

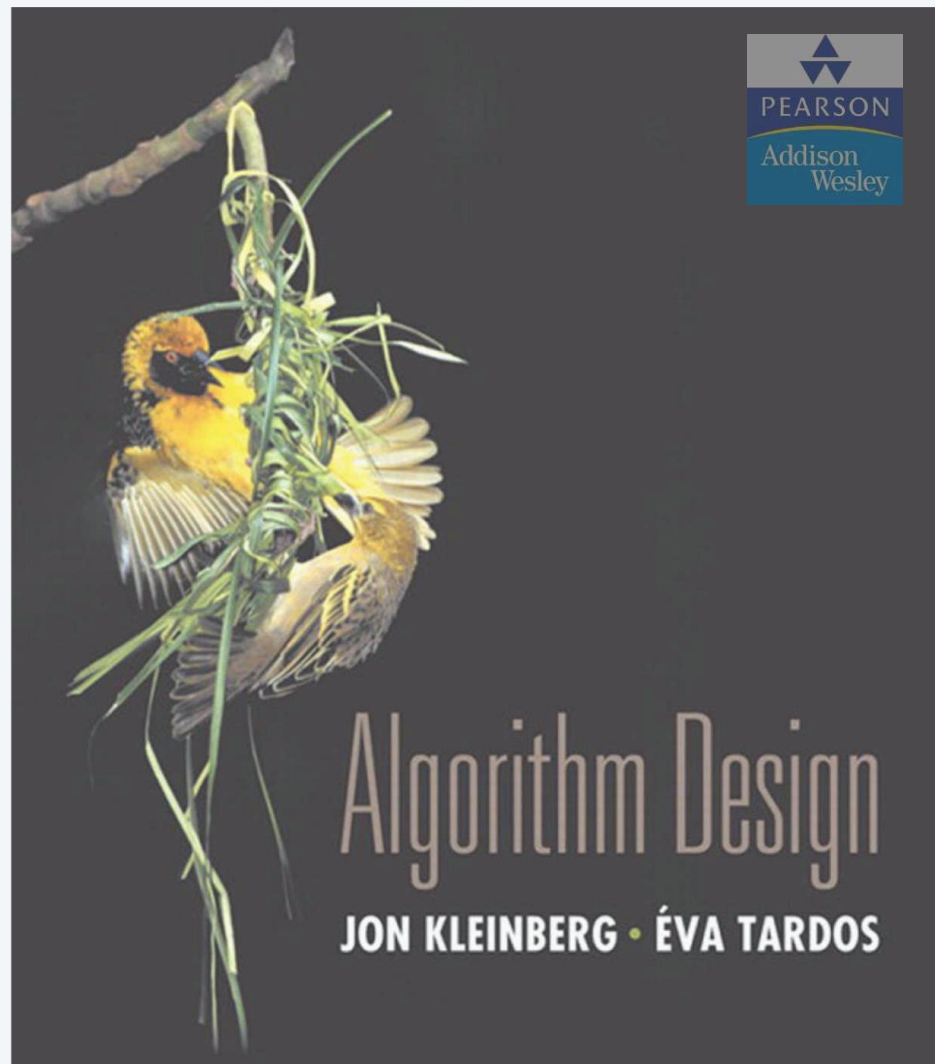
7. NETWORK FLOW I

- ▶ *max-flow and min-cut problems*
- ▶ *Ford–Fulkerson algorithm*
- ▶ *max-flow min-cut theorem*
- ▶ *choosing good augmenting paths*

Lecture slides by Kevin Wayne

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<http://www.cs.princeton.edu/~wayne/kleinberg-tardos>



SECTION 7.1

7. NETWORK FLOW I

- ▶ *max-flow and min-cut problems*
- ▶ *Ford–Fulkerson algorithm*
- ▶ *max-flow min-cut theorem*
- ▶ *choosing good augmenting paths*

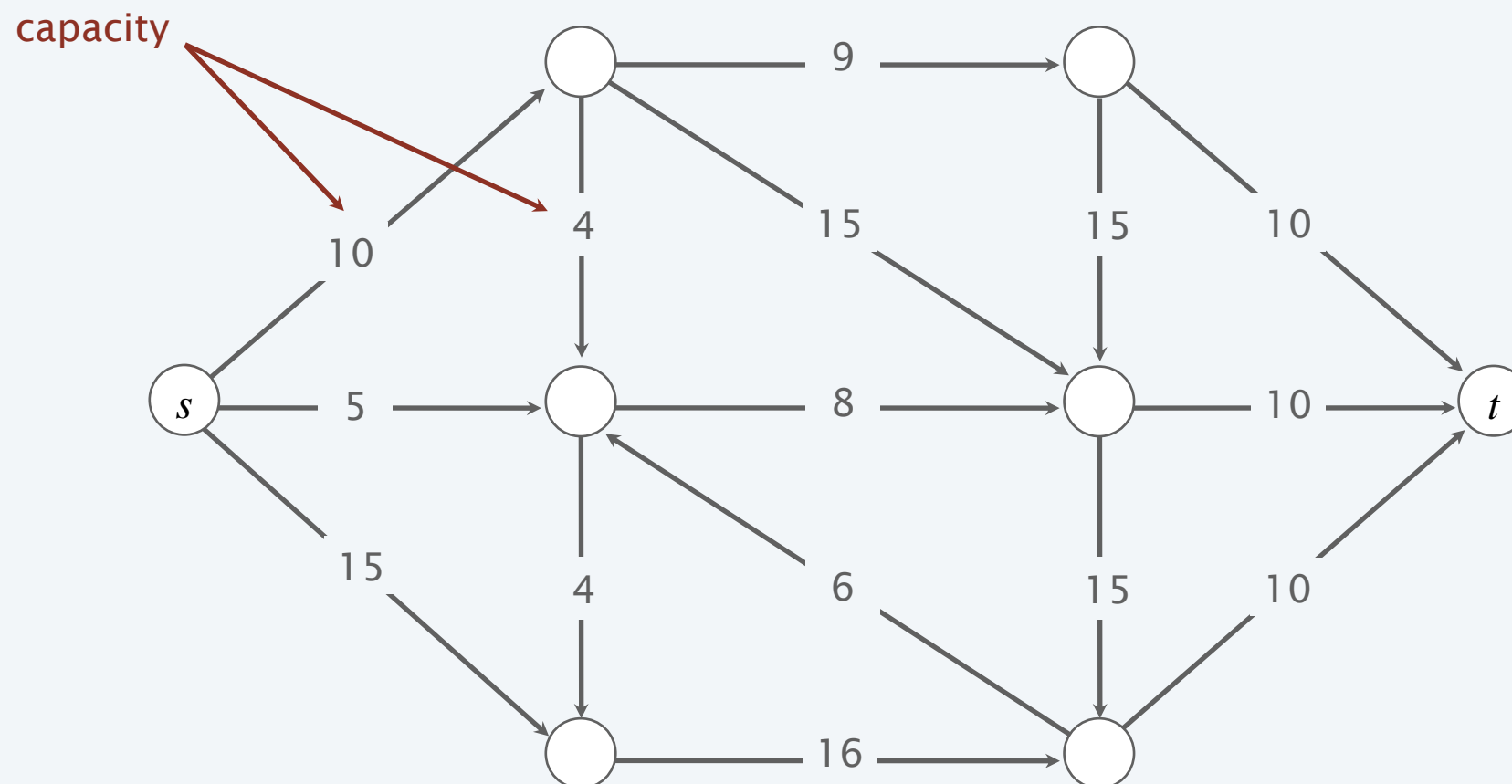
Flow network

A **flow network** is a tuple $G = (V, E, s, t, c)$.

- Digraph (V, E) with source $s \in V$ and sink $t \in V$.
- Capacity $c(e) \geq 0$ for each $e \in E$.

assume all nodes are reachable from s

Intuition. Material flowing through a transportation network; material originates at source and is sent to sink.

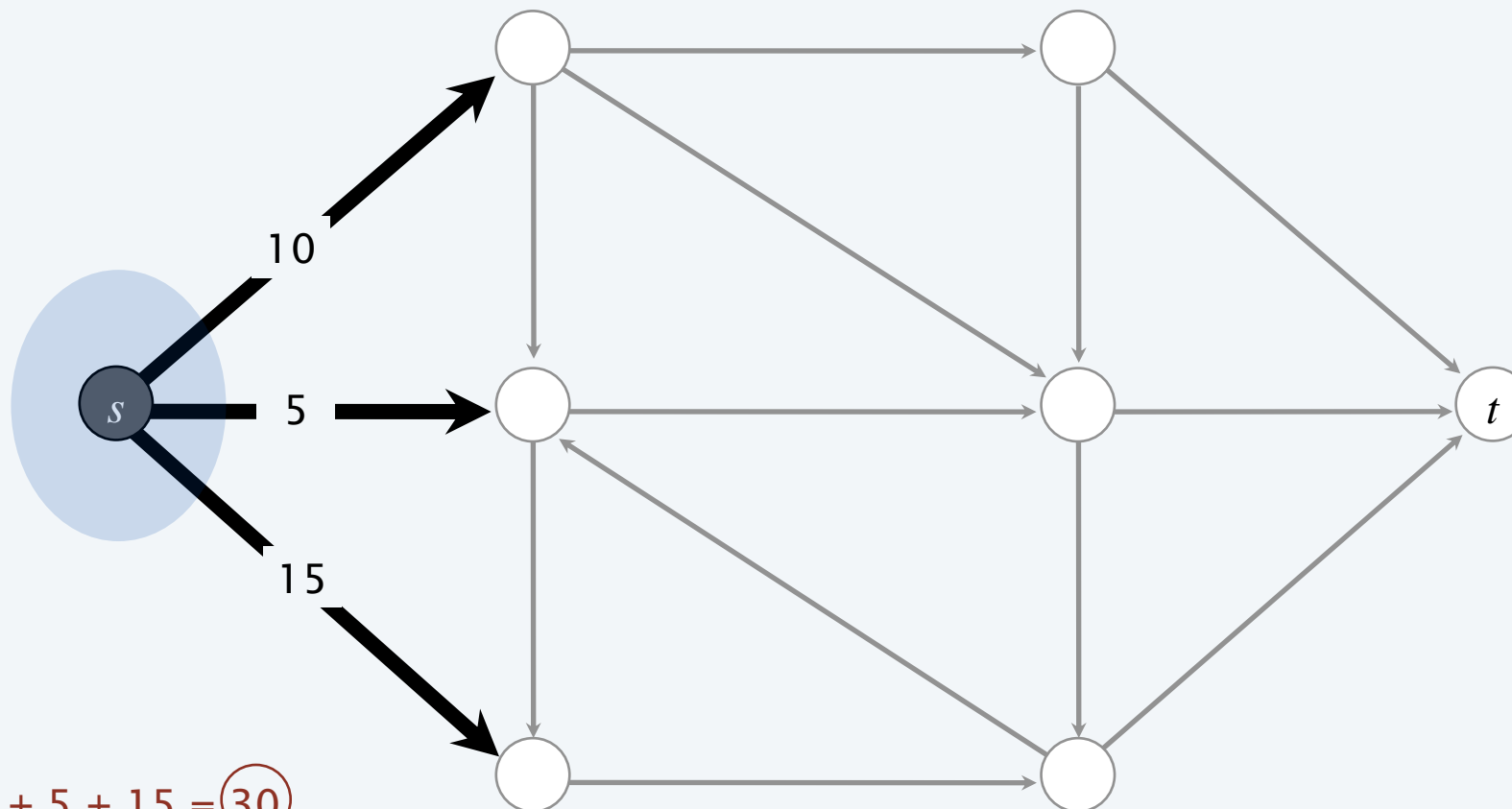


Minimum-cut problem

Def. An *st-cut (cut)* is a partition (A, B) of the nodes with $s \in A$ and $t \in B$.

Def. Its *capacity* is the sum of the capacities of the edges from A to B .

