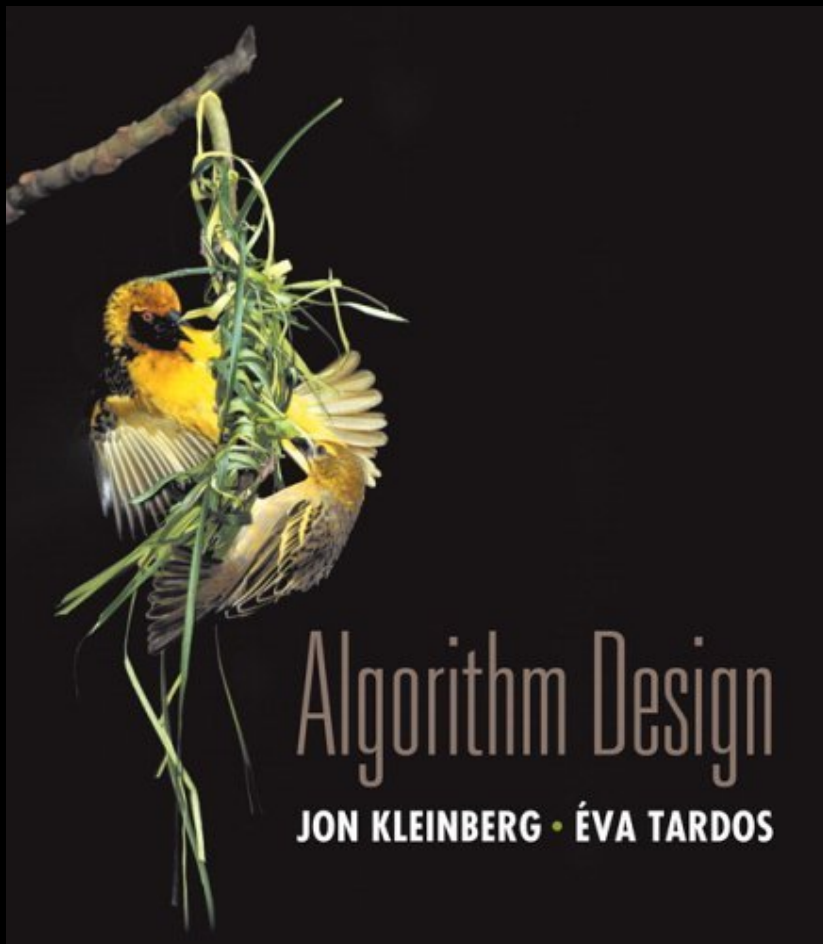


# Chapter 7

## Network Flow

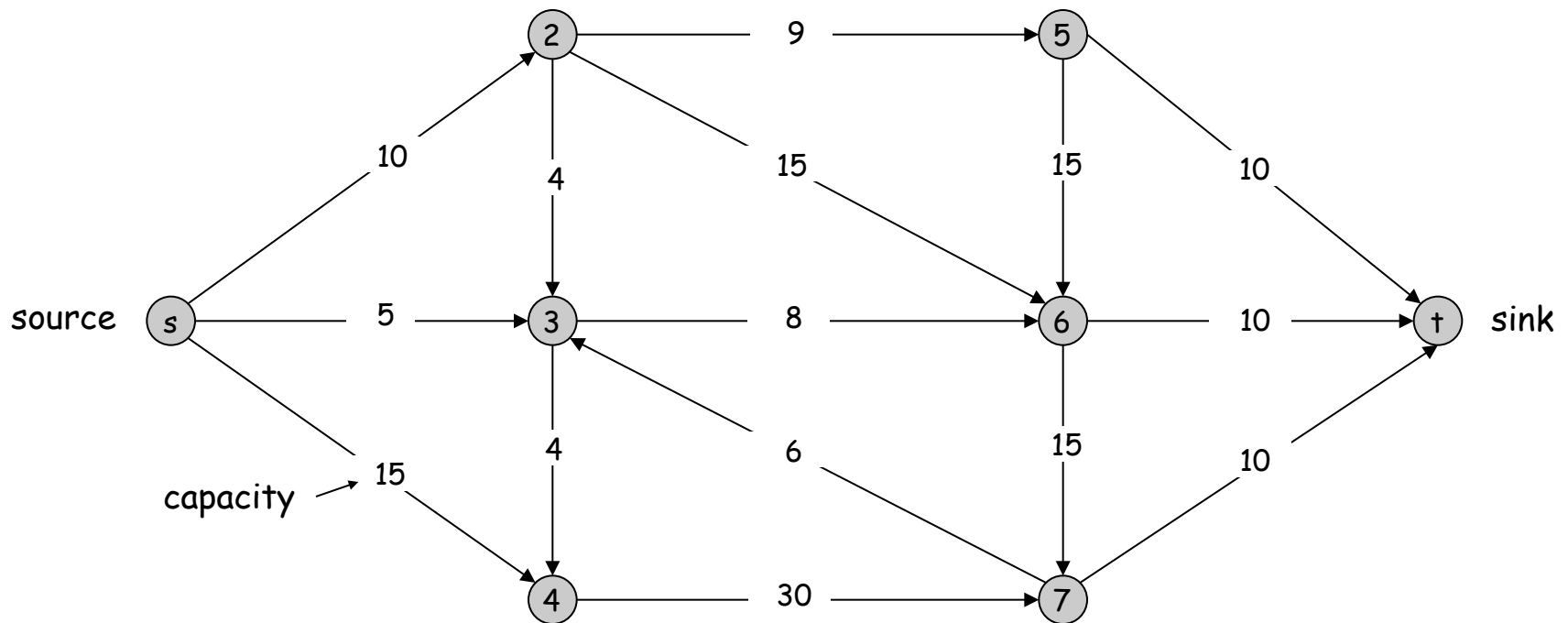


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# Minimum Cut Problem

## 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  $e$ .

