

# Discussion 8

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CS180: Introduction to Algorithms and Complexity

Prof. Mark Burgin

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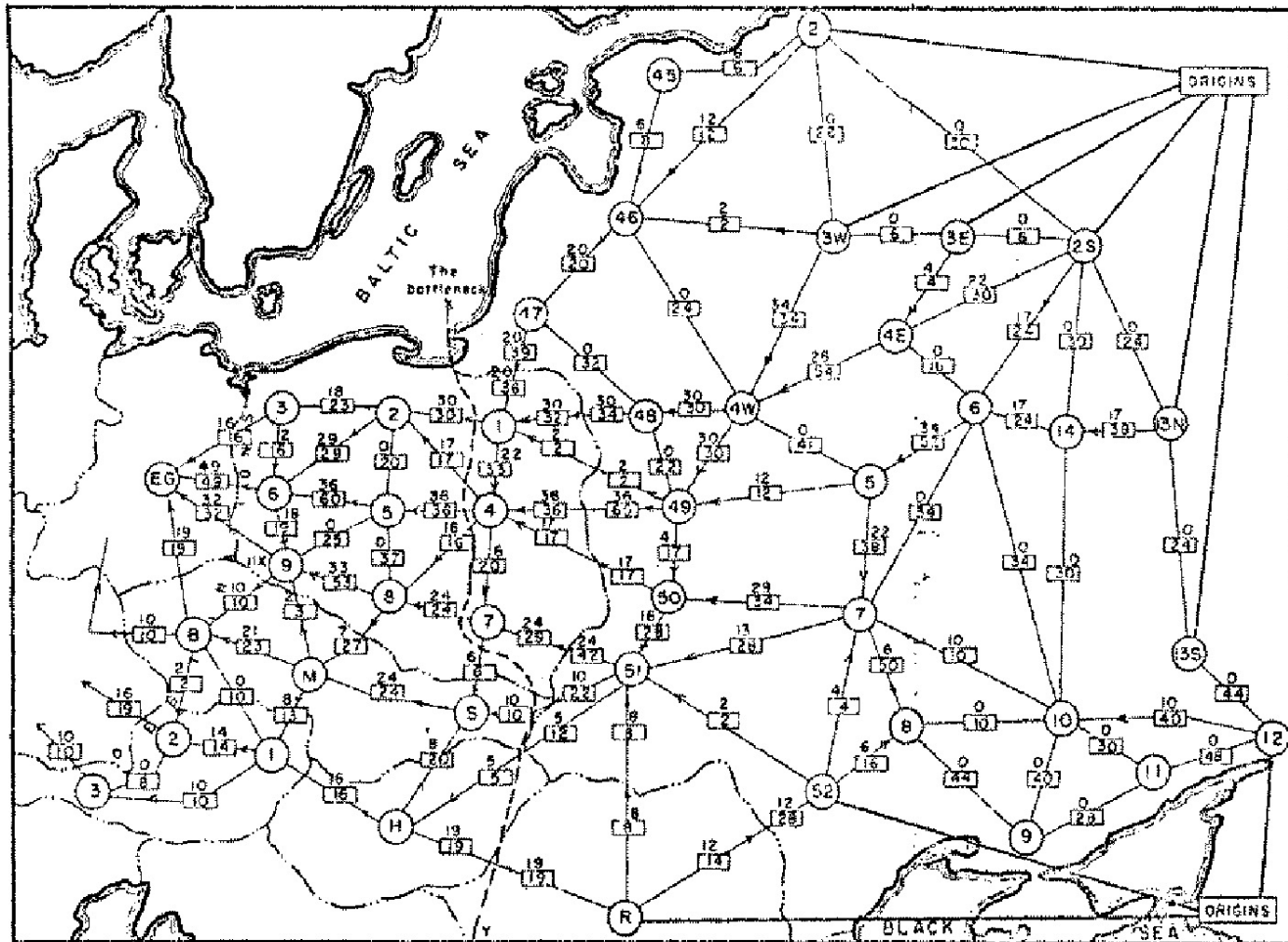
Ling Ding [lingding@cs.ucla.edu](mailto:lingding@cs.ucla.edu)

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# Outline

- Maximum flow minimal cut
- Matching

# Soviet Rail Network, 1955



Reference: *On the history of the transportation and maximum flow problems.*  
Alexander Schrijver in Math Programming, 91: 3, 2002.

# Maximum Flow and Minimum Cut

## Max flow and min cut.

- Two very rich algorithmic problems.
- Cornerstone problems in combinatorial optimization.
- Beautiful mathematical duality.

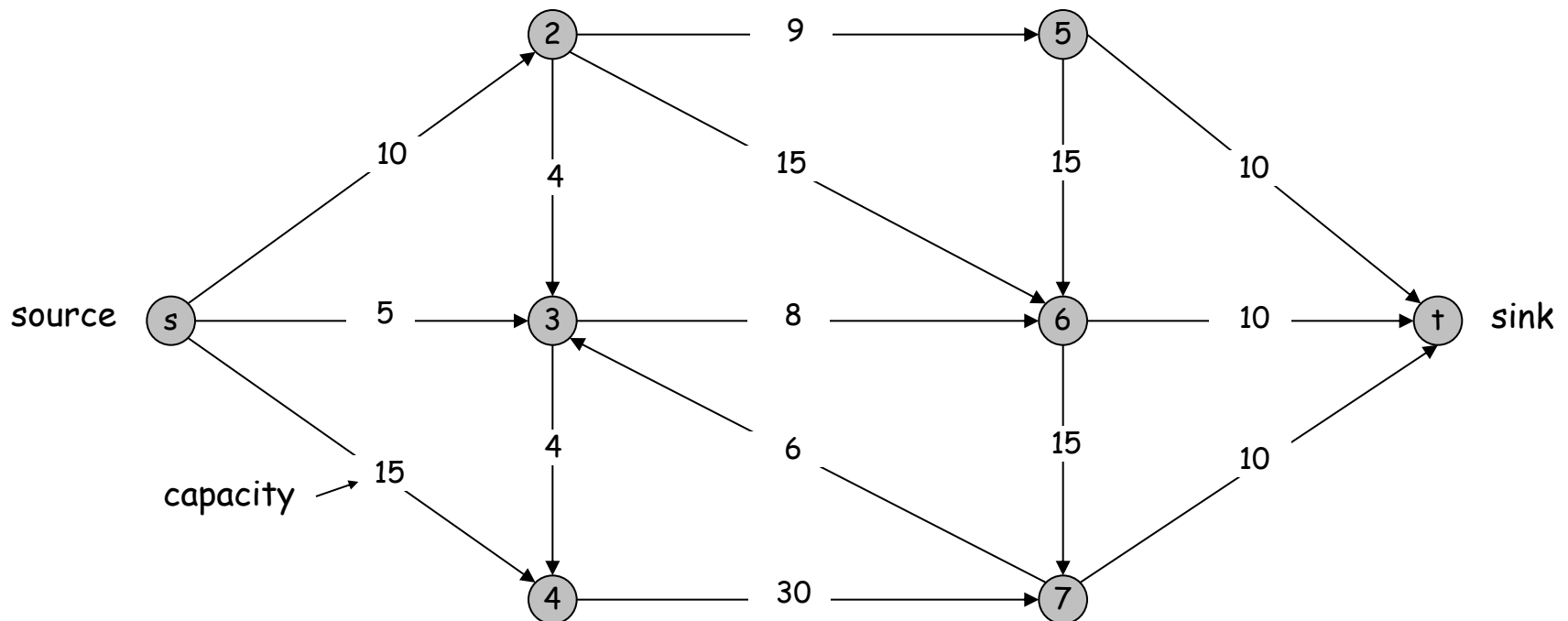
## Nontrivial applications / reductions.

- Data mining.
- Open-pit mining.
- Project selection.
- Airline scheduling.
- Bipartite matching.
- Baseball elimination.
- Image segmentation.
- Network connectivity.
- Network reliability.
- Distributed computing.
- Egalitarian stable matching.
- Security of statistical data.
- Network intrusion detection.
- Multi-camera scene reconstruction.
- Many many more ...

# Minimum Cut Problem

## Flow network.

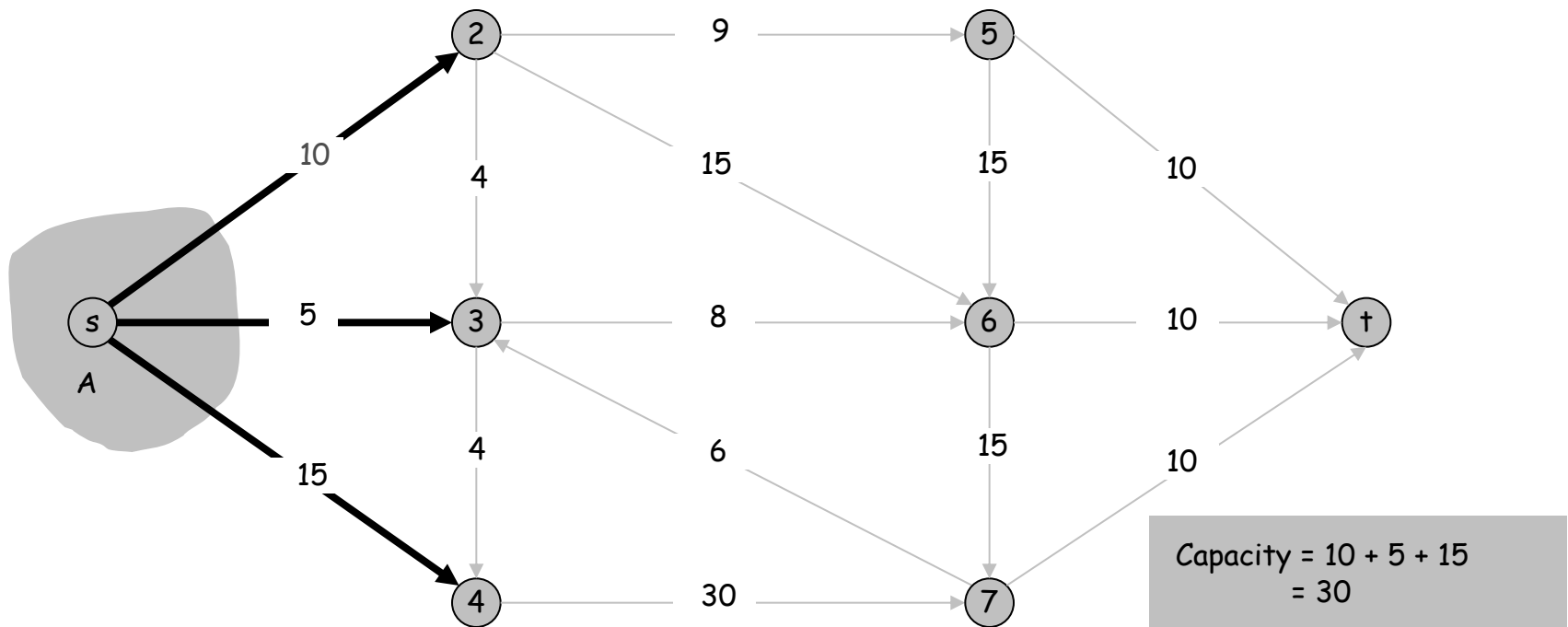
- Abstraction for material **flowing** through the edges.
- $G = (V, E)$  = directed graph, no parallel edges.
- Two distinguished nodes:  $s$  = source,  $t$  = sink.
- $c(e)$  = capacity of edge  $e$ .



# Cuts

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

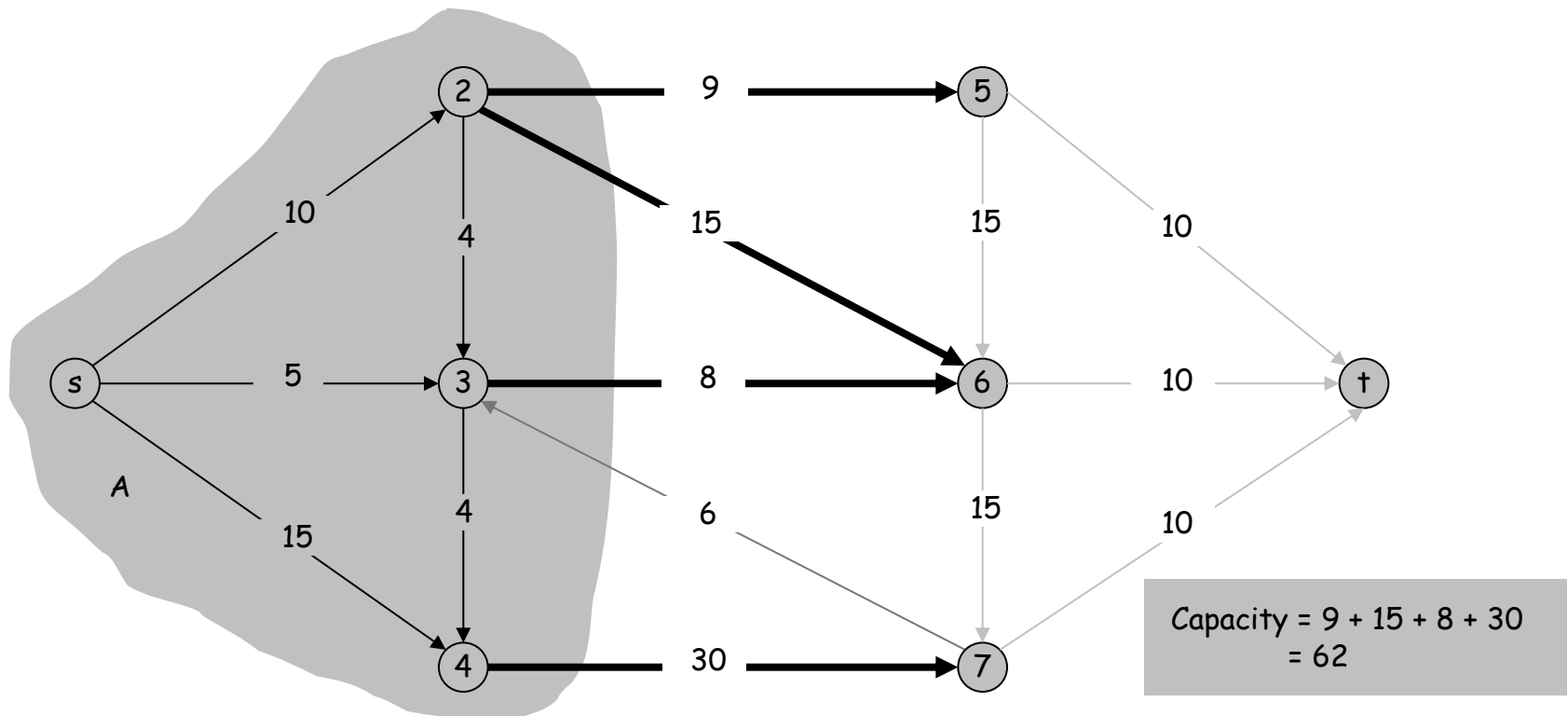
Def. The **capacity** of a cut  $(A, B)$  is:  $cap(A, B) = \sum_{e \text{ out of } A} c(e)$



