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6.4 MAXIMUM FLOW

- ▶ *introduction*
- ▶ *Ford–Fulkerson algorithm*
- ▶ *maxflow–mincut theorem*
- ▶ *analysis of running time*
- ▶ *Java implementation* ← *see textbook*
- ▶ *applications*



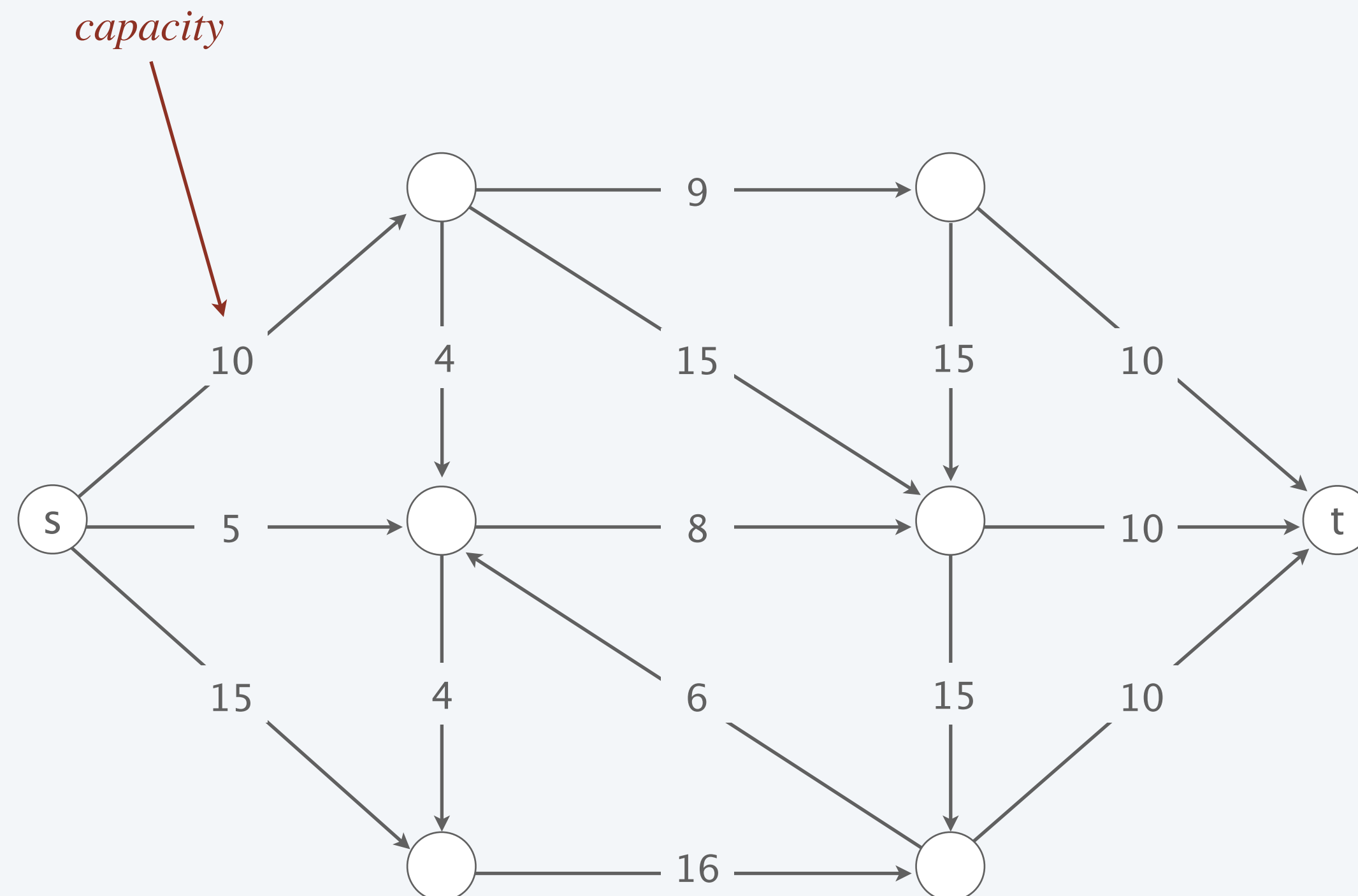
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Mincut problem

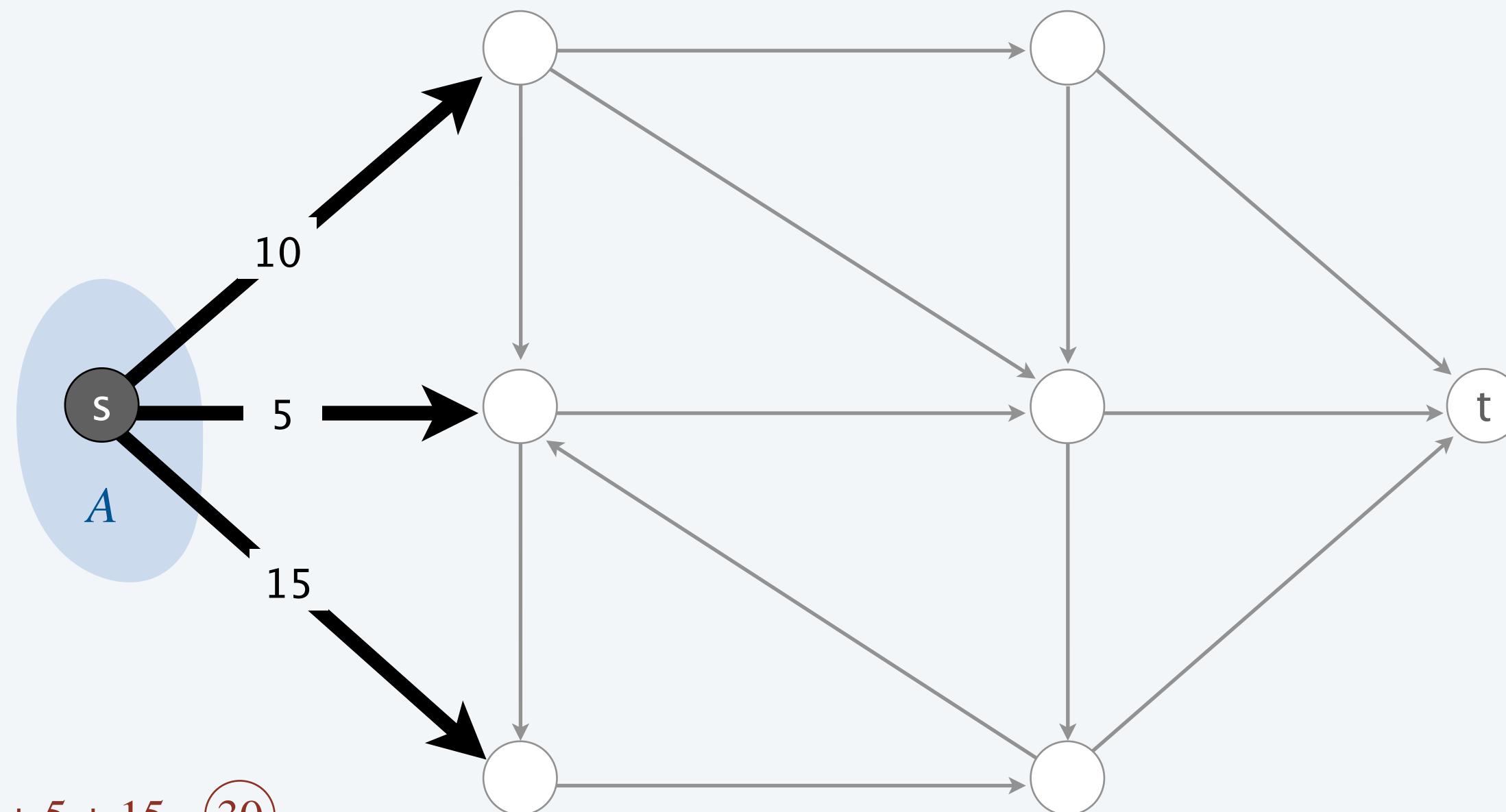
Input. A digraph with positive edge weights (**capacities**), source vertex s , and target vertex t .



Mincut problem

Def. An *st-cut (cut)* is a partition of the vertices into two disjoint sets, with *s* in one set *A* and *t* in the other set *B*.

Def. The *capacity* of a cut (A, B) is the sum of the capacities of the edges from *A* to *B*.

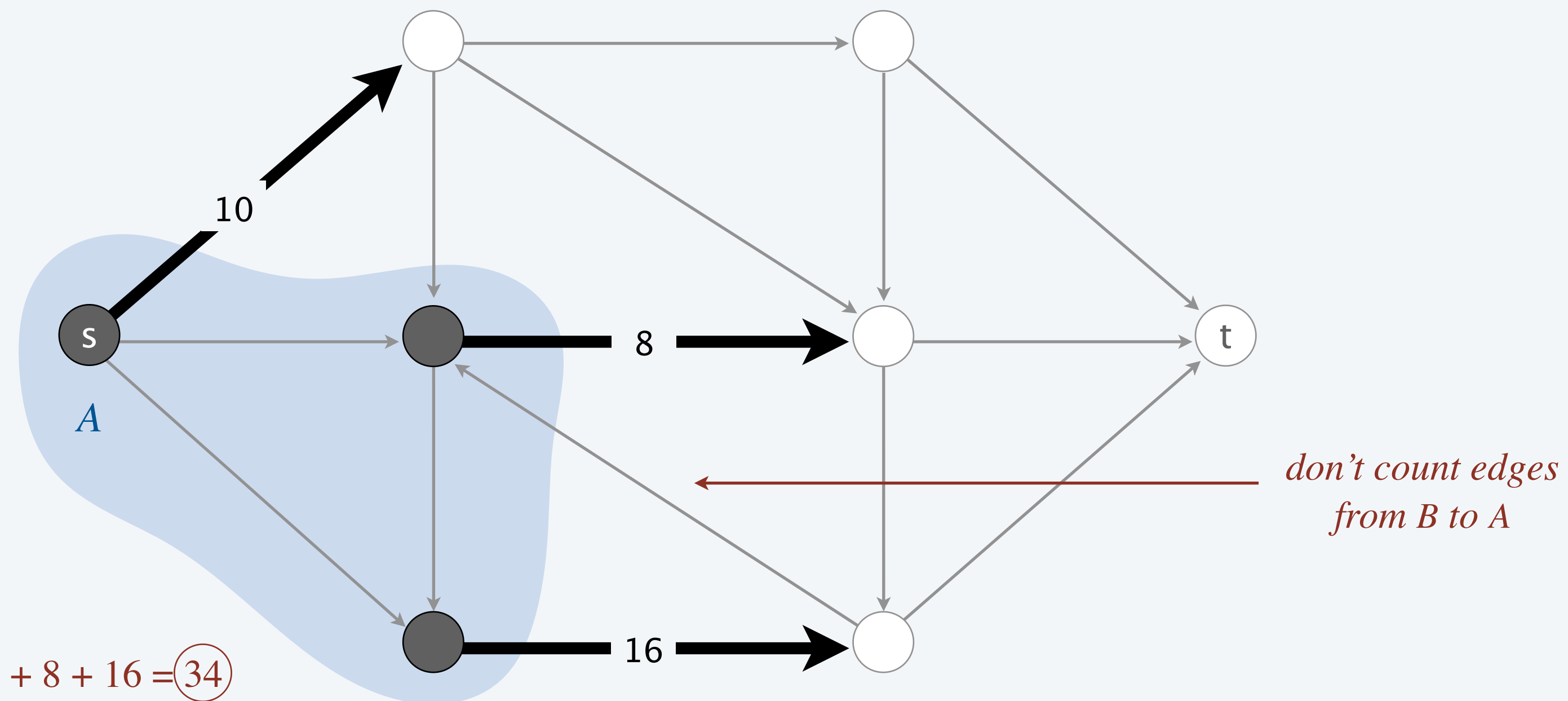


$$\text{capacity} = 10 + 5 + 15 = \textcircled{30}$$

Mincut problem

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Def. The *capacity* of a cut (A, B) is the sum of the capacities of the edges from *A* to *B*.



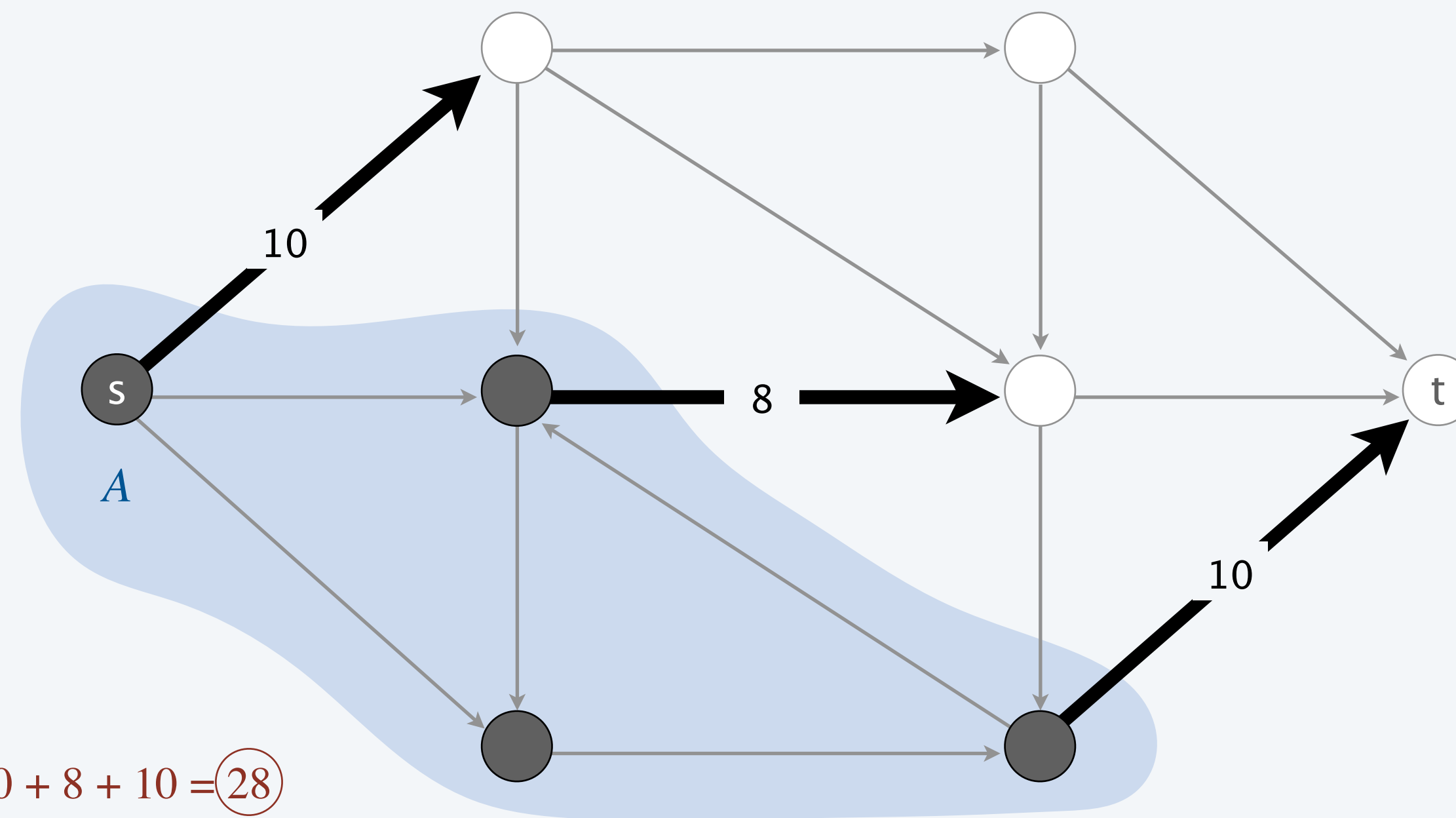
$$\text{capacity} = 10 + 8 + 16 = 34$$

Mincut problem

Def. An *st-cut (cut)* is a partition of the vertices into two disjoint sets, with *s* in one set *A* and *t* in the other set *B*.

Def. The *capacity* of a cut (A, B) is the sum of the capacities of the edges from *A* to *B*.

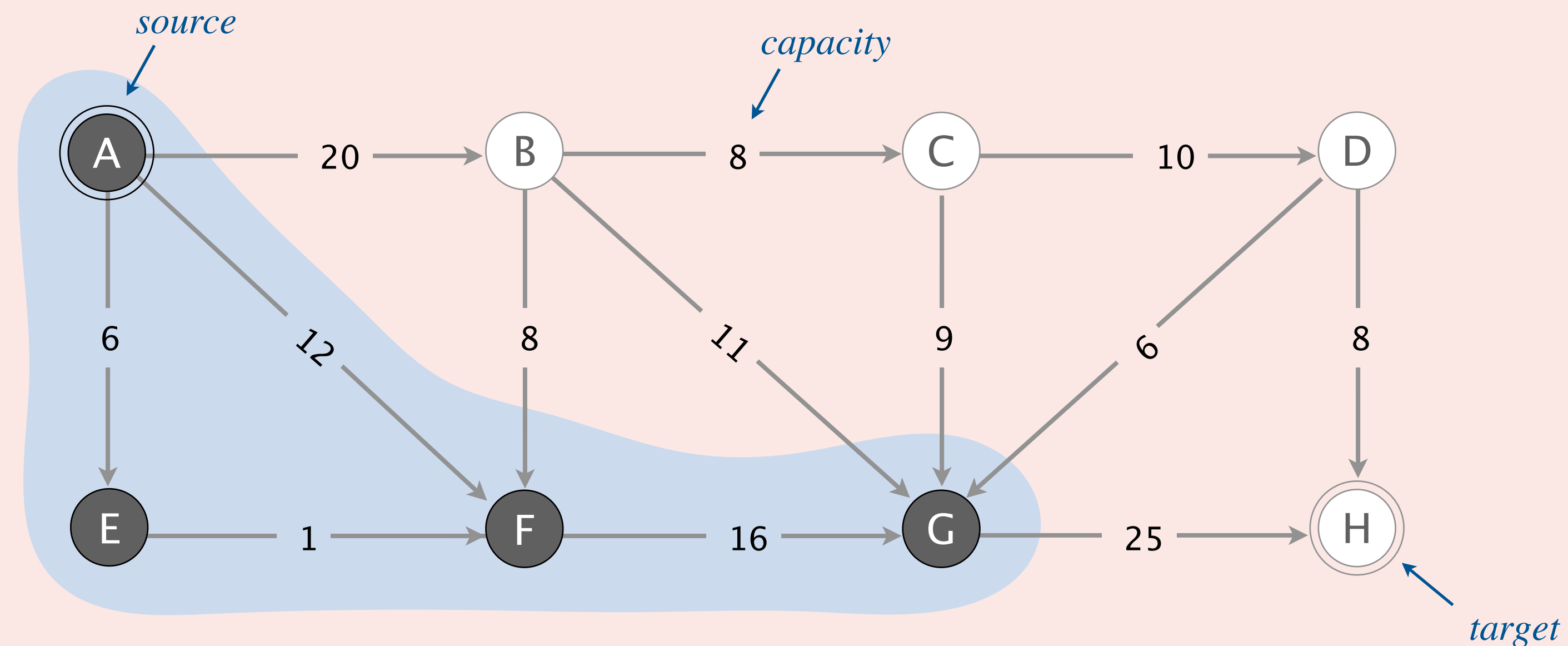
Minimum st-cut (mincut) problem. Find a cut of minimum capacity.