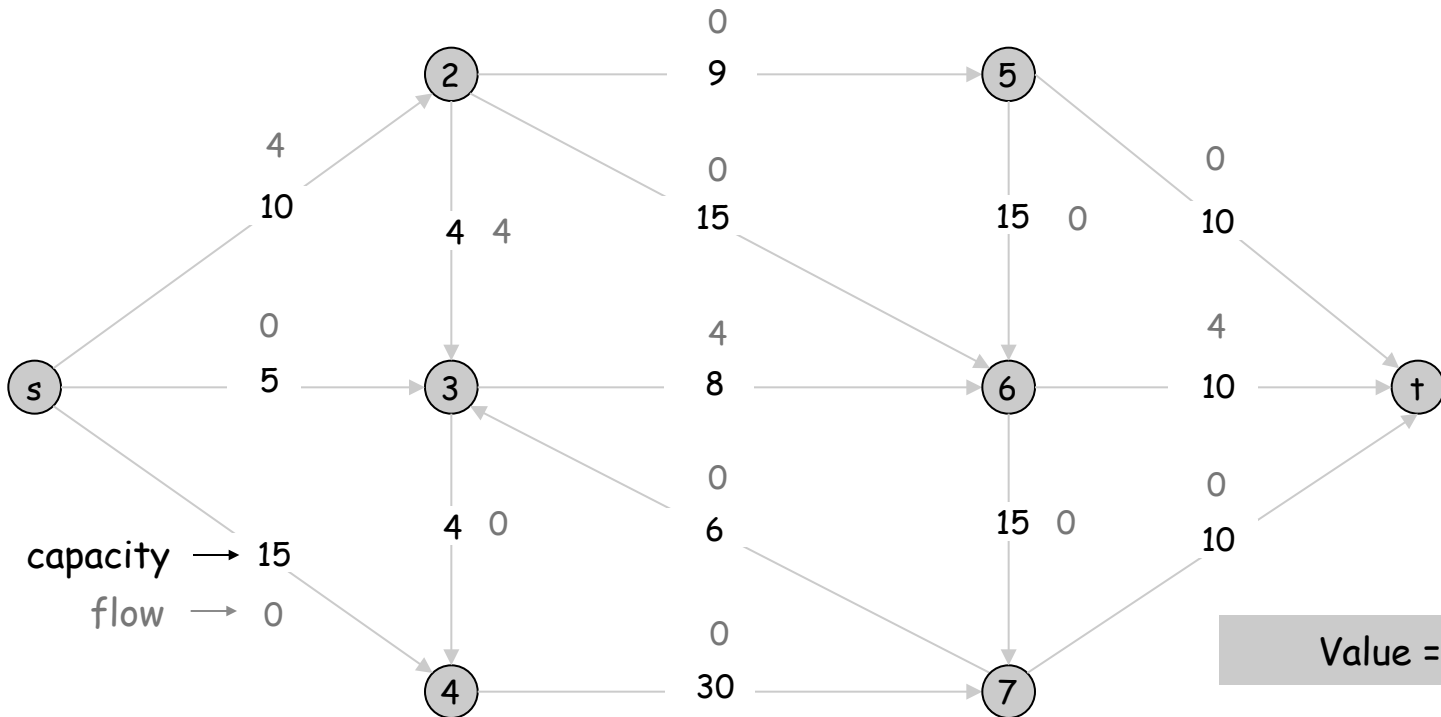


# Flows

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

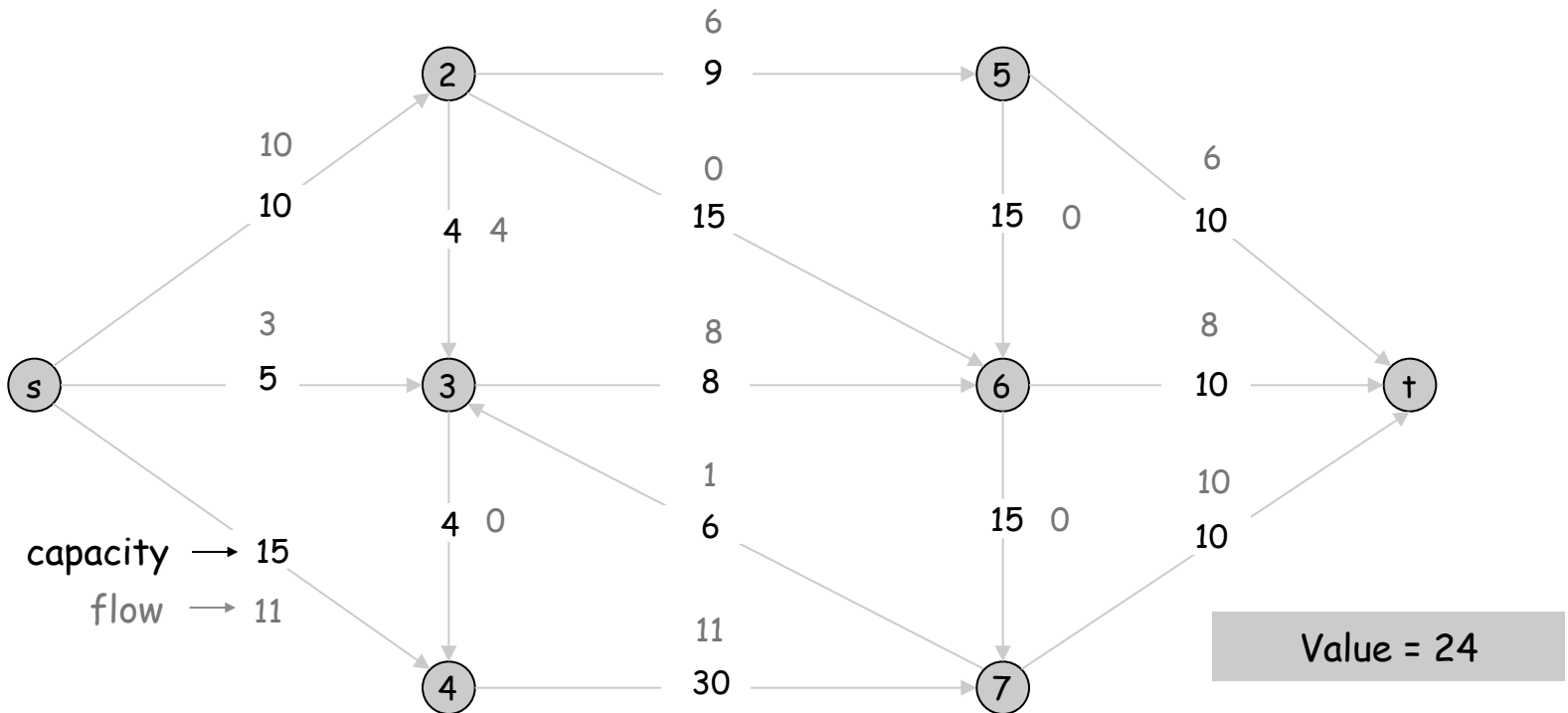


# Flows

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

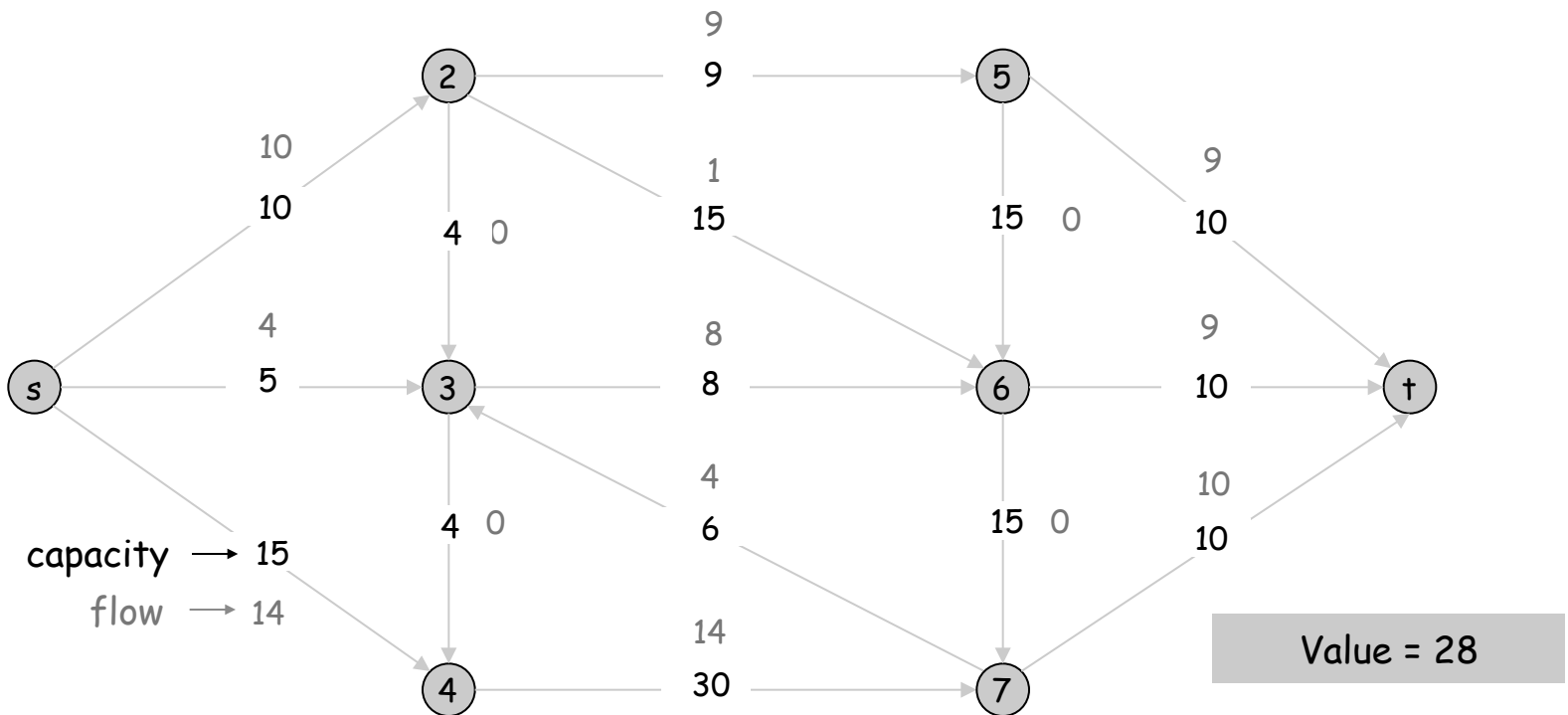