# Cuts

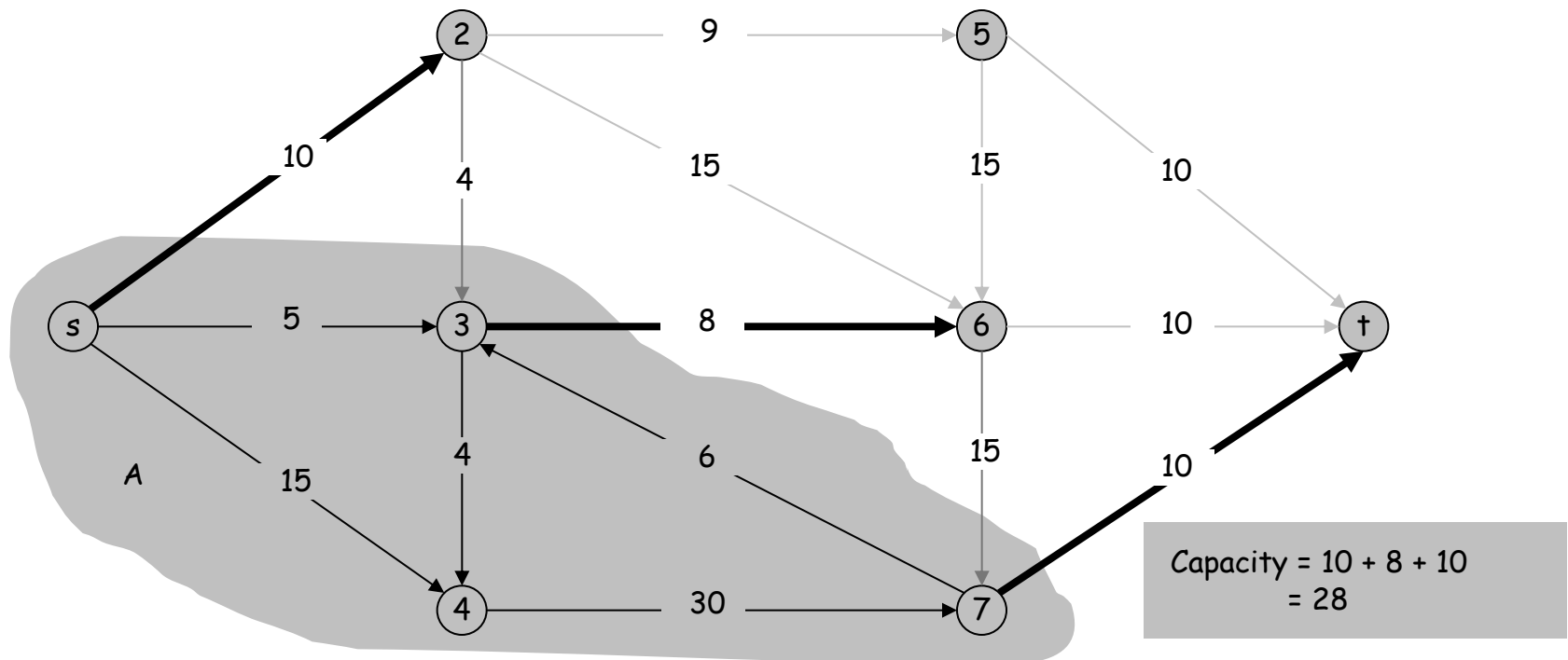
Def. An **s-t cut** is a partition  $(A, B)$  of  $V$  with  $s \in A$  and  $t \in B$ .

Def. The **capacity** of a cut  $(A, B)$  is:  $cap(A, B) = \sum_{e \text{ out of } A} c(e)$



# Minimum Cut Problem

Min s-t cut problem. Find an s-t cut of minimum capacity.



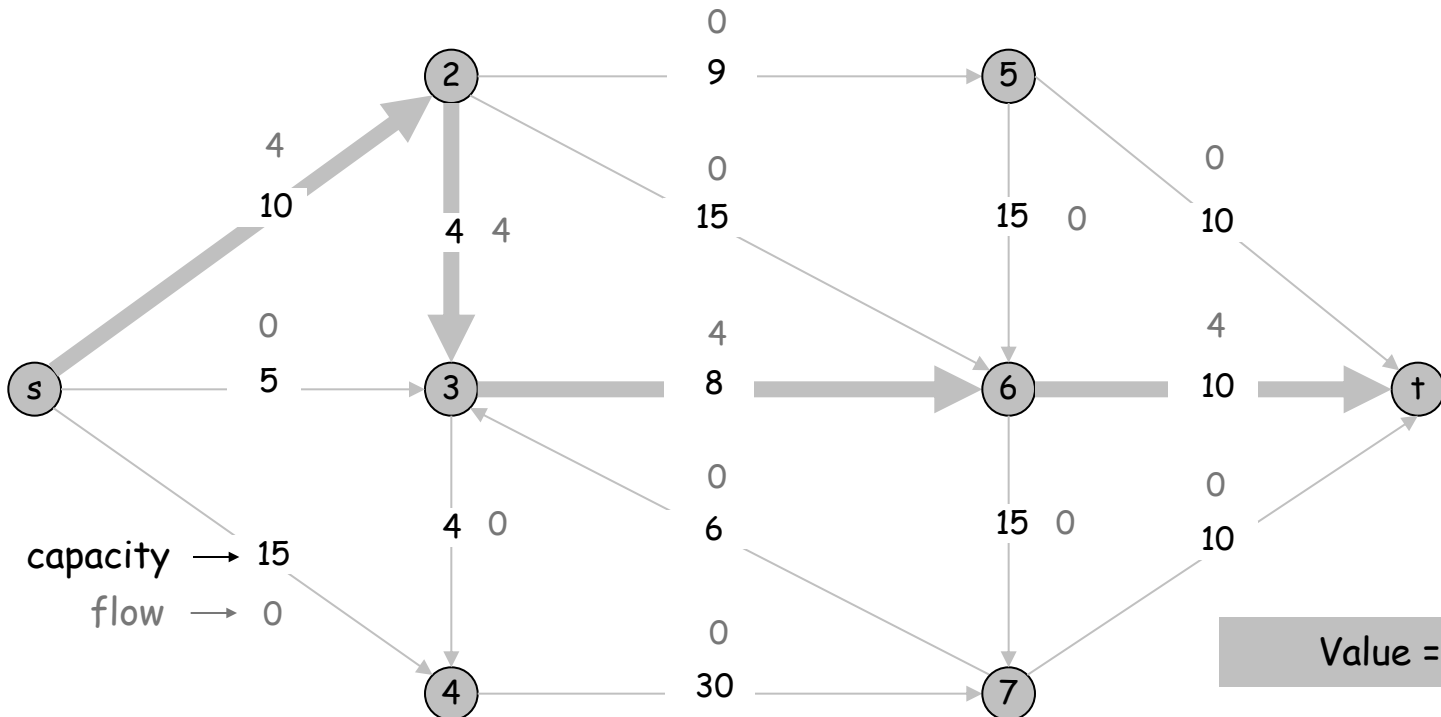


# Flows

**Def.** An **s-t flow** is a function that satisfies:

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

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

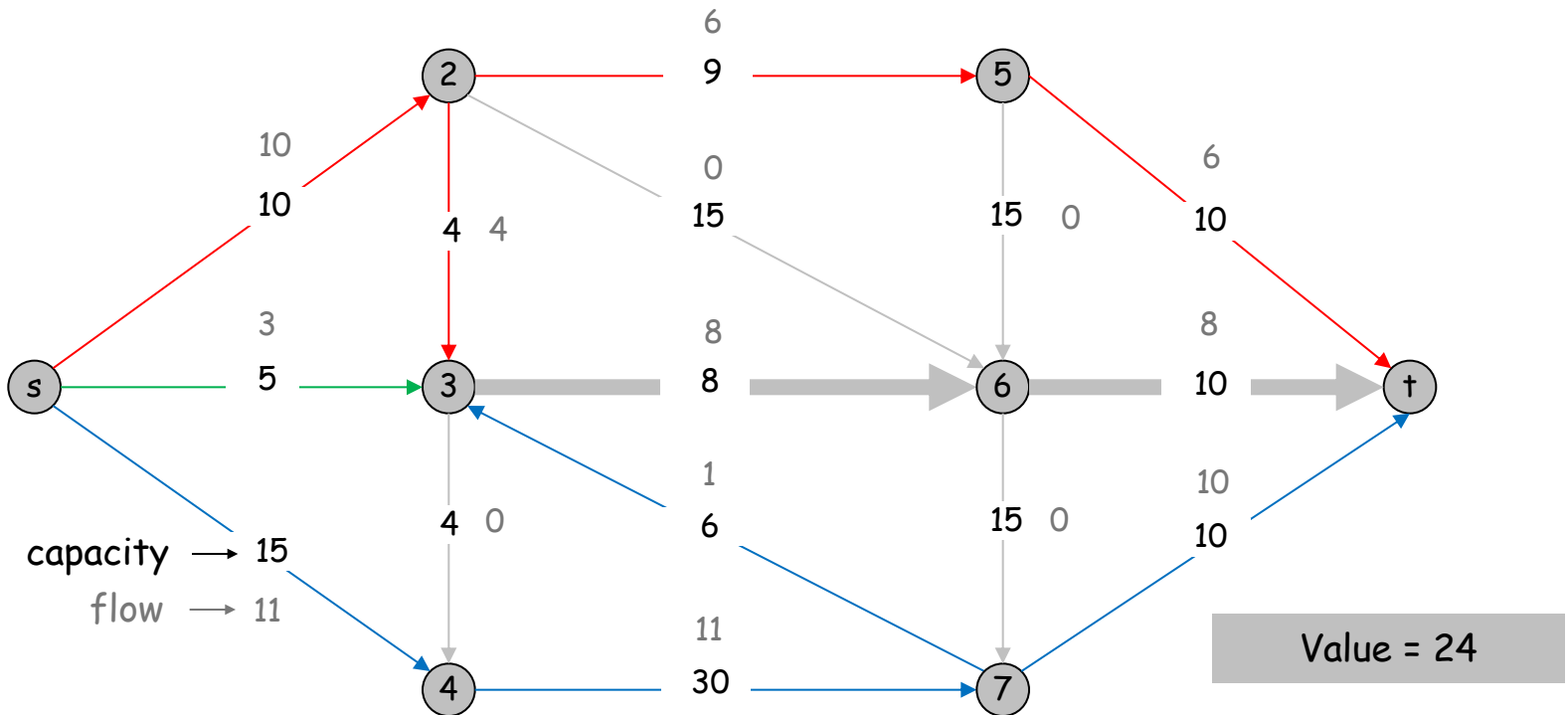


# Flows

Def. An **s-t flow** is a function that satisfies:

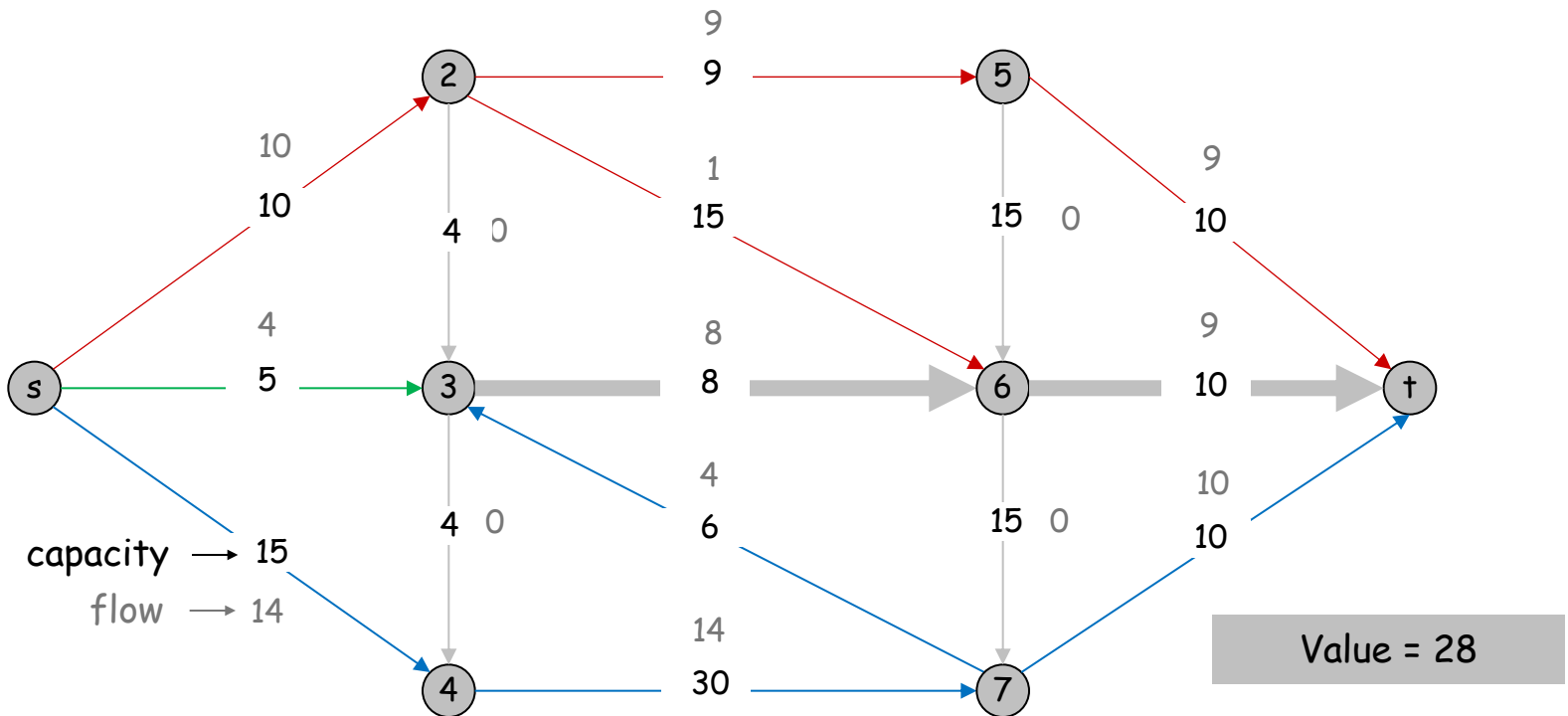
- For each  $e \in E$ :  $0 \leq f(e) \leq c(e)$  [capacity]
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# Maximum Flow Problem