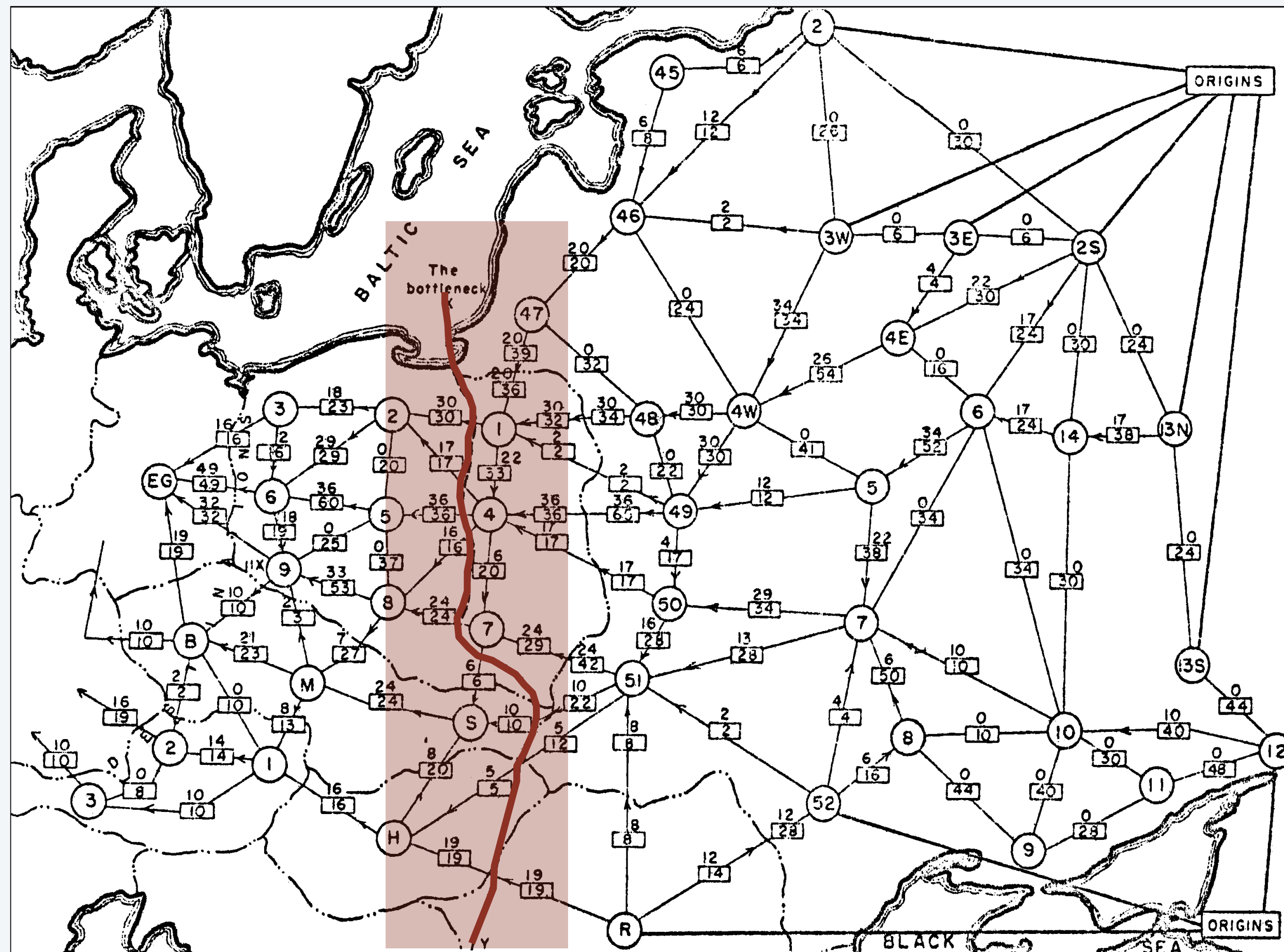
What is the **capacity** of the cut $\{A, E, F, G\}$?

- A. 11 ($20 + 25 - 8 - 11 - 9 - 6$)
- B. 34 ($8 + 11 + 9 + 6$)
- C. 45 ($20 + 25$)
- D. 79 ($20 + 25 + 8 + 11 + 9 + 6$)



Mincut application (RAND 1950s)

“Free world” goal. Disrupt rail network (if Cold War turns into real war).



rail network connecting Soviet Union with Eastern European countries

(map declassified by Pentagon in 1999)



Though maximum flow algorithms have a long history, revolutionary progress is still being made.

BY ANDREW V. GOLDBERG AND ROBERT E. TARJAN

Efficient Maximum Flow Algorithms

gorithms in more detail. We restrict ourselves to basic maximum flow algorithms and do not cover interesting special cases (such as undirected graphs, planar graphs, and bipartite matchings) or generalizations (such as minimum-cost and multi-commodity flow problems).

Before formally defining the maximum flow and the minimum cut problems, we give a simple example of each problem: For the maximum flow example, suppose we have a graph that represents an oil pipeline network from an oil well to an oil depot. Each arc has a capacity, or maximum number of liters per second that can flow through the corresponding pipe. The goal is to find the maximum number of liters per second (maximum flow) that can be shipped from well to depot. For the minimum cut problem, we want to find the set of pipes of the smallest total capacity such that removing the pipes disconnects the oil well from the oil depot (minimum cut).

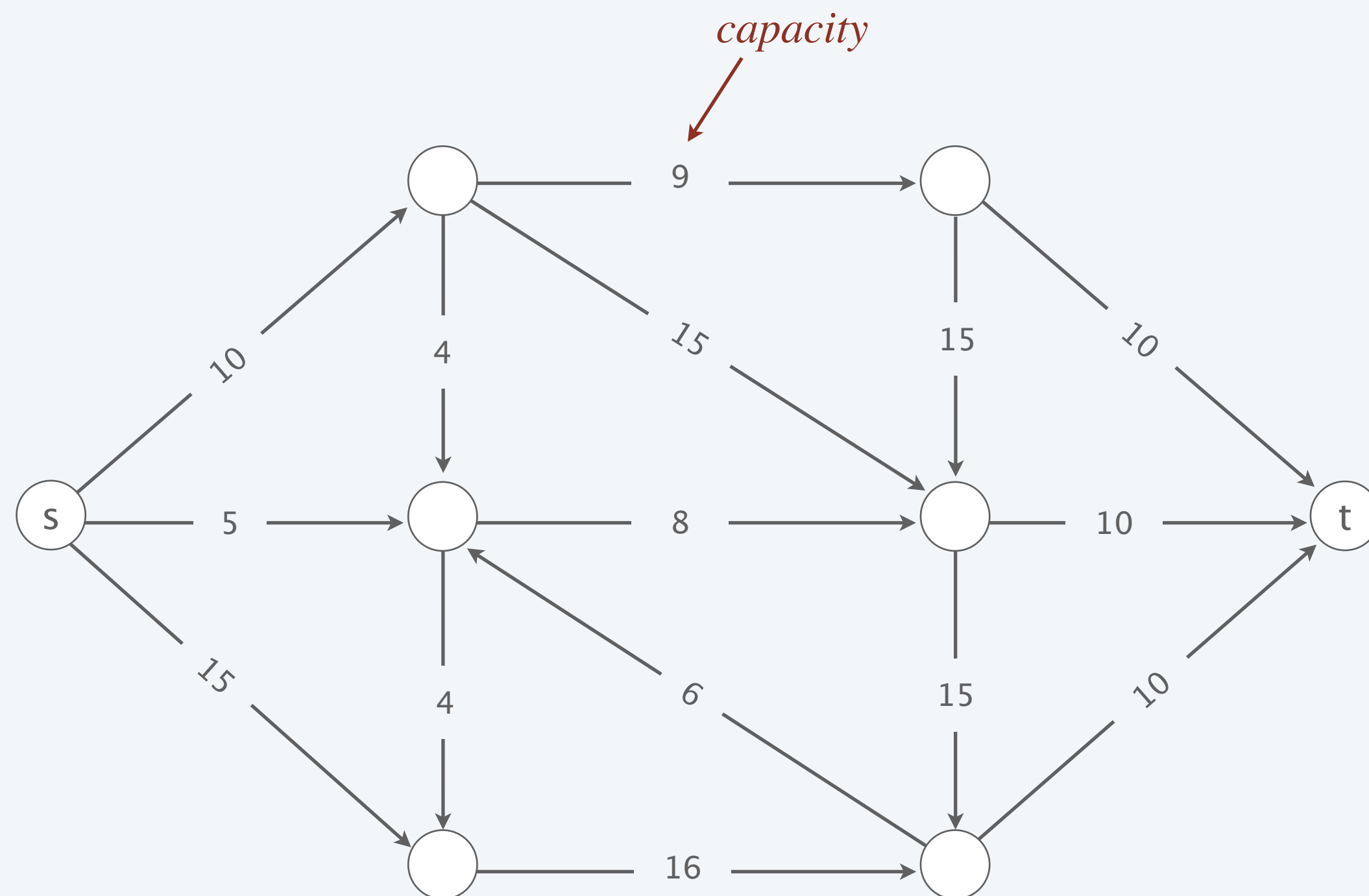
The maximum flow, minimum cut

Efficient Maximum Flow Algorithms by Andrew Goldberg and Bob Tarjan

<https://vimeo.com/100774435>

Maxflow problem

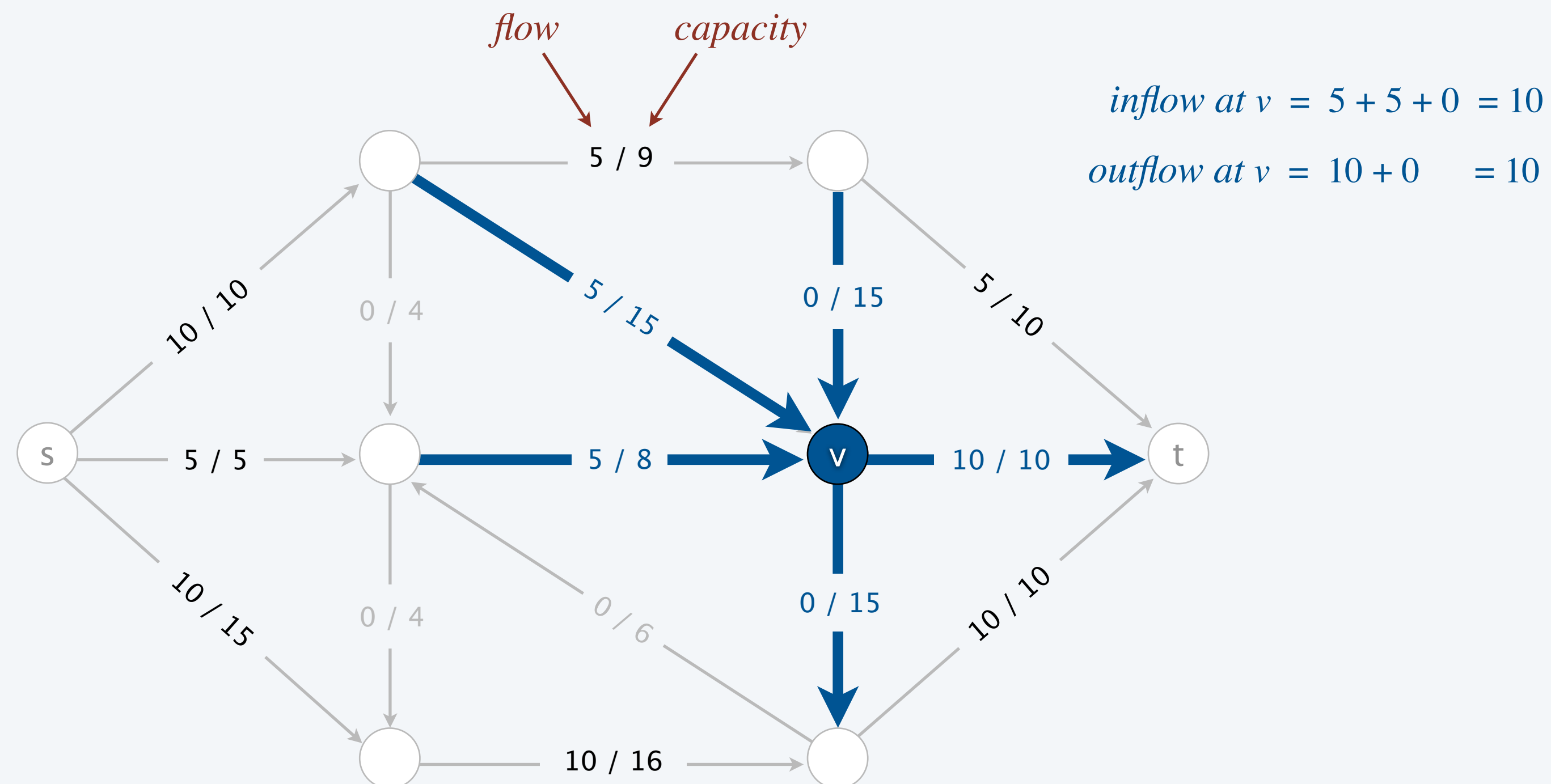
Input. A digraph with positive edge weights (**capacities**), source vertex s , and target vertex t .



Maxflow problem

Def. An *st-flow* (flow) is an assignment of real numbers to the edges such that:

- Capacity constraints: $0 \leq \text{edge's flow} \leq \text{edge's capacity}$.
- Flow conservation constraints: $\text{inflow} = \text{outflow}$ at every vertex (except *s* and *t*).



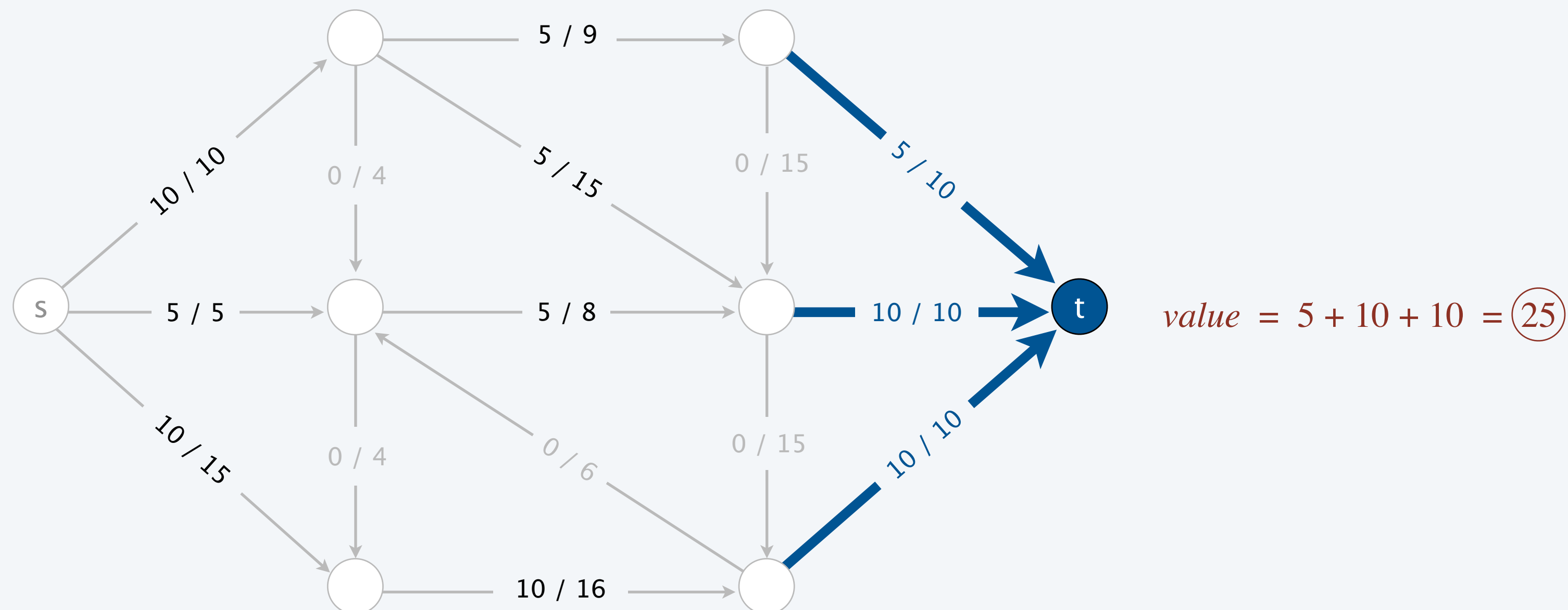
Maxflow problem

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- Flow conservation constraints: $\text{inflow} = \text{outflow}$ at every vertex (except *s* and *t*).

Def. The *value* of a flow is the inflow at *t*.

*we assume no edge leaves *s* or enters *t**



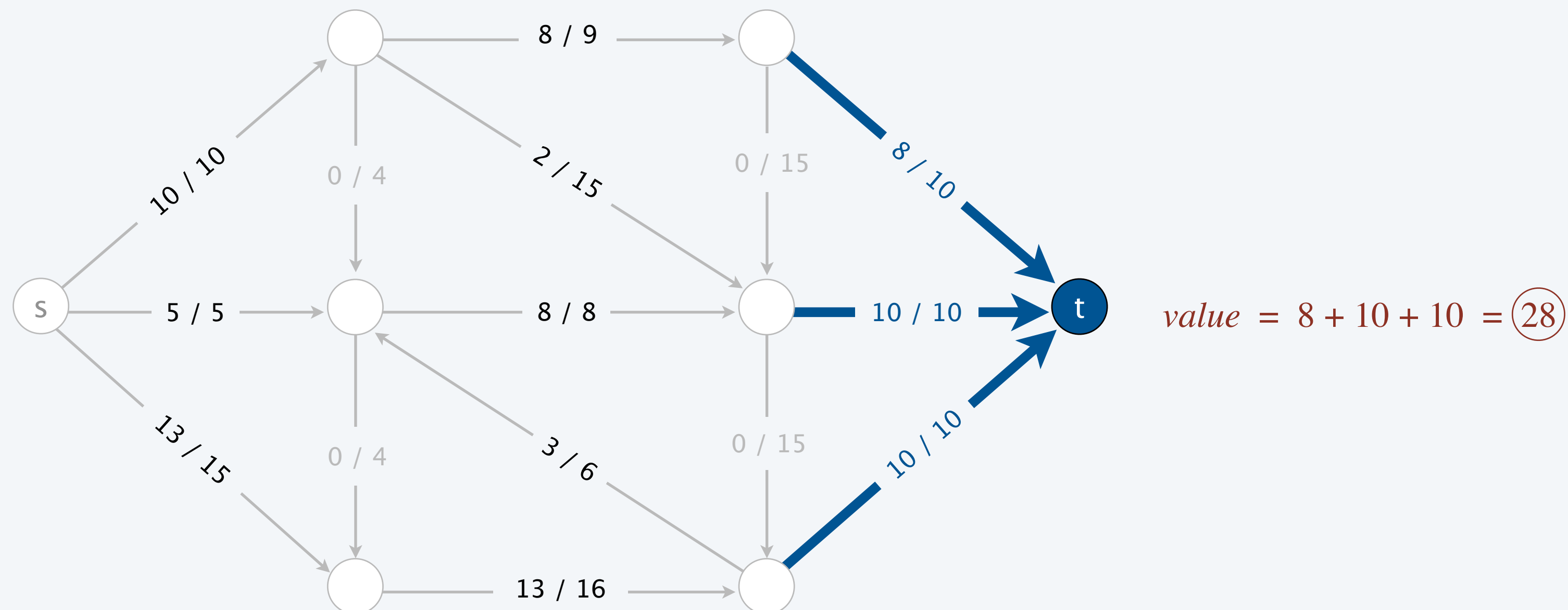
Maxflow problem

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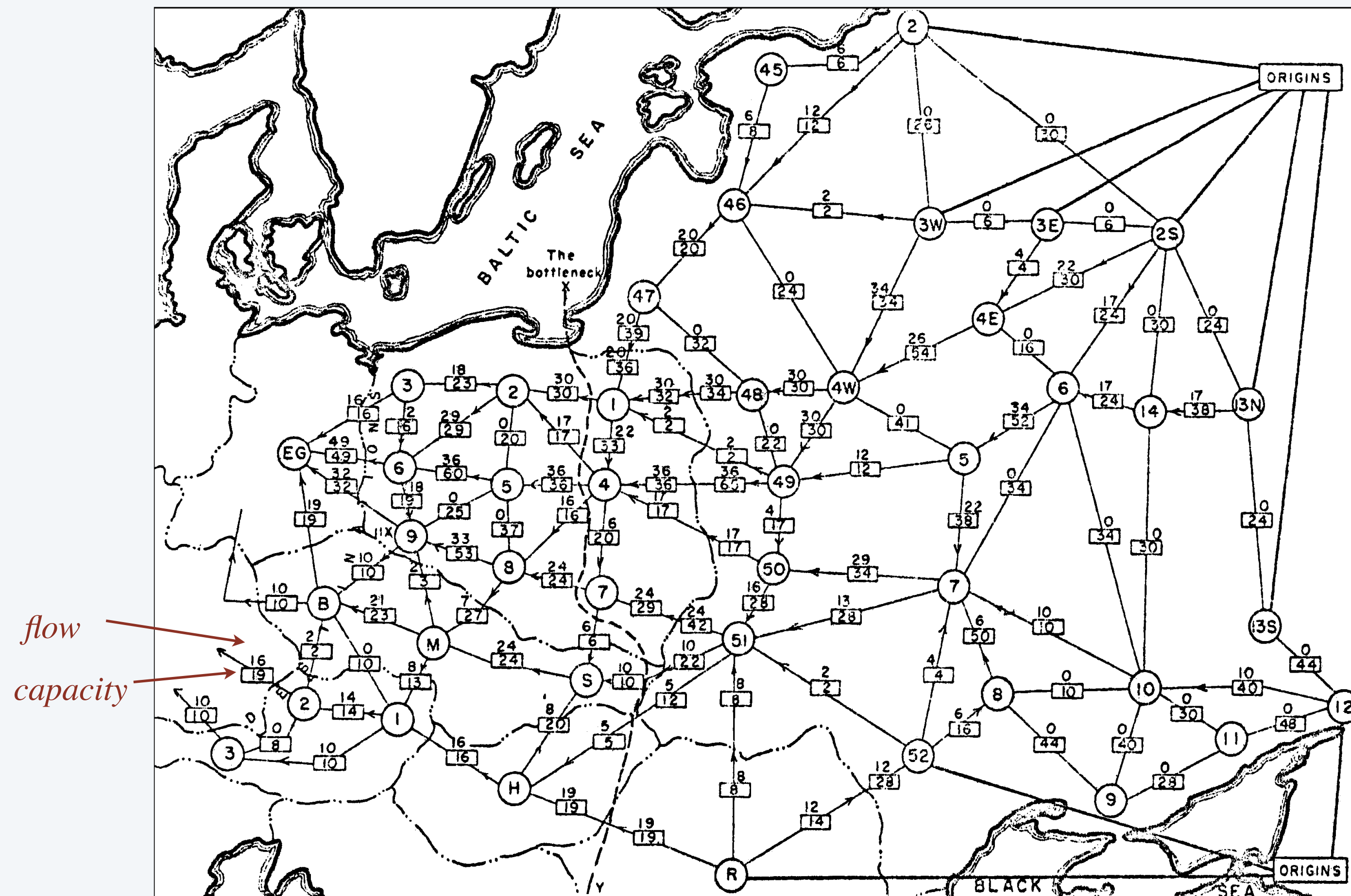
Def. The *value* of a flow is the inflow at *t*.