- For each  $e \in E$ :  $0 \leq f(e) \leq c(e)$  (capacity)
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# Maximum Flow Problem

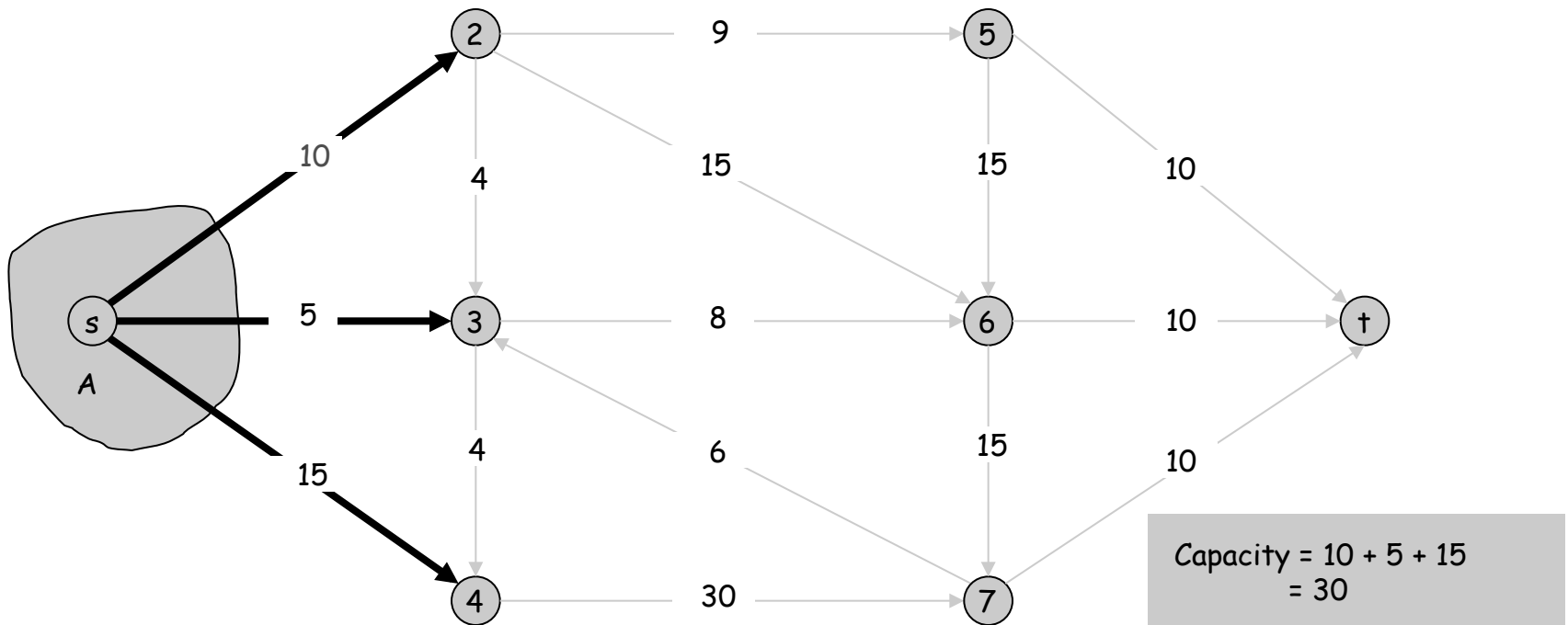
Max flow problem. Find s-t flow of maximum value.



