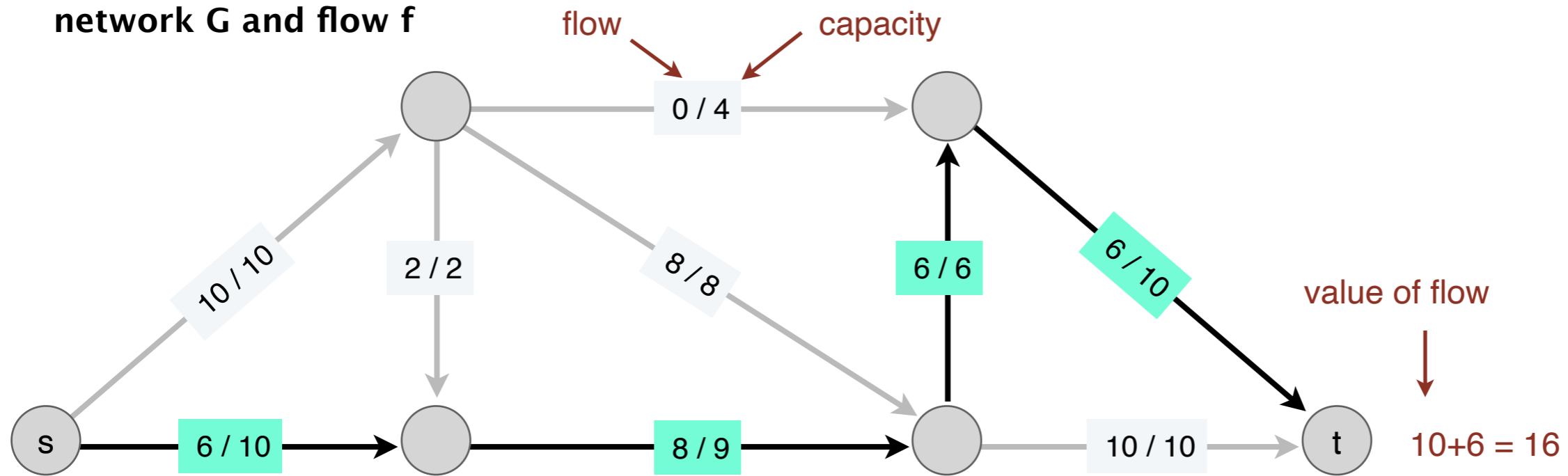


P in residual network G_f

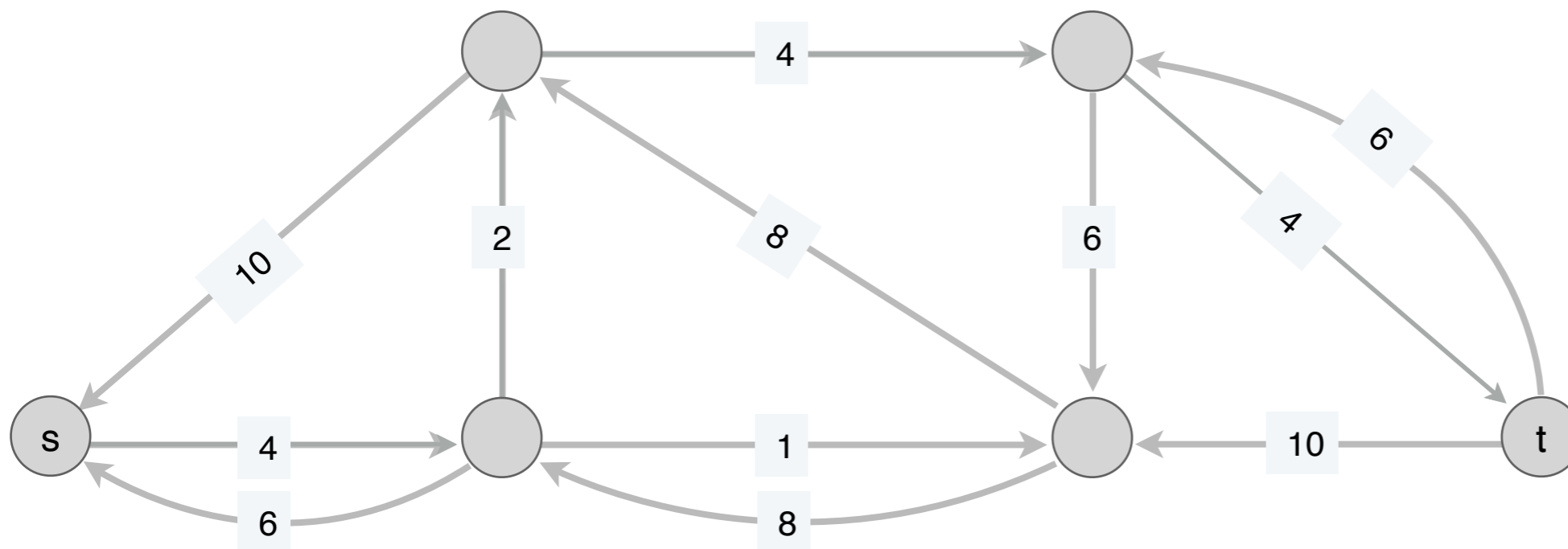


Ford-Fulkerson Example

network G and flow f

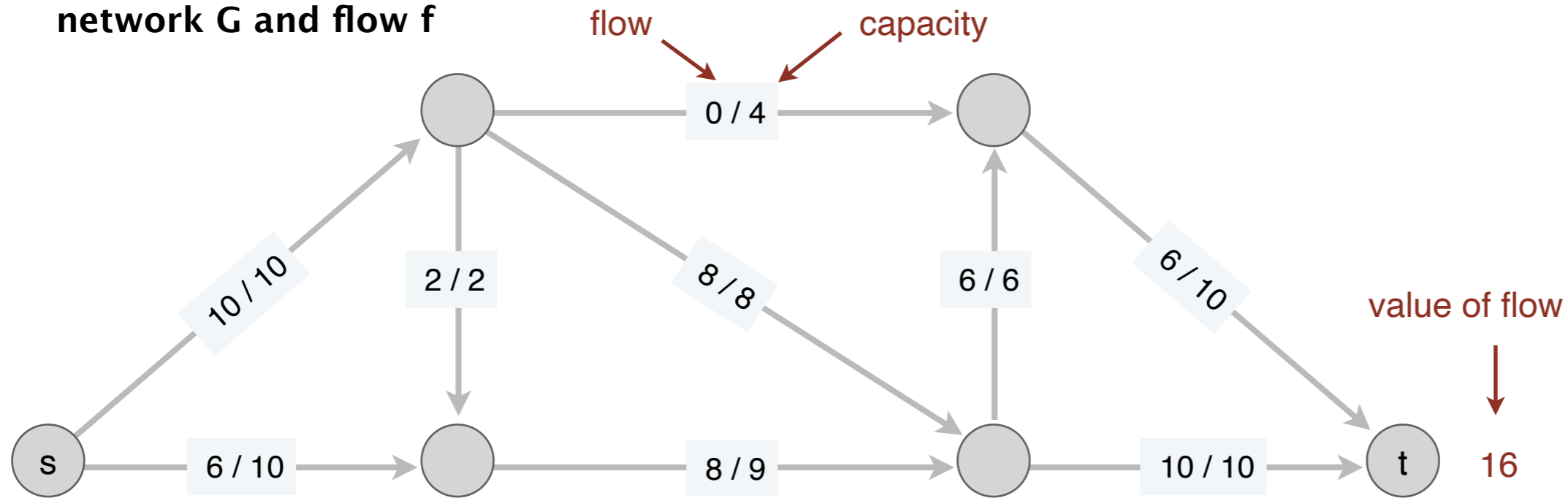


residual network G_f

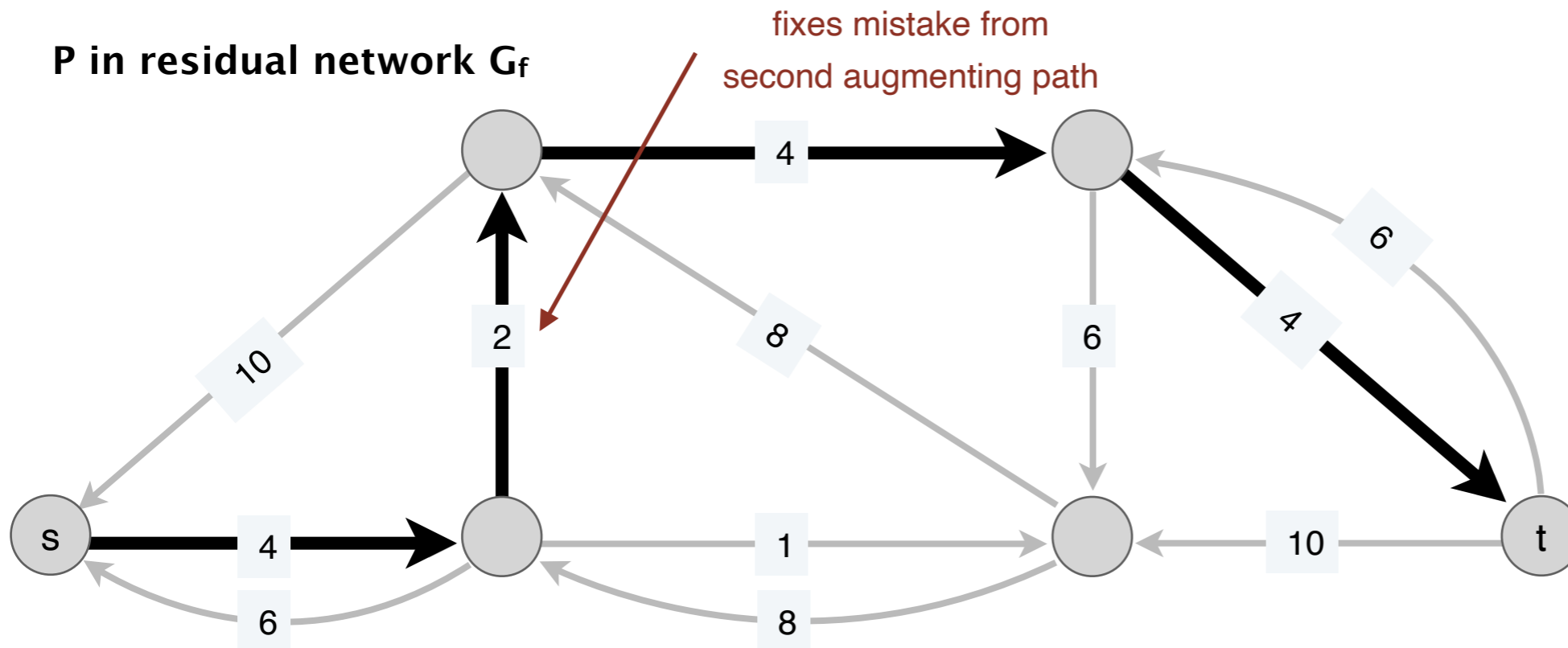