Maximum st-flow (maxflow) problem. Find a flow of maximum value.



Maxflow application (Tolstoï 1930s)

Soviet Union goal. Maximize flow of supplies to Eastern Europe.



rail network connecting Soviet Union with Eastern European countries

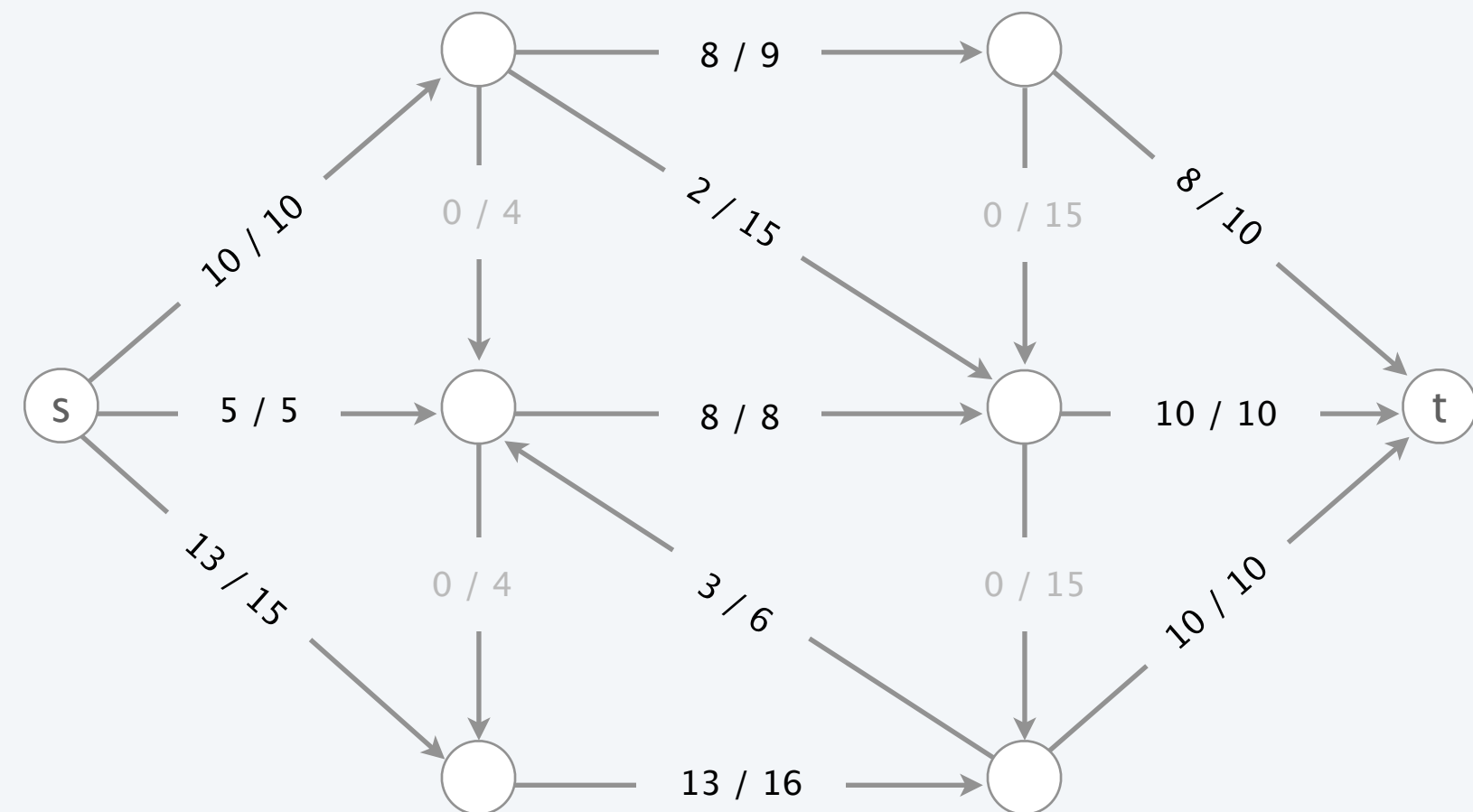
(map declassified by Pentagon in 1999)

Summary

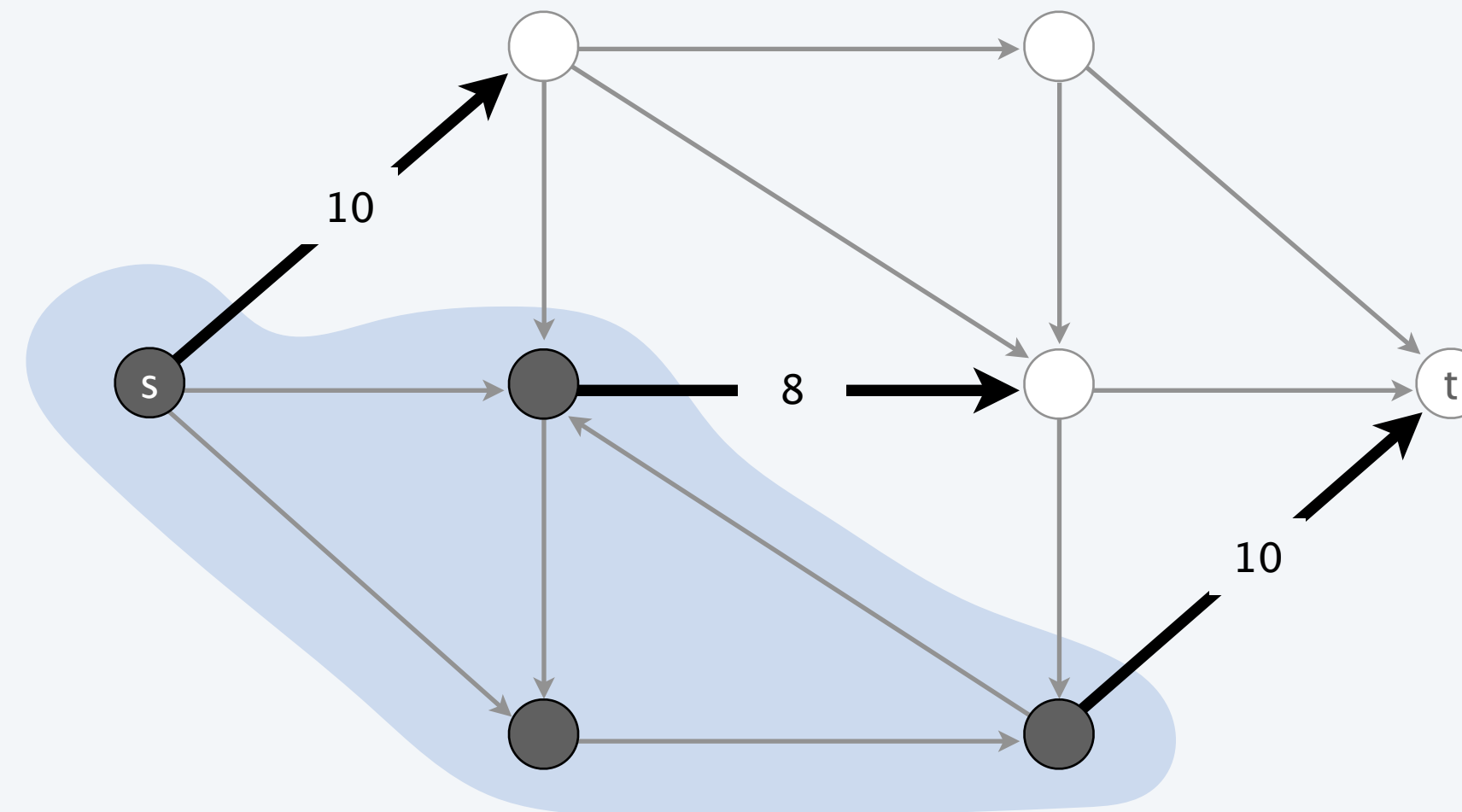
Input. A digraph with positive edge weights, source vertex s , and target vertex t .

Mincut problem. Find a cut of minimum capacity.

Maxflow problem. Find a flow of maximum value.



value of flow = 28



capacity of cut = 28

Remarkable fact. These two problems are dual! [stay tuned]



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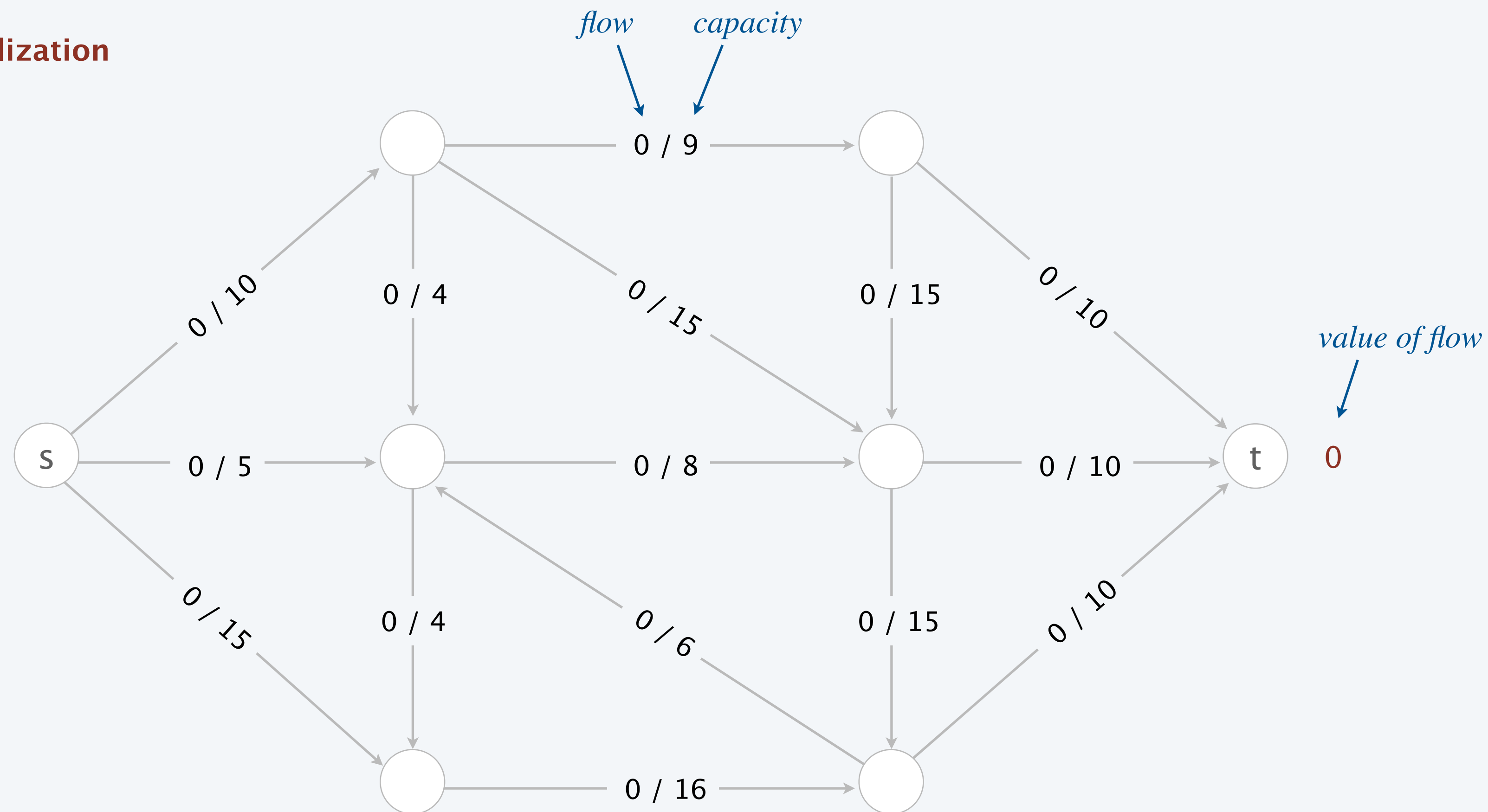
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Ford–Fulkerson algorithm demo

Initialization. Start with 0 flow.

initialization

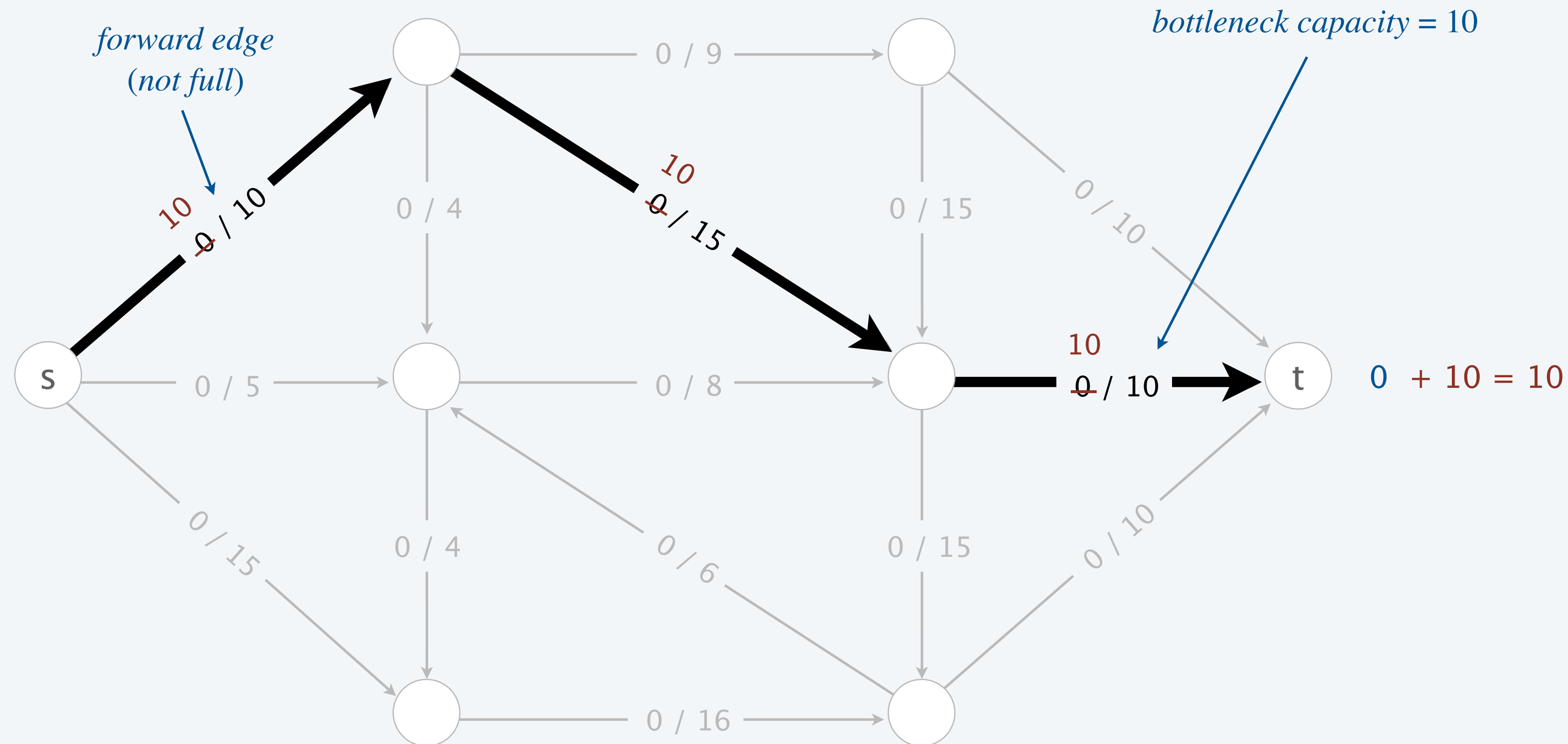


Ford–Fulkerson algorithm demo

Augmenting path. Find an (undirected) path from s to t such that:

- Can increase flow on forward edges (not full). ← *impact: increases value of flow, while maintaining capacity and flow conservation constraints*
- Can decrease flow on backward edge (not empty).

1st augmenting path

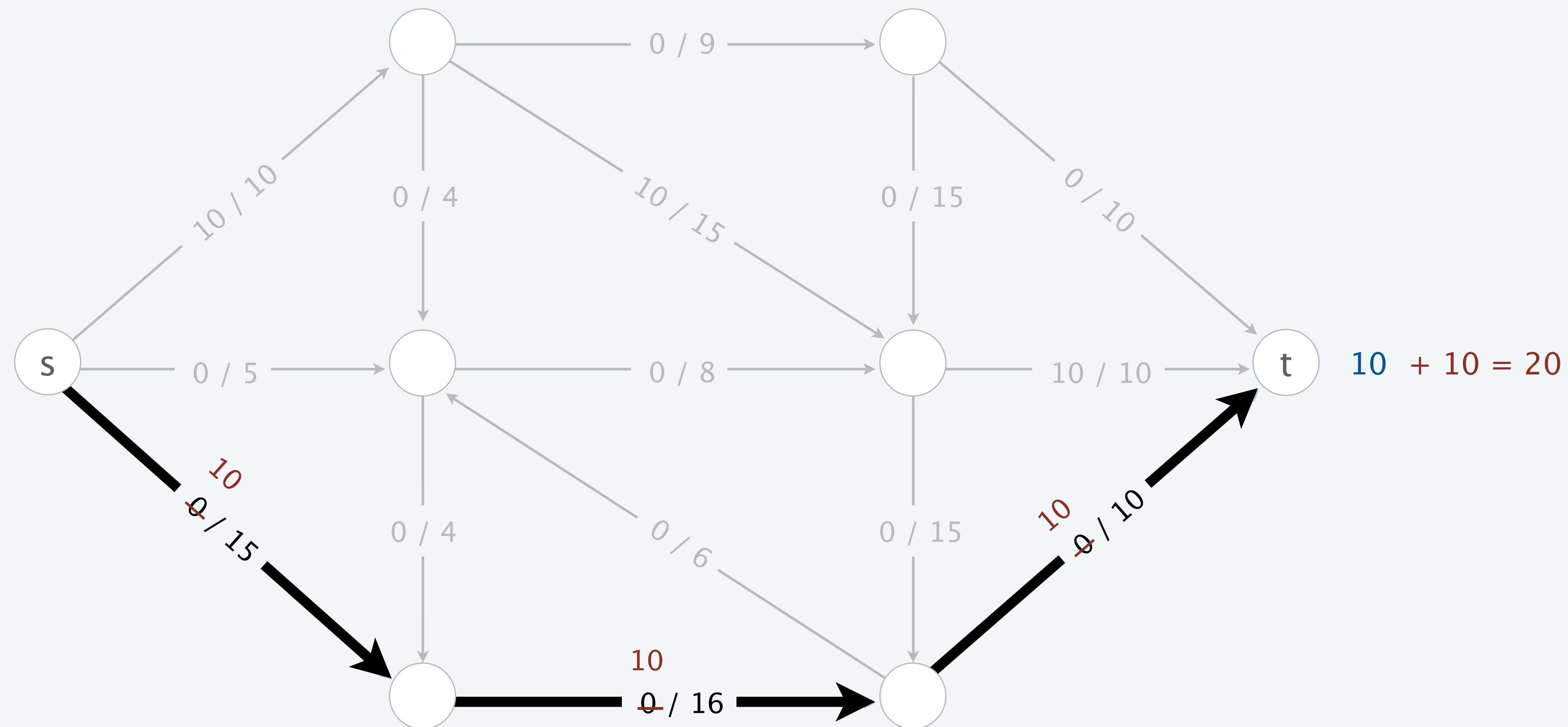


Ford–Fulkerson algorithm demo

Augmenting path. Find an (undirected) path from s to t such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