# Cuts

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

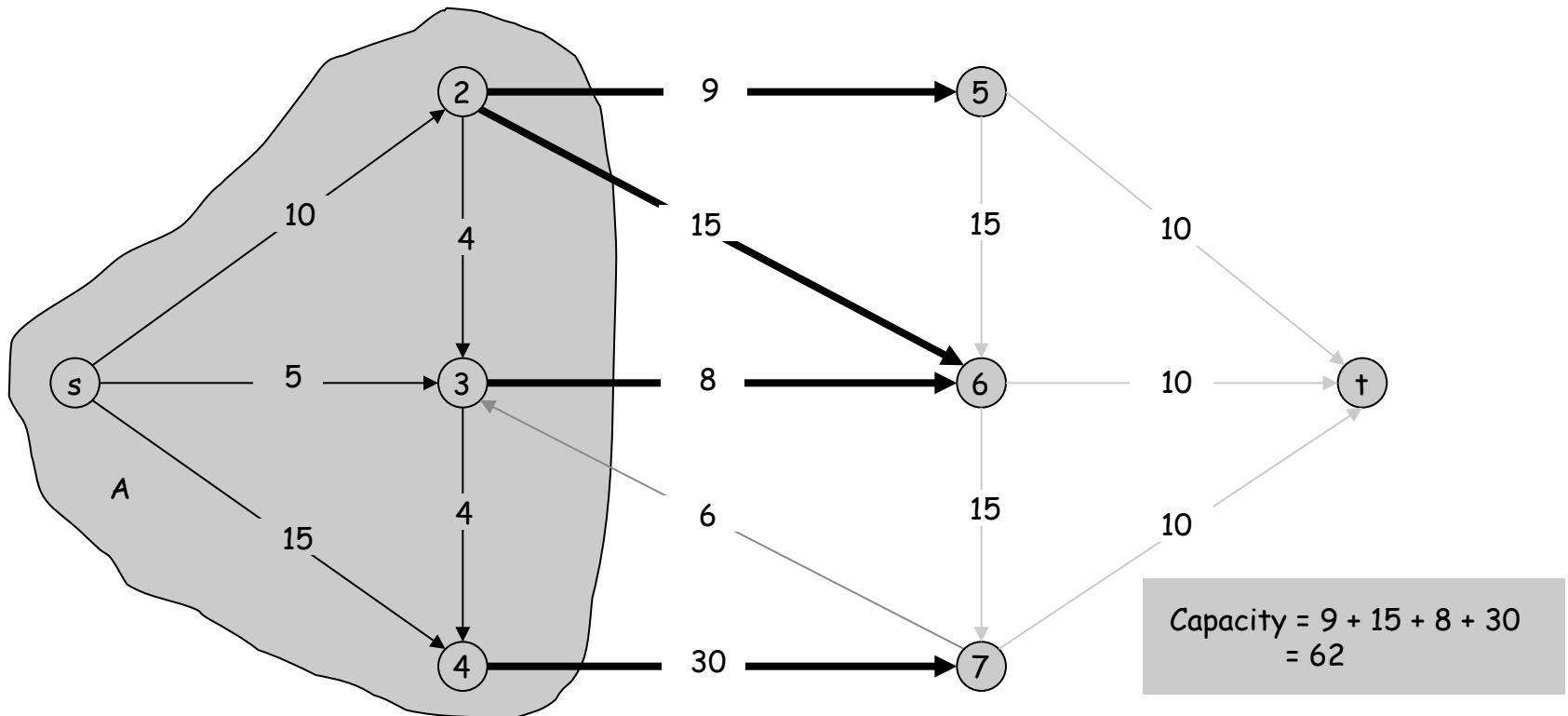
Def. The **capacity** of a cut  $(A, B)$  is:  $cap(A, B) = \sum_{e \text{ out of } A} c(e)$