Ford-Fulkerson Example

network G and flow f

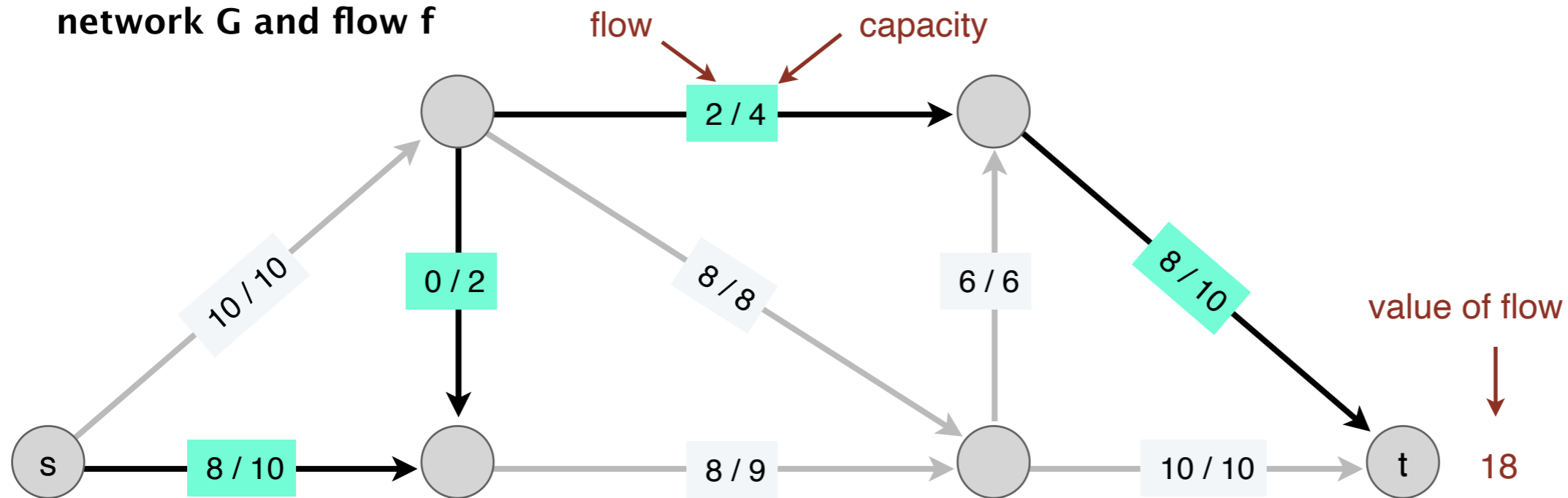


P in residual network G_f

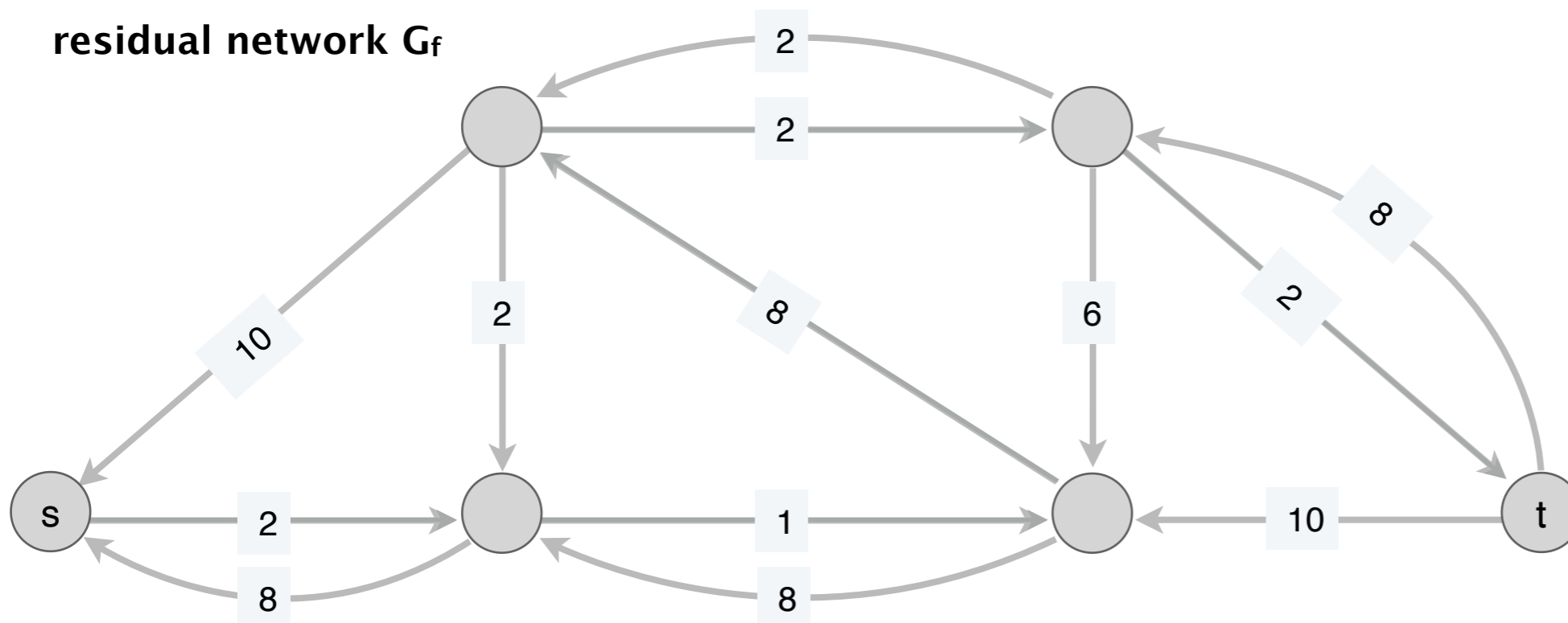


Ford-Fulkerson Example

network G and flow f

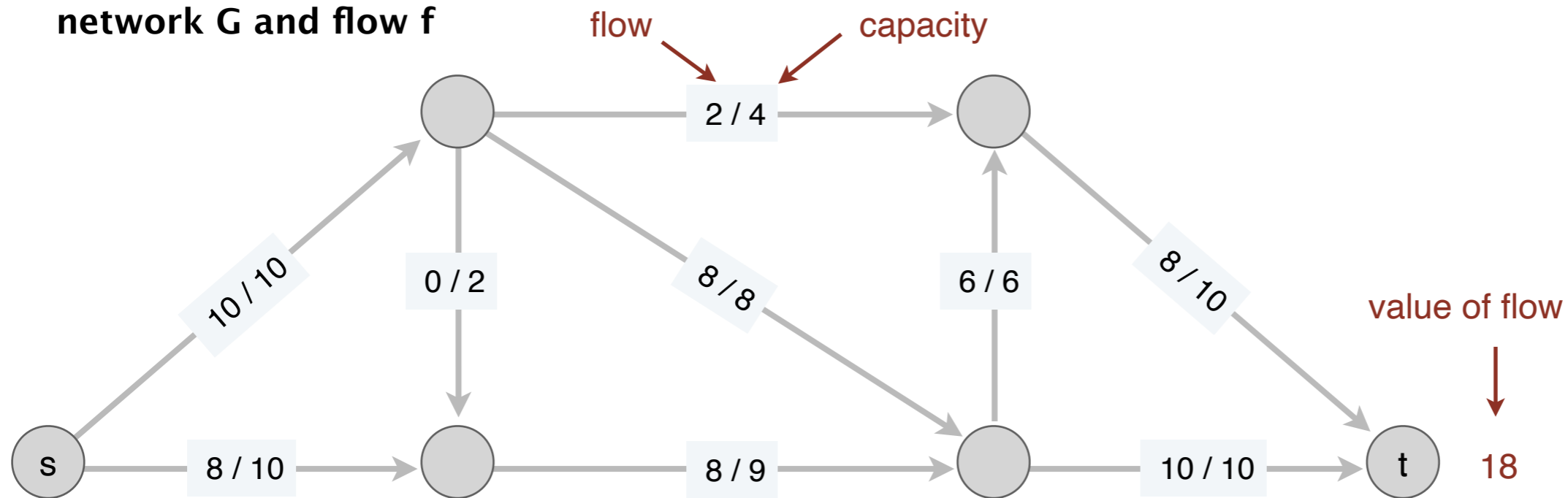


residual network G_f