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



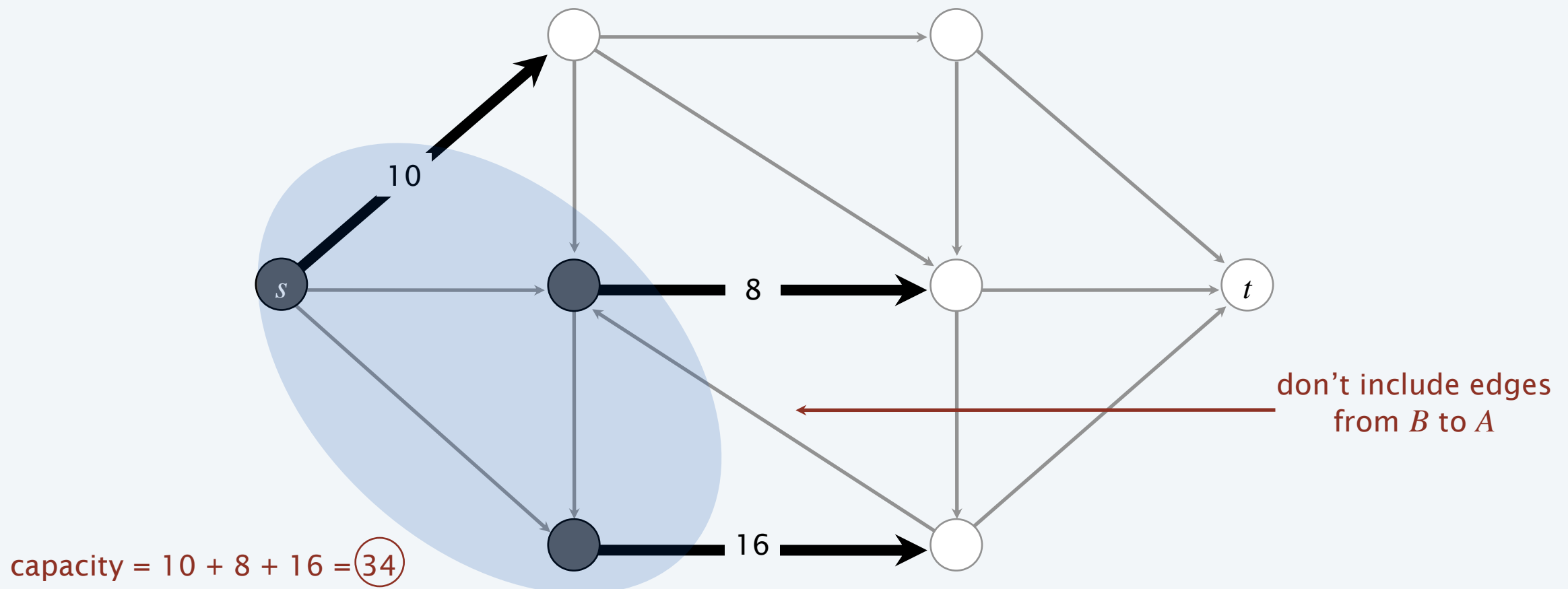
capacity = $10 + 5 + 15 = 30$

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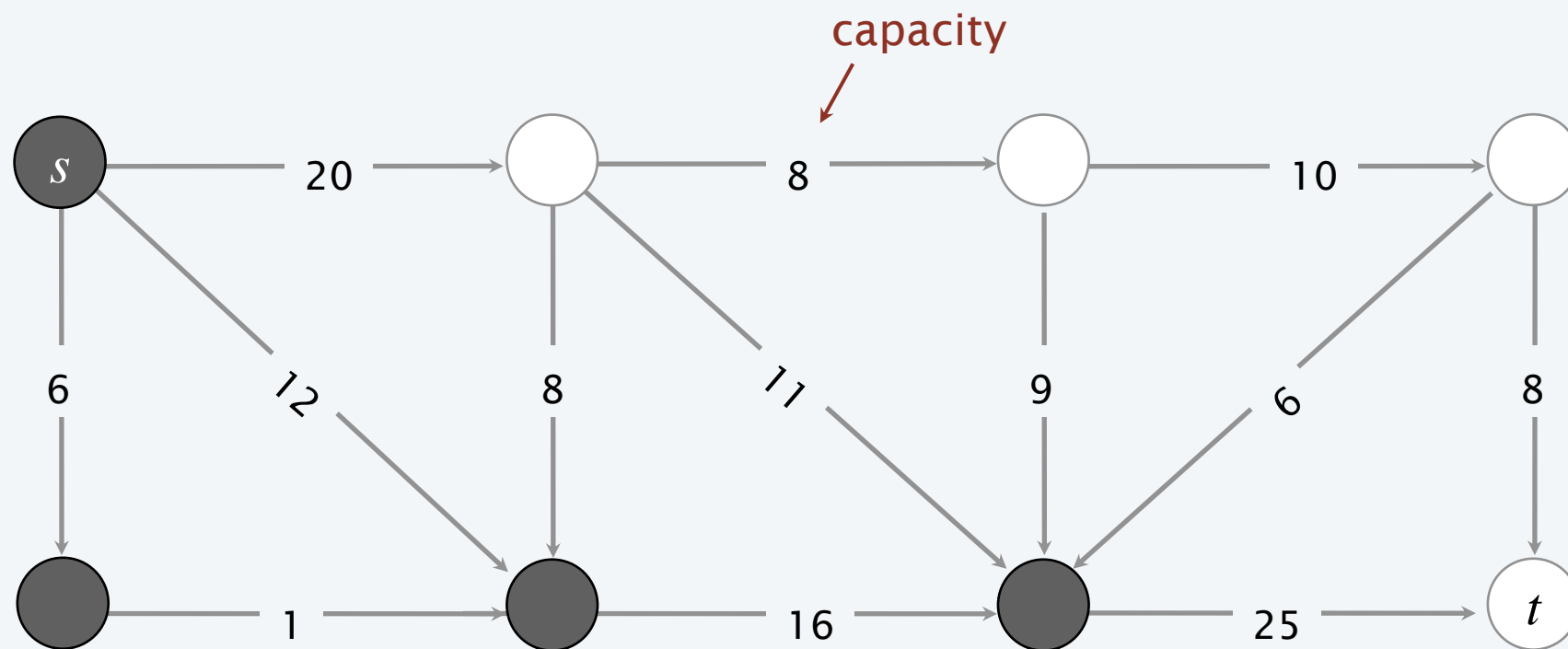
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One more example

Capacity of the given st -cut: $20+25=45$



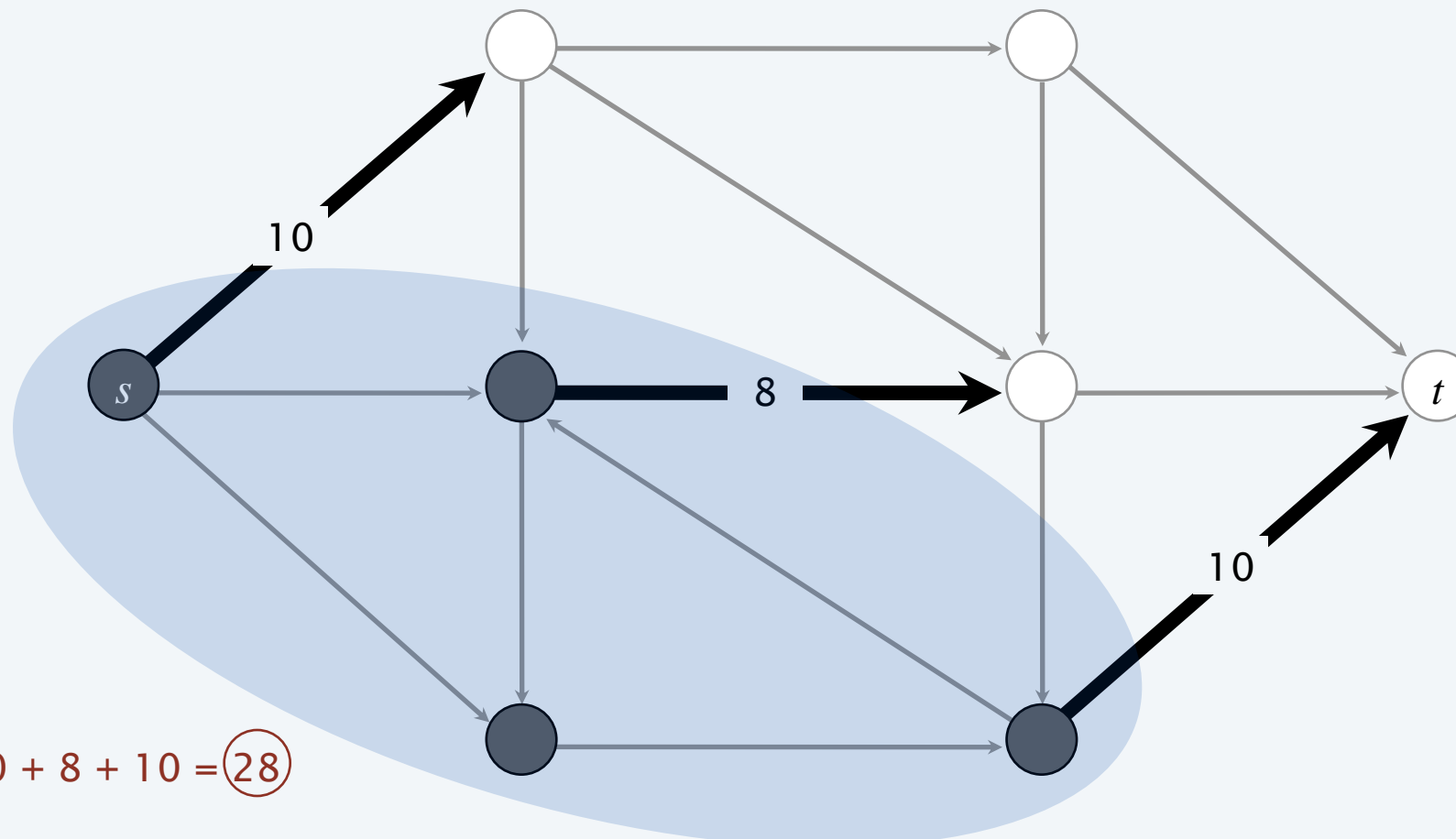
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Min-cut problem. Find a cut of minimum capacity.

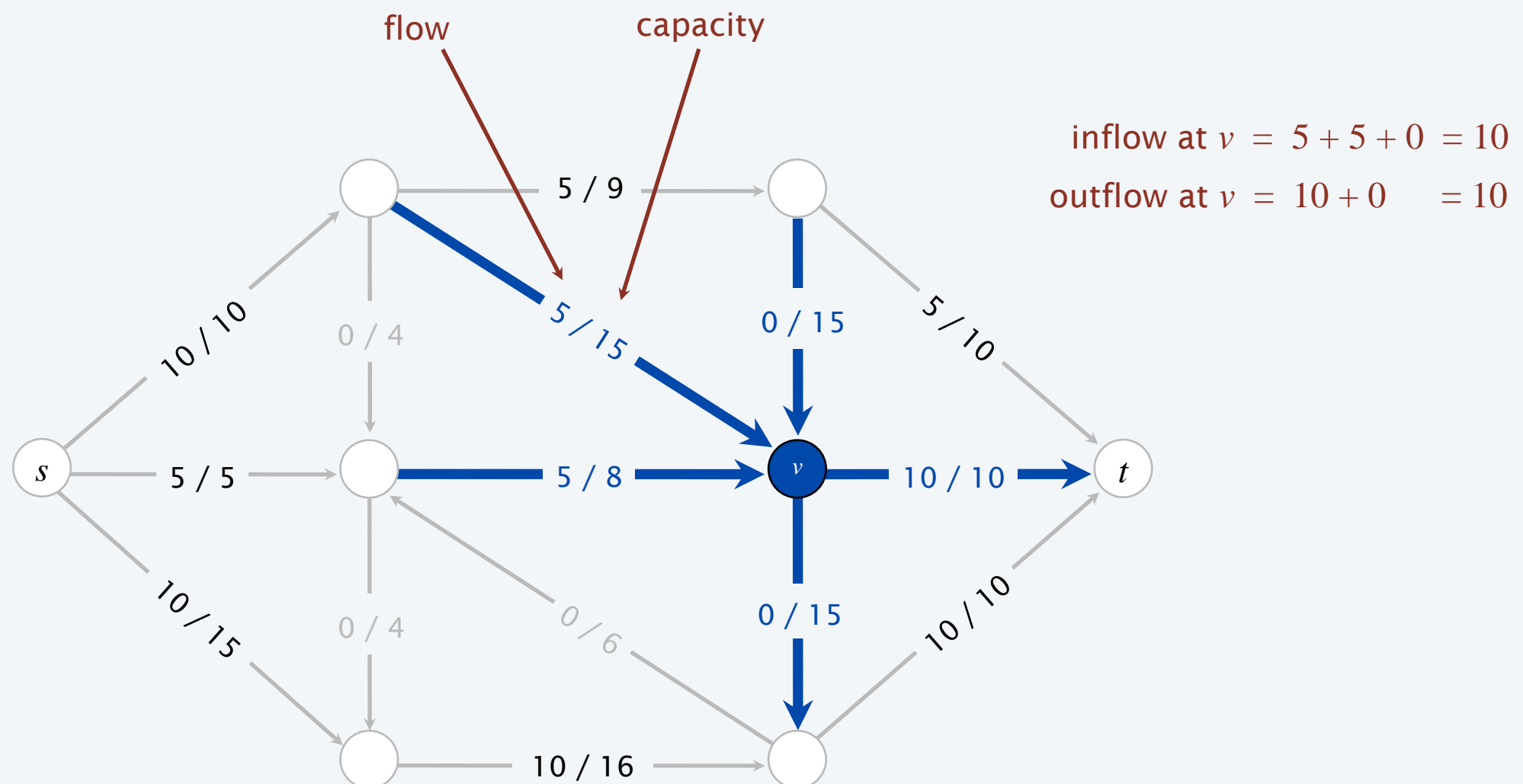


capacity = $10 + 8 + 10 = 28$

Maximum-flow problem

Def. An *st-flow (flow)* f is a function that satisfies:

- For each $e \in E$: $0 \leq f(e) \leq c(e)$ [capacity]
- For each $v \in V - \{s, t\}$: $\sum_{e \text{ in to } v} f(e) = \sum_{e \text{ out of } v} f(e)$ [flow conservation]

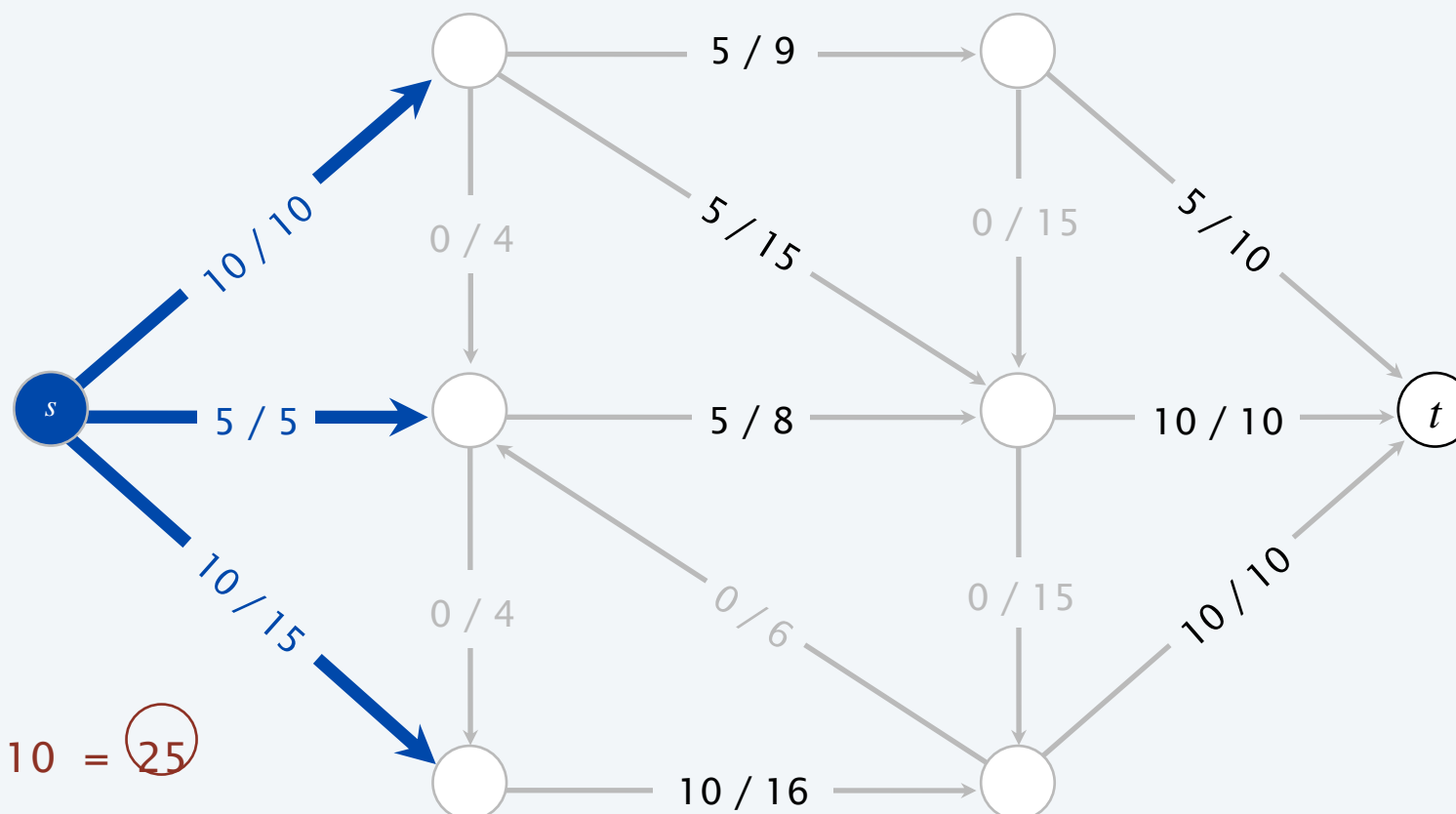


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Def. The *value* of a flow f is: $val(f) = \sum_{e \text{ out of } s} f(e) - \sum_{e \text{ in to } s} f(e)$



$$\text{value} = 5 + 10 + 10 = 25$$

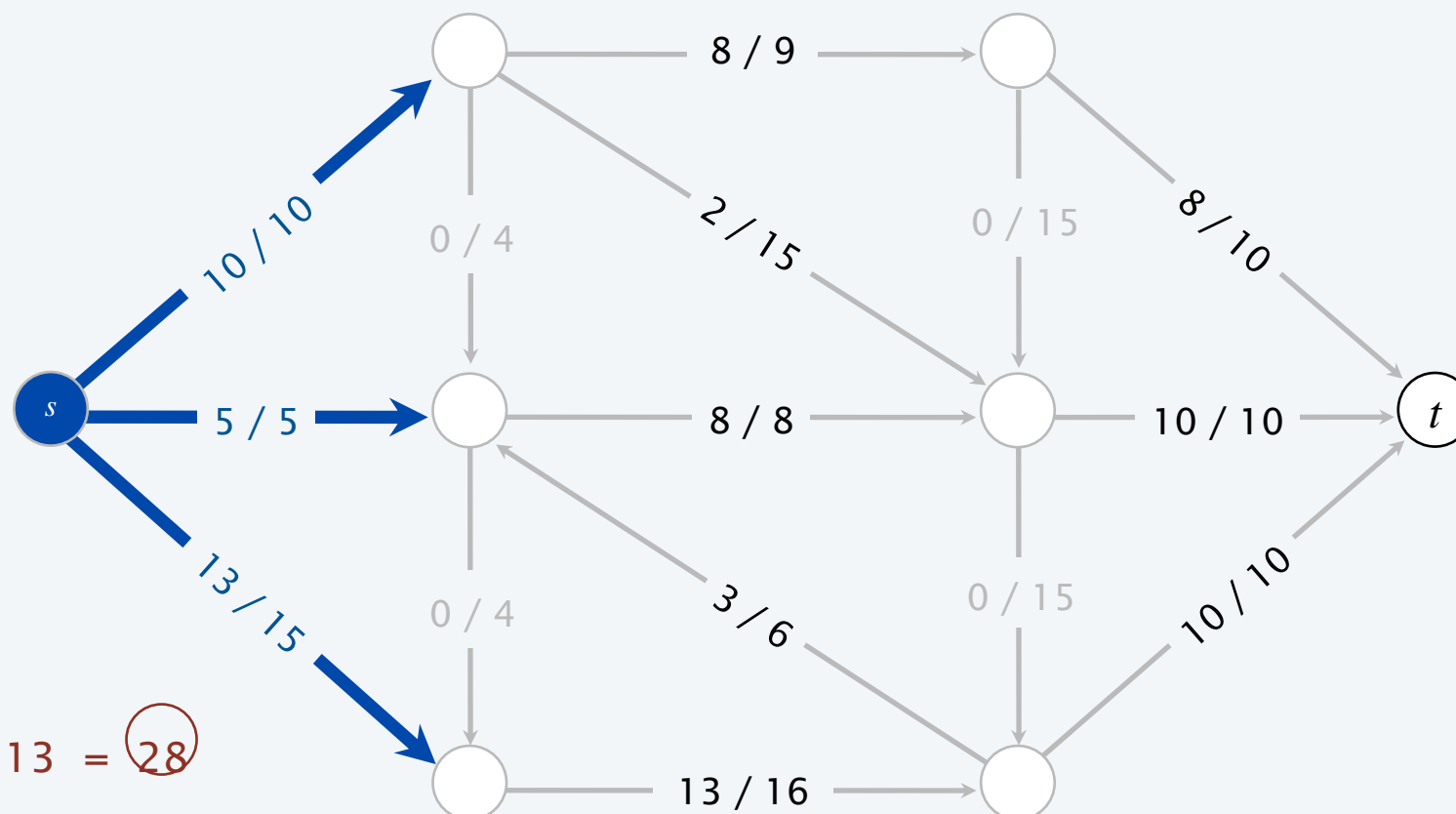
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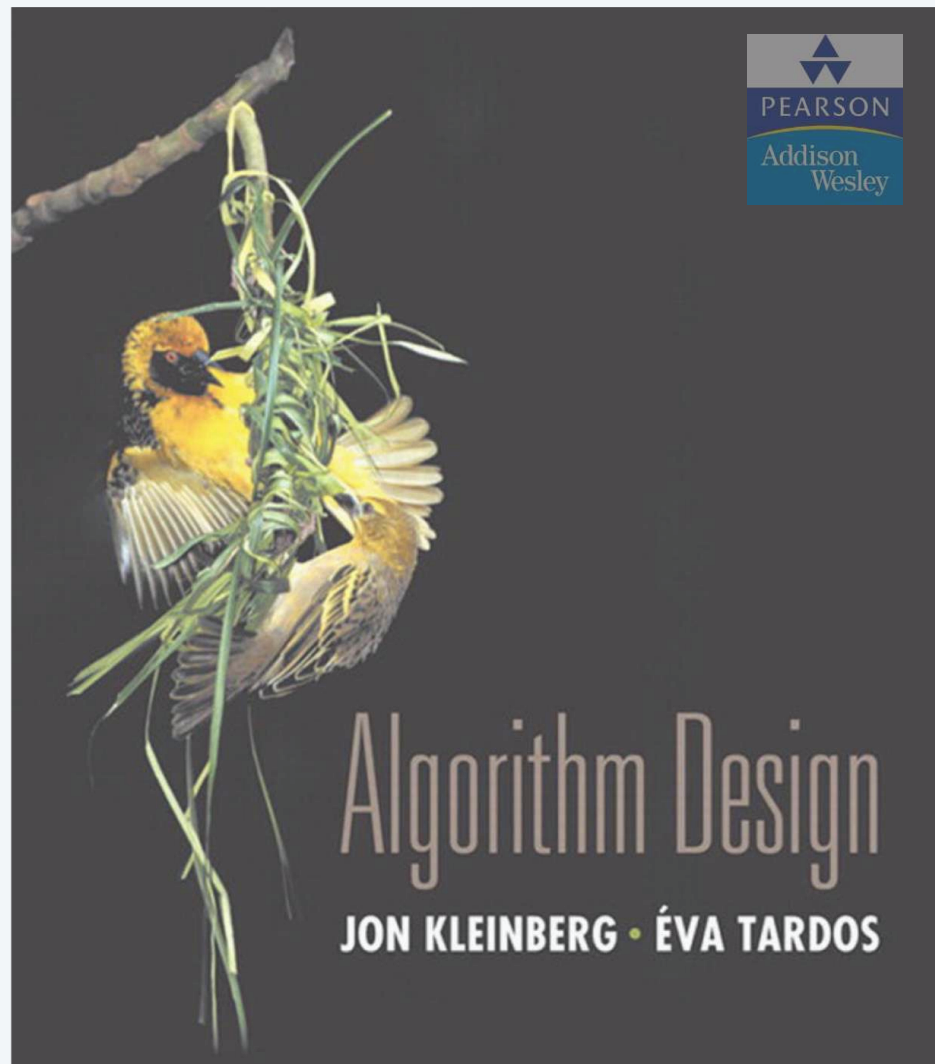
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Max-flow problem. Find a flow of maximum value.