# Cuts

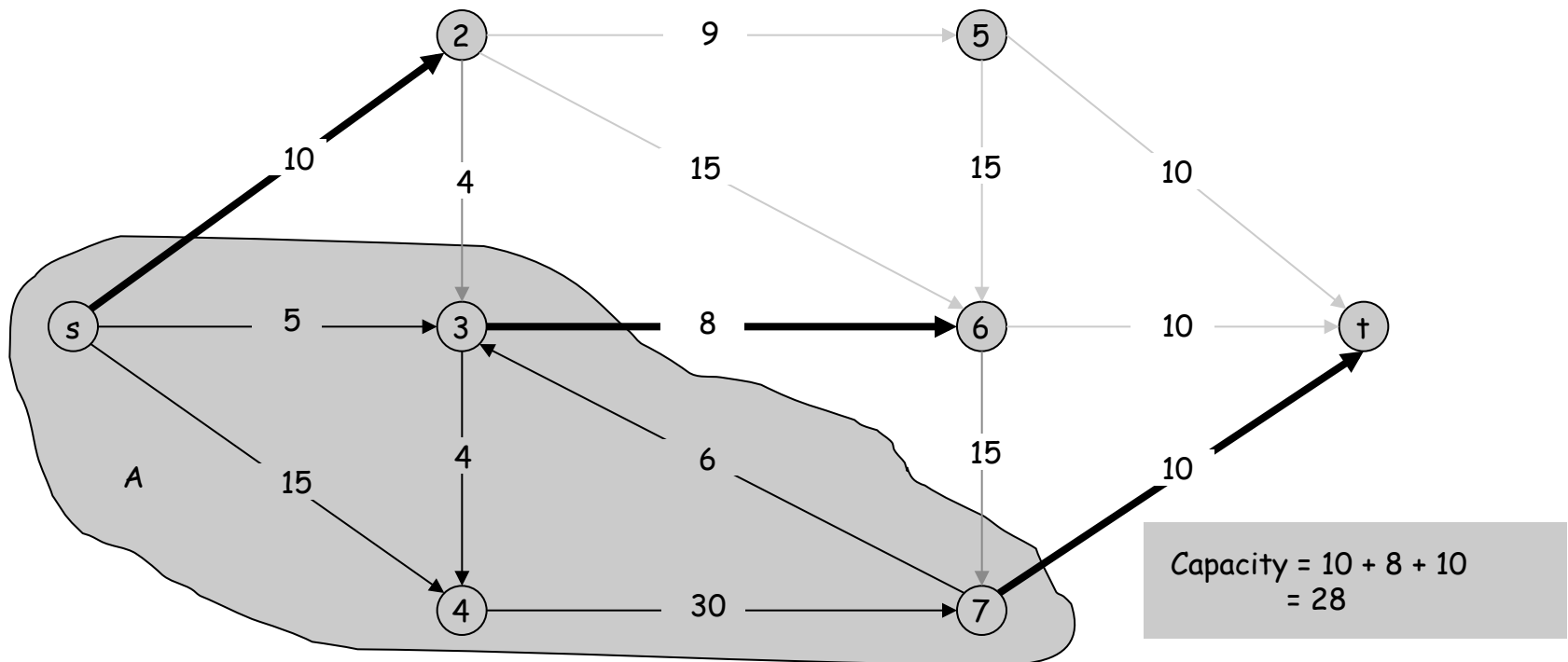
Def. An **s-t cut** is a partition  $(A, B)$  of  $V$  with  $s \in A$  and  $t \in B$ .

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.