2nd augmenting path



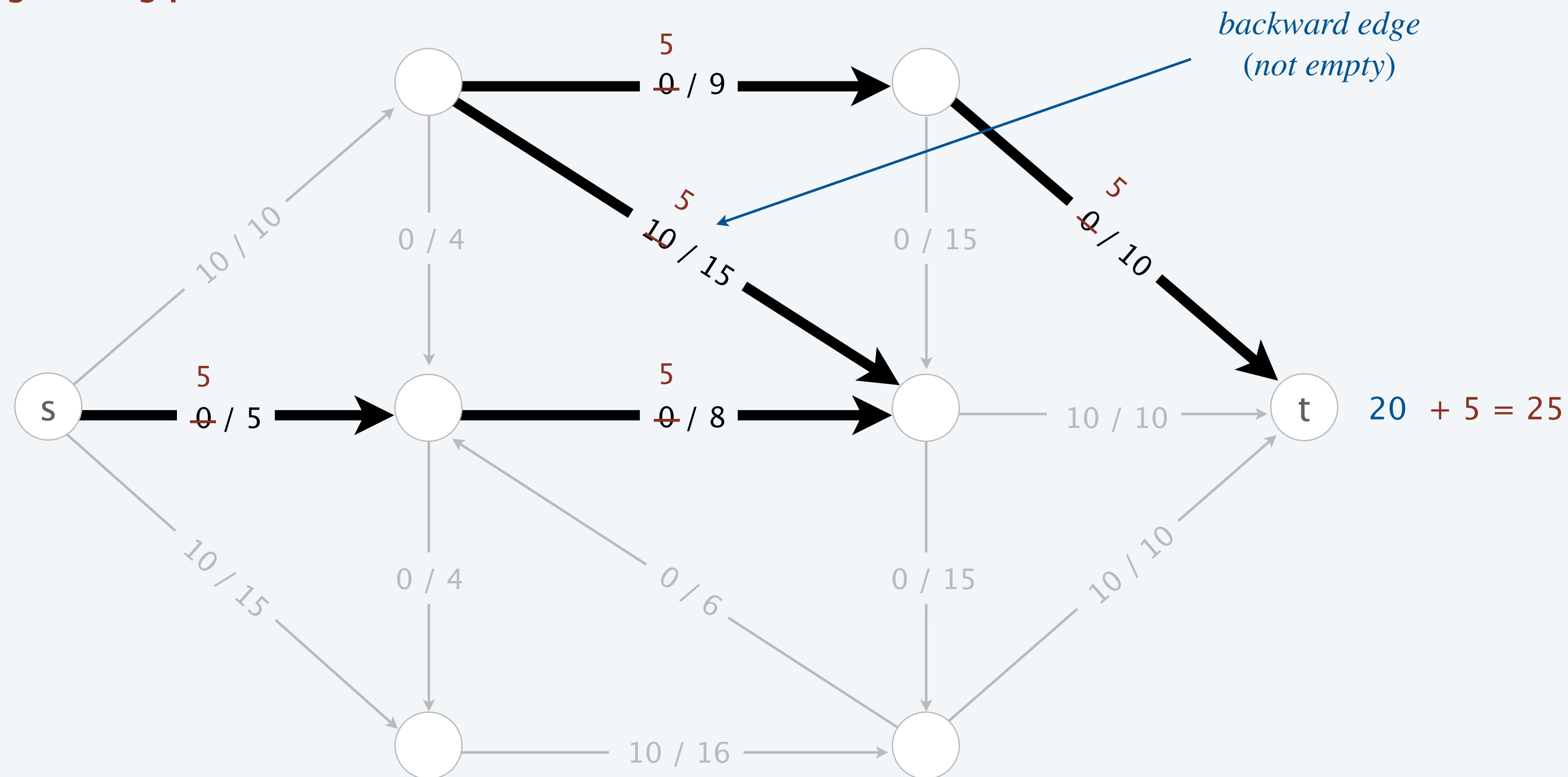
Ford–Fulkerson algorithm demo

Augmenting path. Find an (undirected) path from s to t such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

← *impact: increases value of flow, while maintaining capacity and flow conservation constraints*

3rd augmenting path

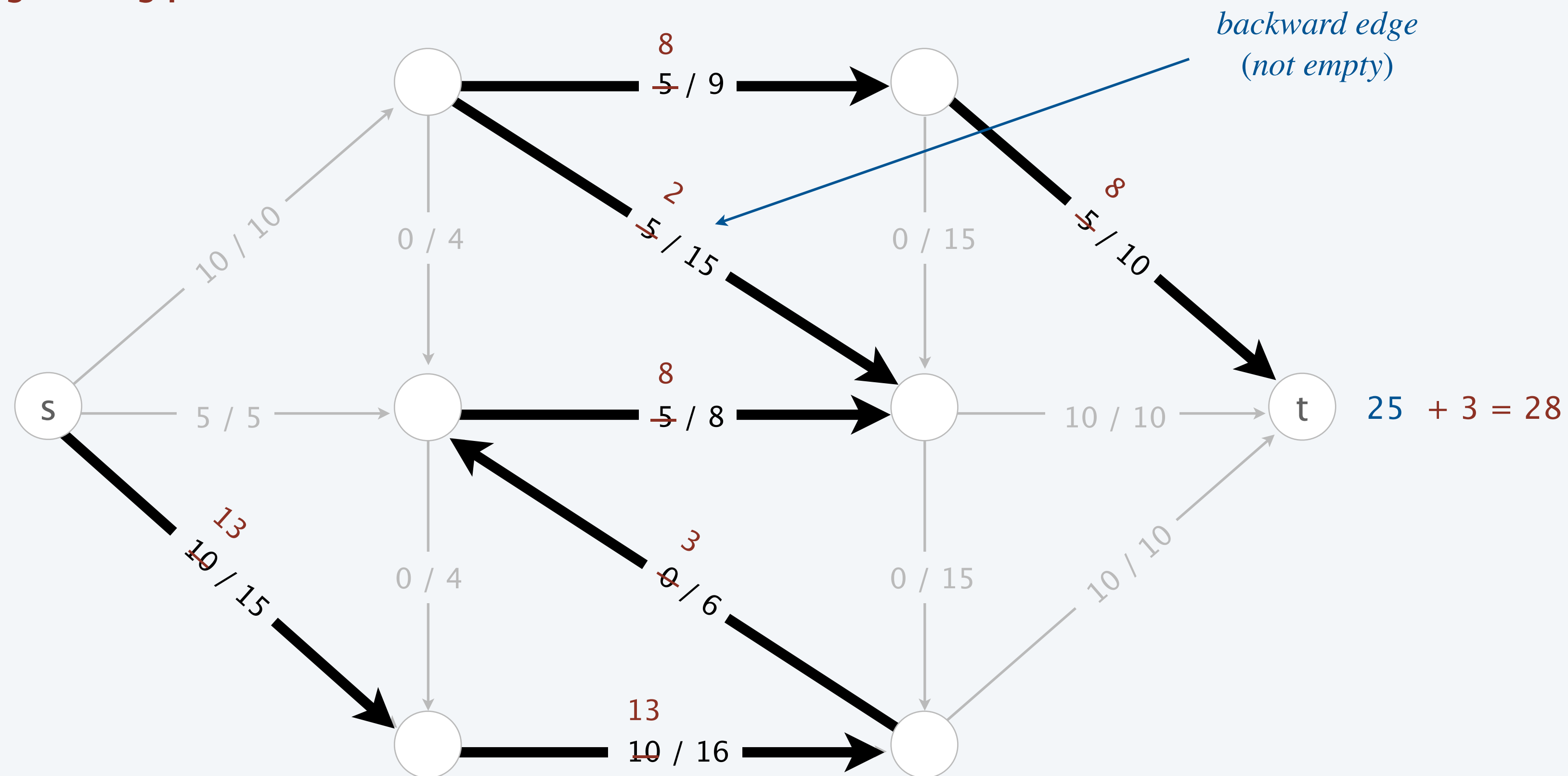


Ford–Fulkerson algorithm demo

Augmenting path. Find an (undirected) path from s to t such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

4th augmenting path

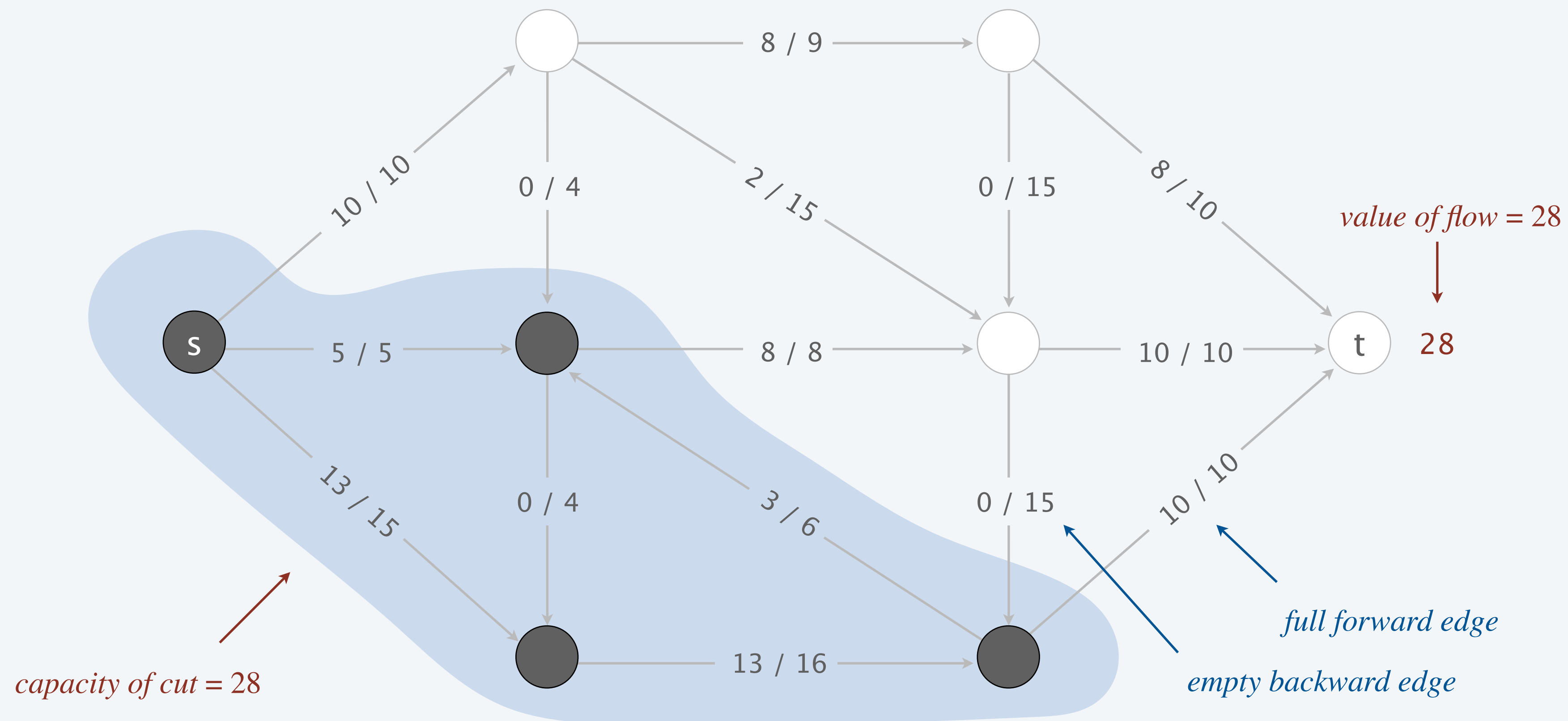


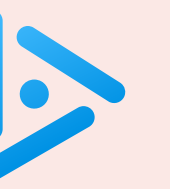
Ford–Fulkerson algorithm demo

Termination. All paths from s to t are blocked by either

- a full forward edge, or
- an empty backward edge.

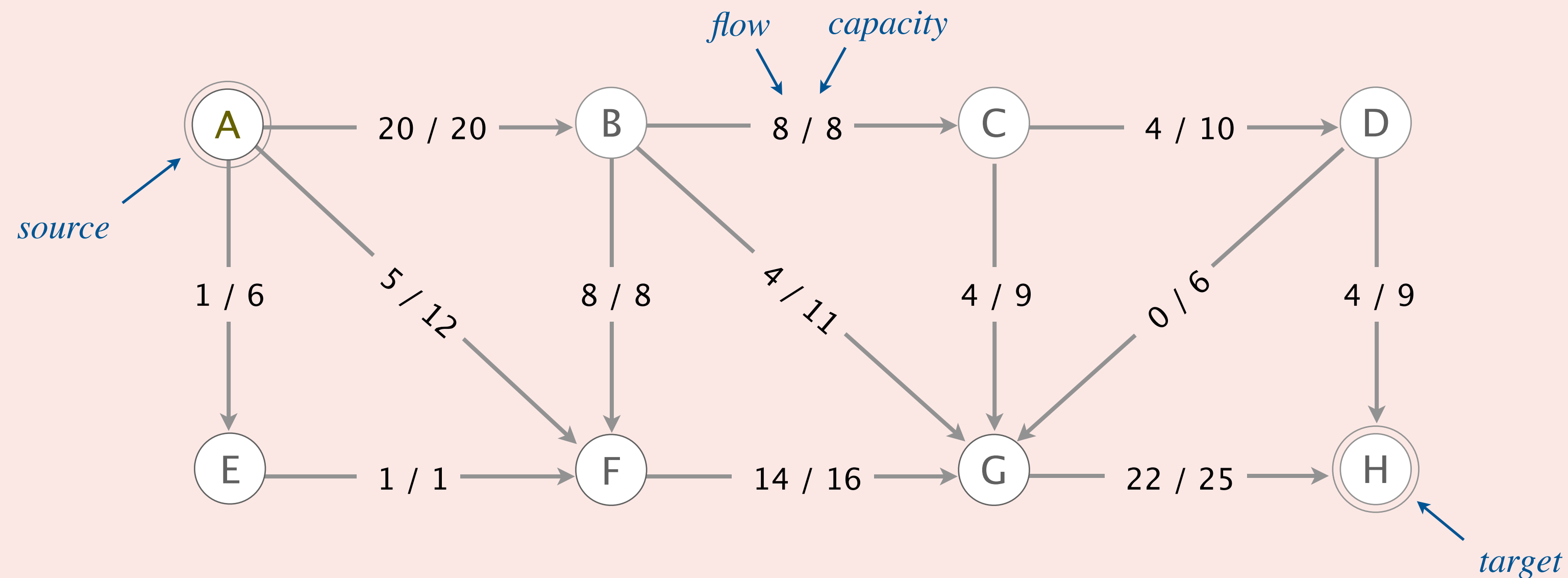
no more augmenting paths

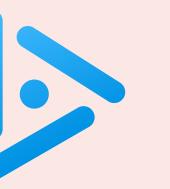




Which is an augmenting path with respect to the given flow?

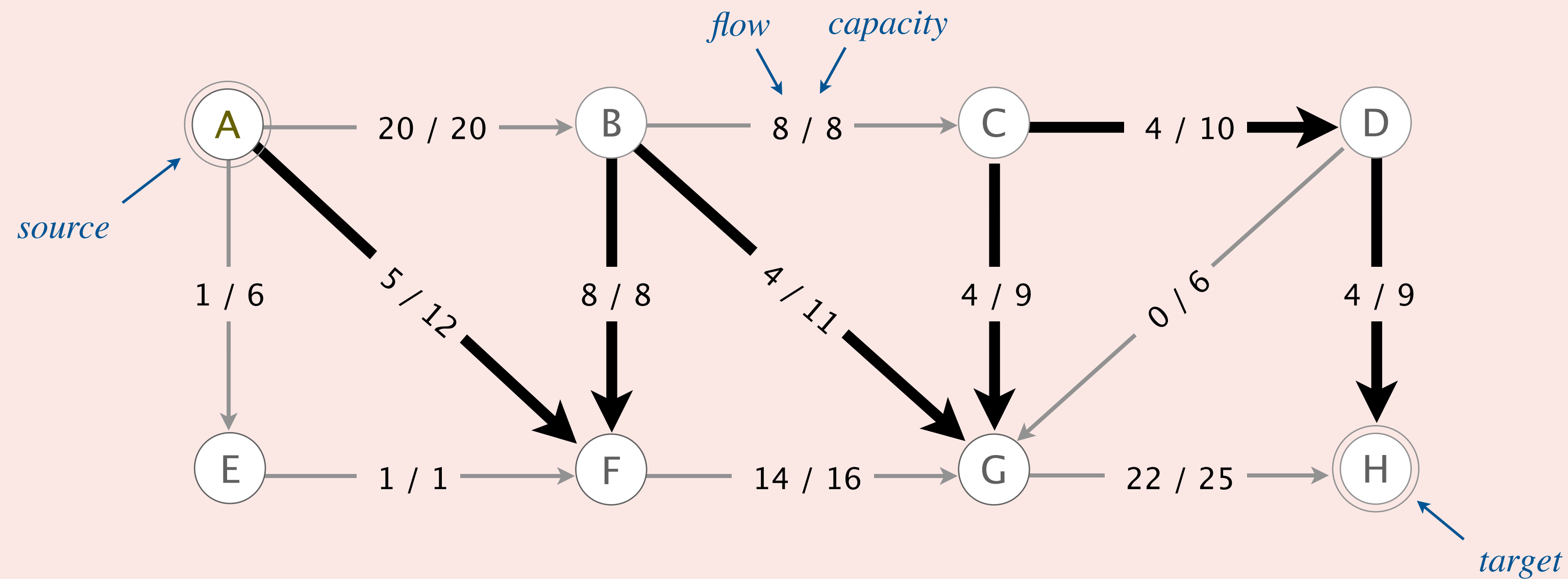
- A. $A \rightarrow F \rightarrow G \rightarrow D \rightarrow H$
- B. $A \rightarrow F \rightarrow B \rightarrow G \rightarrow C \rightarrow D \rightarrow H$
- C. Both A and B.
- D. Neither A nor B.





What is the **bottleneck capacity** of the augmenting path $A \rightarrow F \rightarrow B \rightarrow G \rightarrow C \rightarrow D \rightarrow H$?

- A. 4
- B. 5
- C. 6
- D. 7



Ford–Fulkerson algorithm

Ford–Fulkerson algorithm

Initialize flow $f = 0$.

While there exists an augmenting path:

- find an augmenting path P
 - compute bottleneck capacity of P
 - update flow f on P by bottleneck capacity
-

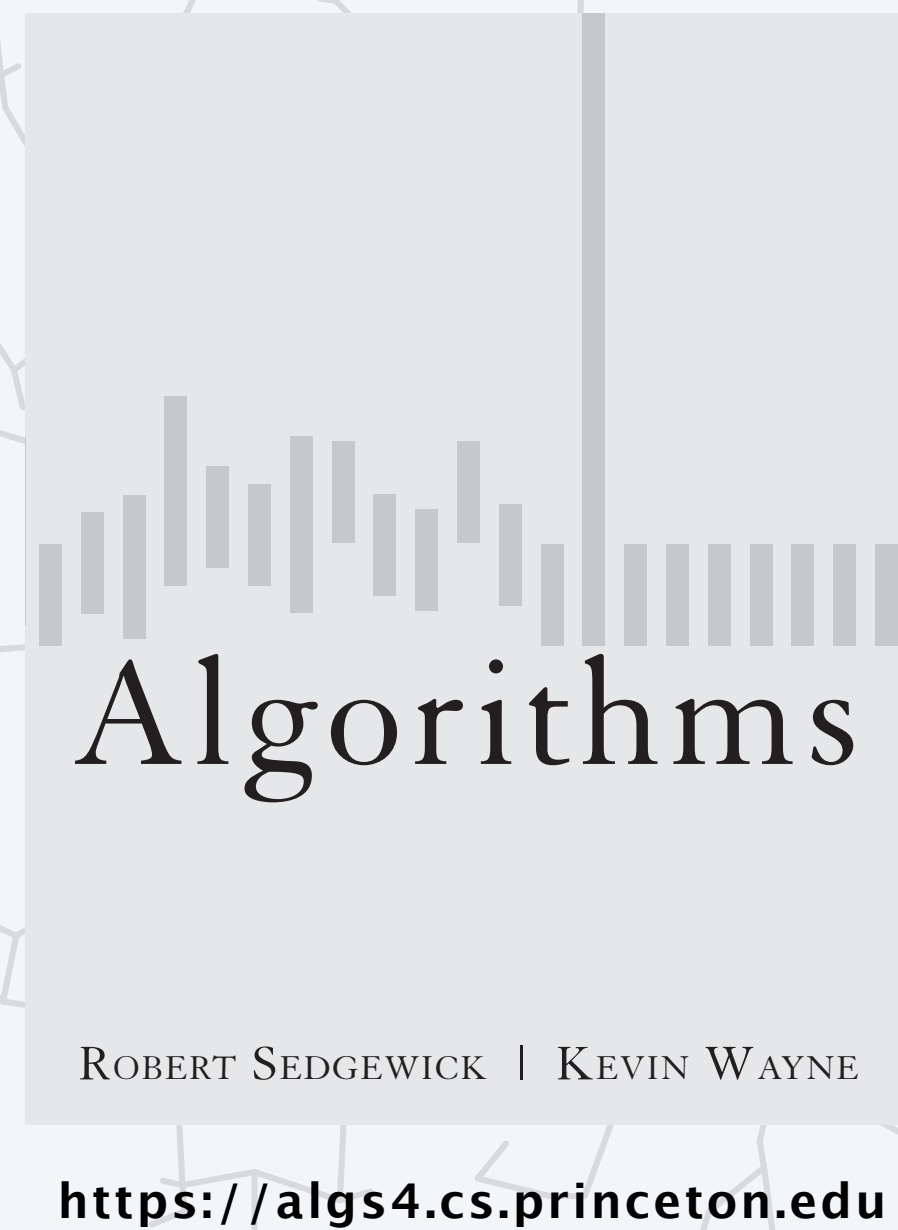
Fundamental questions.