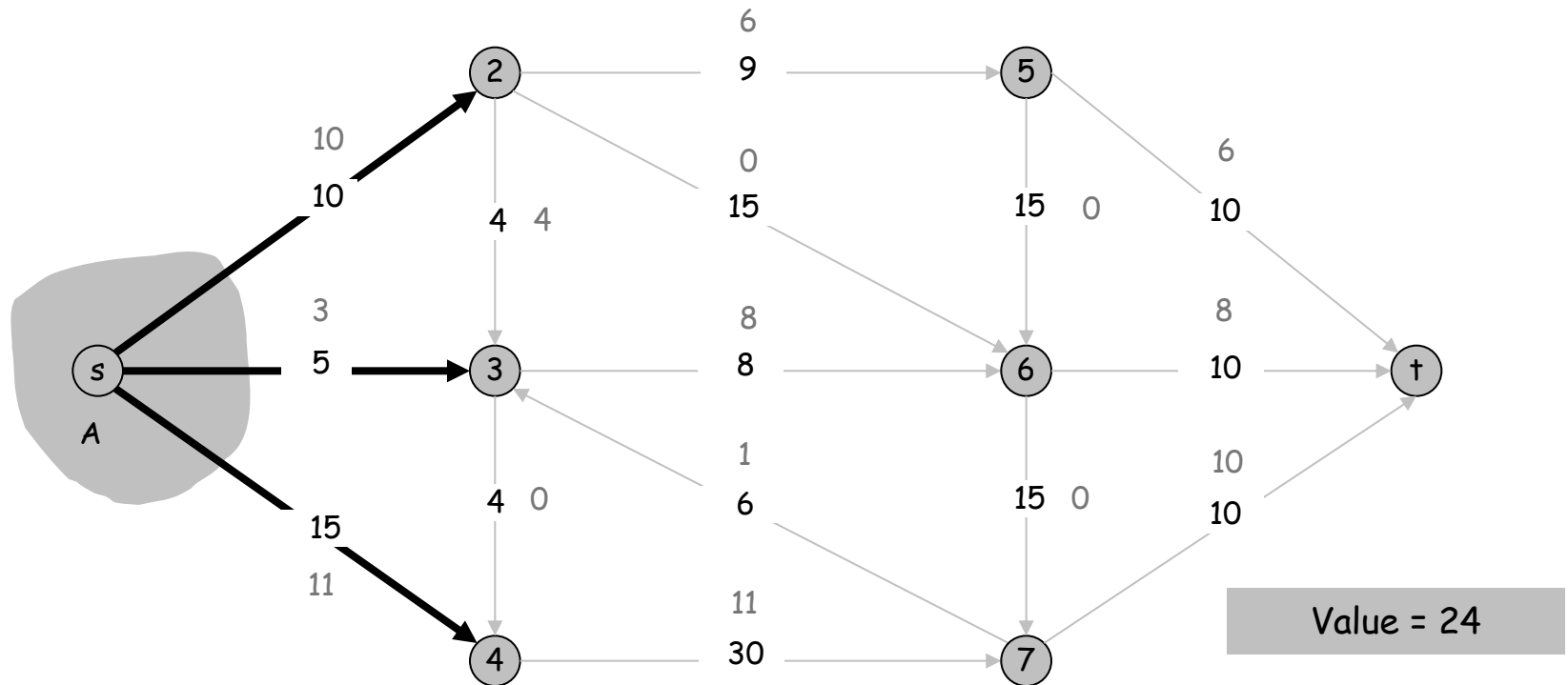
Max flow problem. Find s-t flow of maximum value.



# Flows and Cuts

**Flow value lemma.** Let  $f$  be any flow, and let  $(A, B)$  be any  $s$ - $t$  cut. Then, the net flow sent across the cut is equal to the amount leaving  $s$ .

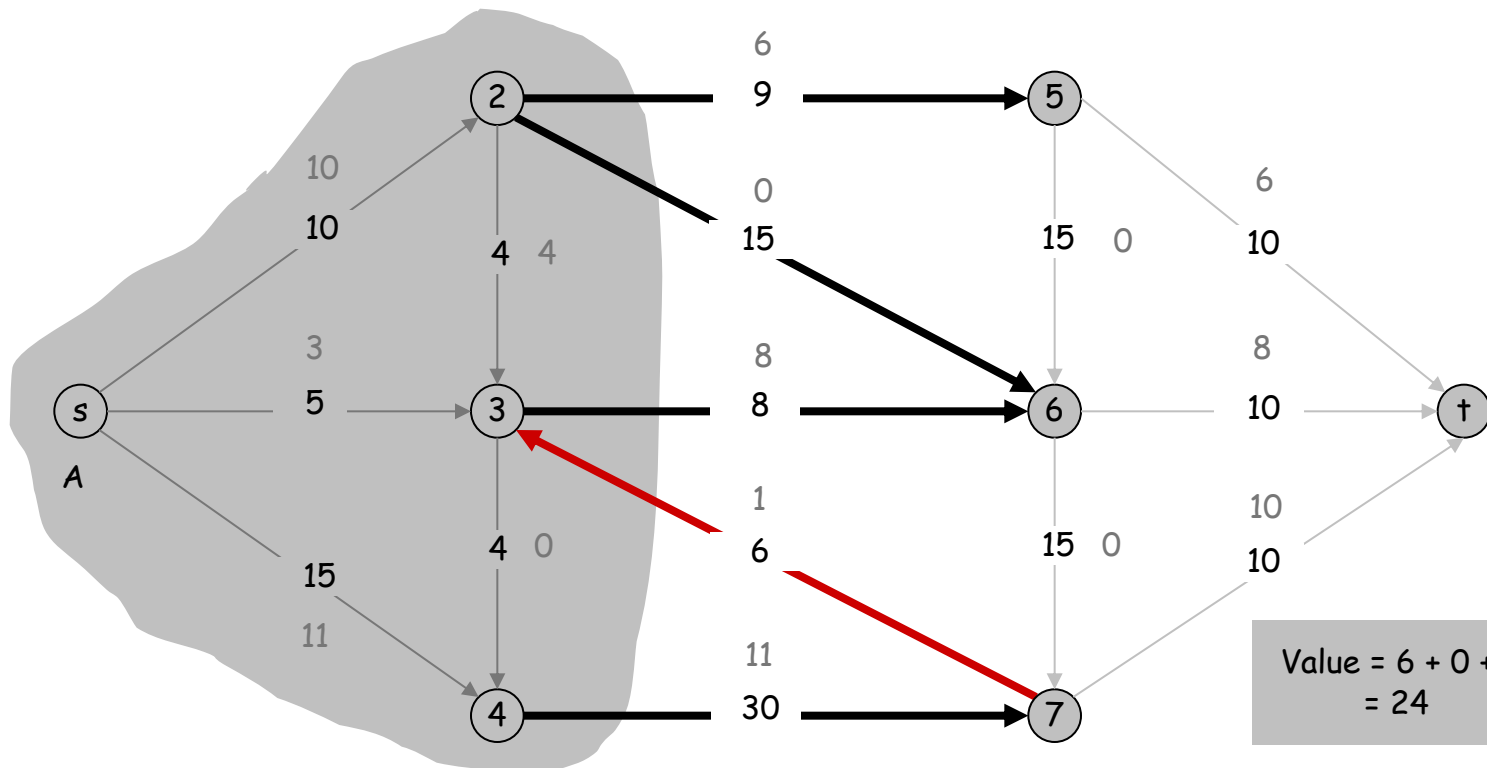
$$\sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e) = v(f)$$



# Flows and Cuts

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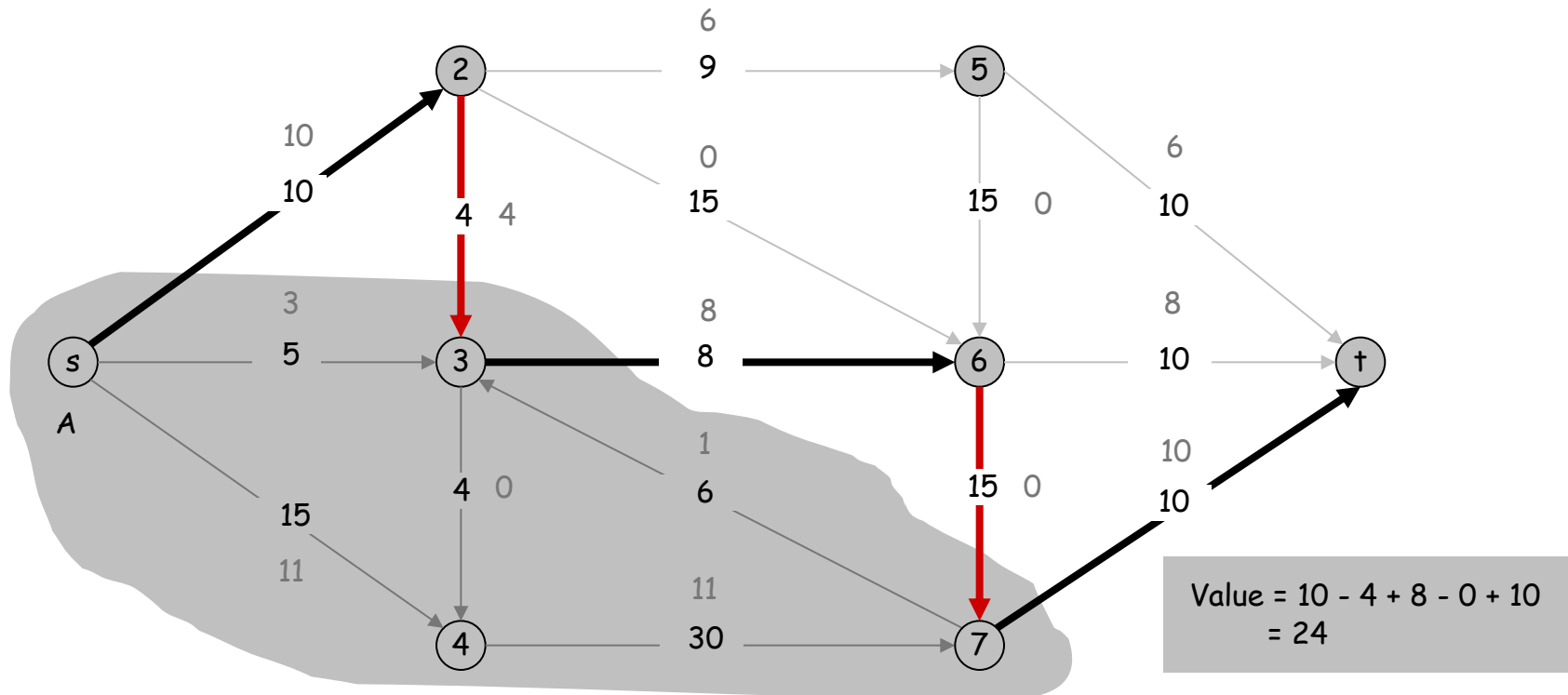
$$\sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e) = v(f)$$



# Flows and Cuts

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



# Flows and Cuts

**Flow value lemma.** Let  $f$  be any flow, and let  $(A, B)$  be any  $s$ - $t$  cut. Then

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

**Pf.**

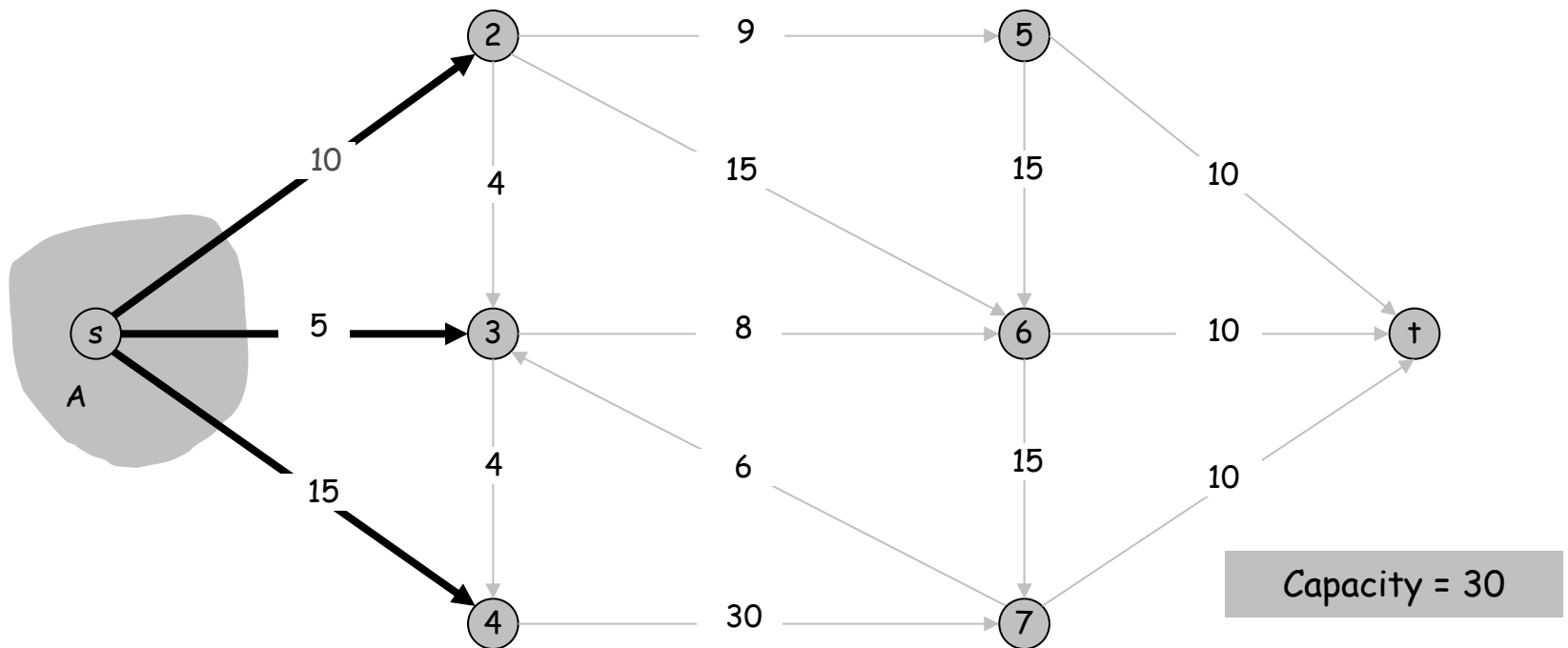
by flow conservation, all terms  
except  $v = s$  are 0

$$\begin{aligned} v(f) &= \sum_{e \text{ out of } s} f(e) \\ &\rightarrow = \sum_{v \in A} \left( \sum_{e \text{ out of } v} f(e) - \sum_{e \text{ in to } v} f(e) \right) \\ &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e). \end{aligned}$$

# Flows and Cuts

**Weak duality.** Let  $f$  be any flow, and let  $(A, B)$  be any  $s$ - $t$  cut. Then the value of the flow is at most the capacity of the cut.