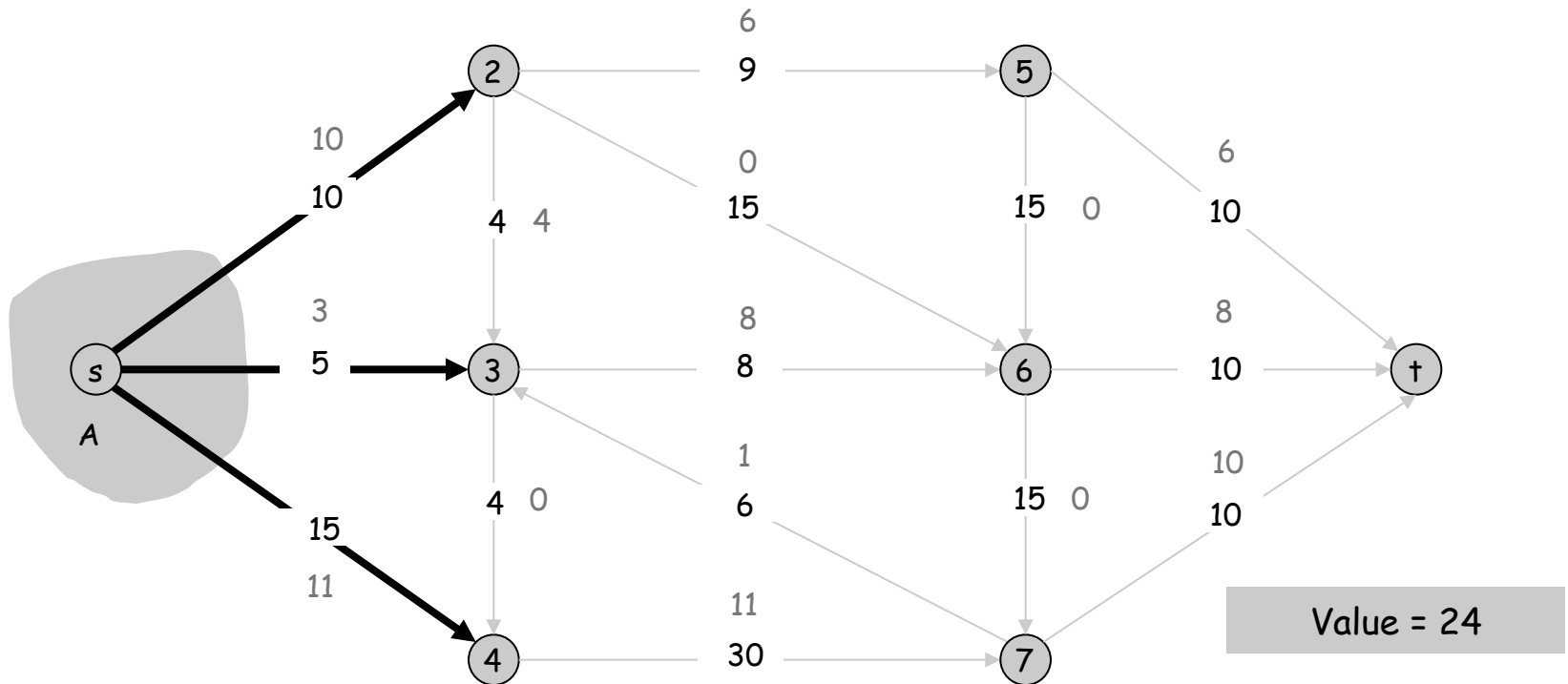




# Flows and Cuts

**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  $s$ .

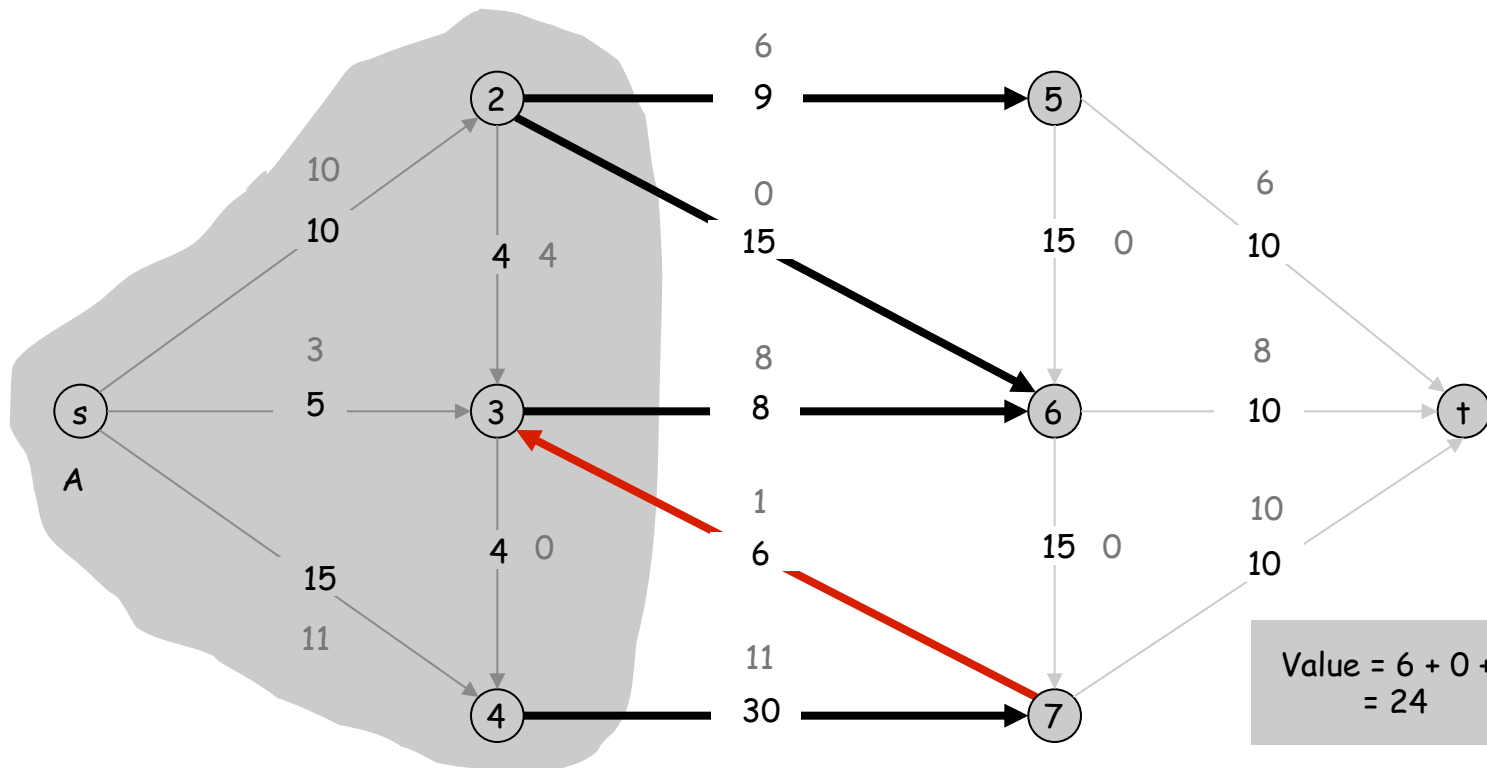
$$\sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e) = v(f)$$



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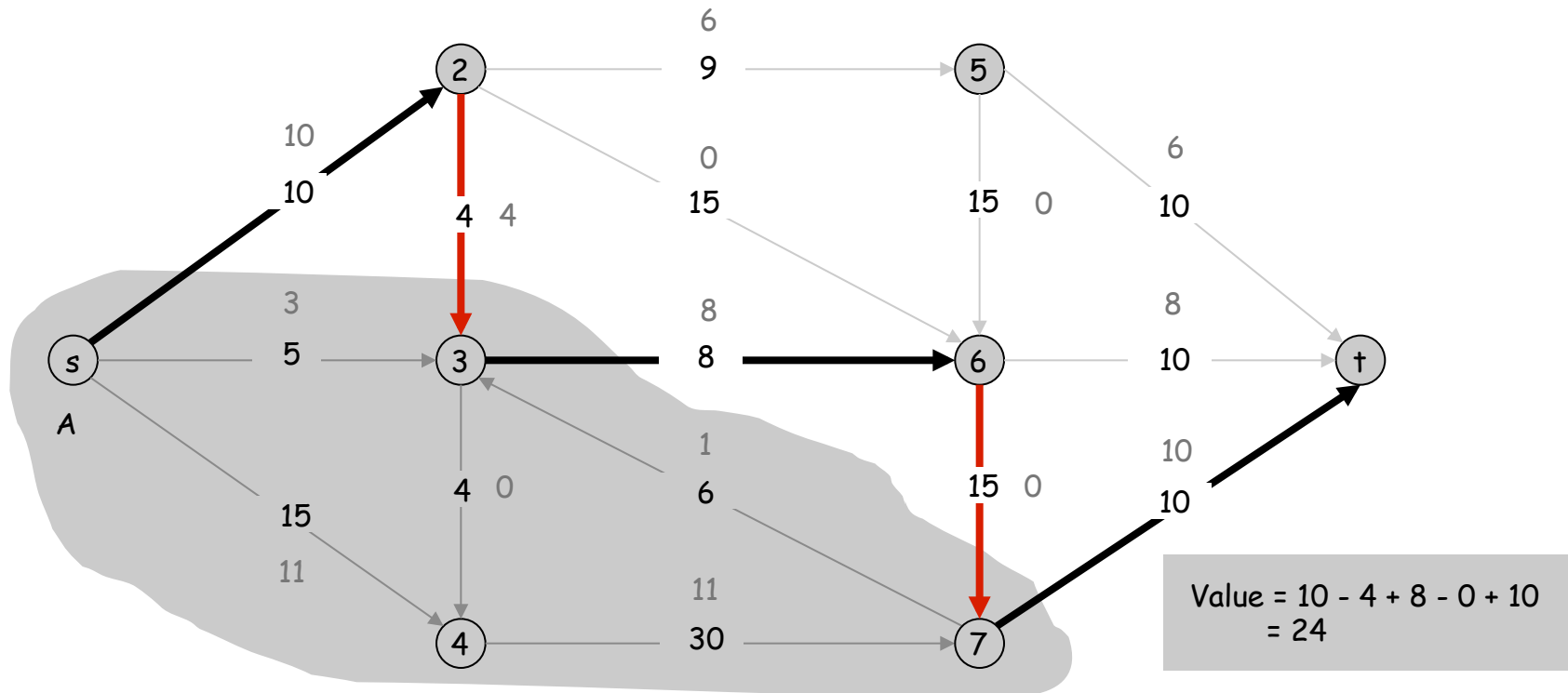
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# Flows and Cuts

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# 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.**