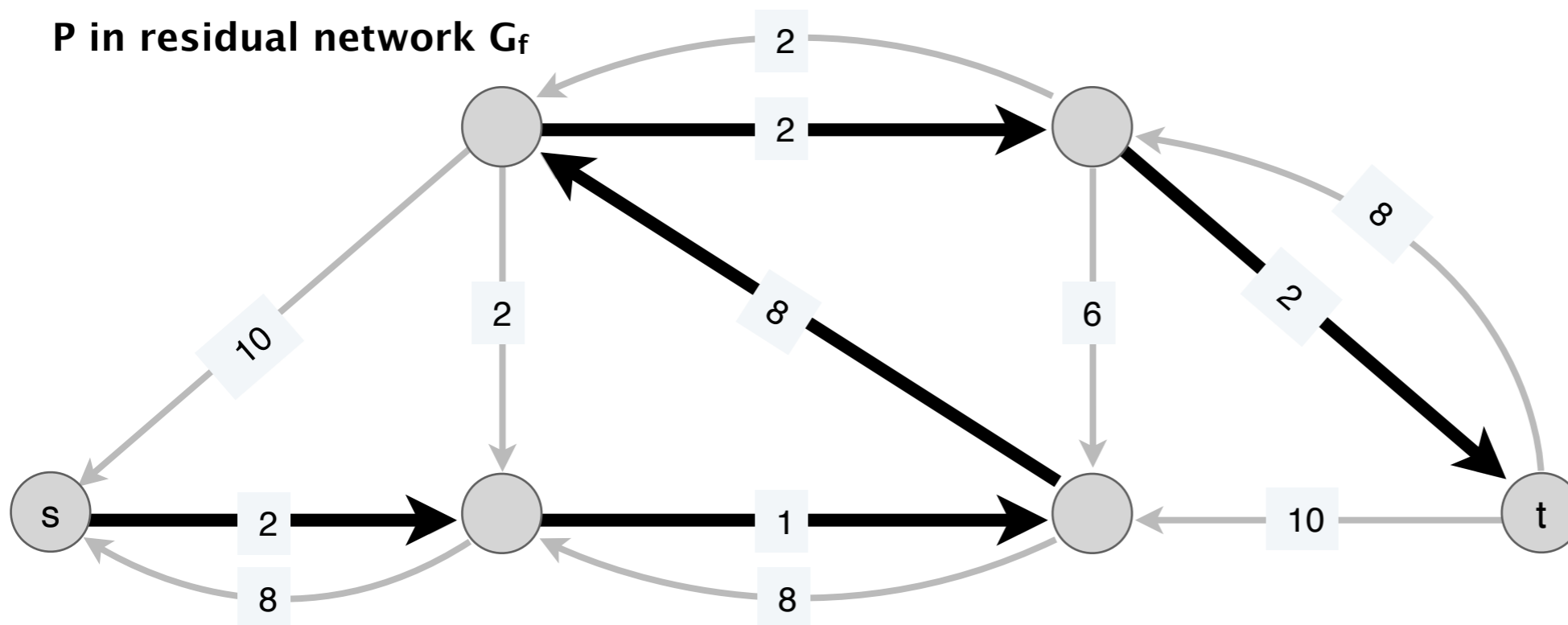


Ford-Fulkerson Example

network G and flow f

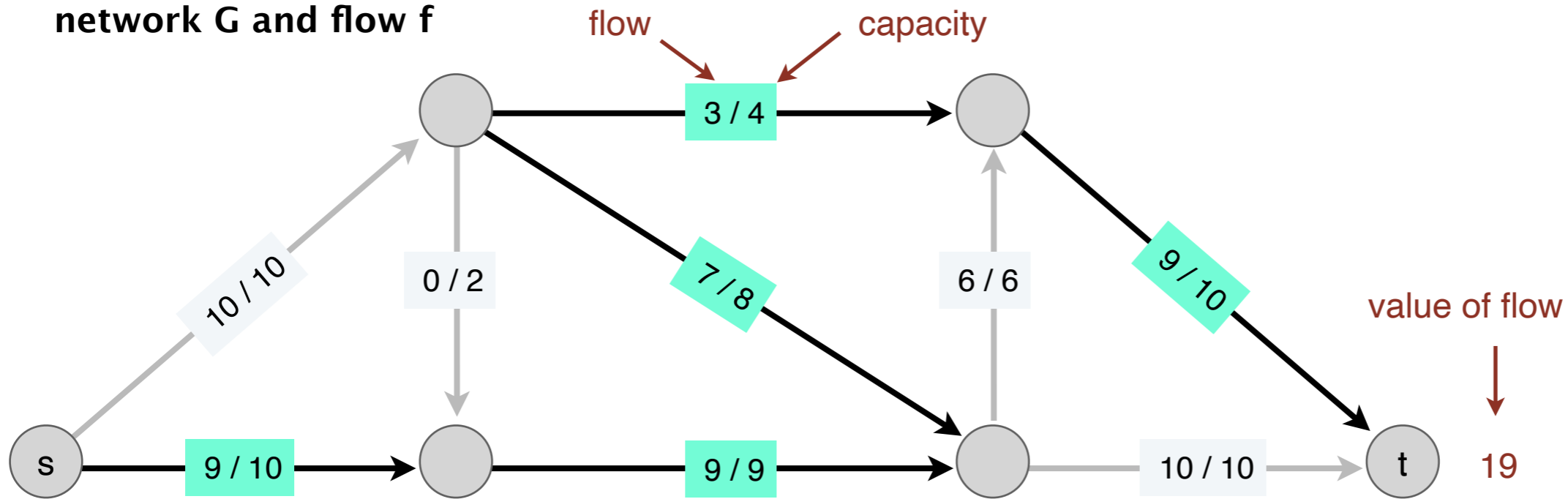


P in residual network G_f

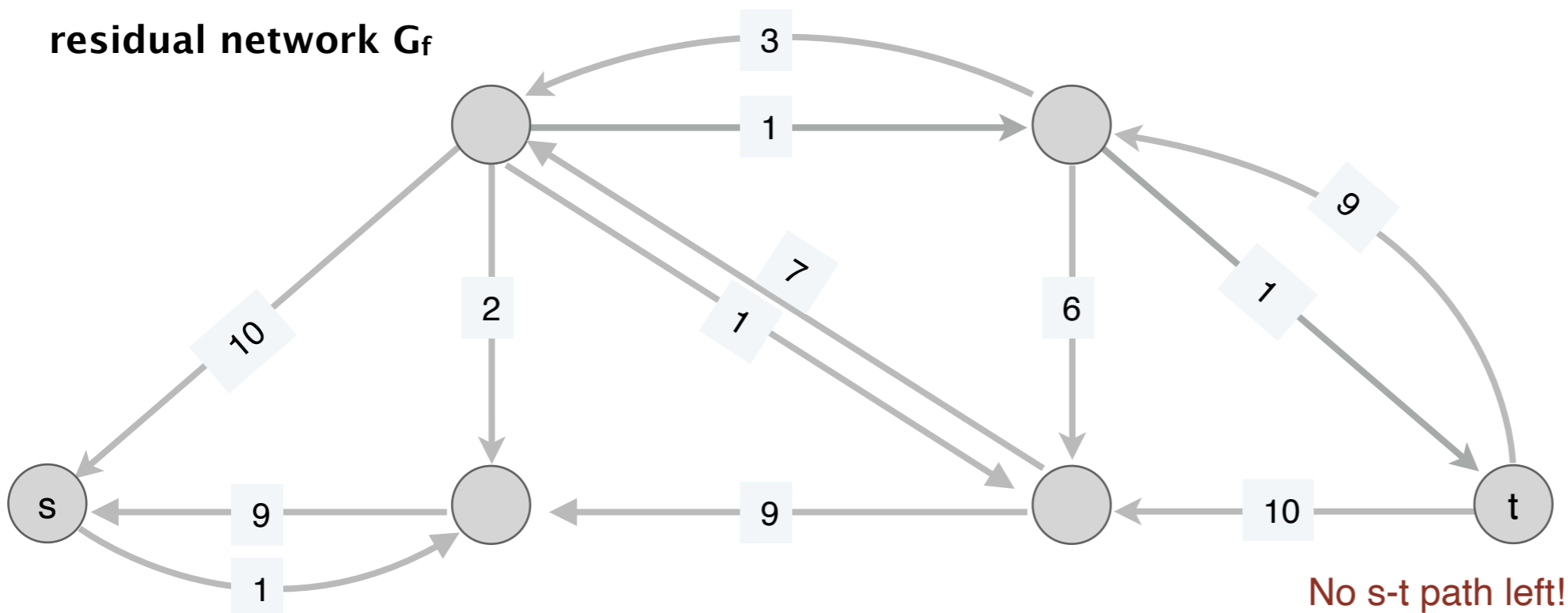


Ford-Fulkerson Example

network G and flow f

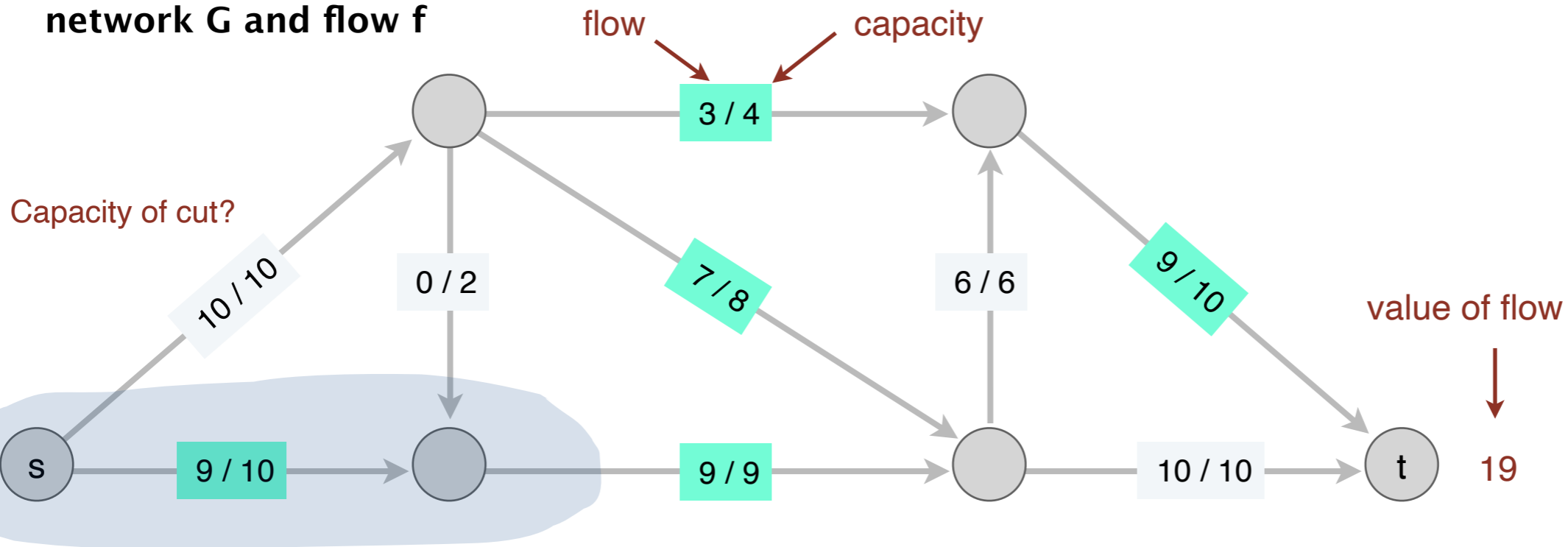