$$\text{value} = 10 + 5 + 13 = 28$$



SECTION 7.1

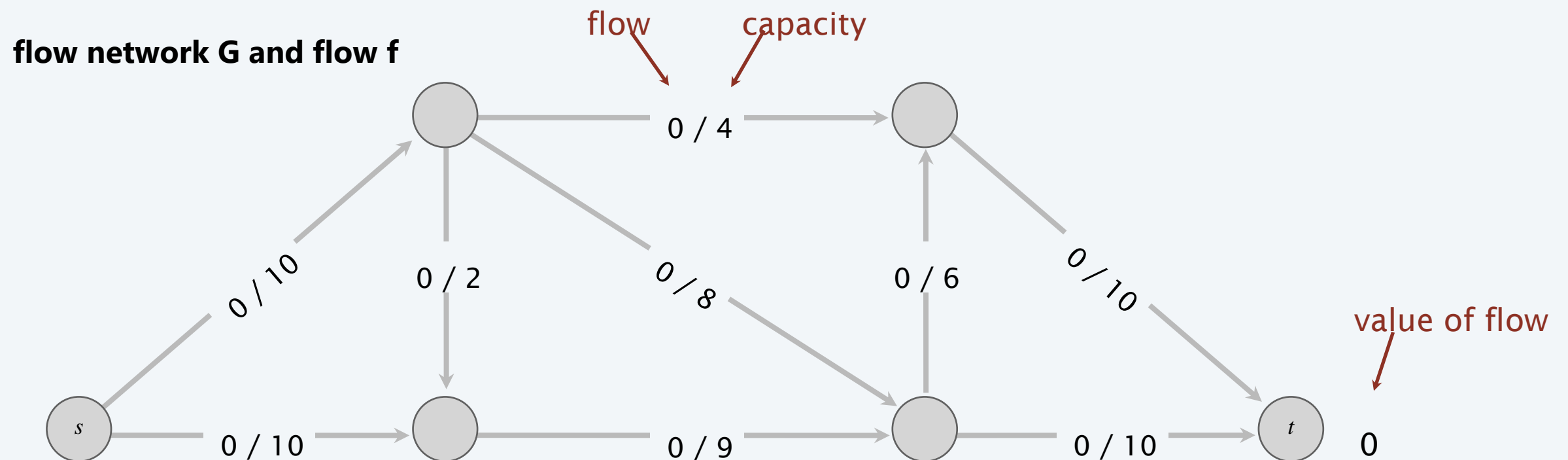
7. NETWORK FLOW I

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Toward a max-flow algorithm

Greedy algorithm.

- Start with $f(e) = 0$ for each edge $e \in E$.
- Find an $s \rightsquigarrow t$ path P where each edge has $f(e) < c(e)$.
- Augment flow along path P .
- Repeat until you get stuck.