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

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

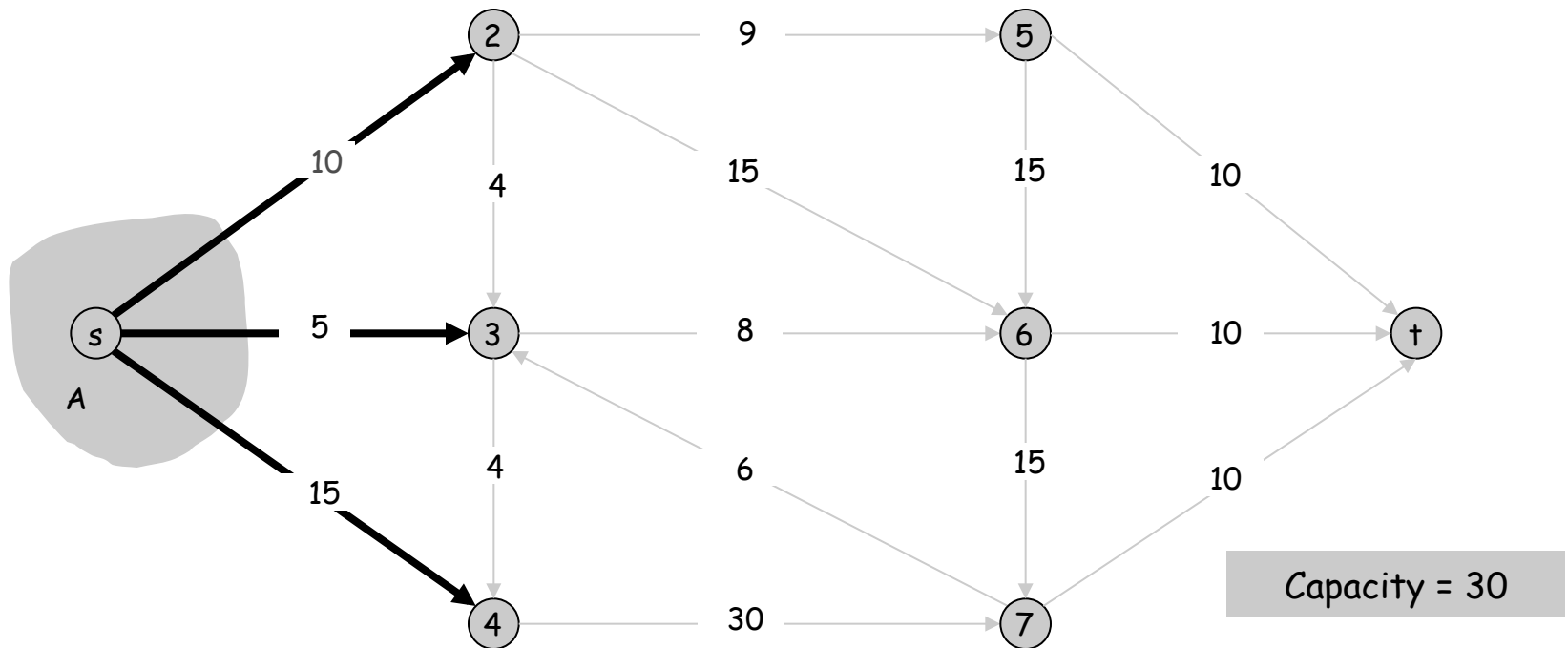
$$\longrightarrow = \sum_{v \in A} \left( \sum_{e \text{ out of } v} f(e) - \sum_{e \text{ in to } v} f(e) \right)$$