Cut capacity = 30  $\Rightarrow$  Flow value  $\leq 30$



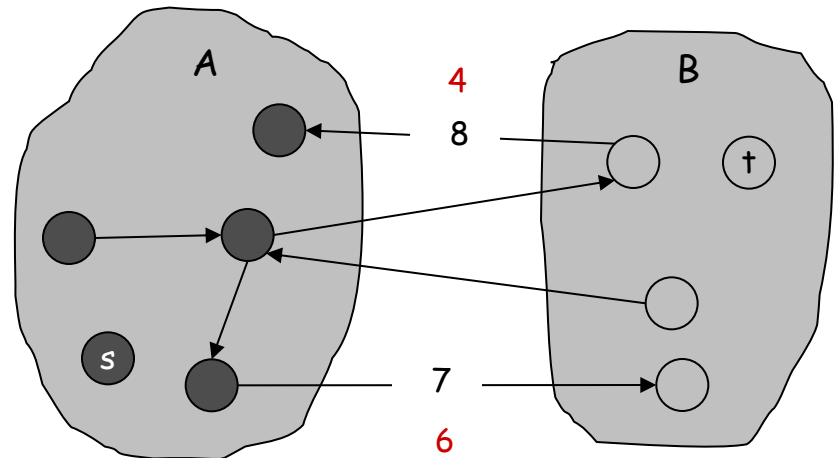


# Flows and Cuts

**Weak duality.** Let  $f$  be any flow. Then, for any  $s$ - $t$  cut  $(A, B)$  we have  $v(f) \leq \text{cap}(A, B)$ .

**Pf.**

$$\begin{aligned} v(f) &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e) \\ &\leq \sum_{e \text{ out of } A} c(e) \\ &\leq \sum_{e \text{ out of } A} c(e) \\ &= \text{cap}(A, B) \quad \blacksquare \end{aligned}$$

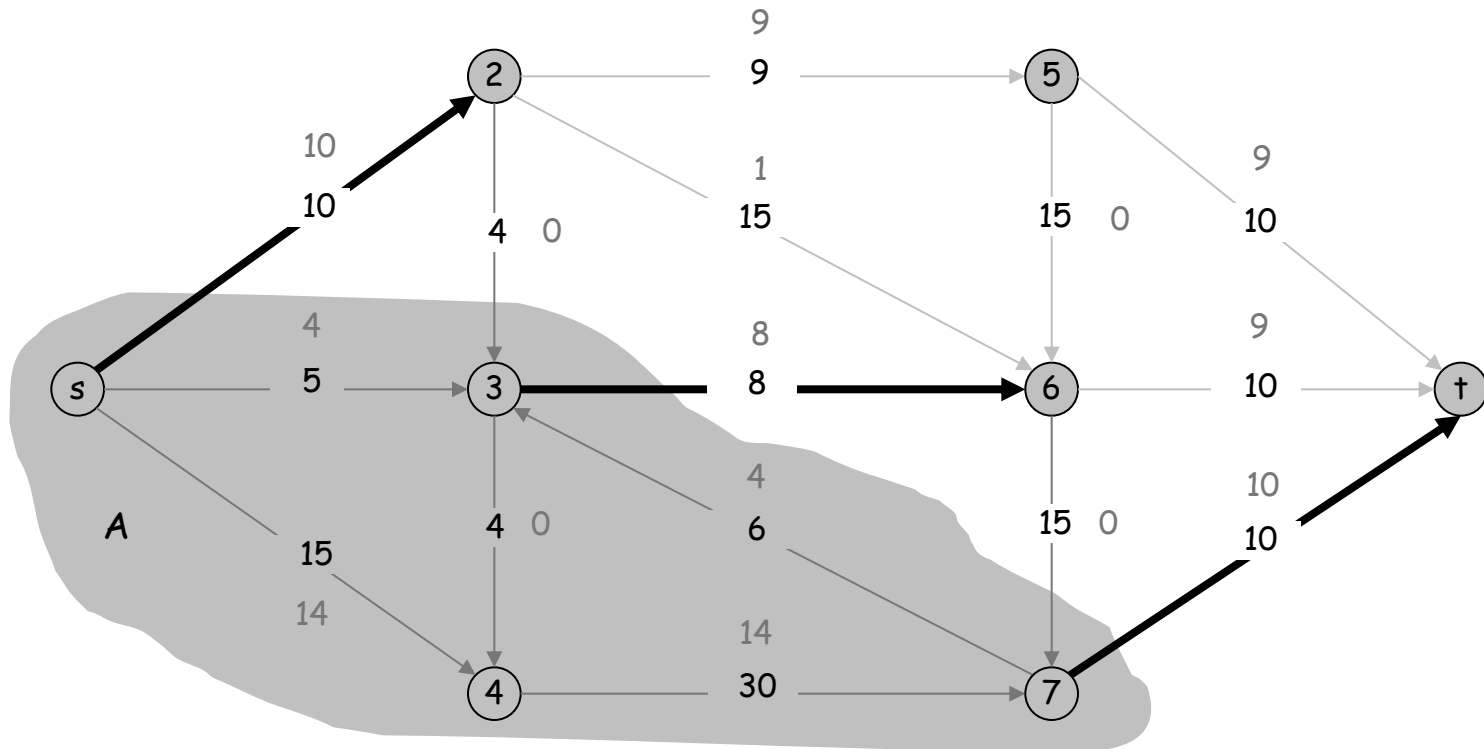


# Certificate of Optimality

**Corollary.** Let  $f$  be any flow, and let  $(A, B)$  be any cut. If  $v(f) = \text{cap}(A, B)$ , then  $f$  is a max flow and  $(A, B)$  is a min cut.

Value of flow = 28

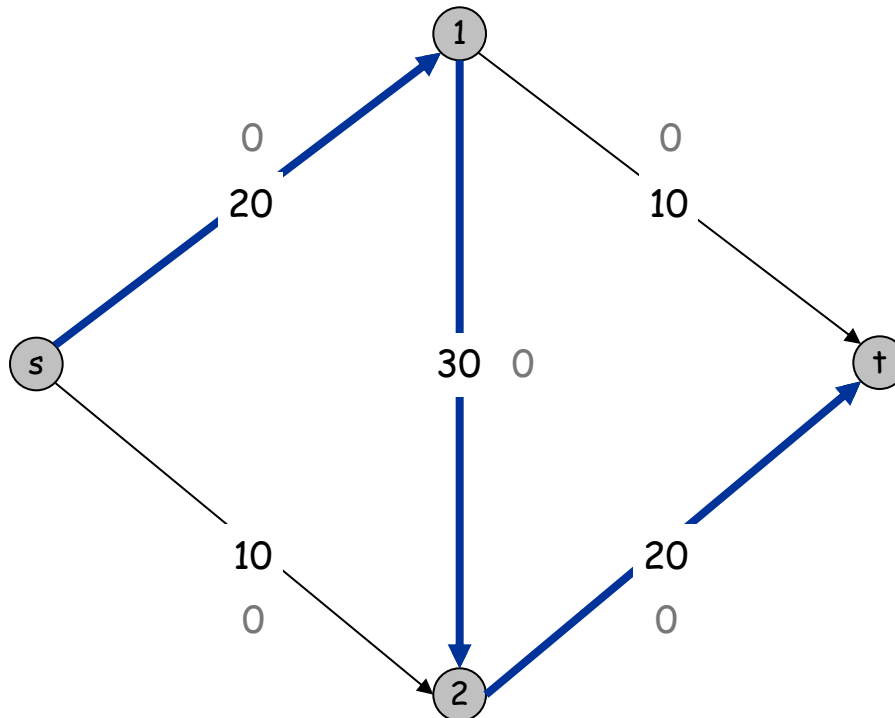
Cut capacity = 28  $\Rightarrow$  Flow value  $\leq 28$



# Towards a Max Flow Algorithm

## Greedy algorithm.

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

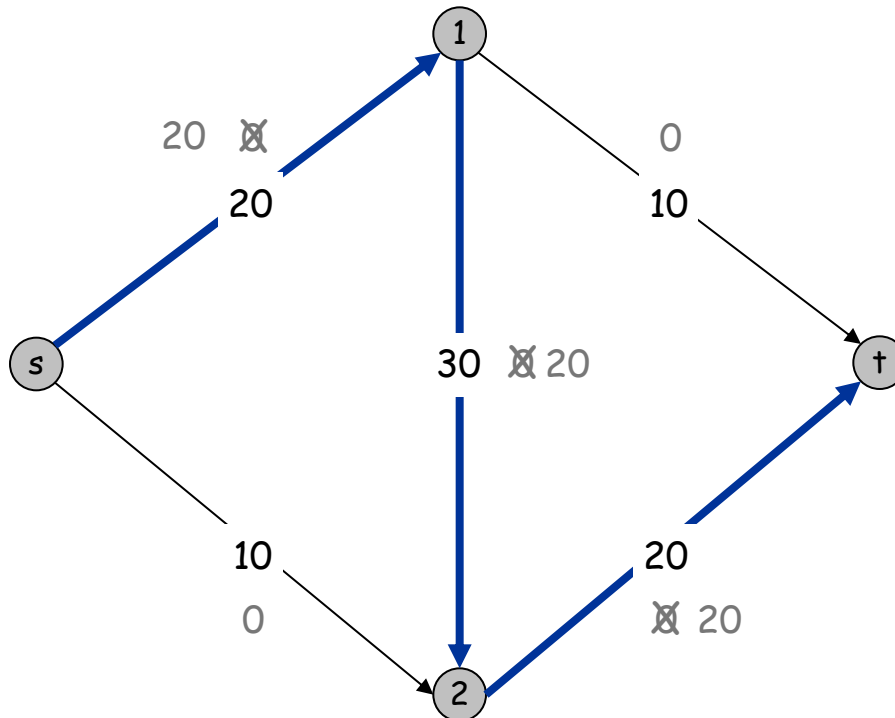


Flow value = 0

# Towards a Max Flow Algorithm

## Greedy algorithm.

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



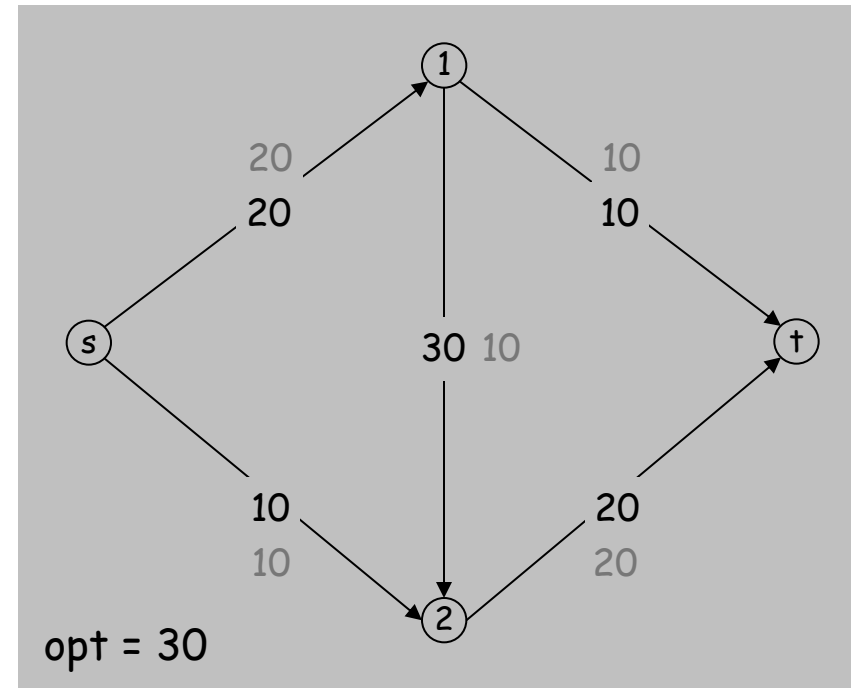
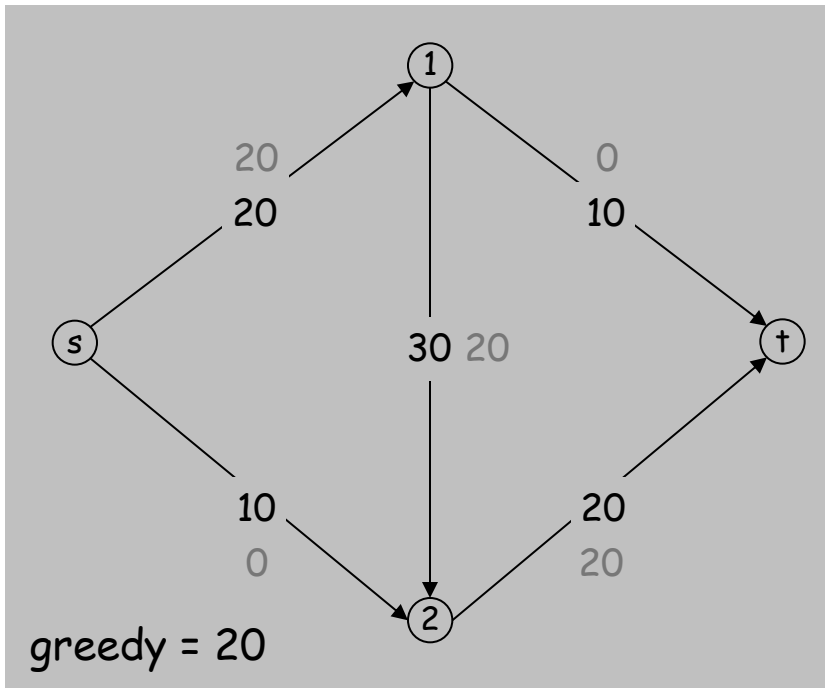
Flow value = 20

# Towards a Max Flow Algorithm

## Greedy algorithm.

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

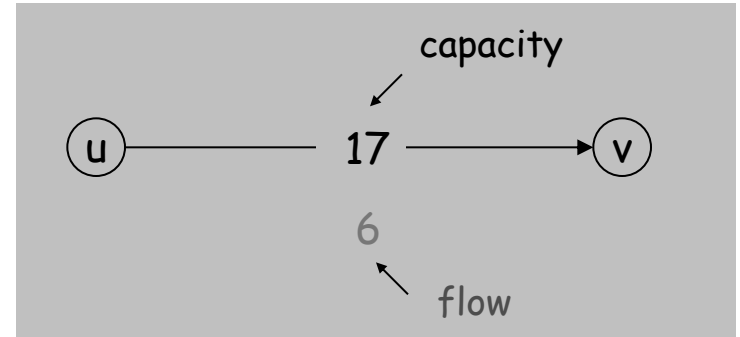
↖ locally optimality  $\nRightarrow$  global optimality



# Residual Graph

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

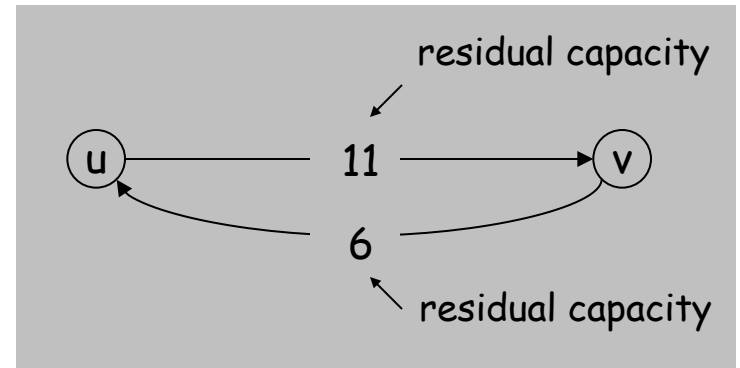
- Flow  $f(e)$ , capacity  $c(e)$ .