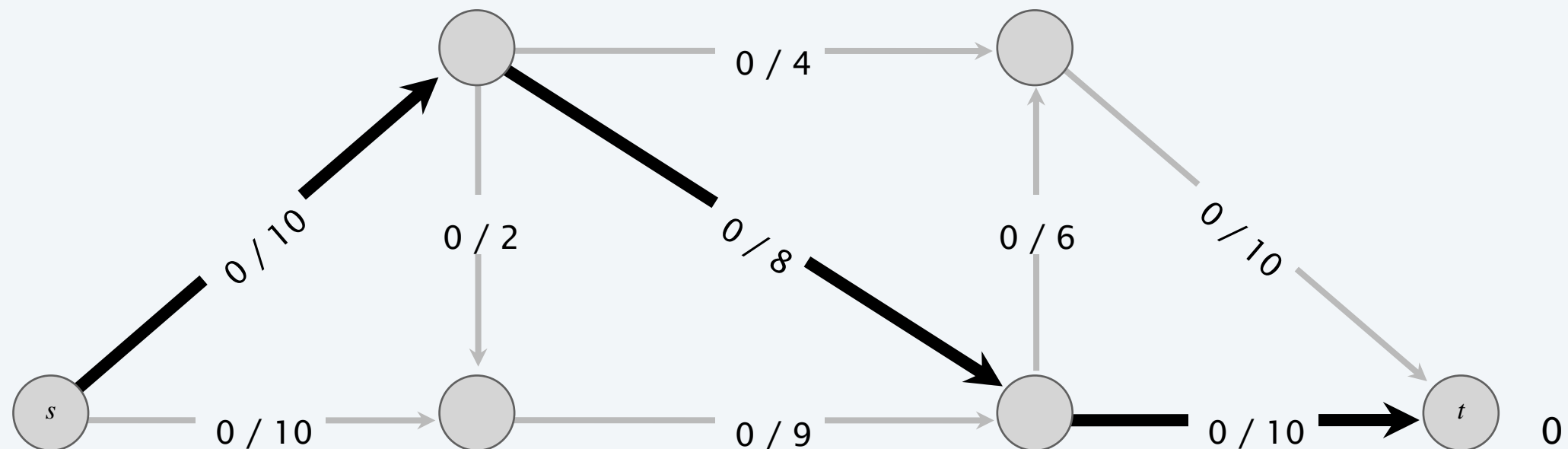


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flow network G and flow f

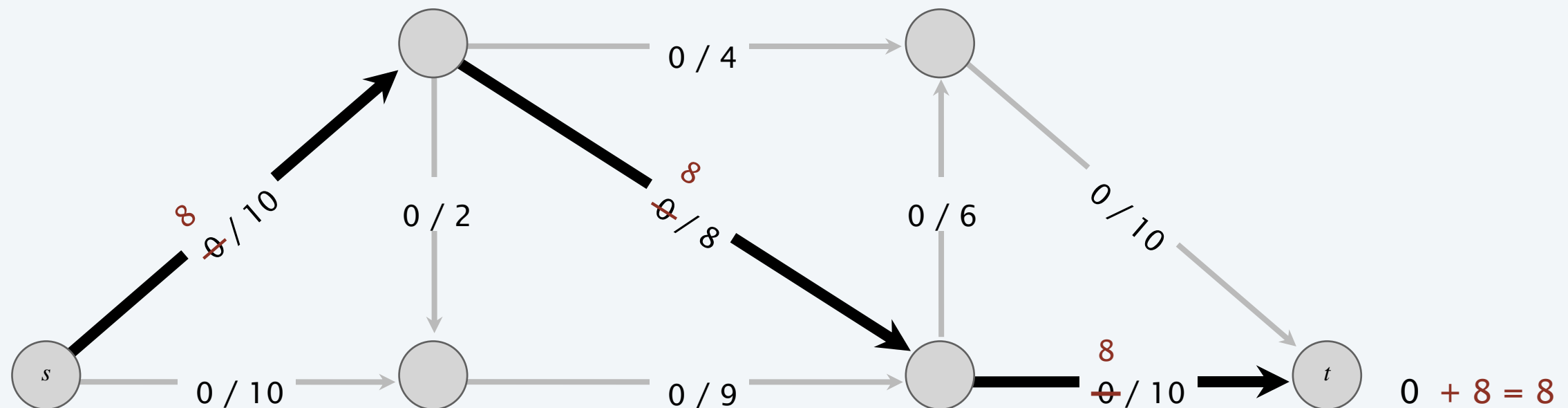


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flow network G and flow f

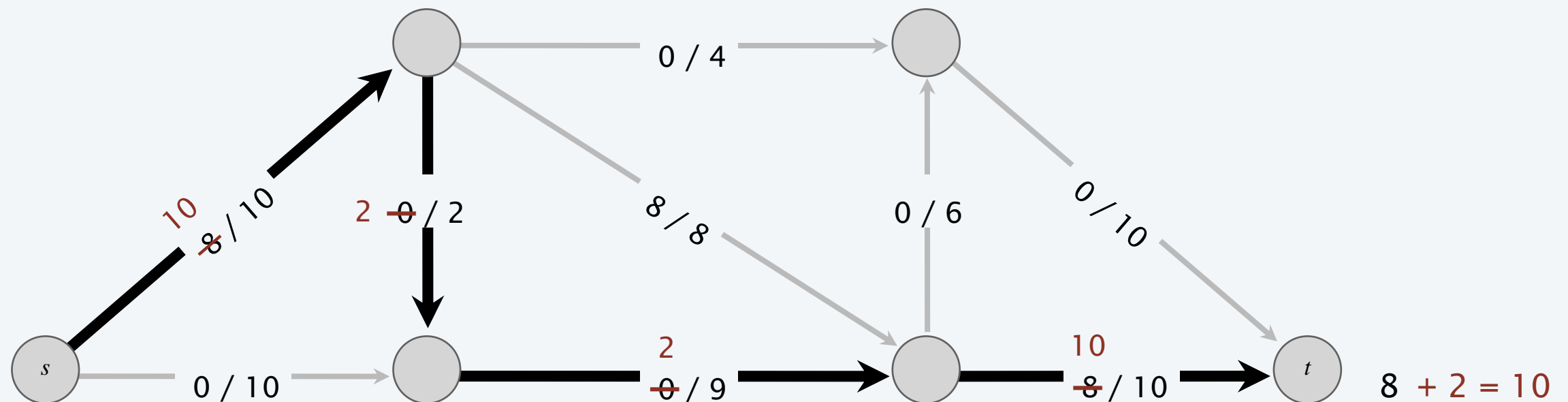


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flow network G and flow f

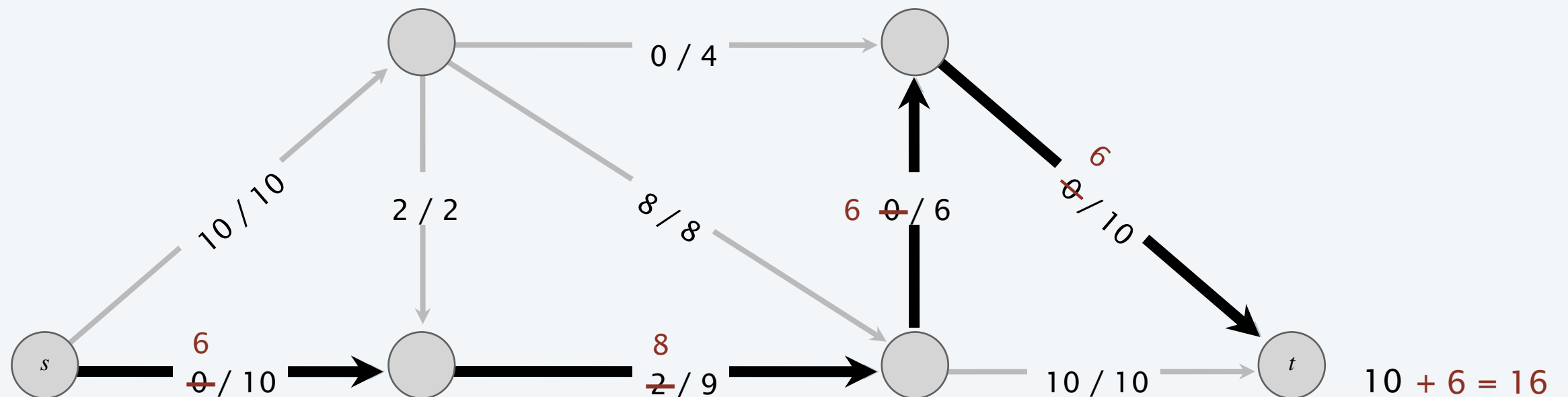


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flow network G and flow f



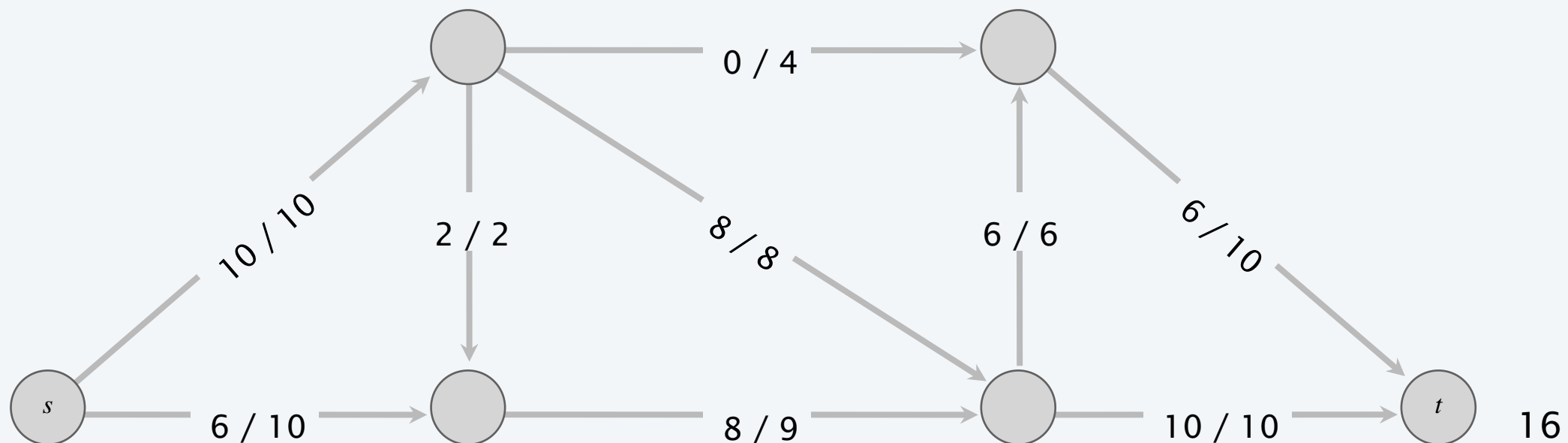
Toward a max-flow algorithm

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- Augment flow along path P .
- Repeat until you get stuck.

ending flow value = 16

flow network G and flow f



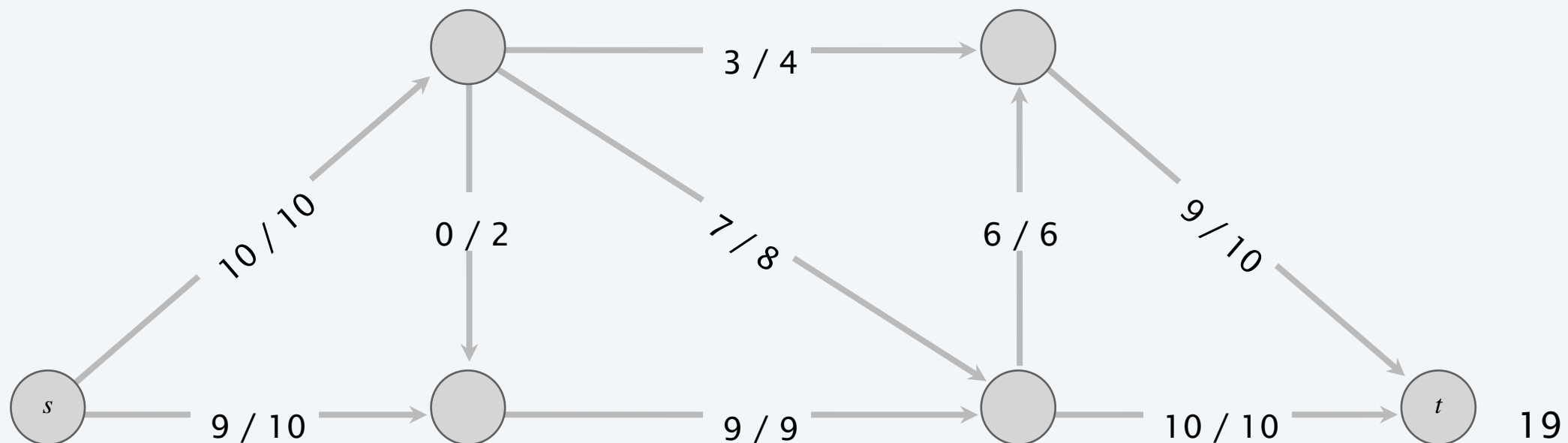
Toward a max-flow algorithm

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- Augment flow along path P .
- Repeat until you get stuck.

but max-flow value = 19

flow network G and flow f



Why the greedy algorithm fails

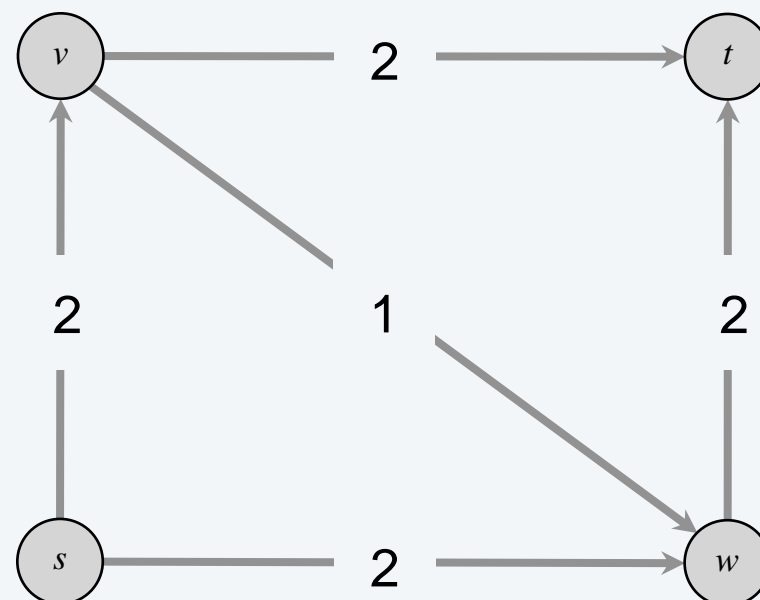
Q. Why does the greedy algorithm fail?

A. Once greedy algorithm increases flow on an edge, it never decreases it.

Ex. Consider flow network G .

- The unique max flow f^* has $f^*(v, w) = 0$.
- Greedy algorithm could choose $s \rightarrow v \rightarrow w \rightarrow t$ as first path.

flow network G



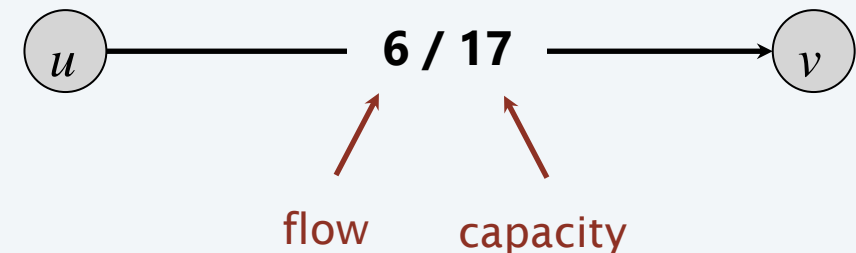
Bottom line. Need some mechanism to “undo” a bad decision.

Residual network

Original edge. $e = (u, v) \in E$.

- Flow $f(e)$.
- Capacity $c(e)$.

original flow network G



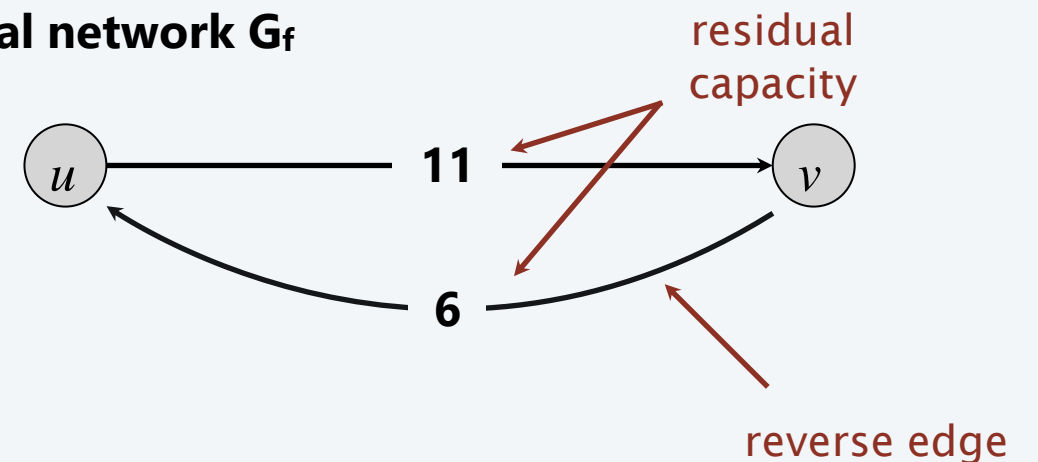
Reverse edge. $e^{\text{reverse}} = (v, u)$.

- “Undo” flow sent.

Residual capacity.

$$c_f(e) = \begin{cases} c(e) - f(e) & \text{if } e \in E \\ f(e^{\text{reverse}}) & \text{if } e^{\text{reverse}} \in E \end{cases}$$

residual network G_f



edges with positive
residual capacity

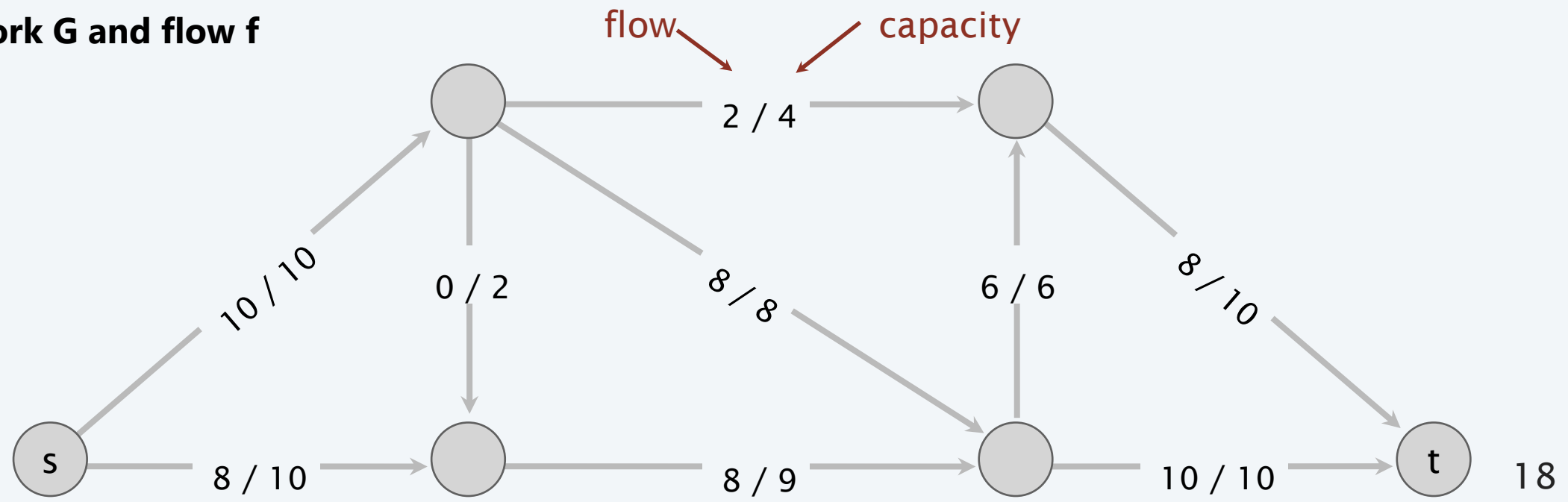
Residual network. $G_f = (V, E_f, s, t, c_f)$.

- $E_f = \{e : f(e) < c(e)\} \cup \{e : f(e^{\text{reverse}}) > 0\}$.
- Key property: f' is a flow in G_f iff $f + f'$ is a flow in G .