residual network G_f

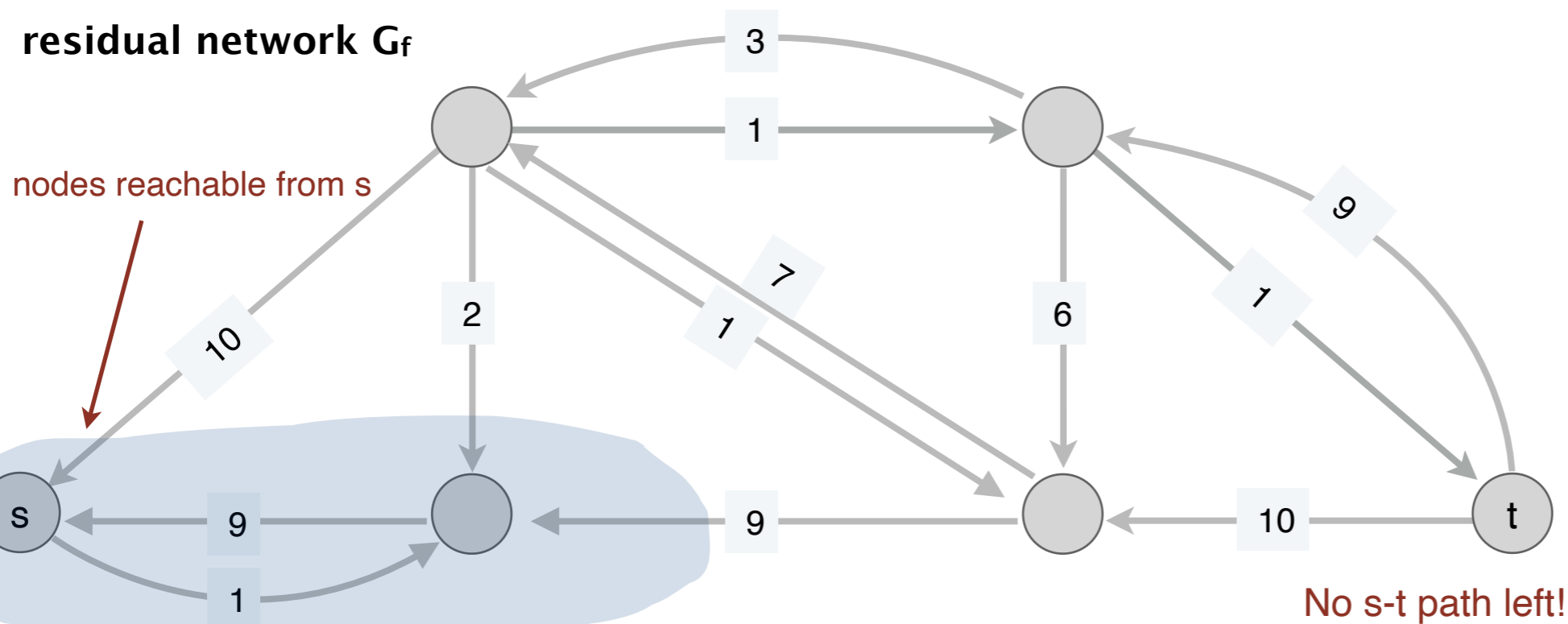


Ford-Fulkerson Example

network G and flow f



residual network G_f



Analysis: Ford-Fulkerson

Analysis Outline

- Feasibility and value of flow:
 - Show that each time we update the flow, we are routing a feasible s - t flow through the network
 - And that value of this flow increases each time by that amount
- Optimality:
 - Final value of flow is the maximum possible
- Running time:
 - How long does it take for the algorithm to terminate?
- Space:
 - How much total space are we using