Residual edge.

- "Undo" flow sent.
- $e = (u, v)$  and  $e^R = (v, u)$ .
- Residual capacity:

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

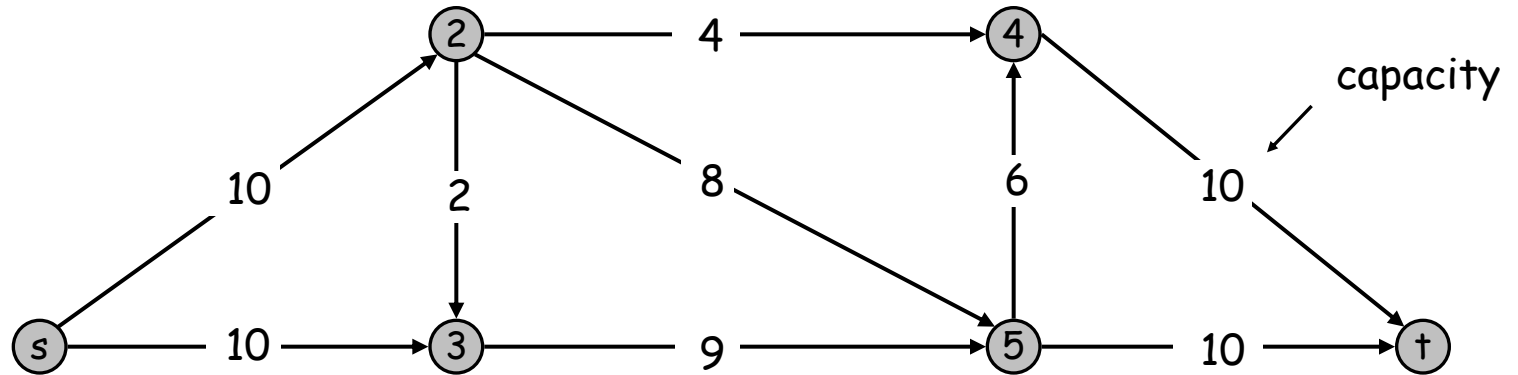


Residual graph:  $G_f = (V, E_f)$ .

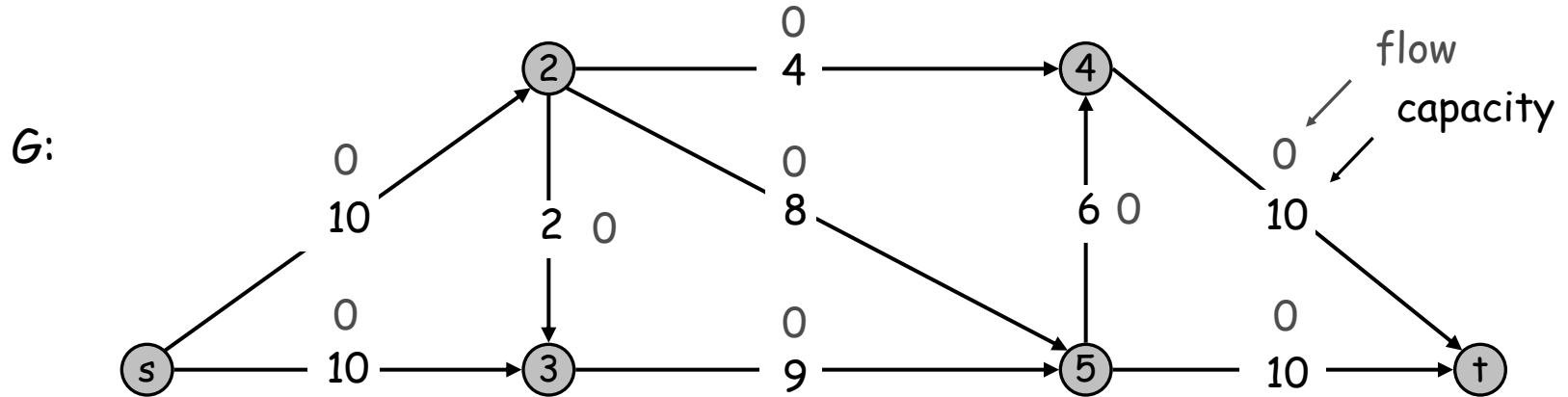
- Residual edges with positive residual capacity.
- $E_f = \{e : f(e) < c(e)\} \cup \{e^R : f(e) > 0\}$ .

# Ford-Fulkerson Algorithm

$G$ :



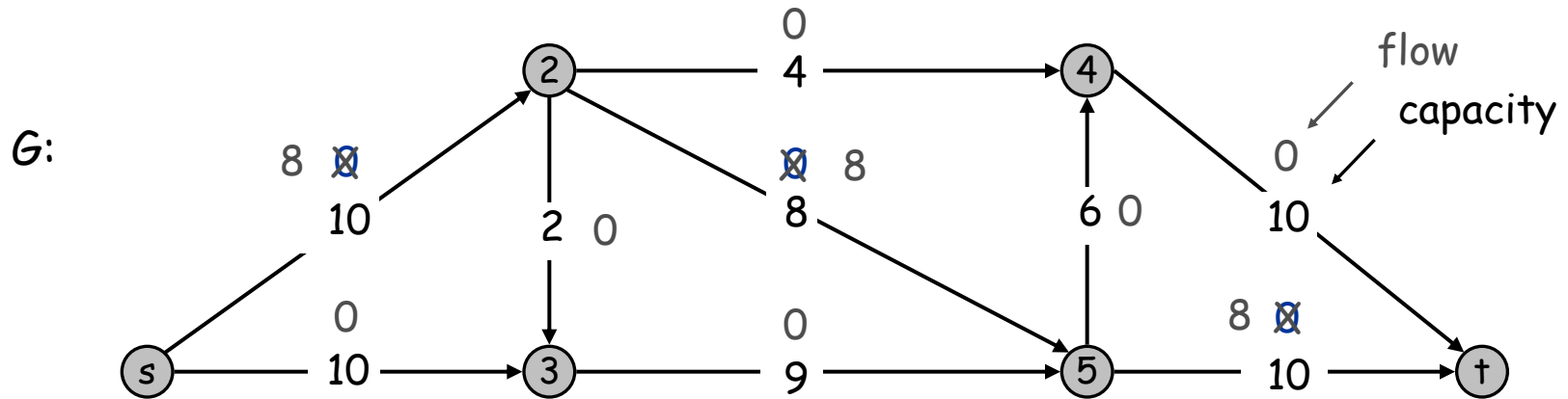
# Ford-Fulkerson Algorithm



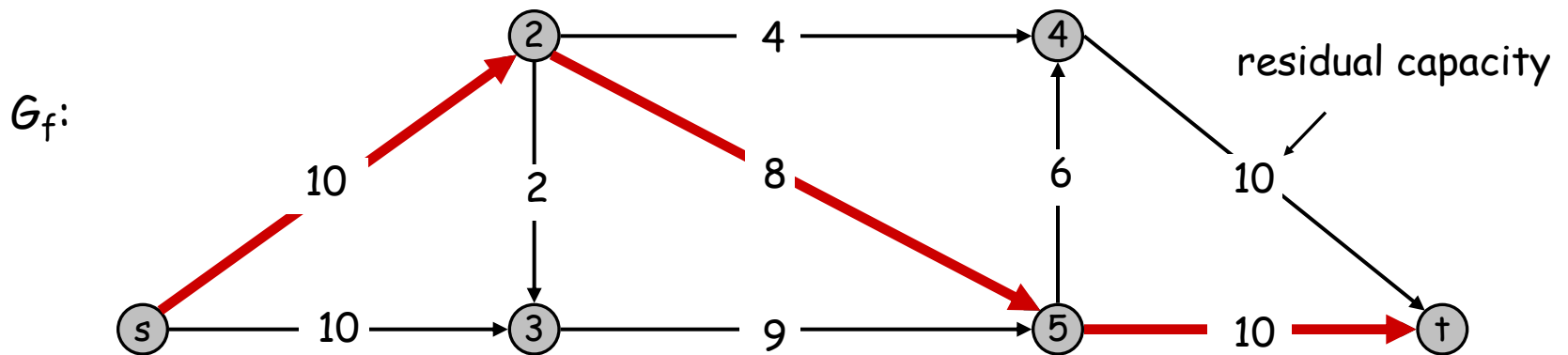
Flow value = 0



# Ford-Fulkerson Algorithm

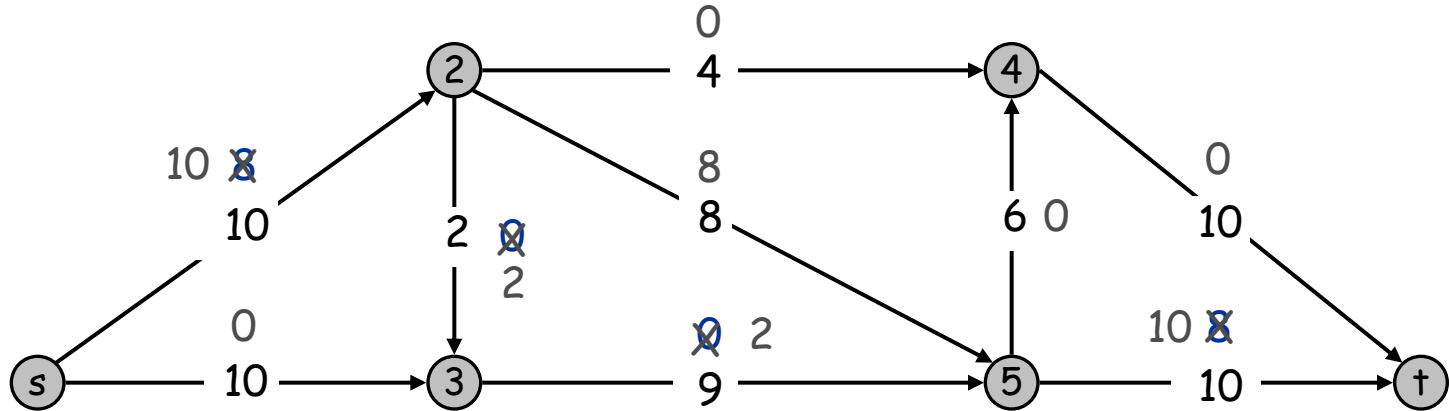


Flow value = 0



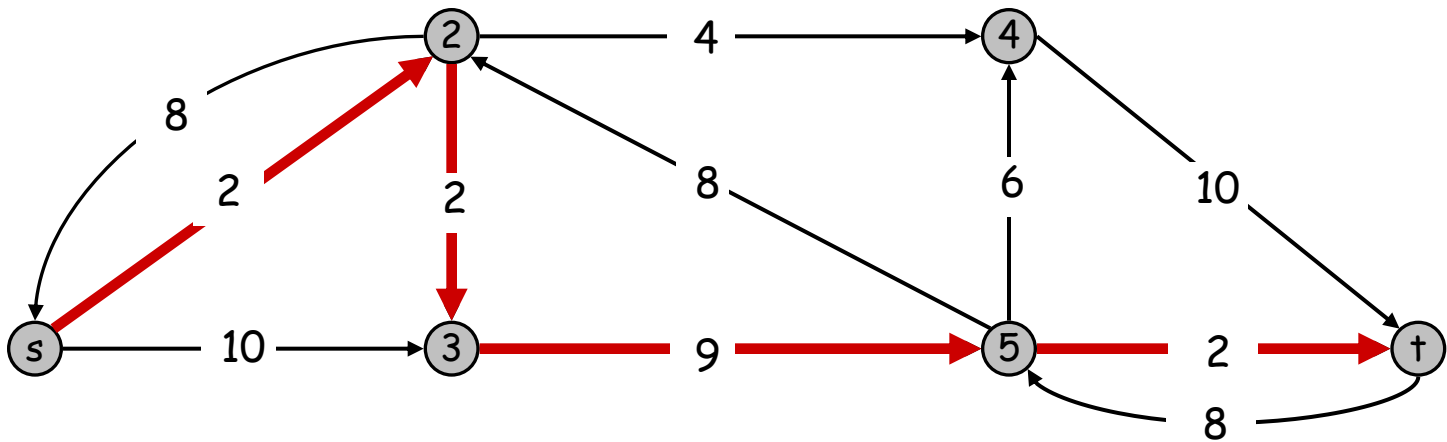
# Ford-Fulkerson Algorithm

$G$ :