Q1. How to find an augmenting path?

Q2. How many augmenting paths?

Q3. Guaranteed to compute a maxflow?

Q4. How to compute a mincut?

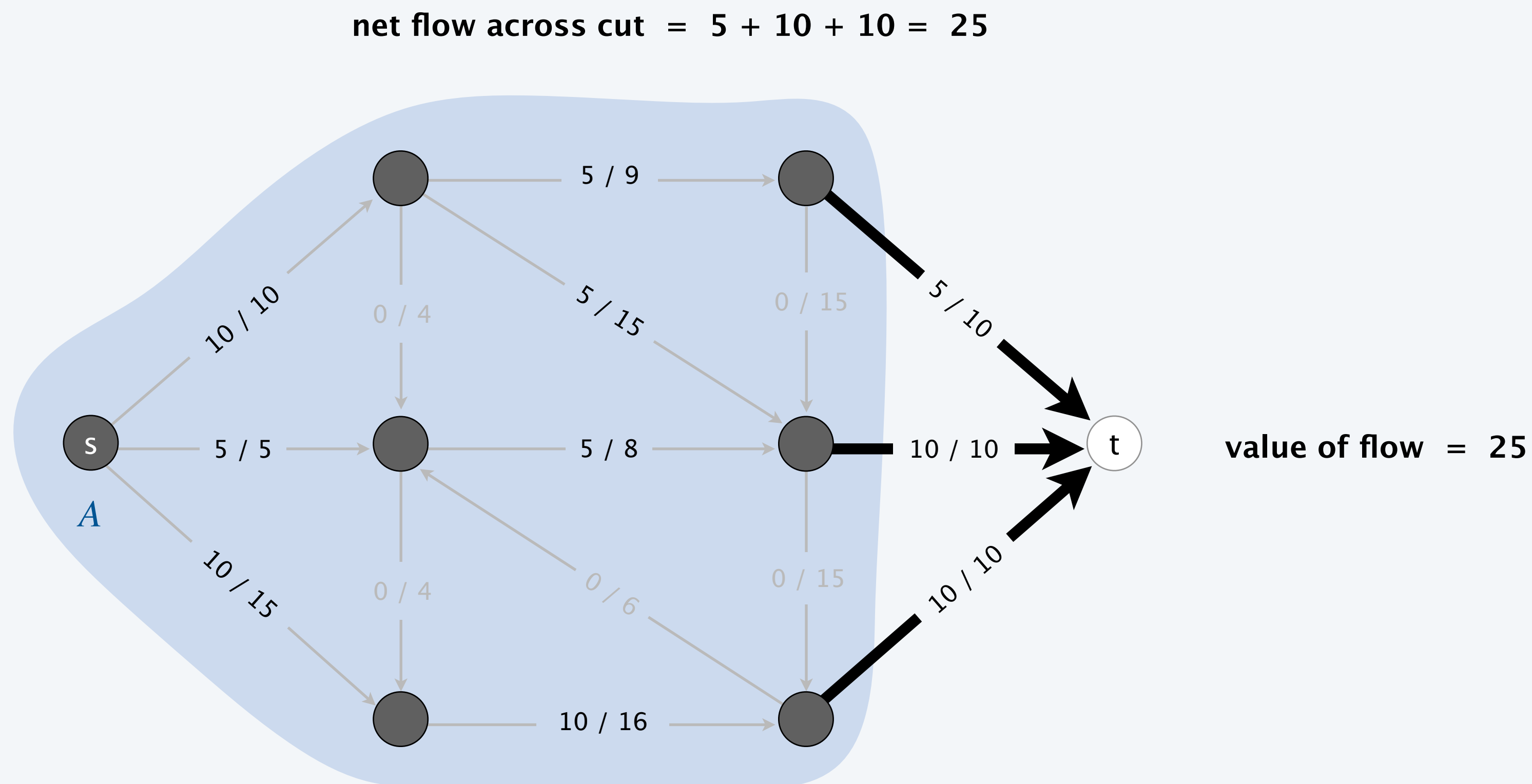


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Relationship between flows and cuts

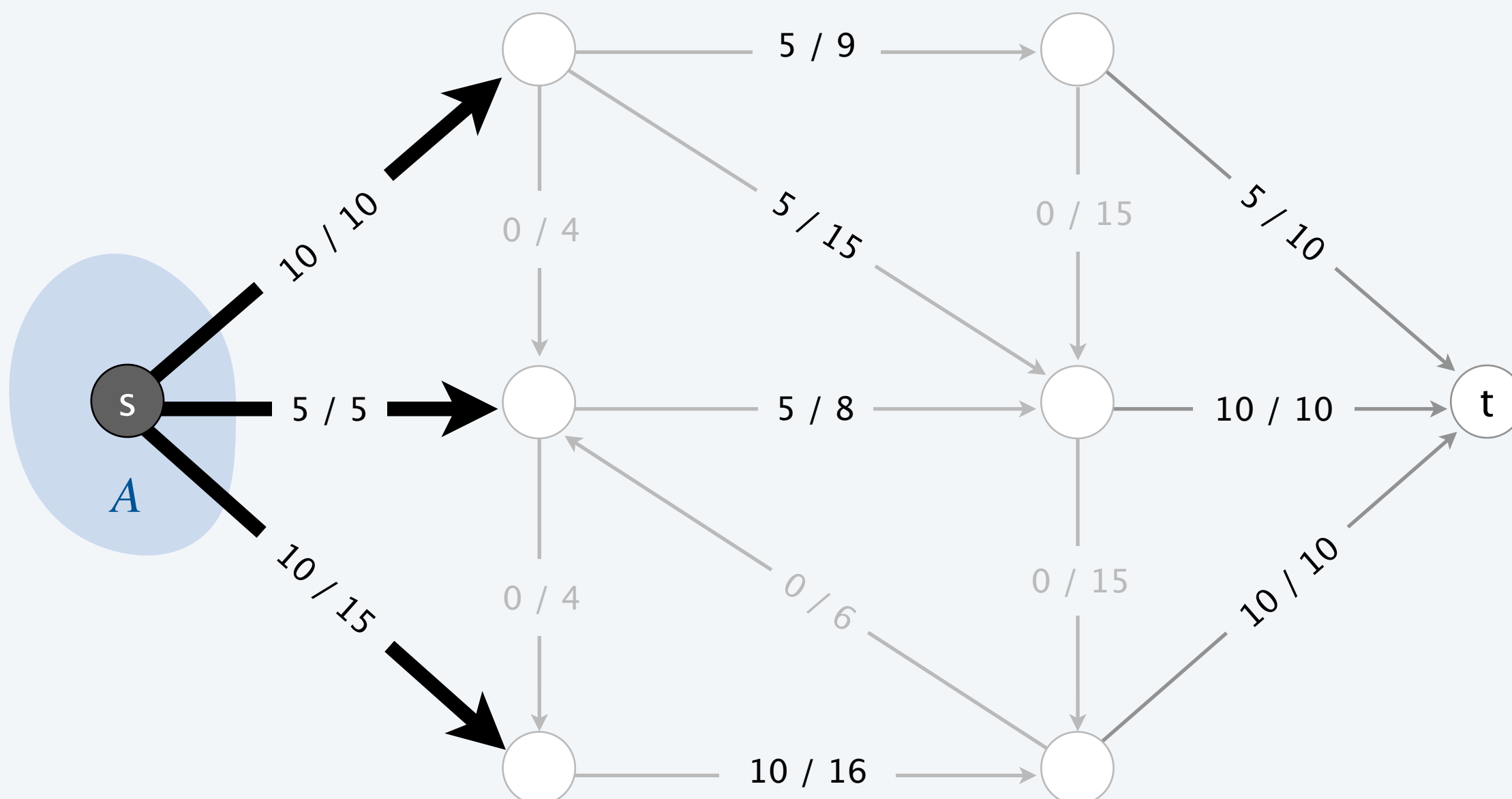
Def. Given a flow f and a cut (A, B) , the **net flow across** the cut is the sum of flows on edges from A to B , minus the sum of flows on edges from B to A .



Relationship between flows and cuts

Def. Given a flow f and a cut (A, B) , the **net flow across** the cut is the sum of flows on edges from A to B , minus the sum of flows on edges from B to A .

$$\text{net flow across cut} = 10 + 5 + 10 = 25$$

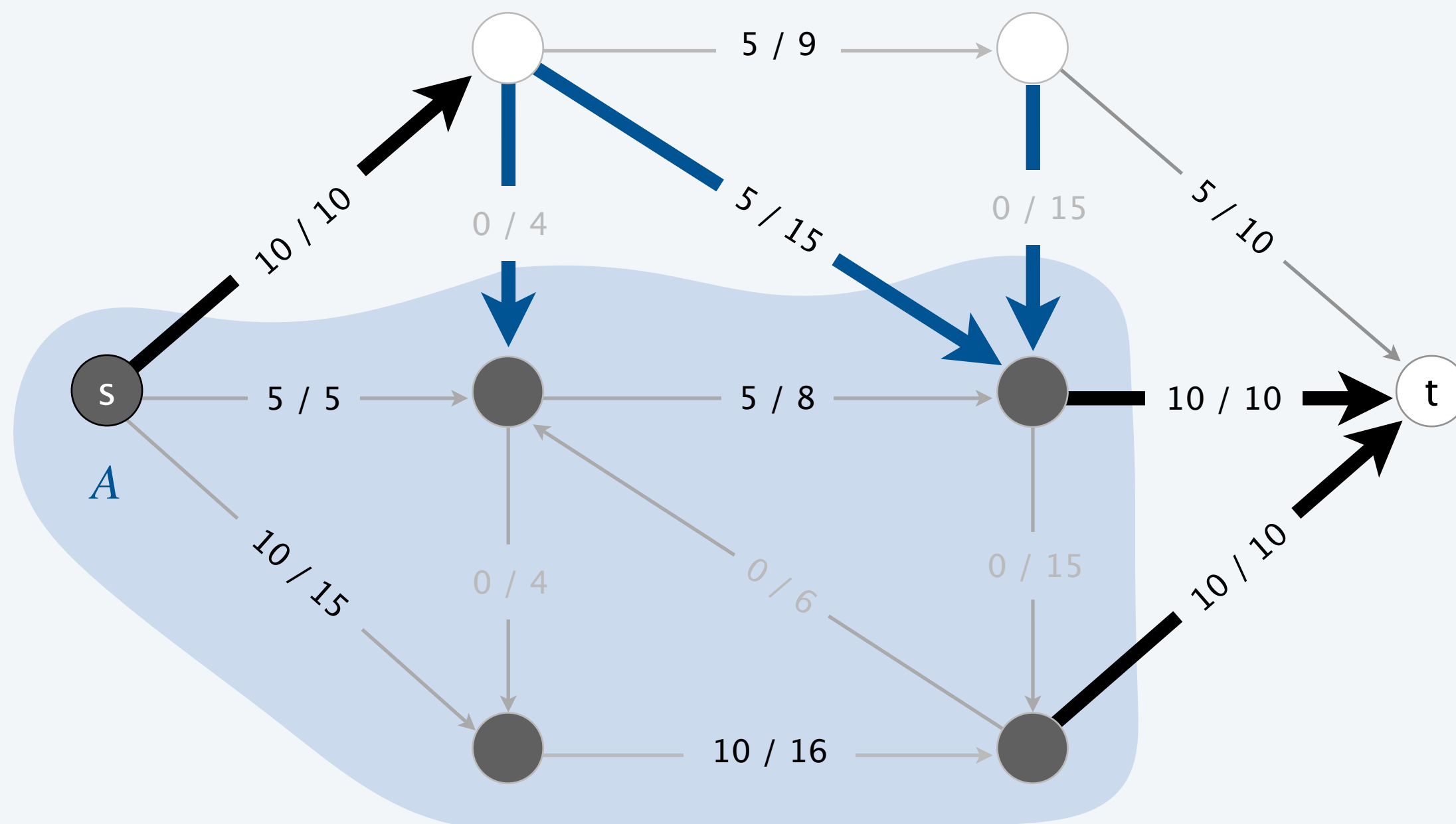


$$\text{value of flow} = 25$$

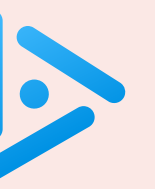
Relationship between flows and cuts

Def. Given a flow f and a cut (A, B) , the **net flow across** the cut is the sum of flows on edges from A to B , minus the sum of flows on edges from B to A .

$$\text{net flow across cut} = (10 + 10 + 10) - (0 + 5 + 0) = 25$$



value of flow = 25



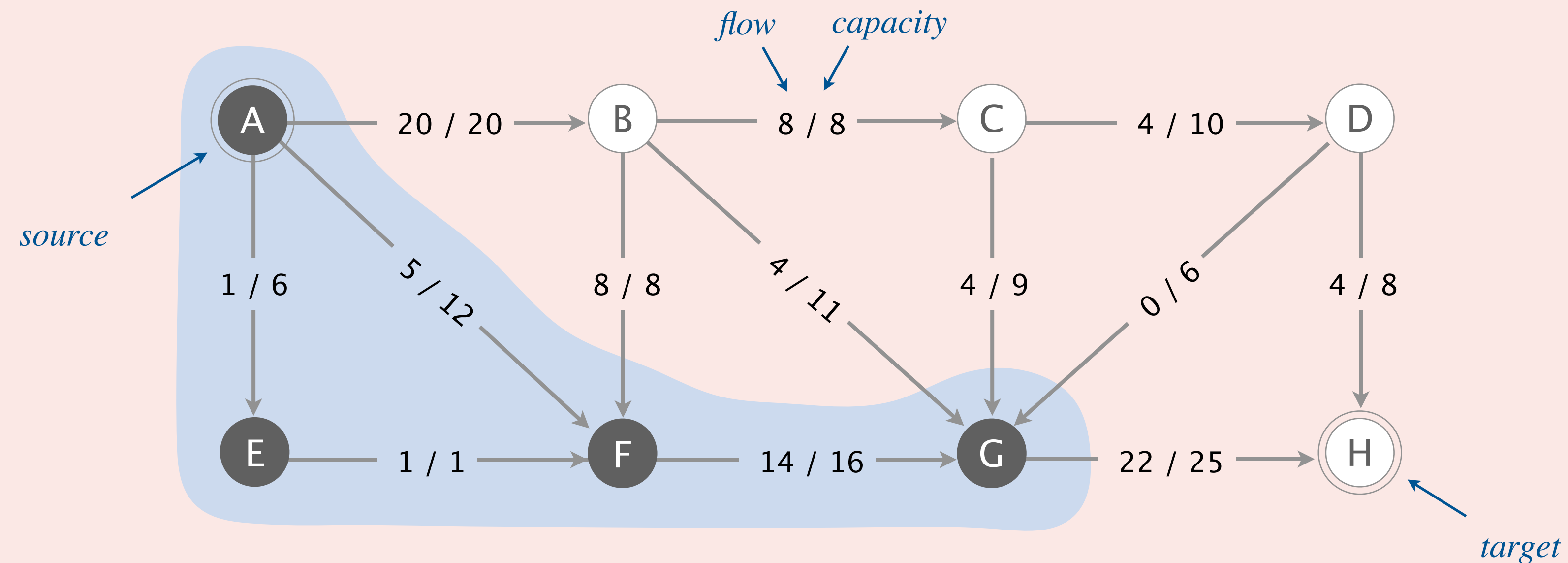
Given the flow f below, what is the **net flow** across the cut $\{A, E, F, G\}$?

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

B. $26 = (20 + 22 - 8 - 4 - 4 - 0)$

C. $42 = (20 + 22)$

D. $45 = (20 + 25)$



Relationship between flows and cuts

Flow-value lemma. Let f be any flow and let (A, B) be any cut.
Then, the net flow across cut $(A, B) = \text{value of flow } f$.

Intuition. Conservation of flow.

Pf. By induction on the number of vertices in B .

- Base case: $B = \{t\}$.
- Induction step: remains true when moving any vertex v from A to B
(because of flow conservation constraint for vertex v)

Corollary. Outflow from $s = \text{inflow to } t = \text{value of flow}$.



we assume no edges incident to s or from t

Relationship between flows and cuts

Weak duality. Let f be any flow and let (A, B) be any cut.

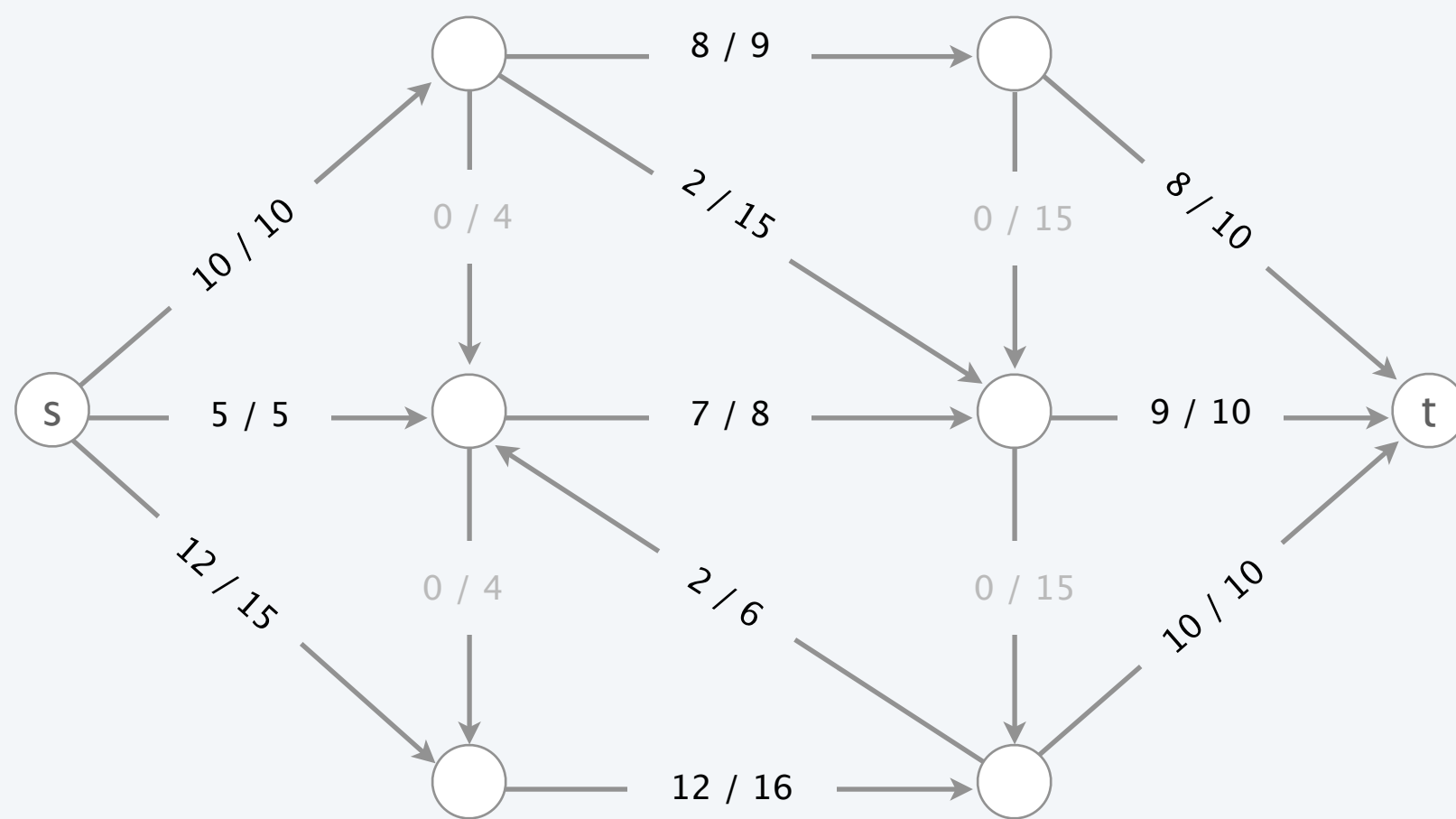
Then, the value of flow $f \leq$ the capacity of cut (A, B) .