Feasibility of Flow

- **Claim.** Let f be a feasible flow in G and let P be an augmenting path in G_f with bottleneck capacity b . Let $f' \leftarrow \text{AUGMENT}(f, P)$, then f' is **a feasible flow**.
- **Proof.** Only need to verify constraints on the edges of P (since $f' = f$ for other edges). Let $e = (u, v) \in P$
 - If e is a forward edge: $f'(e) = f(e) + b$
$$\leq f(e) + (c(e) - f(e)) = c(e)$$
 - If e is a backward edge: $f'(e) = f(e) - b$
$$\geq f(e) - f(e) = 0$$
- Conservation constraint hold on any node in $u \in P$:
 - $f_{in}(u) = f_{out}(u)$, therefore $f'_{in}(u) = f'_{out}(u)$ for both cases

Value of Flow: Making Progress

- **Claim.** Let f be a feasible flow in G and let P be an augmenting path in G_f with bottleneck capacity b . Let $f' \leftarrow \text{AUGMENT}(f, P)$, then $v(f') = v(f) + b$.
- **Proof.**
 - First edge $e \in P$ must be out of s in G_f
 - (P is simple so never visits s again)
 - e must be a forward edge (P is a path from s to t)
 - Thus $f(e)$ increases by b , increasing $v(f)$ by b ■
- Note. Means the algorithm makes forward progress each time!

Optimality

Ford-Fulkerson Optimality

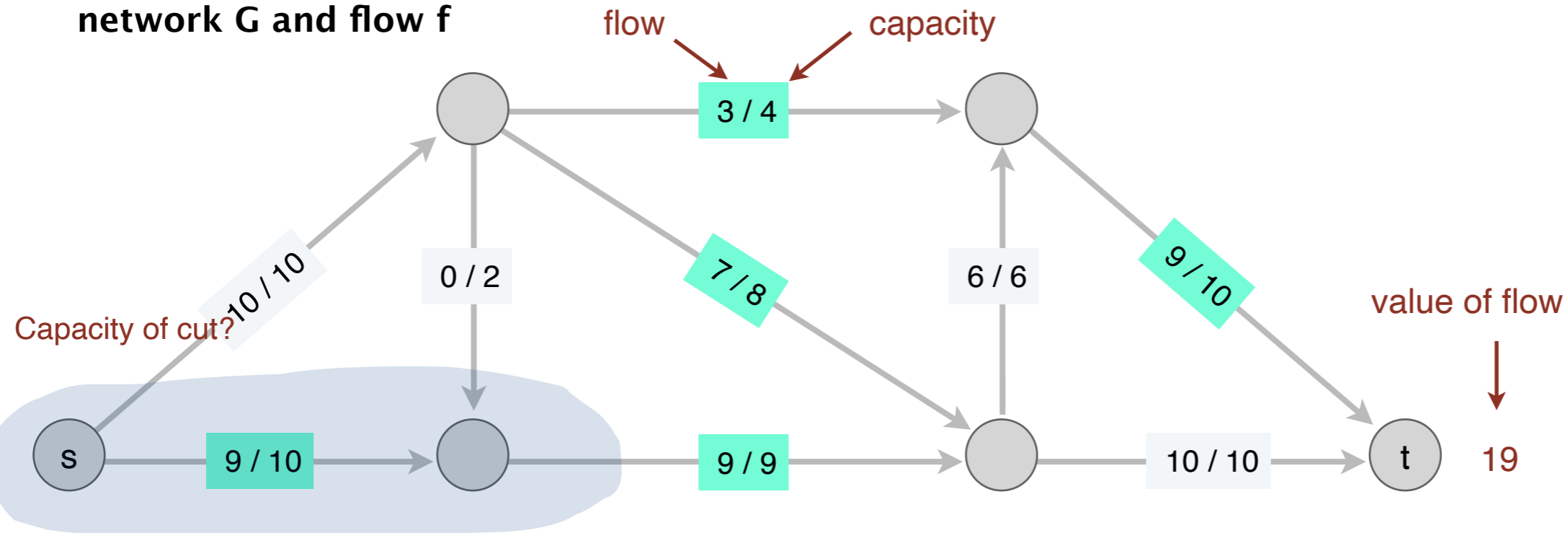
- **Recall:** If f is any feasible s - t flow and (S, T) is any s - t cut then $v(f) \leq c(S, T)$.
- We will show that the Ford-Fulkerson algorithm terminates in a flow that achieves equality, that is,
- Ford-Fulkerson finds a flow f^* and there exists a cut (S^*, T^*) such that, $v(f^*) = c(S^*, T^*)$
- Proving this shows that it finds the maximum flow (and the min cut)
- This also **proves the max-flow min-cut theorem**