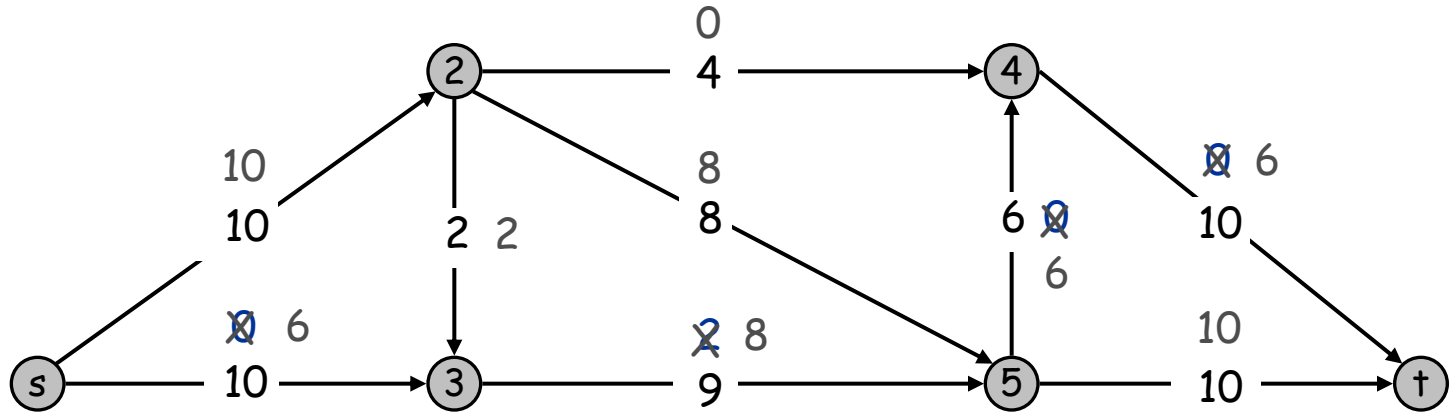
Flow value = 8

$G_f$ :



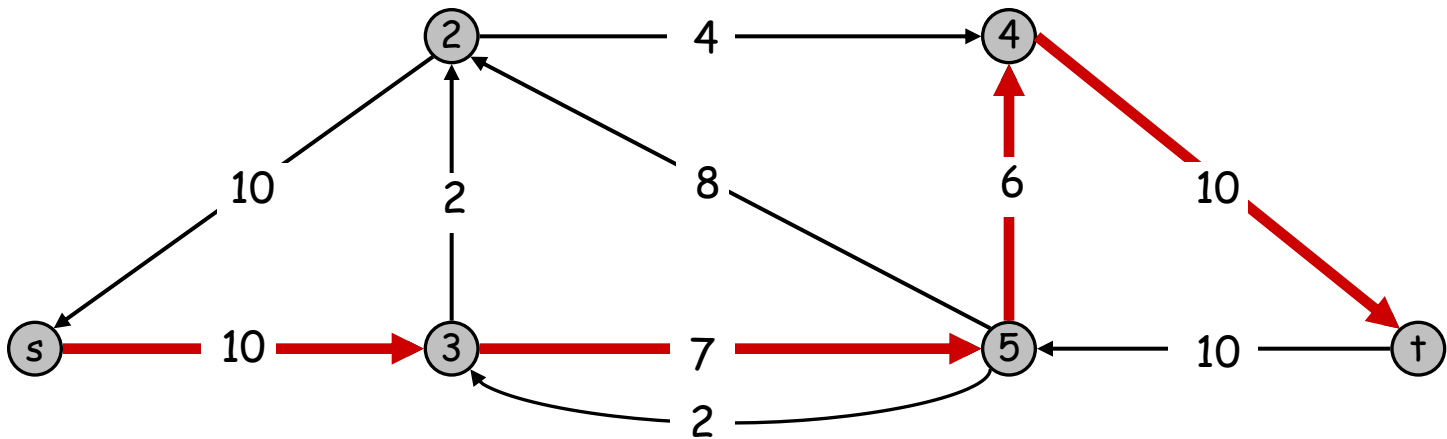
# Ford-Fulkerson Algorithm

$G$ :



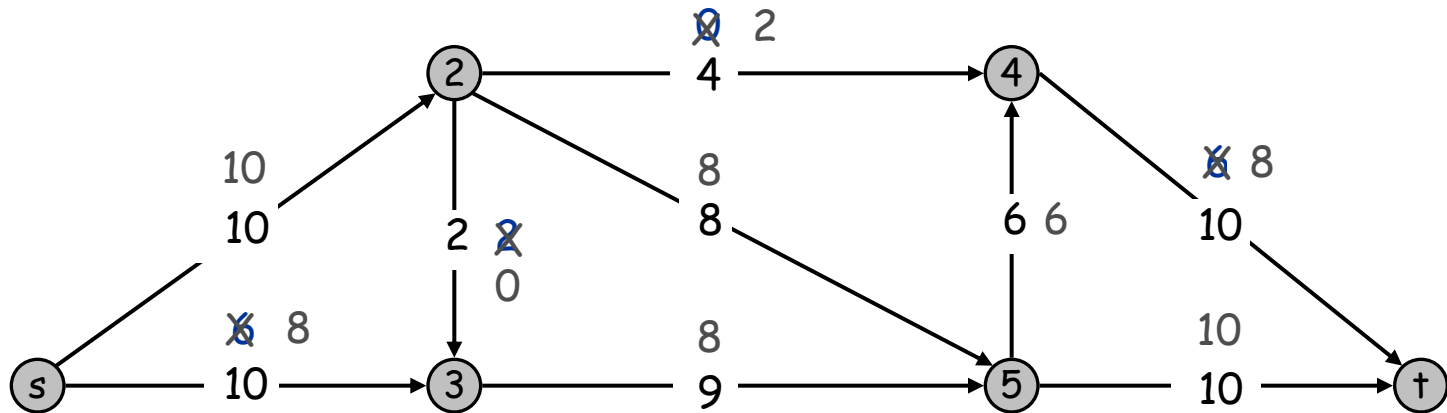
Flow value = 10

$G_f$ :



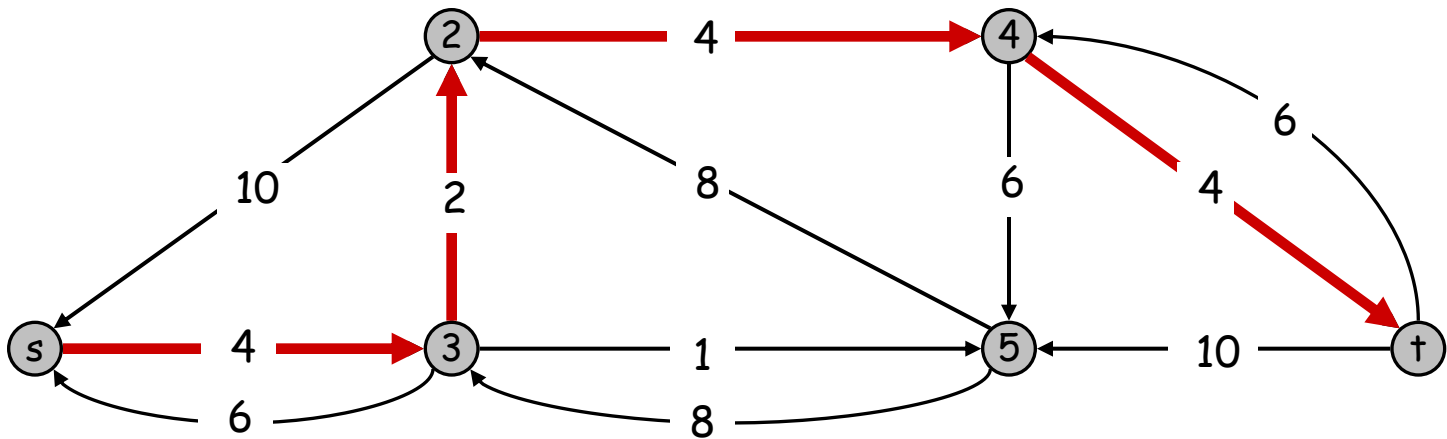
# Ford-Fulkerson Algorithm

$G$ :



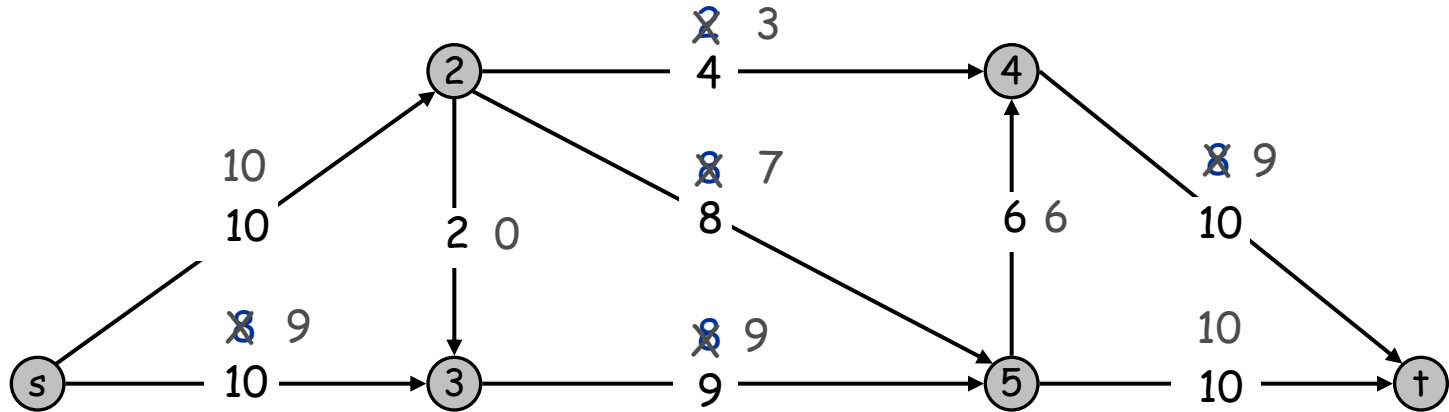
Flow value = 16

$G_f$ :



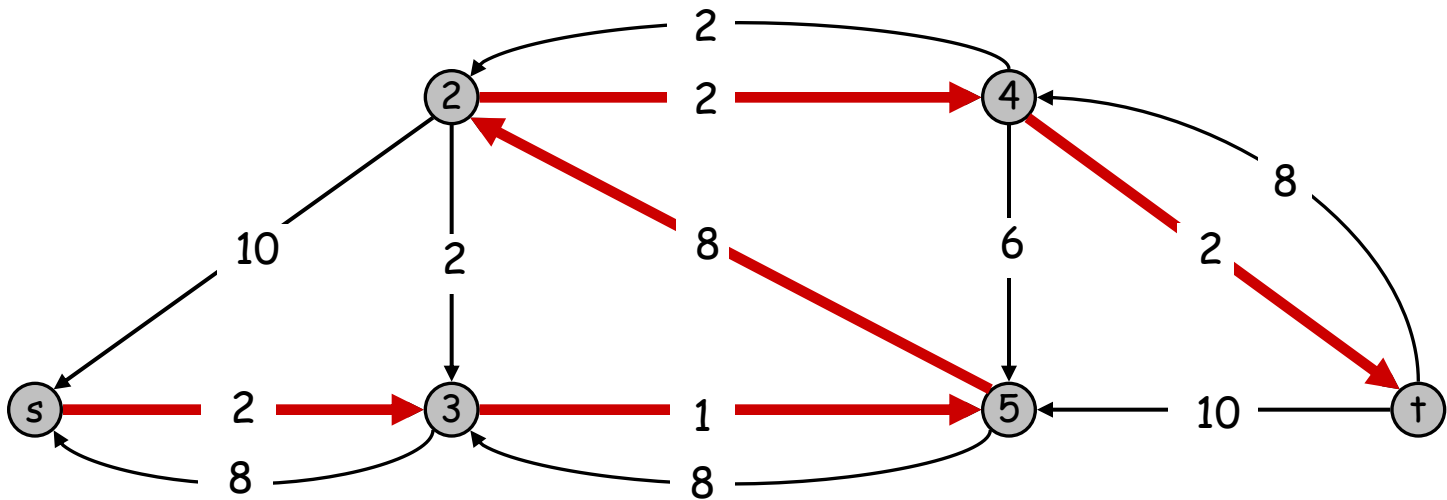
# Ford-Fulkerson Algorithm

$G$ :



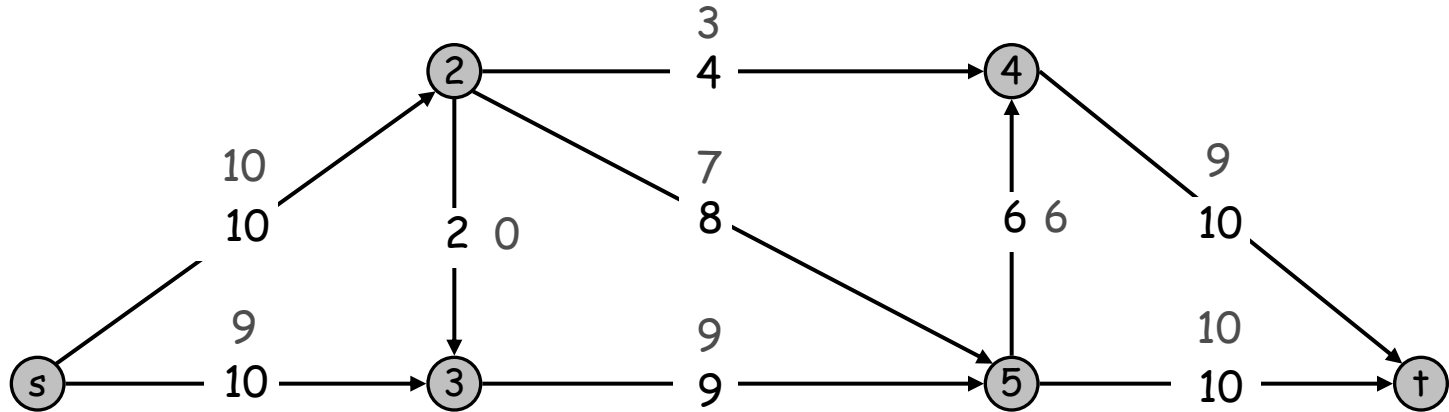
Flow value = 18

$G_f$ :