Pf. Value of flow $f =$ net flow across cut $(A, B) \leq$ capacity of cut (A, B) .

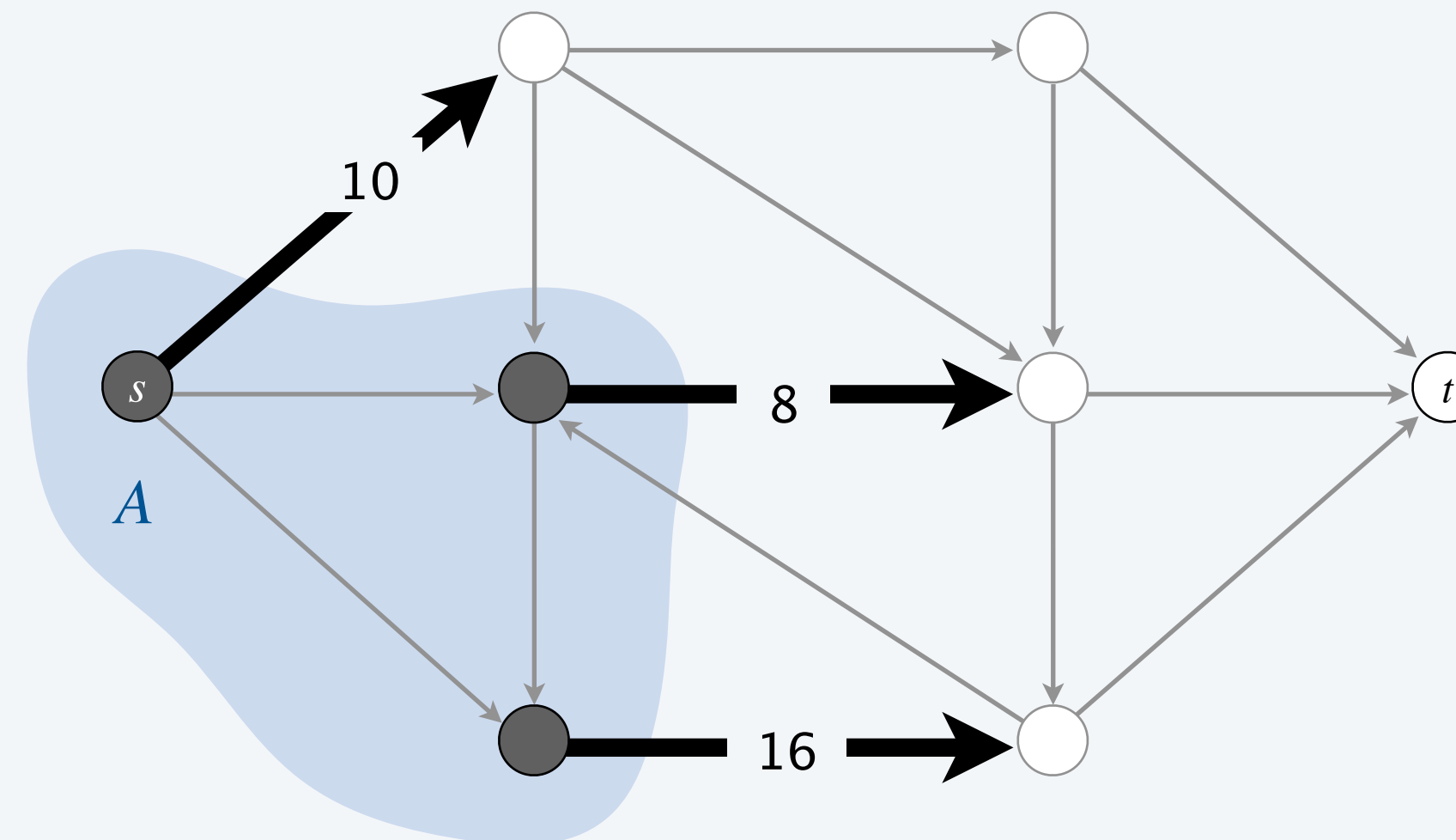
\uparrow
flow-value lemma

\uparrow
flow on each edge from A to B bounded by capacity

Equivalent. Value of maxflow \leq capacity of mincut.



value of flow $f = 27$



capacity of cut $(A, B) = 34$

Maxflow–mincut theorem

Maxflow–mincut theorem. Value of maxflow = capacity of mincut. \longleftarrow “strong duality”

Augmenting path theorem. A flow f is a maxflow if and only if no augmenting paths. \implies *if Ford–Fulkerson terminates, it does so with a maxflow*

Pf. For any flow f , the following three conditions are equivalent:

- i. Flow f is a maxflow.
- ii. There is no augmenting path with respect to flow f .
- iii. There exists a cut whose capacity equals the value of flow f .

Maxflow–mincut theorem

Maxflow–mincut theorem. Value of maxflow = capacity of mincut. $\longleftarrow i \iff iii \text{ and weak duality}$

Augmenting path theorem. A flow f is a maxflow if and only if no augmenting paths. $\longleftarrow i \iff ii$

Pf. For any flow f , the following three conditions are equivalent:

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Maxflow–mincut theorem

Maxflow–mincut theorem. Value of maxflow = capacity of mincut.

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Pf. For any flow f , the following three conditions are equivalent:

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[i \implies ii] We prove contrapositive: $\neg \text{ii} \implies \neg \text{i}$.

- Suppose that there is an augmenting path with respect to flow f .
- Can improve f by updating flow along this path.
- Thus, f is not a maxflow. ■

Maxflow–mincut theorem

Maxflow–mincut theorem. Value of maxflow = capacity of mincut.

Augmenting path theorem. A flow f is a maxflow if and only if no augmenting paths.

Pf. For any flow f , the following three conditions are equivalent:

- i. Flow f is a maxflow.
- ii. There is no augmenting path with respect to flow f .
- iii. There exists a cut whose capacity equals the value of flow f .

[iii \implies i]

- Let (A, B) be a cut whose capacity equals the value of flow f .
 - Then, the value of any flow $f' \leq$ capacity of $(A, B) =$ value of f .
 - Thus, f is a maxflow. ■
- \uparrow
weak duality

\uparrow
by assumption

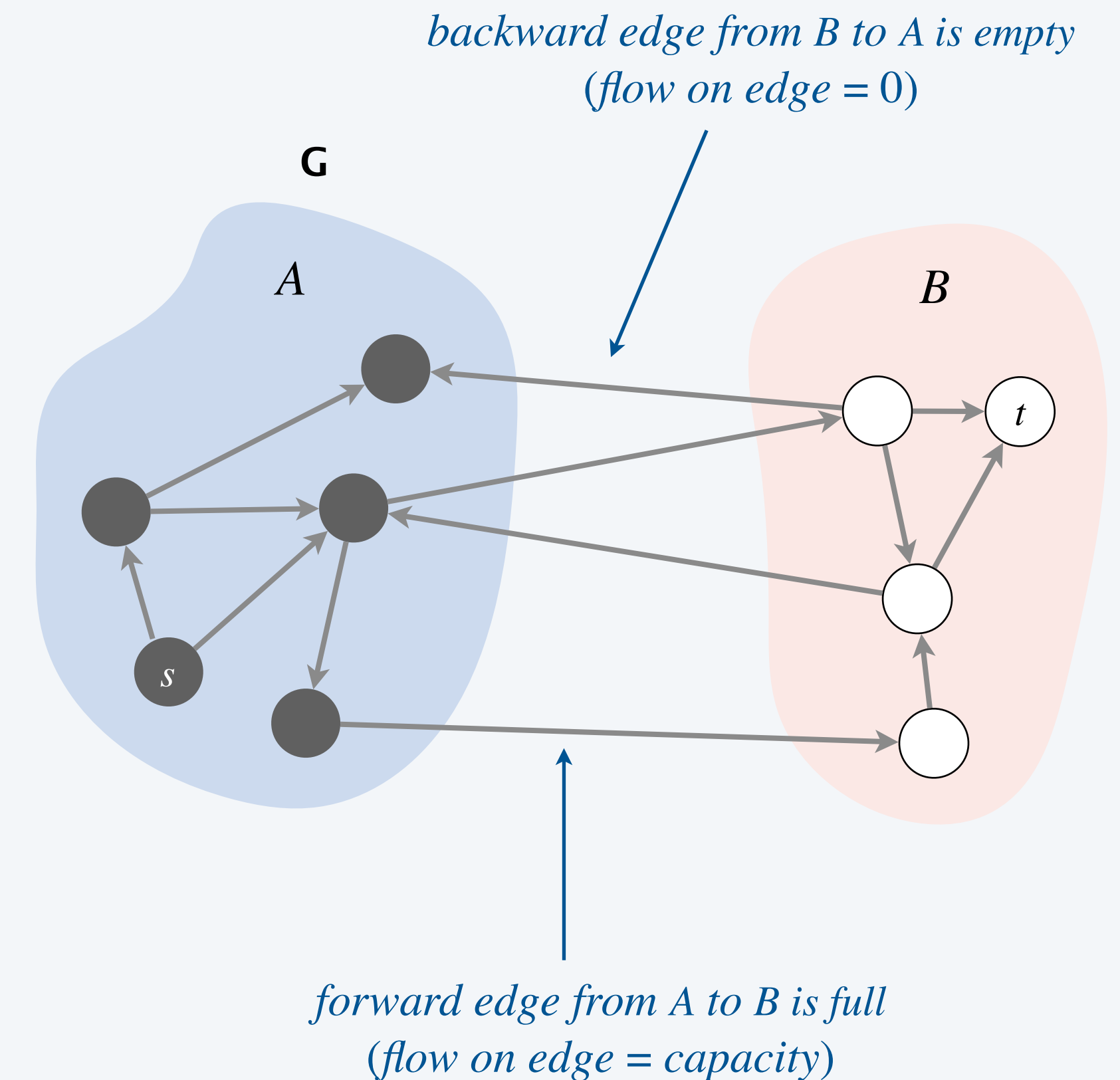
Maxflow–mincut theorem

[ii \implies iii]

- Let f be a flow with no augmenting paths.
- Let A = vertices reachable from s along a path of forward edges (not full) or backward edges (not empty).
- By construction of cut (A, B) , s is in A and t is in B .
- Capacity of cut (A, B) = net flow across cut (A, B)
= value of flow f . ■

*by construction of cut:
all forward edges are full,
all backward edges are empty*

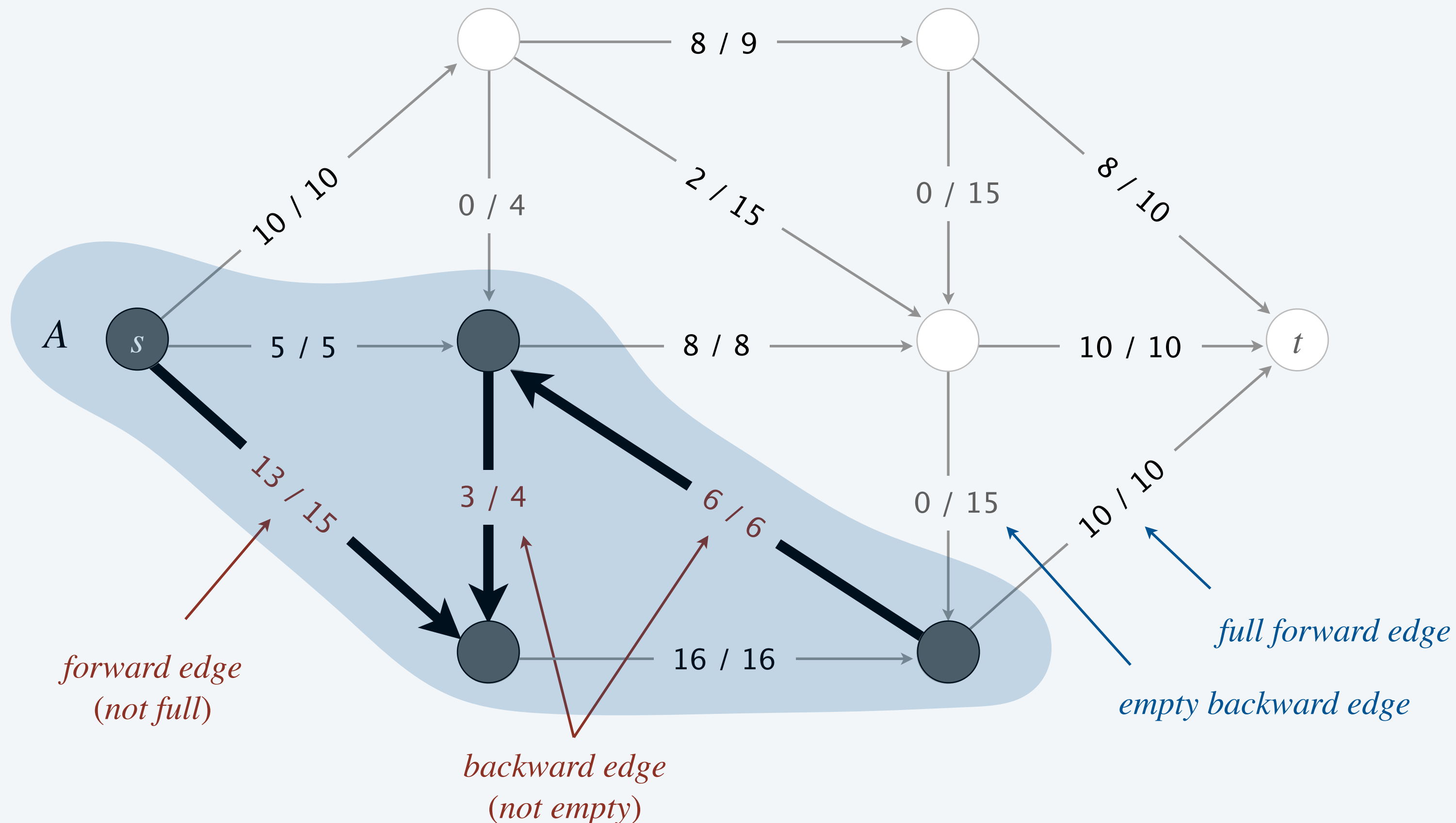
*flow–value
lemma*



Computing a mincut from a maxflow

Q. Given a maxflow f , how to compute a mincut (A, B) ?

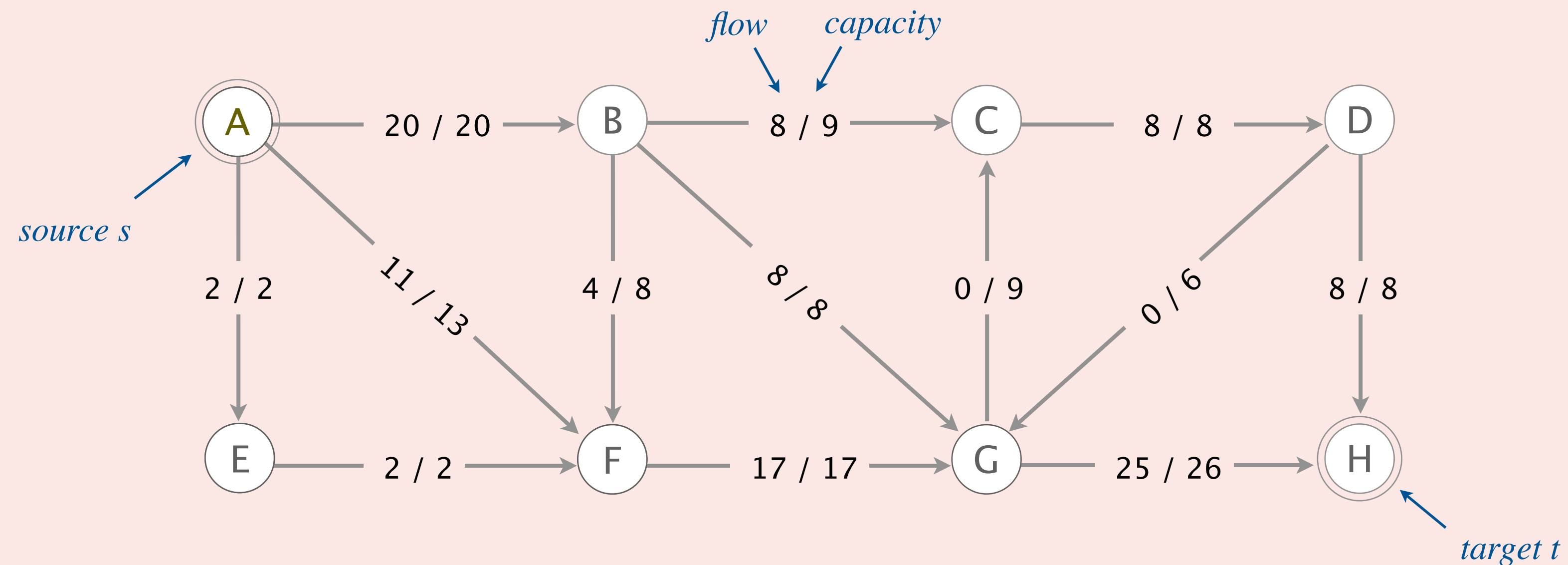
- Augmenting path theorem \implies no augmenting paths with respect to f .
- Use BFS/DFS to compute A = vertices reachable from s along a path of forward edges (not full) or backward edges (not empty).
- As just proved, capacity of cut (A, B) = value of flow $f \implies (A, B)$ is a mincut.

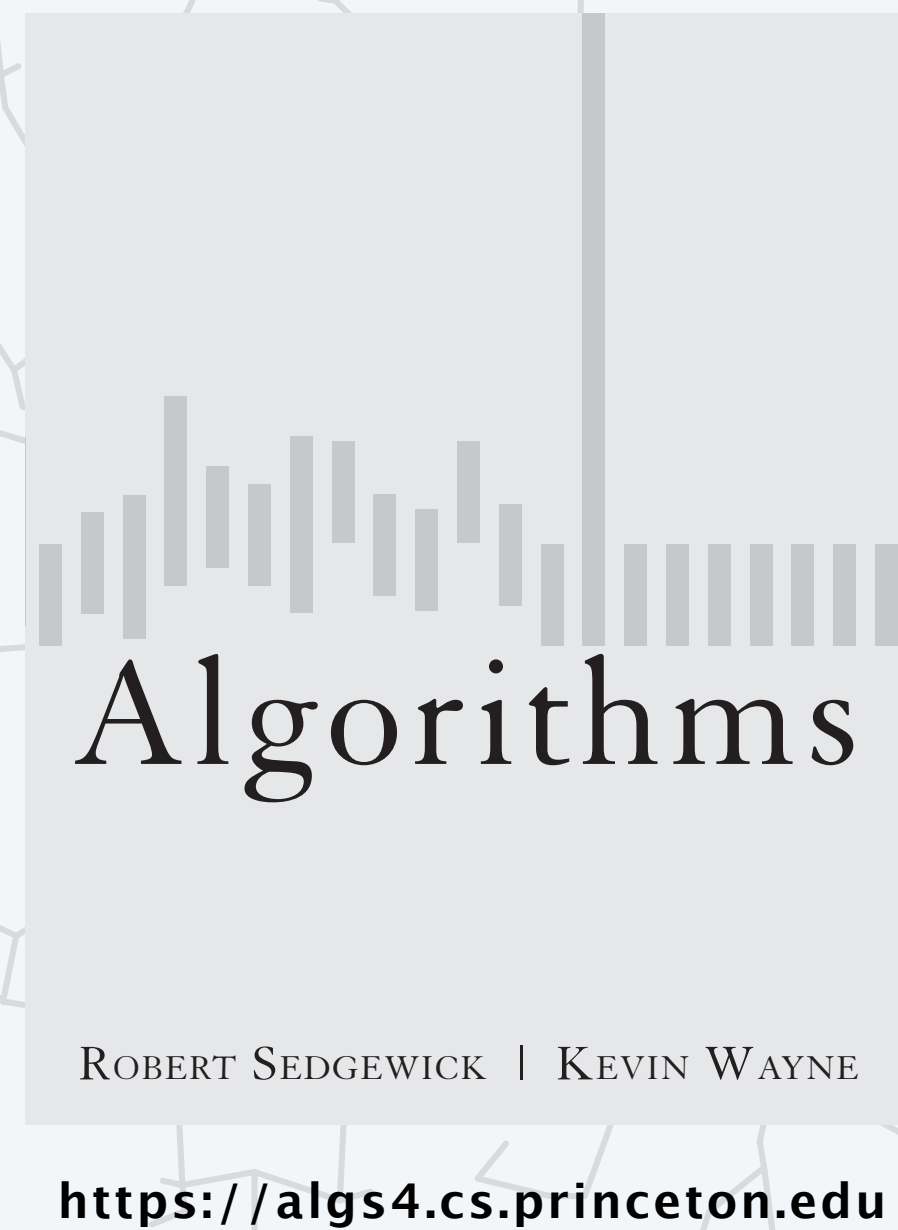




Given the following maxflow, which vertices define a mincut?