Ford-Fulkerson Optimality

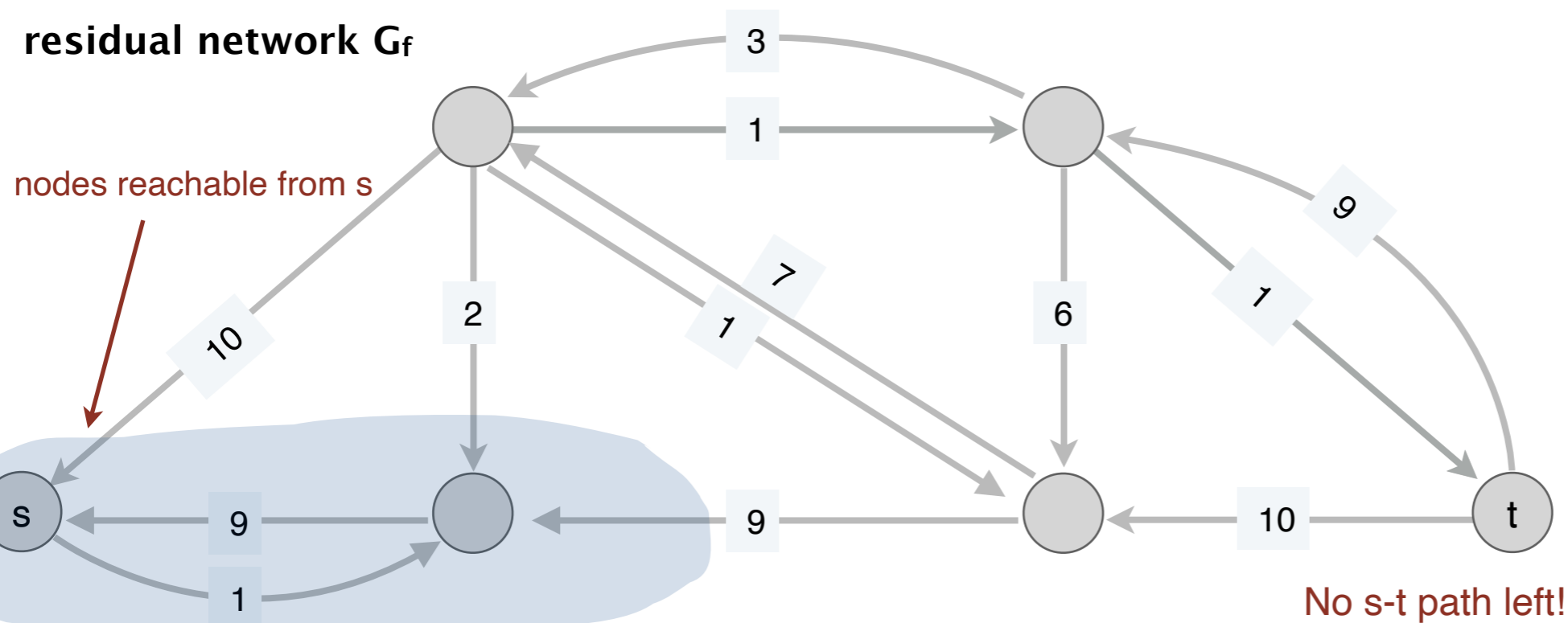
- **Lemma.** Let f be a s - t flow in G such that there is no augmenting path in the residual graph G_f , then there exists a cut (S^*, T^*) such that $v(f) = c(S^*, T^*)$.
- **Proof.**
- Let $S^* = \{v \mid v \text{ is reachable from } s \text{ in } G_f\}$, $T^* = V - S^*$
- Is this an s - t cut?
 - $s \in S, t \in T, S \cup T = V$ and $S \cap T = \emptyset$
- Consider an edge $e = u \rightarrow v$ with $u \in S^*, v \in T^*$, then what can we say about $f(e)$?

Recall: Ford-Fulkerson Example

network G and flow f



residual network G_f



Ford-Fulkerson Optimality

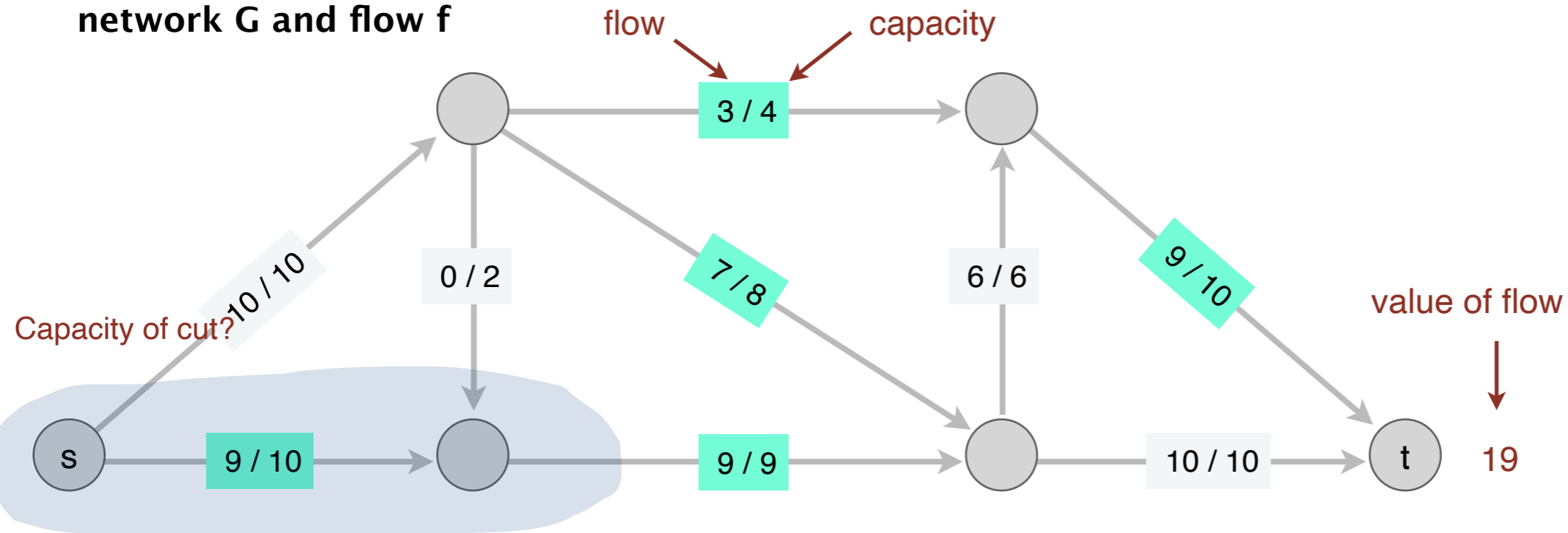
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- **Proof.**
- Let $S^* = \{v \mid v \text{ is reachable from } s \text{ in } G_f\}$, $T^* = V - S^*$
- Is this an s - t cut?
 - $s \in S, t \in T, S \cup T = V$ and $S \cap T = \emptyset$
- Consider an edge $e = u \rightarrow v$ with $u \in S^*, v \in T^*$, then what can we say about $f(e)$?
 - $f(e) = c(e)$