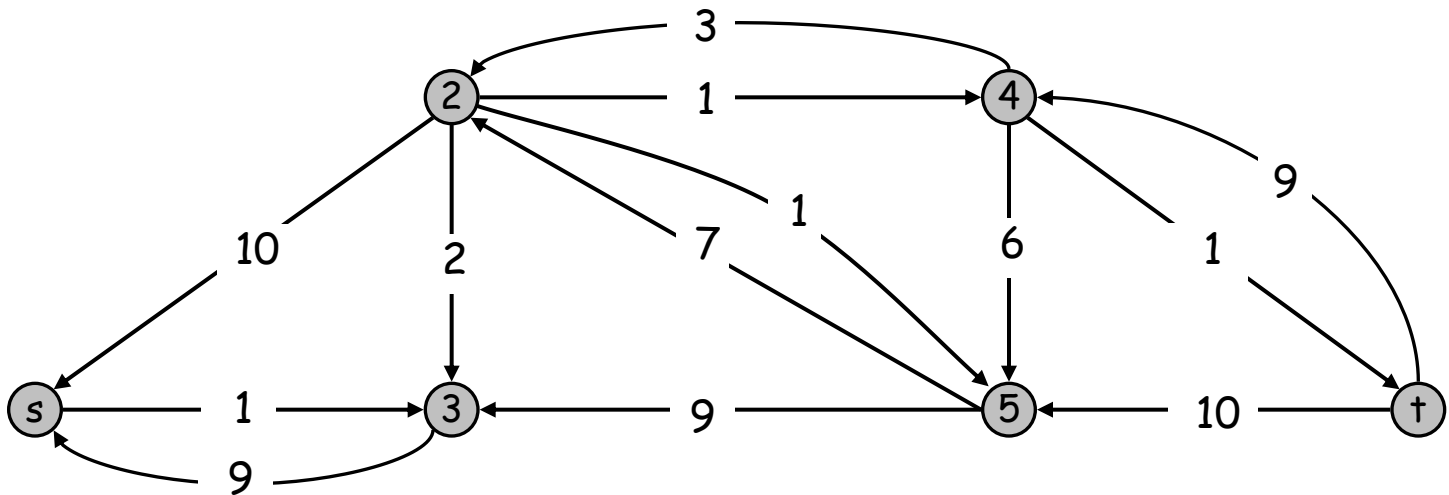
# Ford-Fulkerson Algorithm

$G$ :

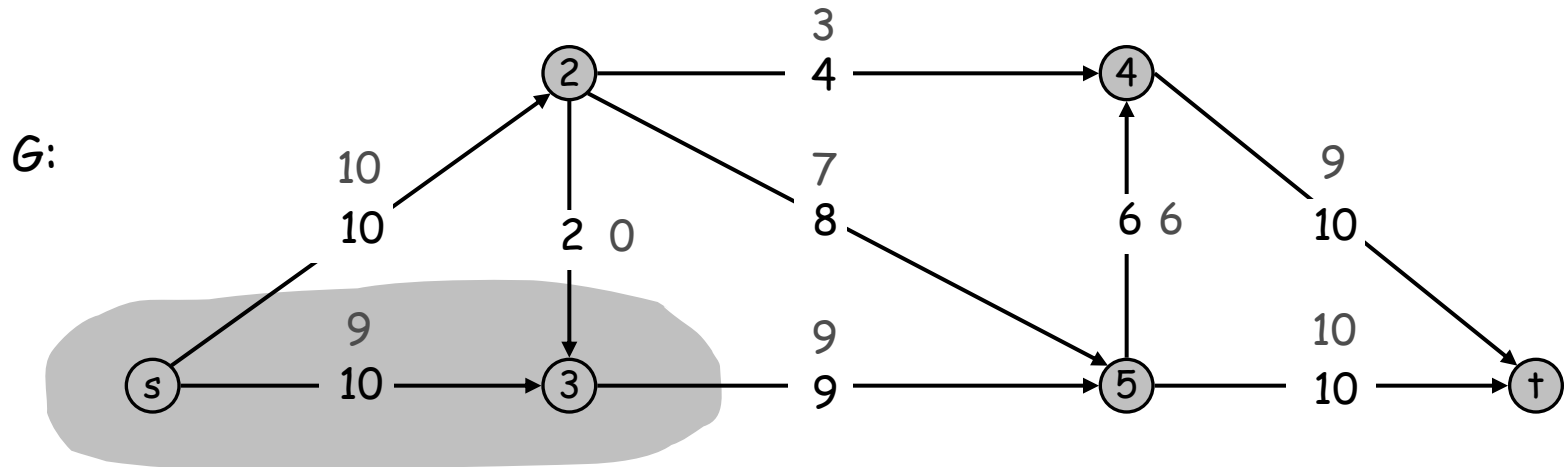


Flow value = 19

$G_f$ :

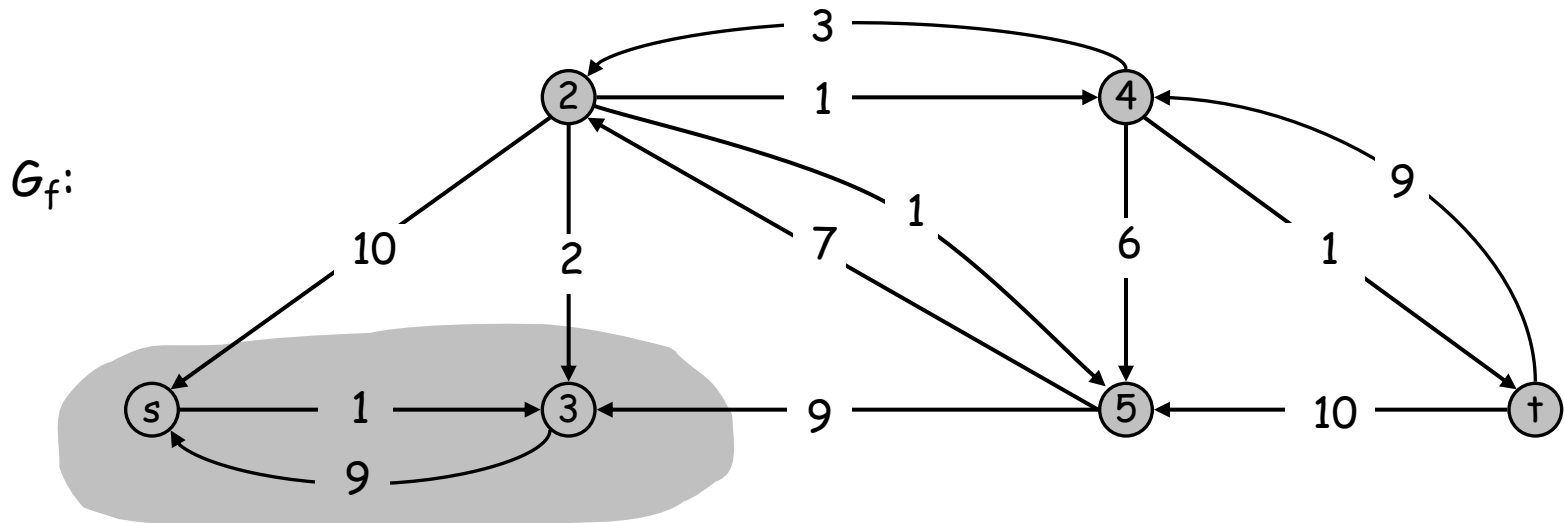


# Ford-Fulkerson Algorithm



Cut capacity = 19

Flow value = 19



# Ford-Fulkerson: A Greedy Max Flow Algorithm

```
Augment(f, c, P) {  
    b ← bottleneck(P)  
    foreach e ∈ P {  
        if (e ∈ E) f(e) ← f(e) + b  
        else      f(eR) ← f(eR) - b  
    }  
    return f  
}
```

forward edge

reverse edge

A path with non-full forward edge

Or non-zero reverse edge

```
Ford-Fulkerson(G, s, t, c) {  
    foreach e ∈ E f(e) ← 0  
    Gf ← residual graph  
  
    while (there exists augmenting path P) {  
        f ← Augment(f, c, P)  
        update Gf  
    }  
    return f  
}
```



# Max-Flow Min-Cut Theorem

**Augmenting path theorem.** Flow  $f$  is a max flow iff there are no augmenting paths.

**Max-flow min-cut theorem.** [Elias-Feinstein-Shannon 1956, Ford-Fulkerson 1956]  
The value of the max flow is equal to the value of the min cut.

**Pf.** We prove both simultaneously by showing TFAE:

- (i) There exists a cut  $(A, B)$  such that  $v(f) = \text{cap}(A, B)$ .
- (ii) Flow  $f$  is a max flow.
- (iii) There is no augmenting path relative to  $f$ .

**(i)  $\Rightarrow$  (ii)** This was the corollary to weak duality lemma.

**(ii)  $\Rightarrow$  (iii)** We show contrapositive.

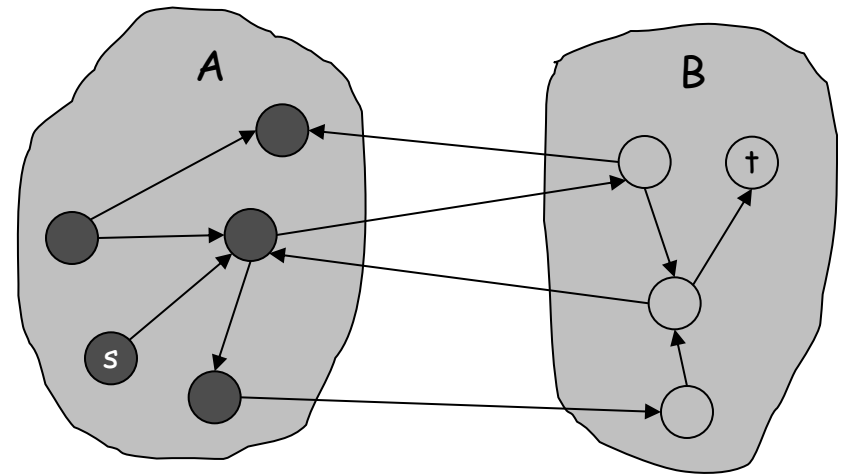
- Let  $f$  be a flow. If there exists an augmenting path, then we can improve  $f$  by sending flow along path.

# Proof of Max-Flow Min-Cut Theorem

(iii)  $\Rightarrow$  (i)

- Let  $f$  be a flow with no augmenting paths.
- Let  $A$  be set of vertices reachable from  $s$  in residual graph.
- By definition of  $A$ ,  $s \in A$ .
- By definition of  $f$ ,  $t \notin A$ .
- 

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



original network

Second equality holds since otherwise there will be backward edge.

# Running Time

**Assumption.** All capacities are integers between 1 and  $C$ .

**Invariant.** Every flow value  $f(e)$  and every residual capacity  $c_f(e)$  remains an integer throughout the algorithm.

**Theorem.** The algorithm terminates in at most  $v(f^*) \leq nC$  iterations.

**Pf.** Each augmentation increase value by at least 1. ■

**Corollary.** If  $C = 1$ , Ford-Fulkerson runs in  $O(mn)$  time.

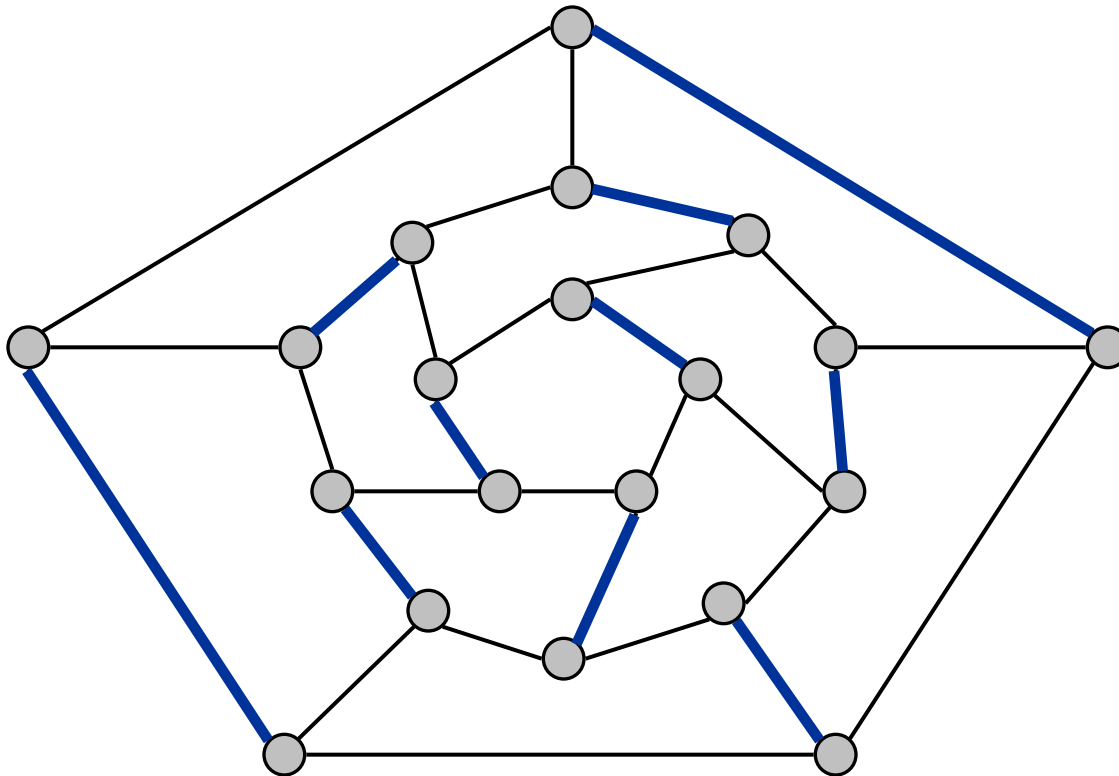
**Integrality theorem.** If all capacities are integers, then there exists a max flow  $f$  for which every flow value  $f(e)$  is an integer.

**Pf.** Since algorithm terminates, theorem follows from invariant. ■

# Matching

## Matching.

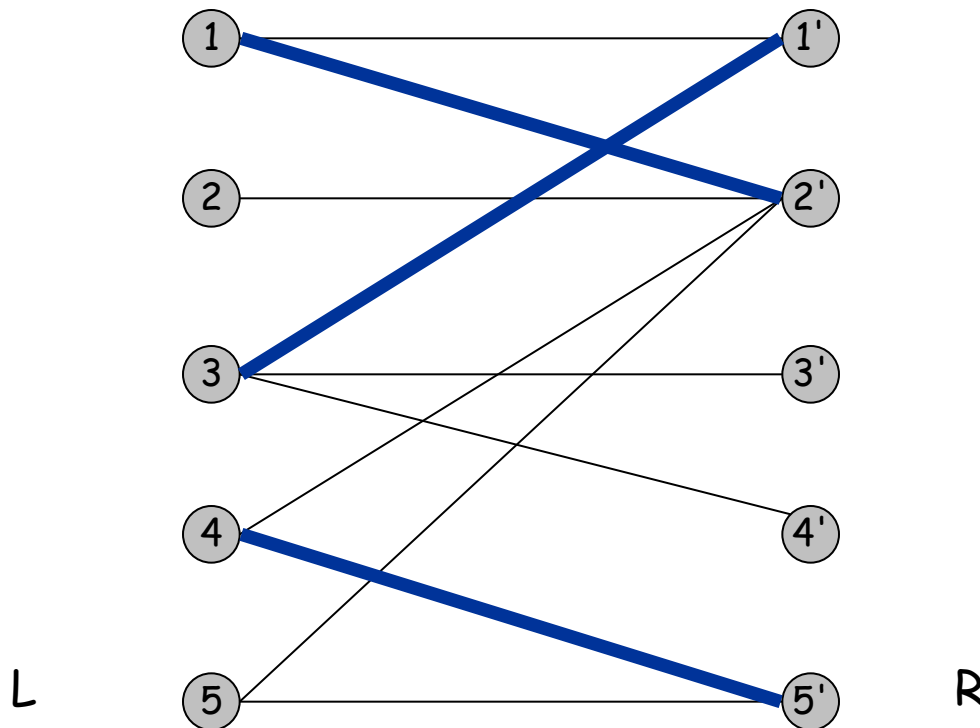
- Input: undirected graph  $G = (V, E)$ .
- $M \subseteq E$  is a **matching** if each node appears in at most edge in  $M$ .
- Max matching: find a max cardinality matching.