- A. $\{ A, F \}$.
- B. $\{ A, B, C, F \}$.
- C. $\{ A, B, C, E, F \}$.
- D. None of the above.





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Ford–Fulkerson algorithm analysis (with integer capacities)

Important special case. Edge capacities are integers between 1 and U .

Invariant. The flow is **integral** throughout Ford–Fulkerson.

flow on each edge is an integer

Pf. Bottleneck capacity is an integer.

Flow on an edge increases/decreases by bottleneck capacity. ■

Proposition. Number of augmentations \leq value of maxflow $\leq EU$.

Pf. Each augmentation increases the value of the flow by at least one. ■

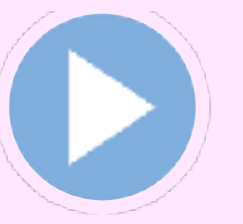
Integrality theorem. There exists an integral maxflow.

critical for some applications (ahead)

Pf.

- Proposition \implies Ford–Fulkerson terminates.
- Augmenting path theorem \implies terminates with a maxflow.
- Invariant \implies maxflow is integral. ■

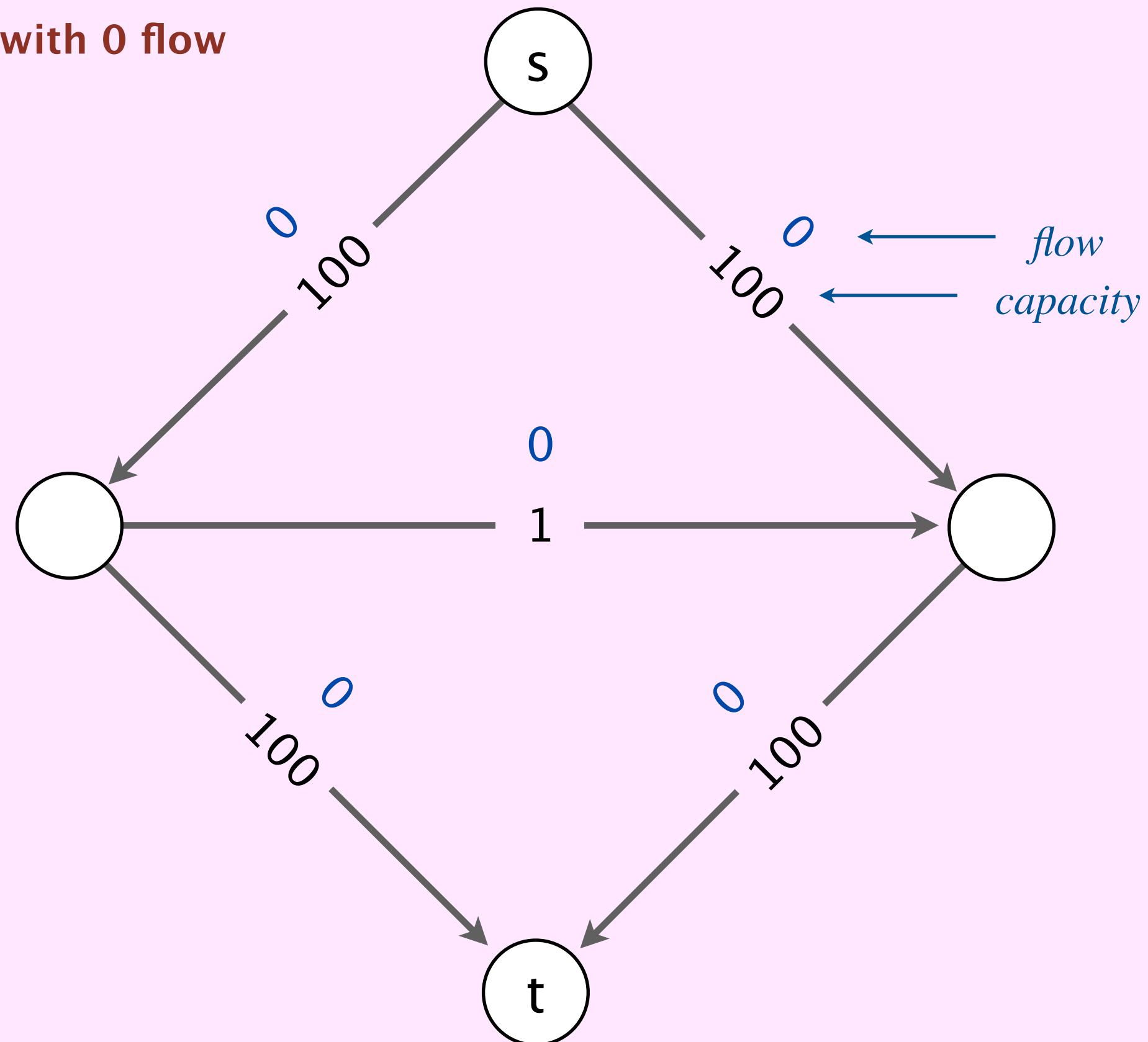
Bad case for Ford–Fulkerson



Bad news. Number of augmenting paths can be very large.

exponential in input size
($V, E, \log U$)

initialize with 0 flow



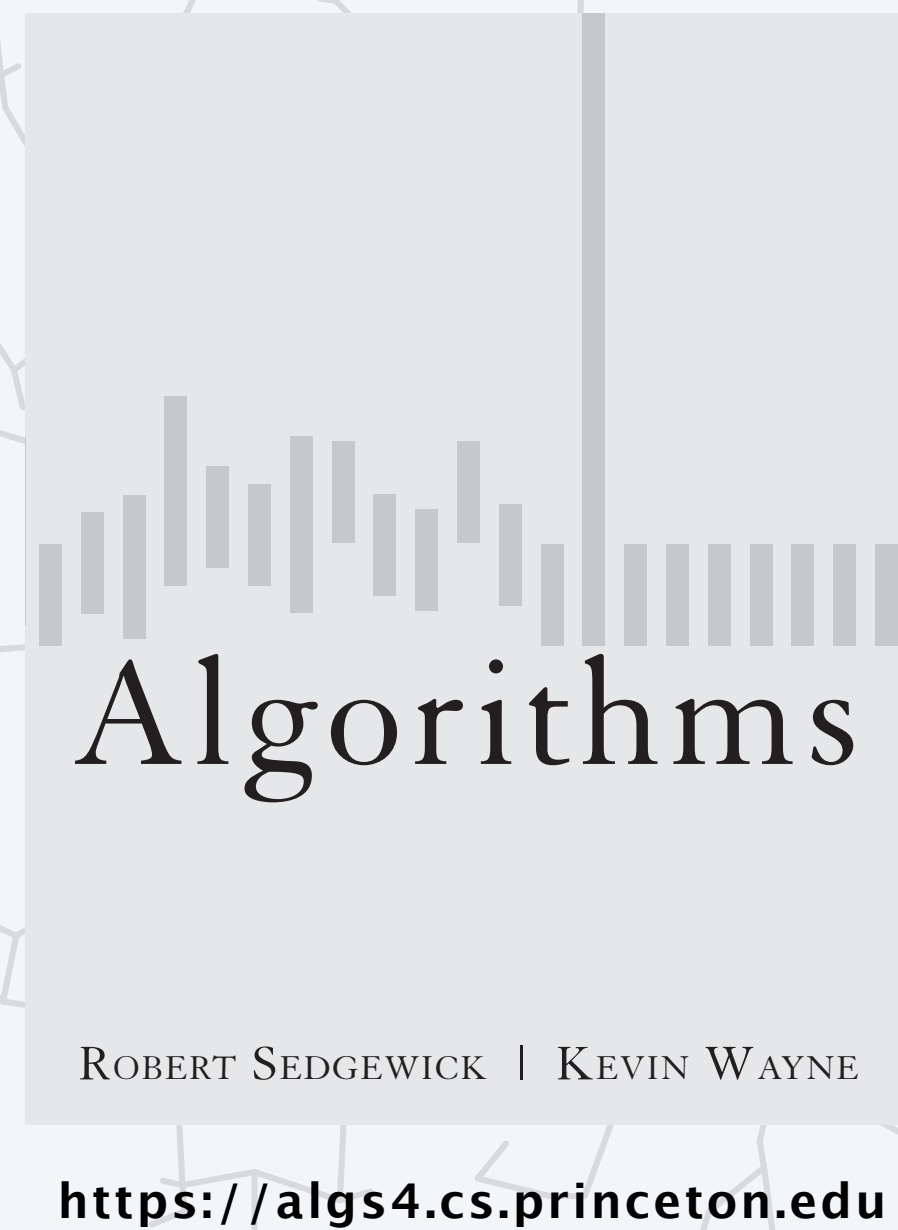
How to choose augmenting paths?

Bad news. Some choices lead to exponential-time algorithms.

Good news. Clever choices lead to polynomial-time algorithms. ← *polynomial in input size*
(V, E, log U)

augmenting path	number of iterations	implementation
DFS path	$\leq E U$	<i>stack</i>
random path	$\leq E U$	<i>randomized queue</i>
shortest path (fewest edges)	$\leq \frac{1}{2} E V$	<i>queue</i>
fattest path (max bottleneck capacity)	$\leq E \ln(E U)$	<i>priority queue</i>

flow network with V vertices, E edges, and integer capacities between 1 and U



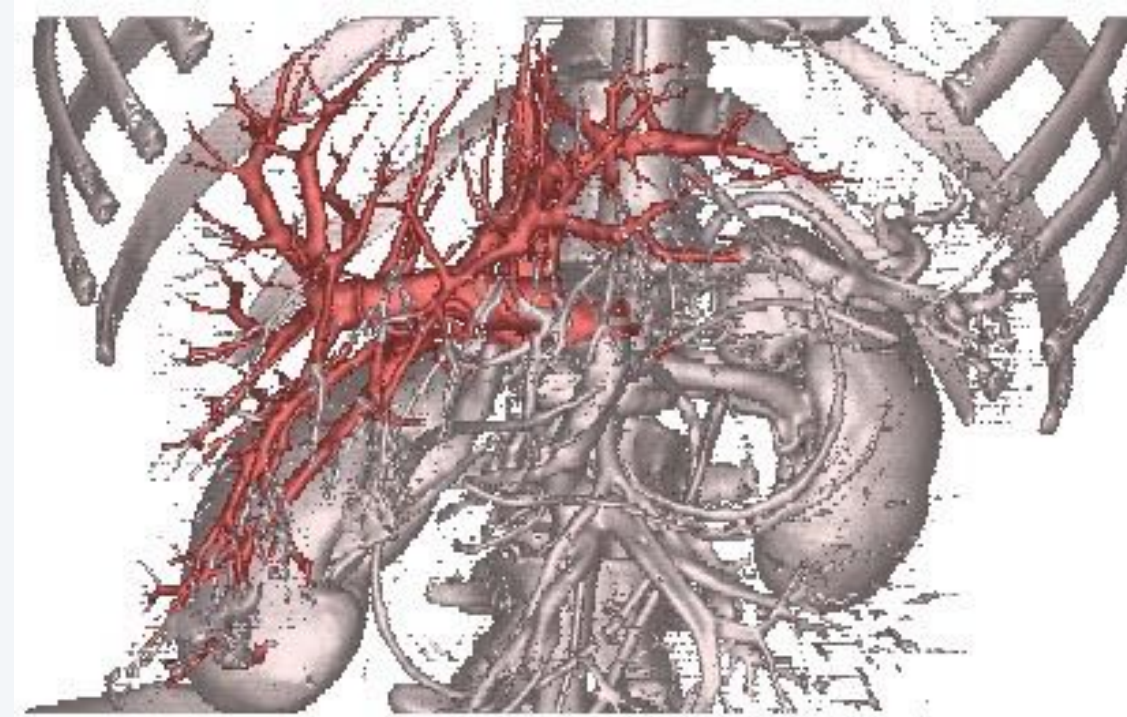
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Maxflow and mincut applications

Maxflow/mincut is a widely applicable problem-solving model.

- Data mining.
- Open-pit mining.
- Bipartite matching.
- Network reliability.
- Baseball elimination.
- Image segmentation.
- Network connectivity.
- Distributed computing.
- Security of statistical data.
- Egalitarian stable matching.
- Multi-camera scene reconstruction.
- Sensor placement for homeland security.
- Many, many, more.



liver and hepatic vascularization segmentation

Bipartite matching problem

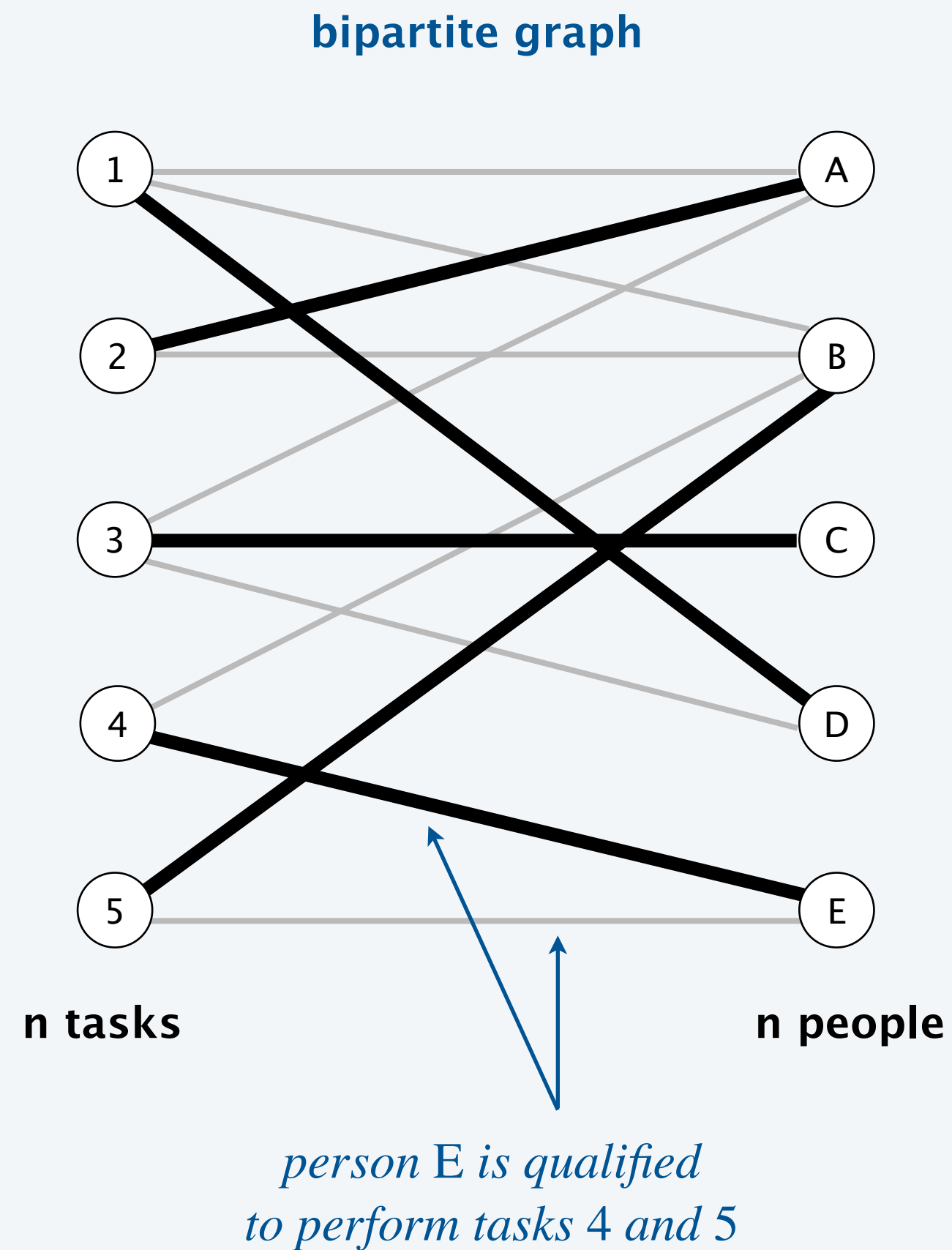
Problem. Given n people and n tasks, assign the tasks to people so that:

- Every task is assigned to a **qualified** person.
- Every person is assigned to exactly one task.



Bipartite matching problem

Problem. Given a **bipartite graph**, find a **perfect matching** (if one exists).



perfect matching

1-D

2-A

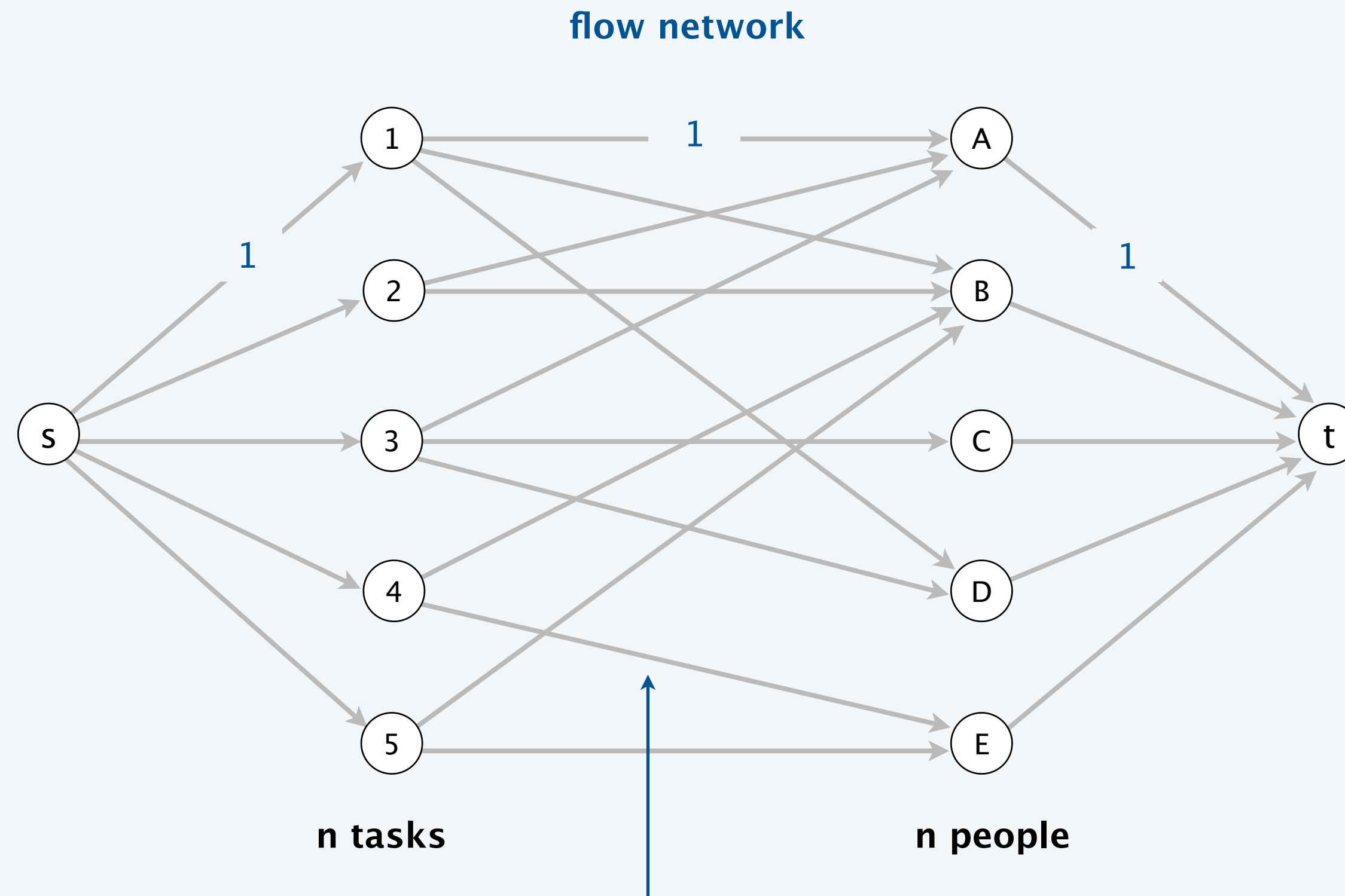
3-C

4-E

5-B

Maxflow formulation of bipartite matching

- Create source s , target t , one vertex i for each task, and one vertex p for each person.
- Add edge from s to each task i of capacity 1.
- Add edge from each person p to t of capacity 1.
- Add edge from task i to qualified person p of capacity 1.