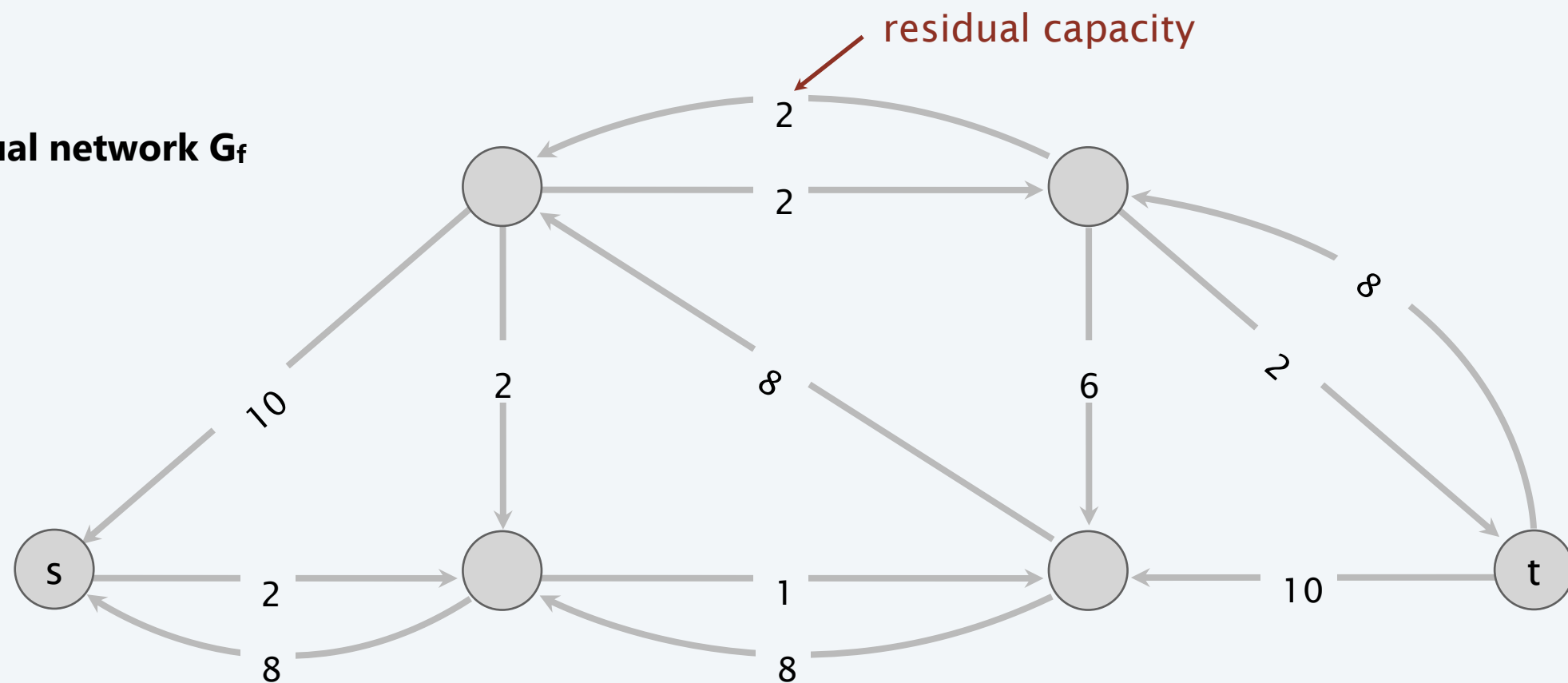
where flow on a reverse edge
negates flow on
corresponding forward edge

Residual network: an example

network **G** and flow **f**



residual network **G_f**



Augmenting path

Def. An **augmenting path** is a simple $s \rightsquigarrow t$ path in the residual network G_f .

Def. The **bottleneck capacity** of an augmenting path P is the minimum residual capacity of any edge in P .

Key property. Let f be a flow and let P be an augmenting path in G_f . Then, after calling $f' \leftarrow \text{AUGMENT}(f, c, P)$, the resulting f' is a flow and $\text{val}(f') = \text{val}(f) + \text{bottleneck}(G_f, P)$.

AUGMENT(f, c, P)

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

FOREACH edge $e \in P$:

IF ($e \in E$) $f(e) \leftarrow f(e) + \delta$.

ELSE $f(e^{\text{reverse}}) \leftarrow f(e^{\text{reverse}}) - \delta$.

RETURN f .

Ford–Fulkerson algorithm

Ford–Fulkerson augmenting path algorithm.

- Start with $f(e) = 0$ for each edge $e \in E$.
- Find an $s \rightsquigarrow t$ path P in the residual network G_f .
- Augment flow along path P .
- 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, c, P).

Update G_f .

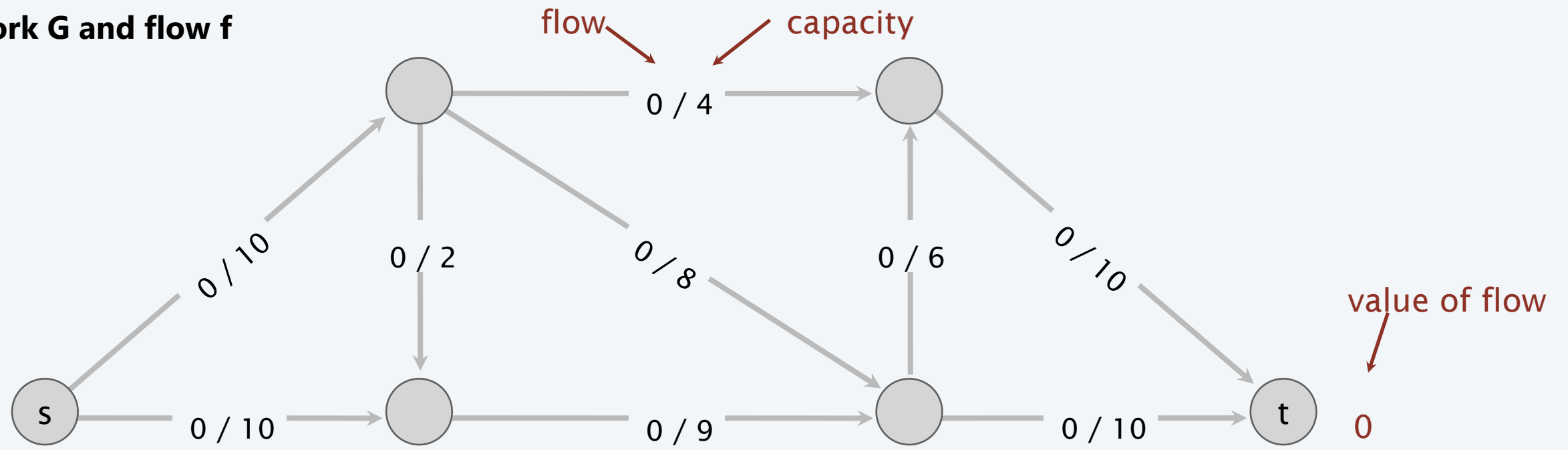
RETURN f .