Ford-Fulkerson Optimality

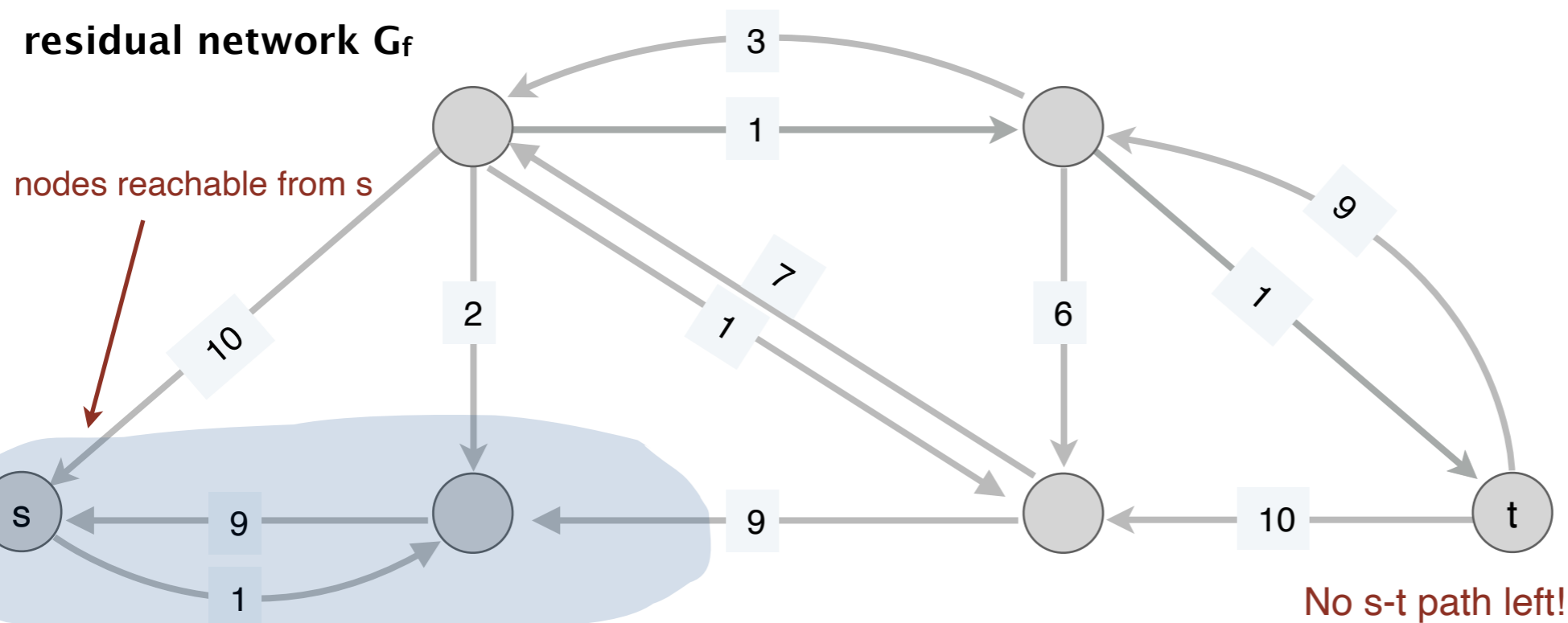
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- **Proof. (Cont.)**
- Let $S^* = \{v \mid v \text{ is reachable from } s \text{ in } G_f\}$, $T^* = V - S^*$
- Is this an s - t cut?
 - $s \in S, t \in T, S \cup T = V$ and $S \cap T = \emptyset$
- Consider an edge $e = w \rightarrow v$ with $v \in S^*, w \in T^*$, then what can we say about $f(e)$?

Recall: Ford-Fulkerson Example

network G and flow f



residual network G_f



Ford-Fulkerson Optimality

- **Lemma.** Let f be a s - t flow in G such that there is no augmenting path in the residual graph G_f , then there exists a cut (S^*, T^*) such that $v(f) = c(S^*, T^*)$.
- **Proof. (Cont.)**
- Let $S^* = \{v \mid v \text{ is reachable from } s \text{ in } G_f\}$, $T^* = V - S^*$
- Is this an s - t cut?
 - $s \in S, t \in T, S \cup T = V$ and $S \cap T = \emptyset$
- Consider an edge $e = w \rightarrow v$ with $v \in S^*, w \in T^*$, then what can we say about $f(e)$?
 - $f(e) = 0$

Ford-Fulkerson Optimality

- **Lemma.** Let f be a s - t flow in G such that there is no augmenting path in the residual graph G_f , then there exists a cut (S^*, T^*) such that $v(f) = c(S^*, T^*)$.
- **Proof. (Cont.)**
- Let $S^* = \{v \mid v \text{ is reachable from } s \text{ in } G_f\}$, $T^* = V - S^*$
- Thus, all edges leaving S^* are completely saturated and all edges entering S^* have zero flow
- $v(f) = f_{out}(S^*) - f_{in}(S^*) = f_{out}(S^*) = c(S^*, T^*)$ ■
- **Corollary.** Ford-Fulkerson returns the maximum flow.

Ford-Fulkerson Algorithm

Running Time

Ford-Fulkerson Performance

FORD-FULKERSON(G)

FOREACH edge $e \in E$: $f(e) \leftarrow 0$.

$G_f \leftarrow$ residual network of G with respect to flow f .

WHILE (there exists an s - t path P in G_f)

$f \leftarrow$ AUGMENT(f, P).

Update G_f .

RETURN f .

- Does the algorithm terminate?
- Can we bound the number of iterations it does?
- Running time?

Ford-Fulkerson Running Time

- Recall we proved that with each call to AUGMENT, we increase **value of flow** by $b = \text{bottleneck}(G_f, P)$
- **Assumption.** Suppose all capacities $c(e)$ are integers.
- **Integrality invariant.** Throughout Ford–Fulkerson, every edge flow $f(e)$ and corresponding residual capacity is an integer. Thus $b \geq 1$.
- Let $C = \max_u c(s \rightarrow u)$ be the maximum capacity among edges leaving the source s .
- It must be that $v(f) \leq (n - 1)C$
- Since, $v(f)$ increases by $b \geq 1$ in each iteration, it follows that FF algorithm terminates in at most $v(f) = O(nC)$ iterations.

Ford-Fulkerson Performance

FORD-FULKERSON(G)

FOREACH edge $e \in E$: $f(e) \leftarrow 0$.

$G_f \leftarrow$ residual network of G with respect to flow f .

WHILE (there exists an $s \rightsquigarrow t$ path P in G_f)

$f \leftarrow$ AUGMENT(f, P).

Update G_f .

RETURN f .

- Operations in each iteration?
 - Find an augmenting path in G_f
 - Augment flow on path
 - Update G_f

Ford-Fulkerson Running Time

- **Claim.** Ford-Fulkerson can be implemented to run in time $O(nmC)$, where $m = |E| \geq n - 1$ and $C = \max_u c(s \rightarrow u)$.
- **Proof.** Time taken by each iteration:
 - Finding an augmenting path in G_f
 - G_f has at most $2m$ edges, using BFS/DFS takes $O(m + n) = O(m)$ time
 - Augmenting flow in P takes $O(n)$ time
 - Given new flow, we can build new residual graph in $O(m)$ time
- Overall, $O(m)$ time per iteration ■

Midterm Review

- Topics:
 - Dijkstra's/max flow min cut
 - Divide and conquer
 - Dynamic Programming

Midterm Tips

- Review topics according to the practice midterm
- Use lectures, assignments
- *Write what you can for partial solutions*
 - If your DP recurrence isn't right, still write the memoization table, the final solution (given your recurrence), etc
 - If your Divide and Conquer algorithm isn't right, still write the running time as a recurrence and solve it
- 1 page 2 sided "cheat sheet" allowed (probably not too useful)

Review Plan

- First: any questions?
- Then: let's pick some homework questions to do on the board
- I'll have to be a little vague for Assignment 5 due to extensions (for holidays/travel/etc.)