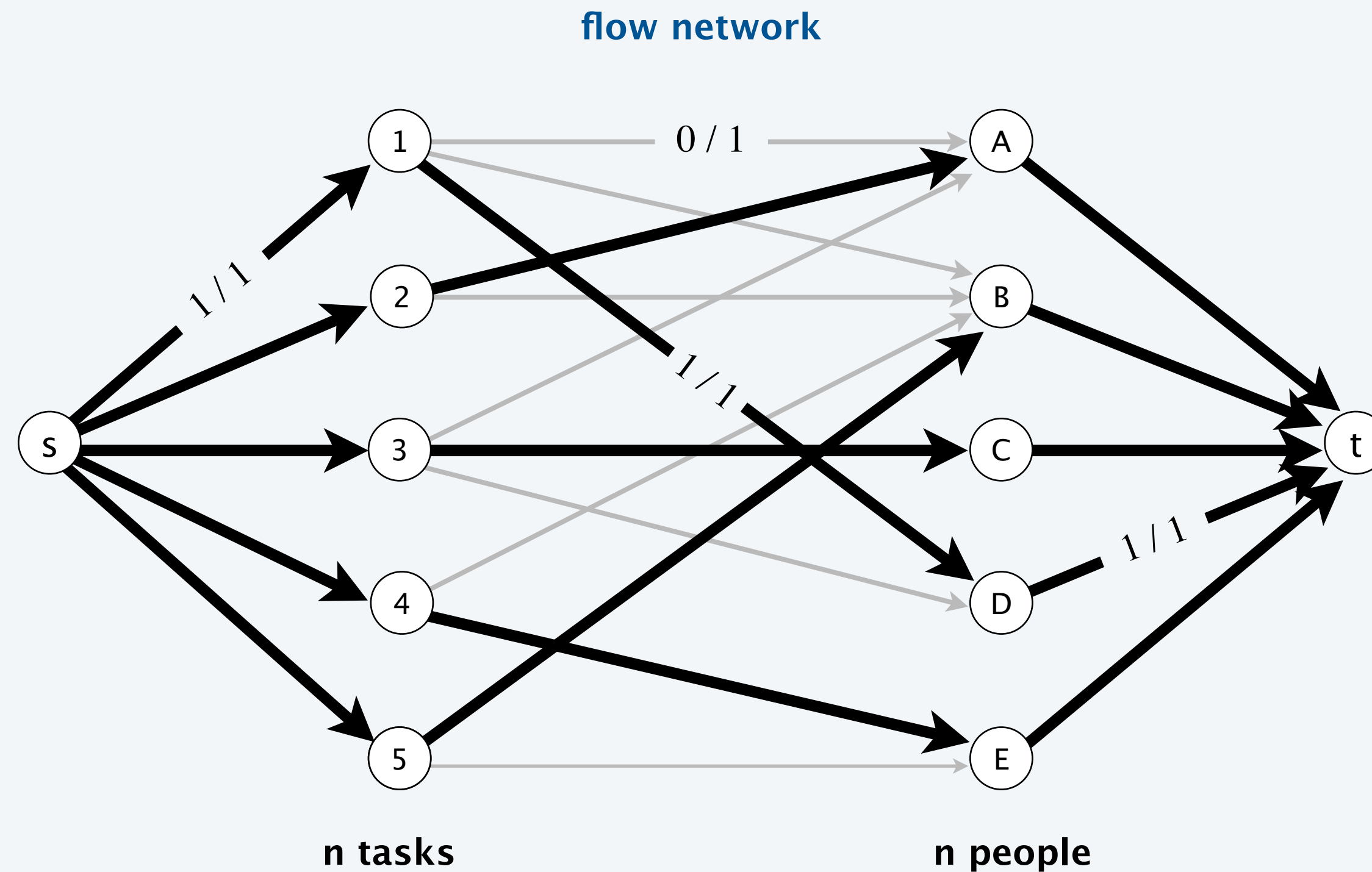


interpretation: flow on edge $4 \rightarrow E = 1$ means assign task 4 to person E

Maxflow formulation of bipartite matching

1-1 correspondence between perfect matchings in bipartite graph and **integral flows** of value n in flow network.

Integrality theorem + 1-1 correspondence \Rightarrow maxflow formulation is correct.





In the worst case, how many iterations (augmenting paths) does the Ford–Fulkerson algorithm perform to find a perfect matching in a bipartite graph with n vertices per side?

A. $\Theta(n)$

B. $\Theta(n^2)$

C. $\Theta(n^3)$

D. $\Theta(n^4)$

Maximum flow algorithms: theory highlights

year	method	worst case	discovered by
1955	<i>augmenting paths</i>	$O(E^2 U)$	Ford–Fulkerson
1970	<i>shortest augmenting paths</i>	$O(E^2 V), O(E V^2)$	Edmonds–Karp, Dinitz
1974	<i>blocking flows</i>	$O(V^3)$	Karzanov
1983	<i>dynamic trees</i>	$O(E V \log V)$	Sleator–Tarjan
1988	<i>push–relabel</i>	$O(E V \log (V^2 / E))$	Goldberg–Tarjan
1998	<i>binary blocking flows</i>	$O(E^{3/2} \log (V^2 / E) \log U)$	Goldberg–Rao
2013	<i>compact networks</i>	$O(E V)$	Orlin
2014	<i>interior-point methods</i>	$\tilde{O}(E V^{1/2} \log U)$	Lee–Sidford
2016	<i>electrical flows</i>	$\tilde{O}(E^{10/7} U^{1/7})$	Mądry
2022	<i>min ratio cycles</i>	$O(E^{1+\varepsilon} \log U)$	CKLPGS
20xx		?????	

max–flow algorithms with E edges, V vertices, and integer capacities between 1 and U

Maximum flow algorithms: practice

Warning. Worst-case running time is generally not useful for predicting (or comparing) the performance of maxflow algorithms in practice.

Often best in practice. Push-relabel method with gap relabeling.

Computer vision. Specialized algorithms for problems with special structure.

On Implementing Push-Relabel Method for the Maximum Flow Problem

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Abstract. We study efficient implementations of the push-relabel method for the maximum flow problem. The resulting codes are faster than the previous codes, and much faster on some problem families. The speedup is due to the combination of heuristics used in our implementations. We also exhibit a family of problems for which the running time of all known methods seem to have a roughly quadratic growth rate.



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Theory and Methodology

Computational investigations of maximum flow algorithms

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Summary

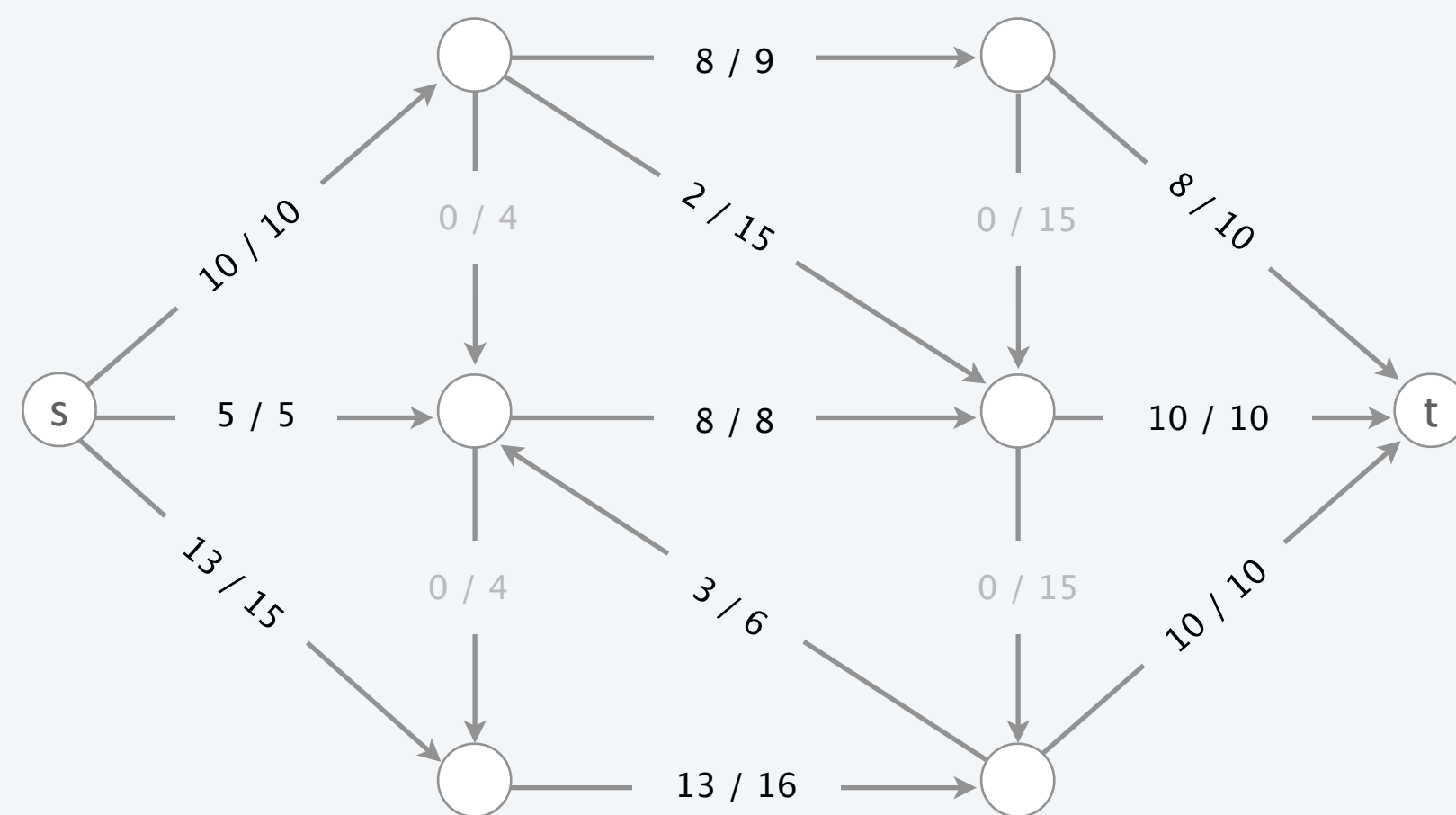
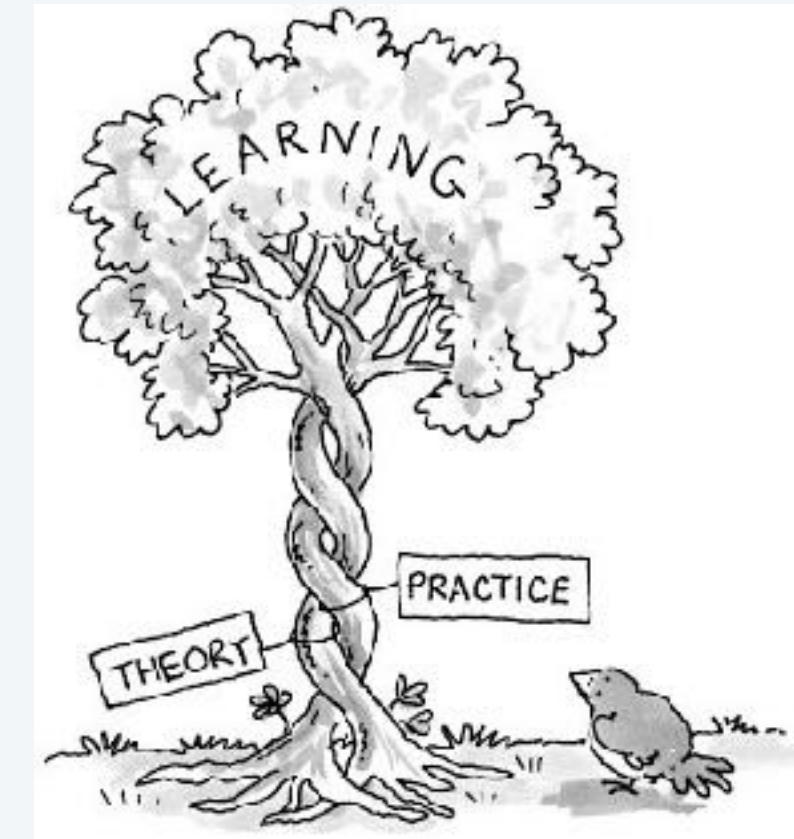
Mincut problem. Find a cut of minimum capacity.

Maxflow problem. Find a flow of maximum value.

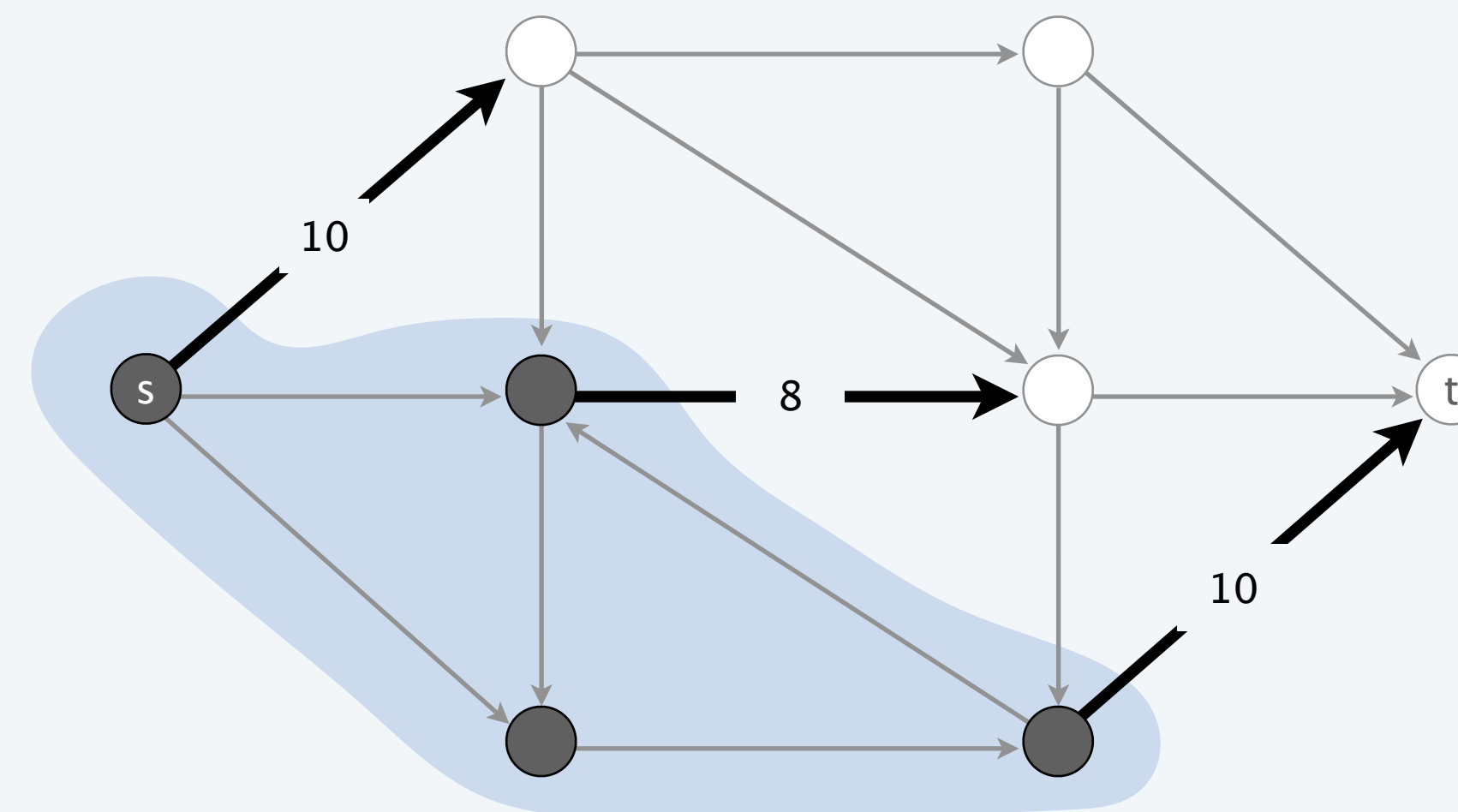
Duality. Value of maxflow = capacity of mincut.

Proven successful approaches.

- Ford-Fulkerson (various augmenting-path strategies).
- Preflow-push (various versions).



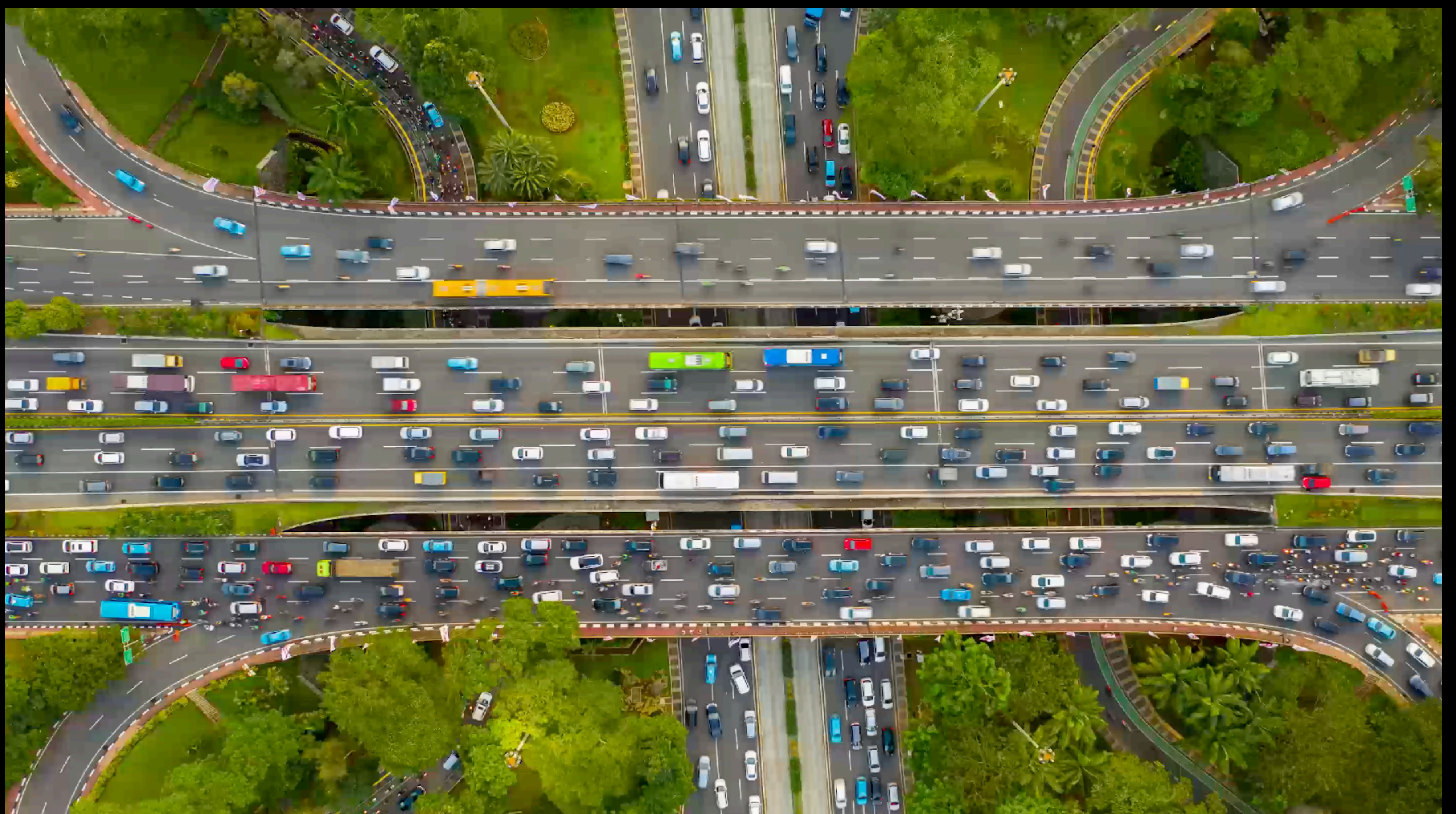
value of flow = 28



capacity of cut = 28

Credits

image	source	license
<i>Warsaw Pact Rail Network</i>	<u>RAND Corporation</u>	
<i>Efficient Max Flow Algorithms</i>	<u>Communications of the ACM</u>	
<i>Liver Segmentation</i>	S. Esneault, T. Pham, K. Torres	
<i>Workers</i>	<u>pictofigo.com</u>	<u>Pictofigo</u>
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<i>Traffic Flows</i>	<u>Quanta Magazine</u>	



Researchers Achieve 'Absurdly Fast' Algorithm for Network Flow

<https://www.quantamagazine.org/researchers-achieve-absurdly-fast-algorithm-for-network-flow-20220608>