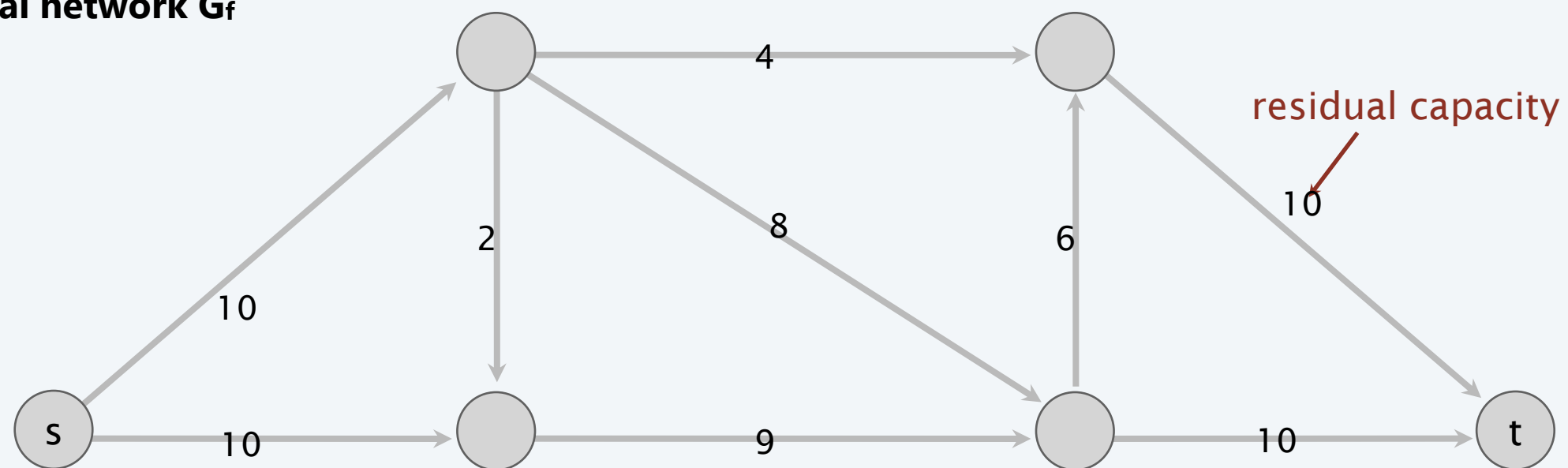
augmenting path

Ford–Fulkerson algorithm demo

network **G** and flow **f**

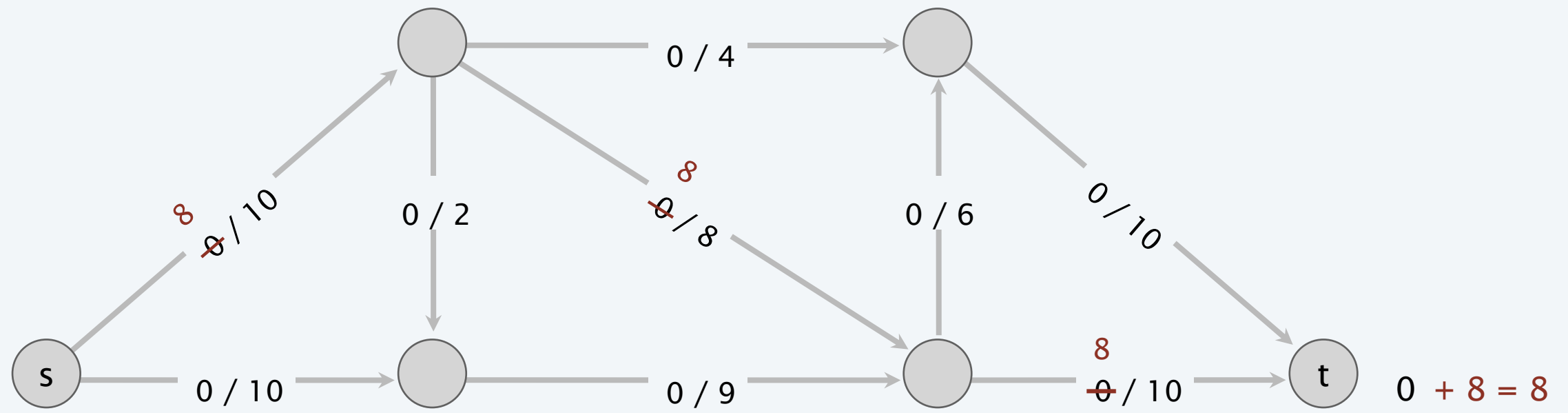


residual network **G_f**

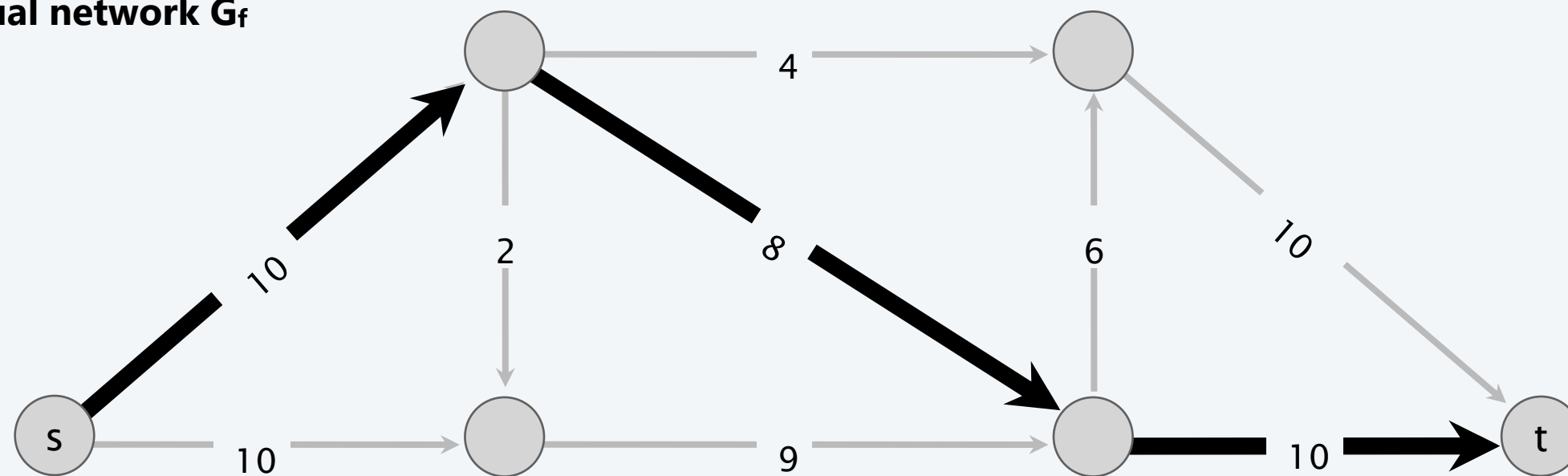


Ford–Fulkerson algorithm demo

network **G** and flow **f**

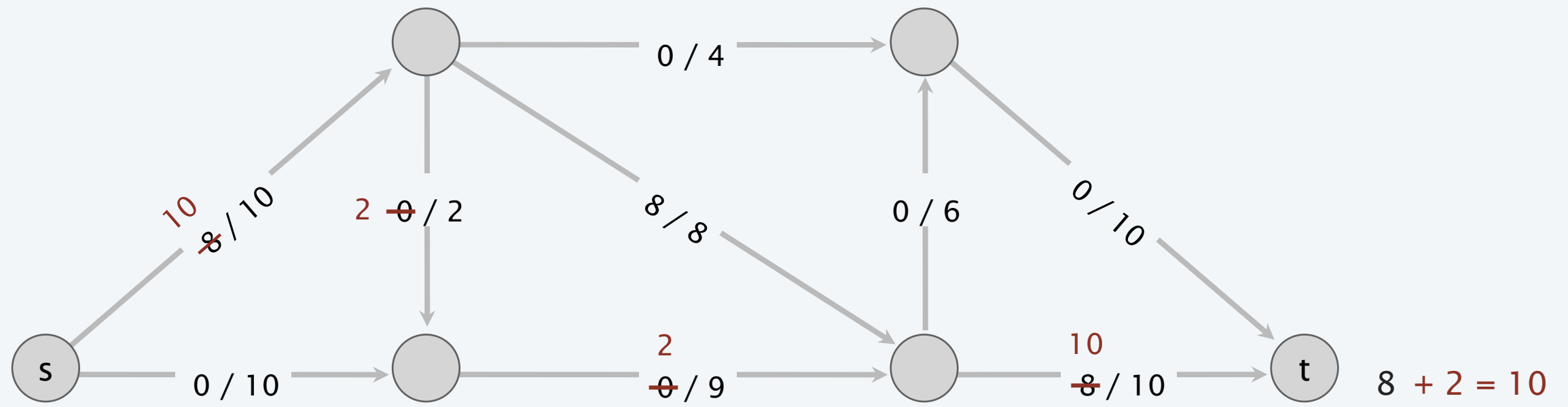


residual network **G_f**

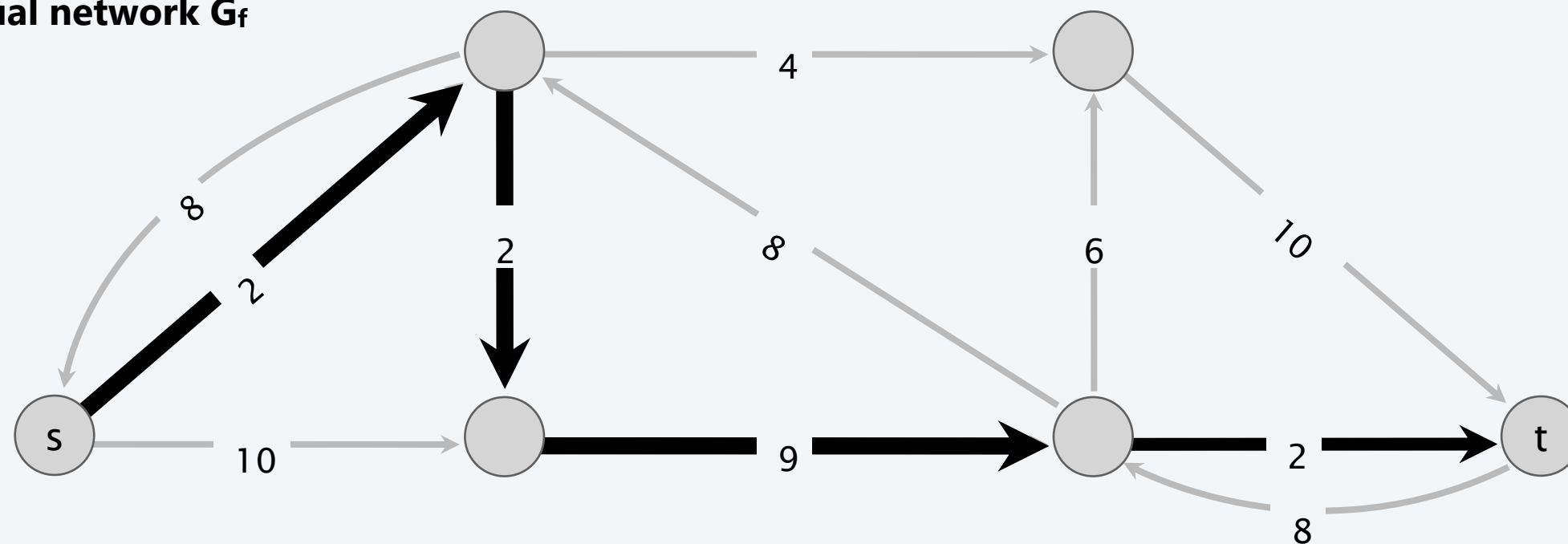


Ford–Fulkerson algorithm demo

network **G** and flow **f**

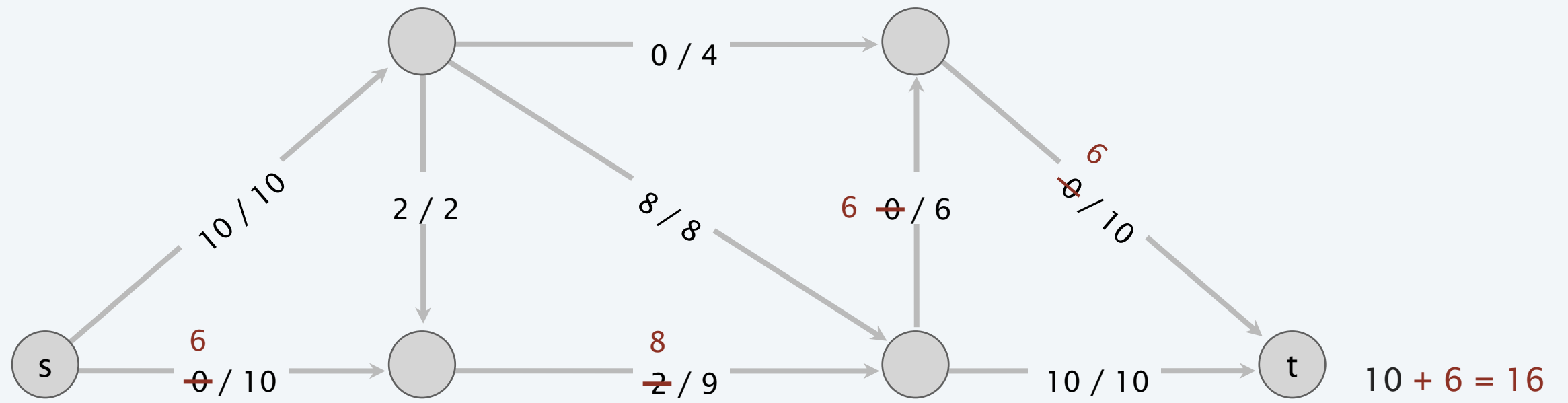


residual network **G_f**

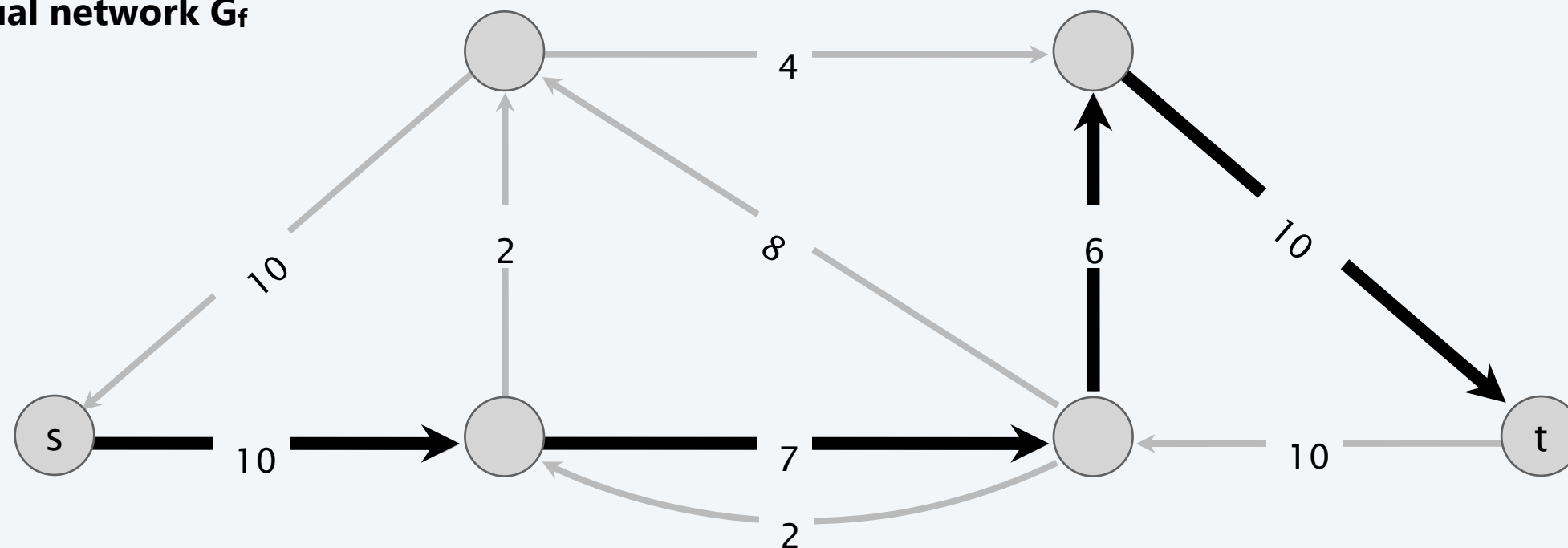