# Bipartite Matching

## Bipartite matching.

- Input: undirected, **bipartite** graph  $G = (L \cup R, E)$ .
- $M \subseteq E$  is a **matching** if each node appears in at most one edge in  $M$ .
- Max matching: find a max cardinality matching.

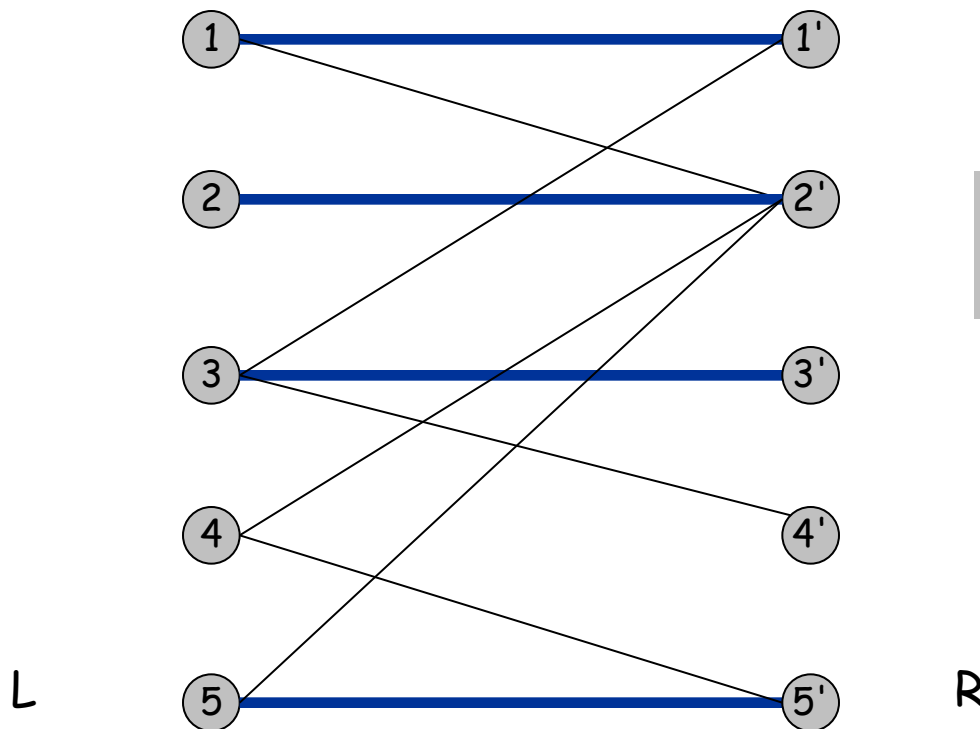


matching  
1-2', 3-1', 4-5'

# Bipartite Matching

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- Input: undirected, **bipartite** graph  $G = (L \cup R, E)$ .
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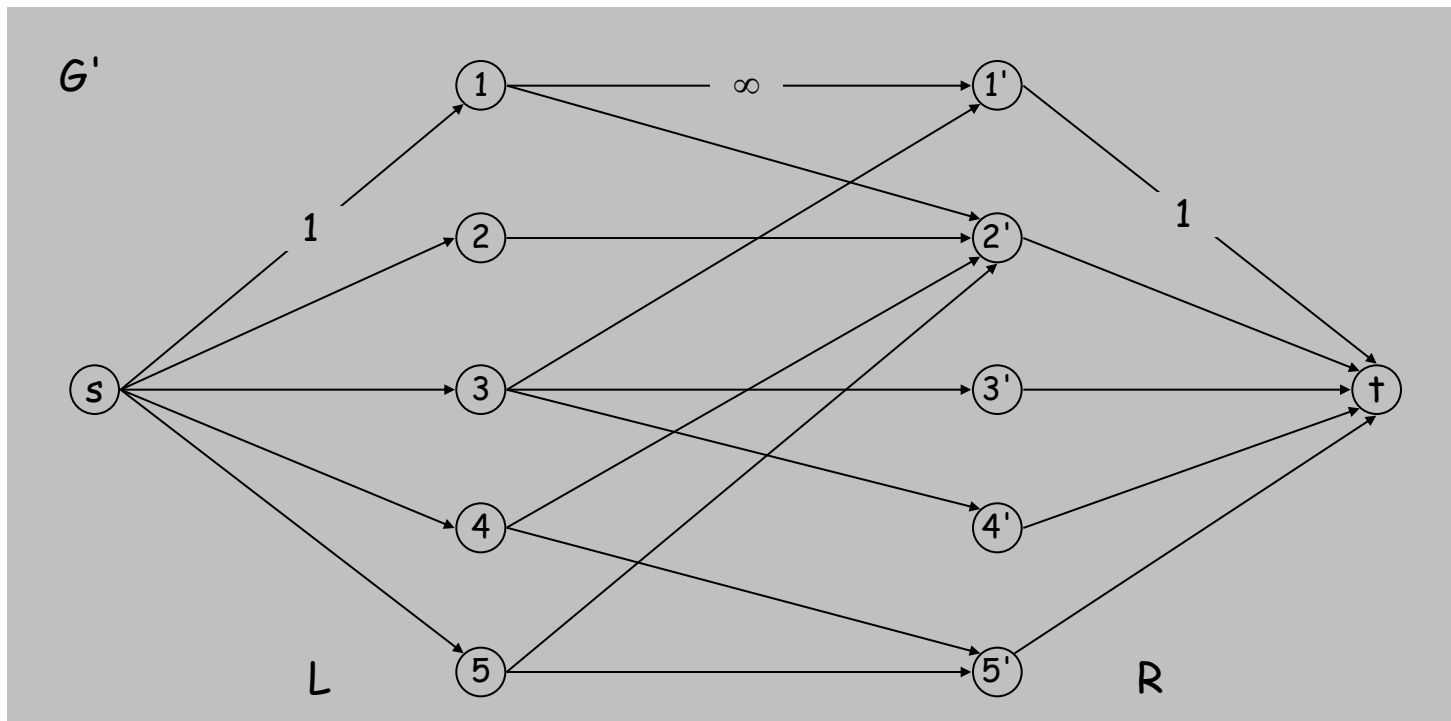


max matching  
1-1', 2-2', 3-3' 4-4'

# Bipartite Matching

## Max flow formulation.

- Create digraph  $G' = (L \cup R \cup \{s, t\}, E')$ .
- Direct all edges from  $L$  to  $R$ , and assign infinite (or unit) capacity.
- Add source  $s$ , and unit capacity edges from  $s$  to each node in  $L$ .
- Add sink  $t$ , and unit capacity edges from each node in  $R$  to  $t$ .

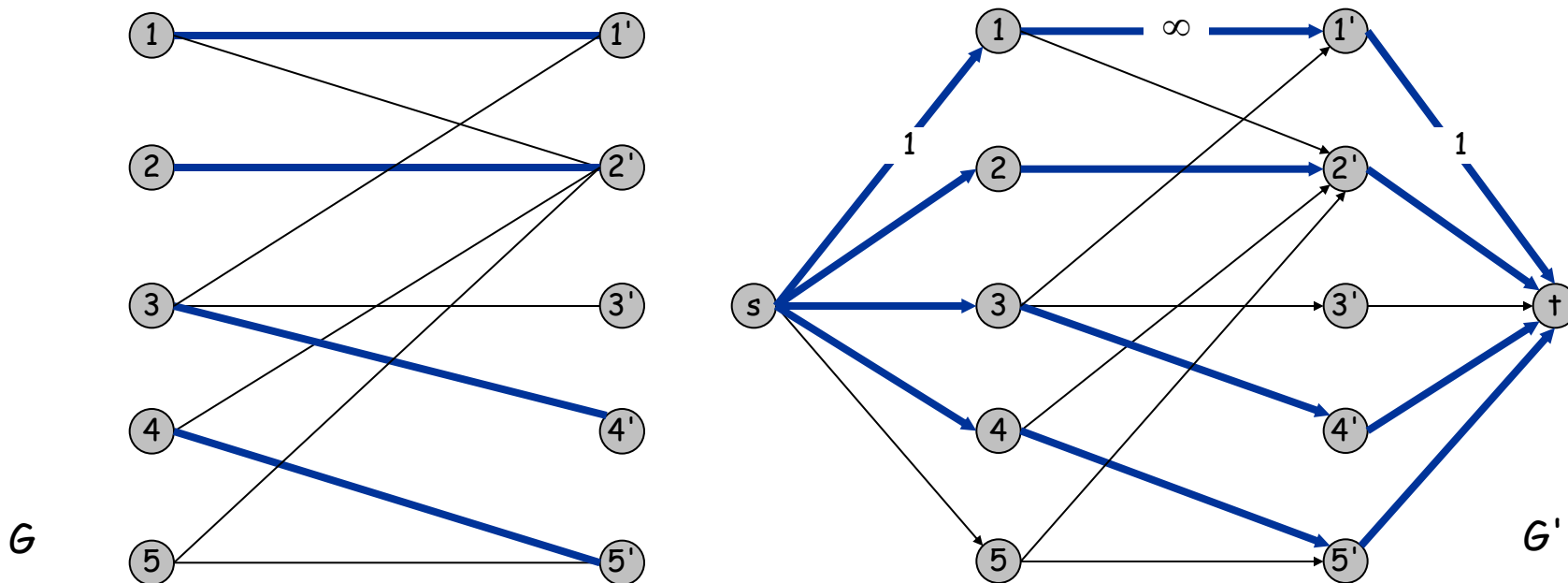


# Bipartite Matching: Proof of Correctness

**Theorem.** Max cardinality matching in  $G$  = value of max flow in  $G'$ .

**Pf.**  $\leq$

- Given max matching  $M$  of cardinality  $k$ .
- Consider flow  $f$  that sends 1 unit along each of  $k$  paths.
- $f$  is a flow, and has cardinality  $k$ . ▪



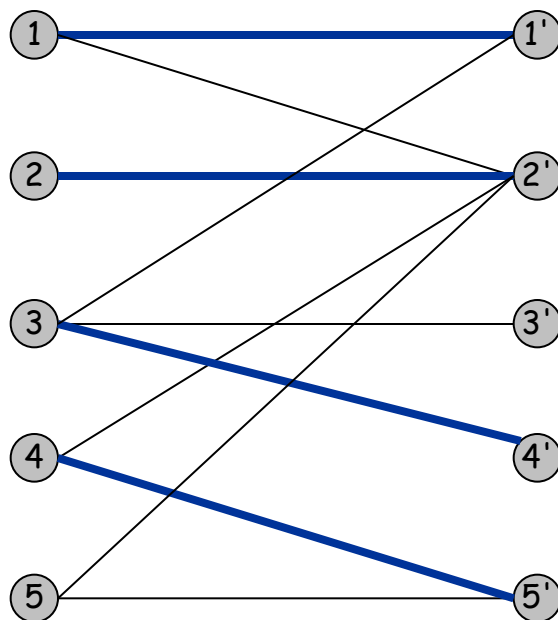
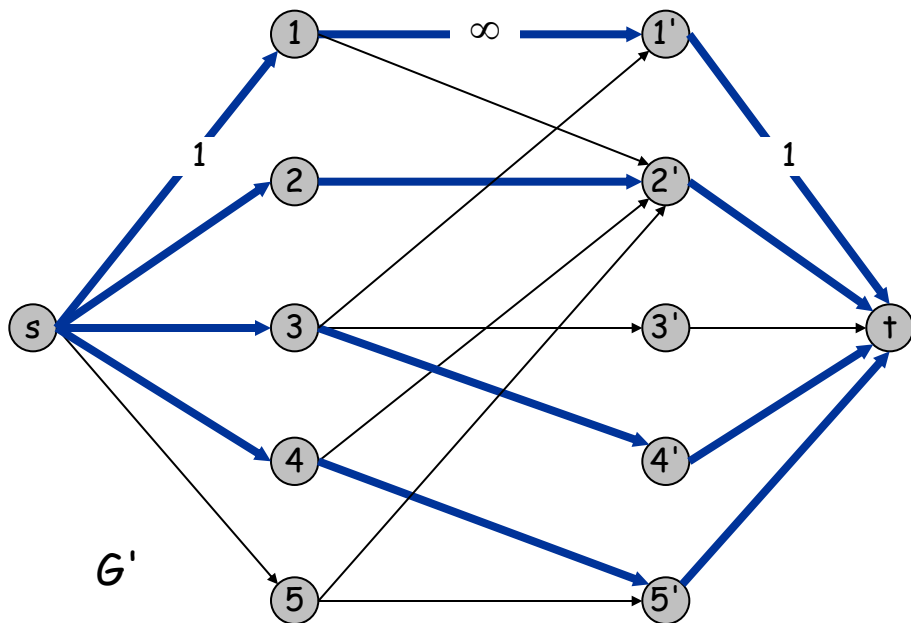


# Bipartite Matching: Proof of Correctness

**Theorem.** Max cardinality matching in  $G$  = value of max flow in  $G'$ .

**Pf.**  $\geq$

- Let  $f$  be a max flow in  $G'$  of value  $k$ .
- Integrality theorem  $\Rightarrow$   $k$  is integral and can assume  $f$  is 0-1.
- Consider  $M$  = set of edges from  $L$  to  $R$  with  $f(e) = 1$ .
  - each node in  $L$  and  $R$  participates in at most one edge in  $M$
  - $|M| = k$ : consider cut  $(L \cup s, R \cup t)$  ▪



## Related Questions

<https://leetcode.com/problems/maximum-students-taking-exam/>