Ford–Fulkerson algorithm demo

network **G** and flow **f**

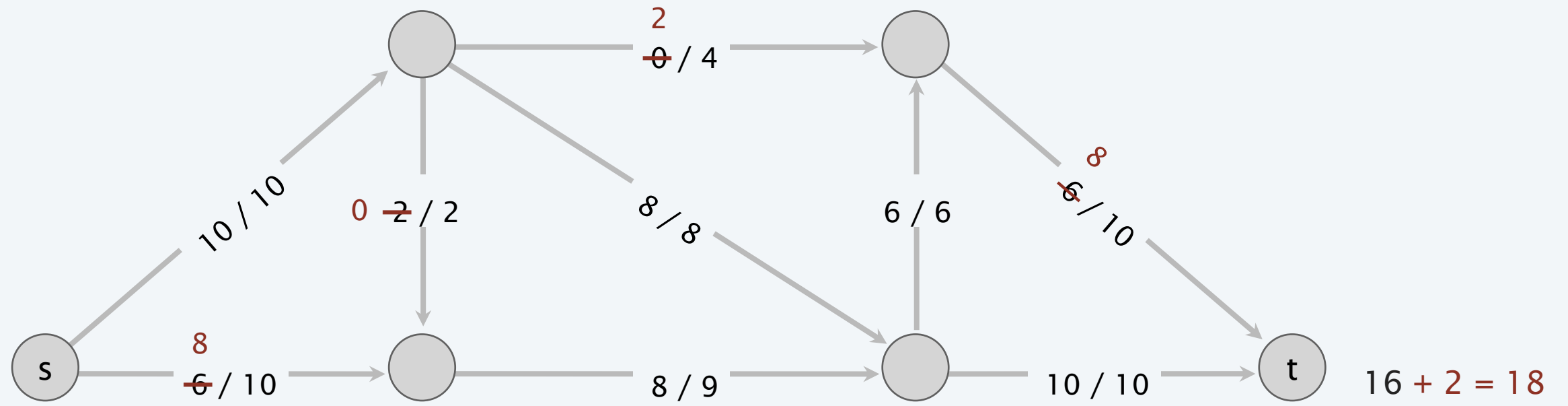


residual network **G_f**

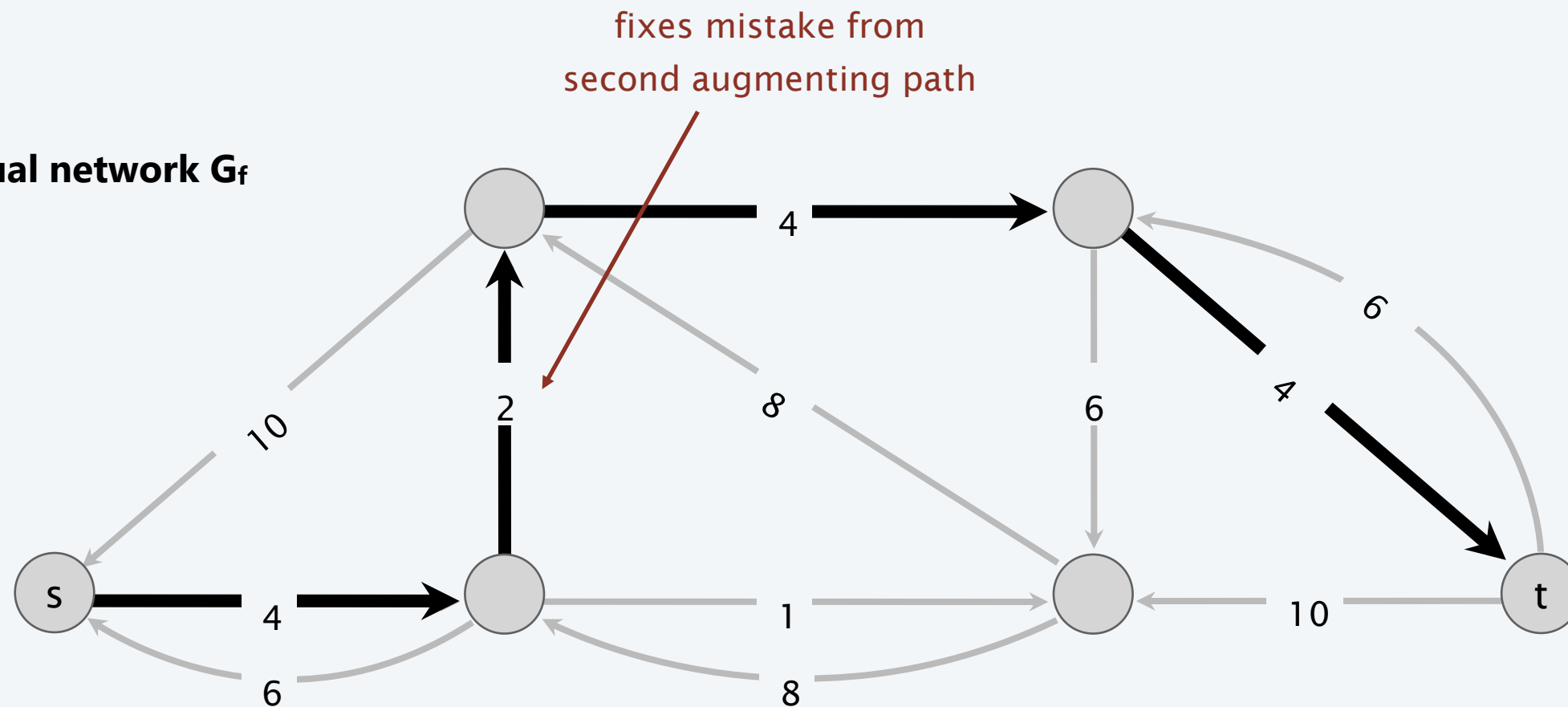


Ford–Fulkerson algorithm demo

network **G** and flow **f**

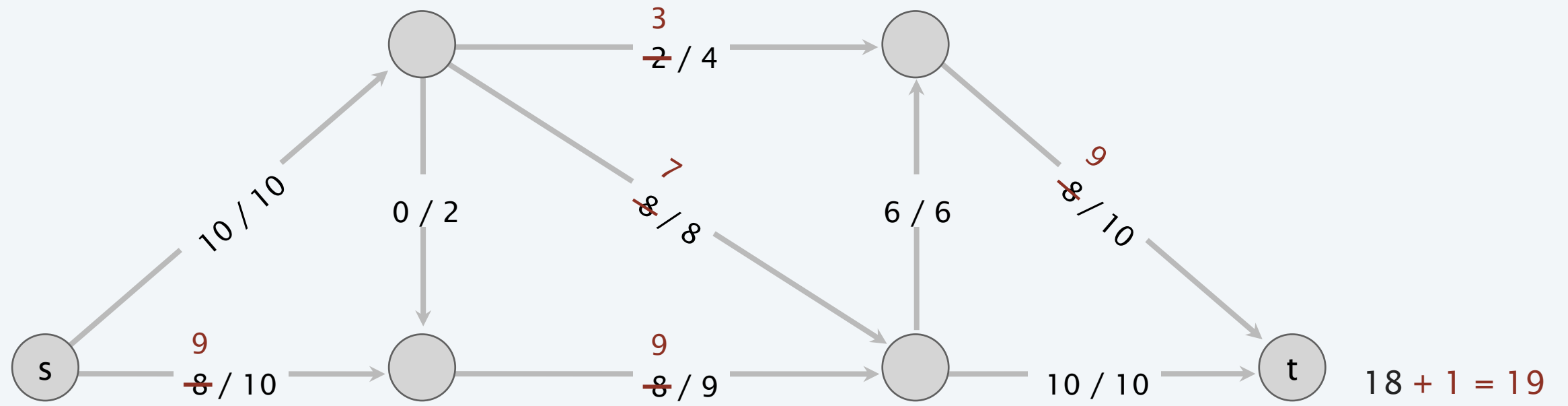


residual network **G_f**

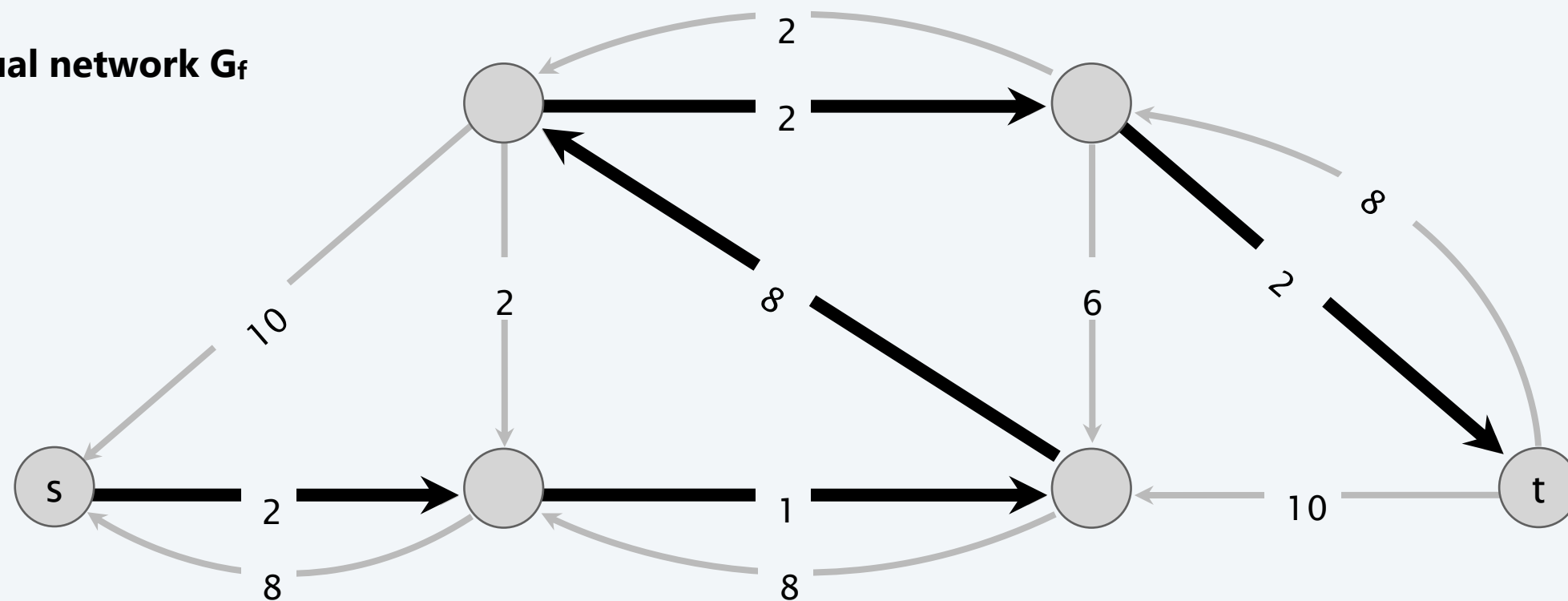


Ford–Fulkerson algorithm demo

network **G** and flow **f**

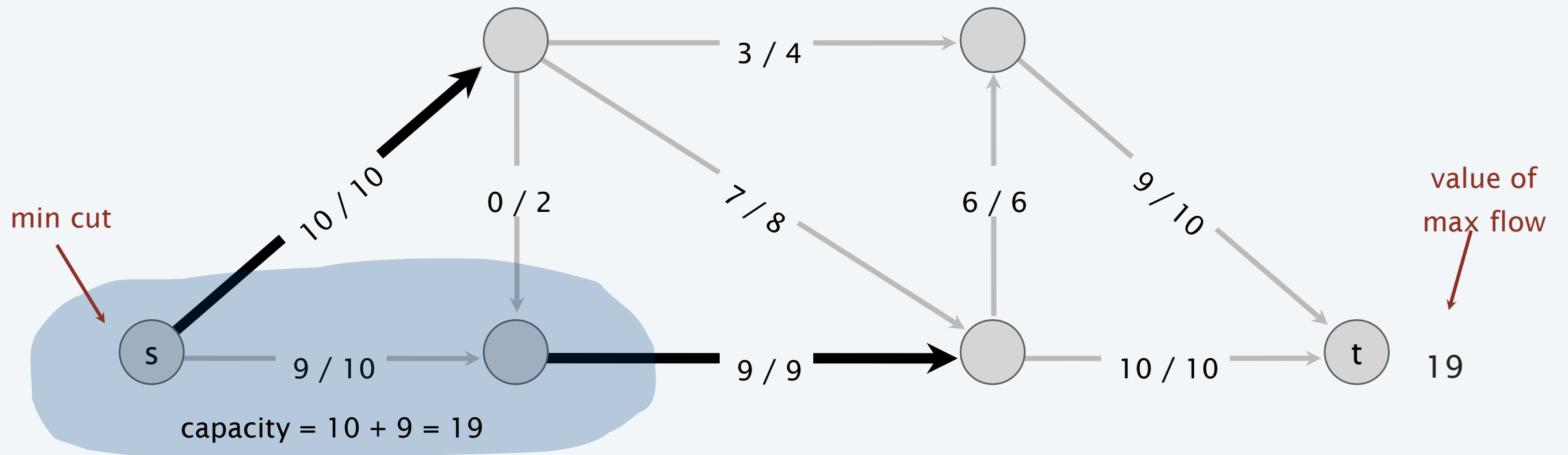


residual network **G_f**



Ford–Fulkerson algorithm demo

network **G** and flow **f**



residual network **G_f**