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

# 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  $\leq 30$

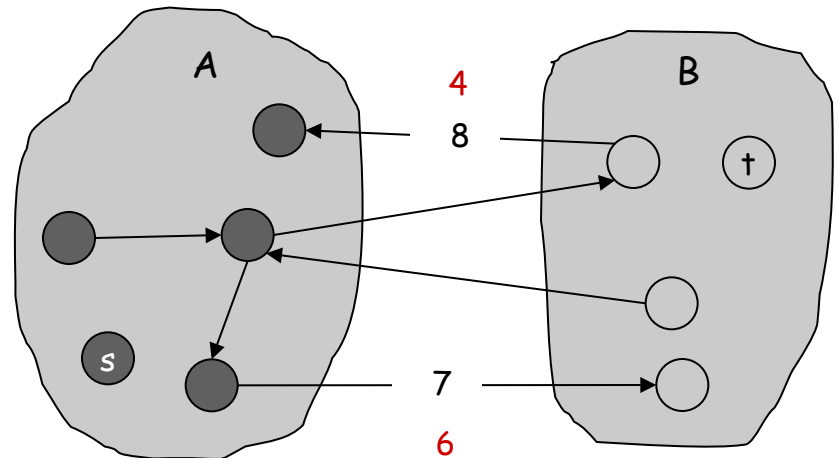


# Flows and Cuts

**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.**

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

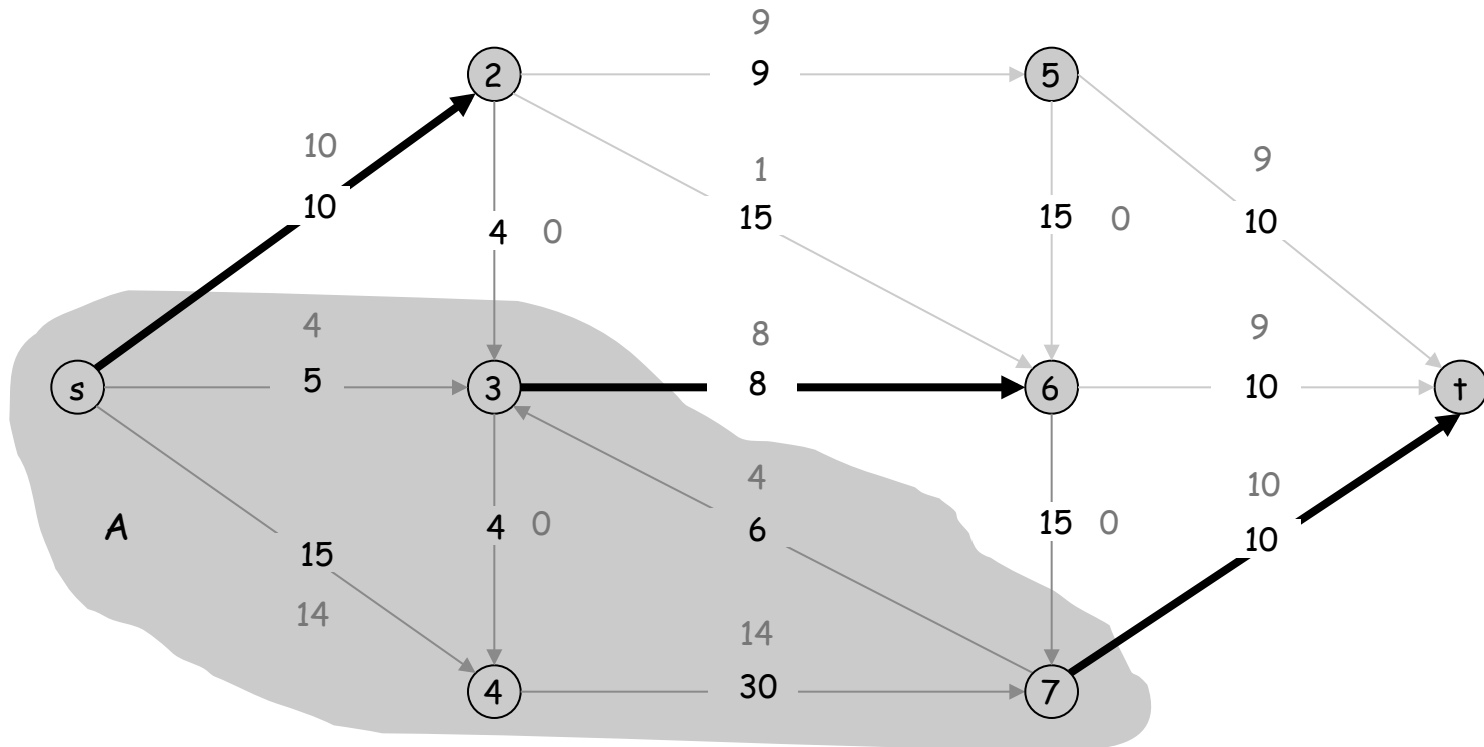


# Certificate of Optimality

**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

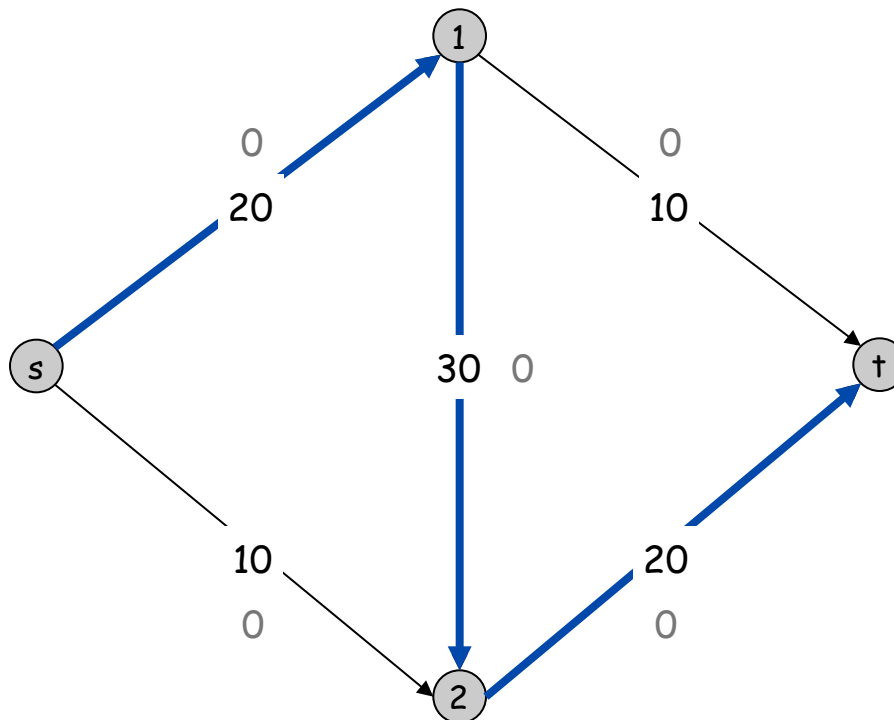
Cut capacity = 28  $\Rightarrow$  Flow value  $\leq 28$



# 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.



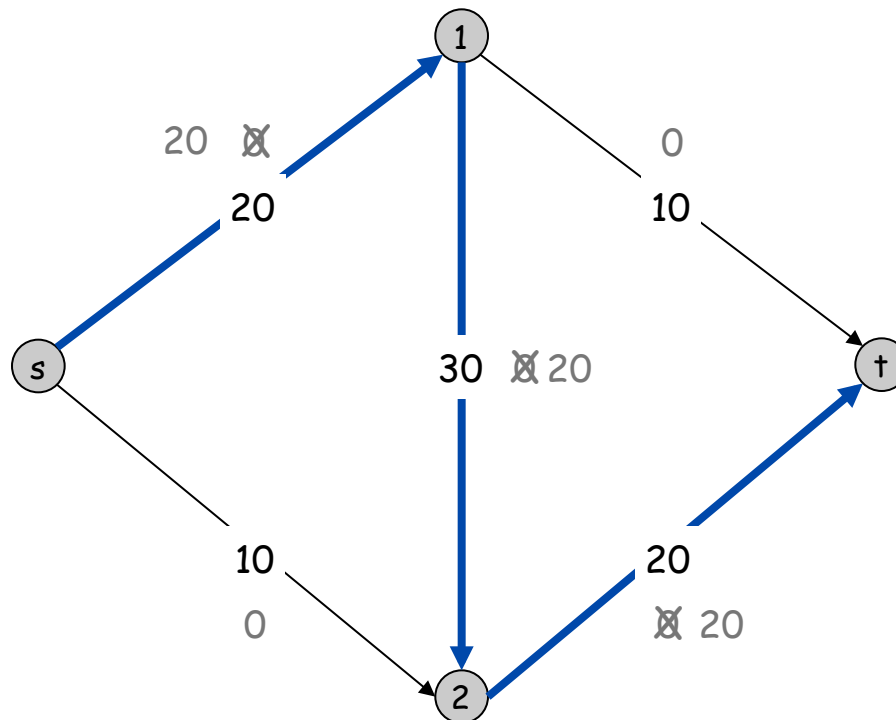
Flow value = 0



# 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

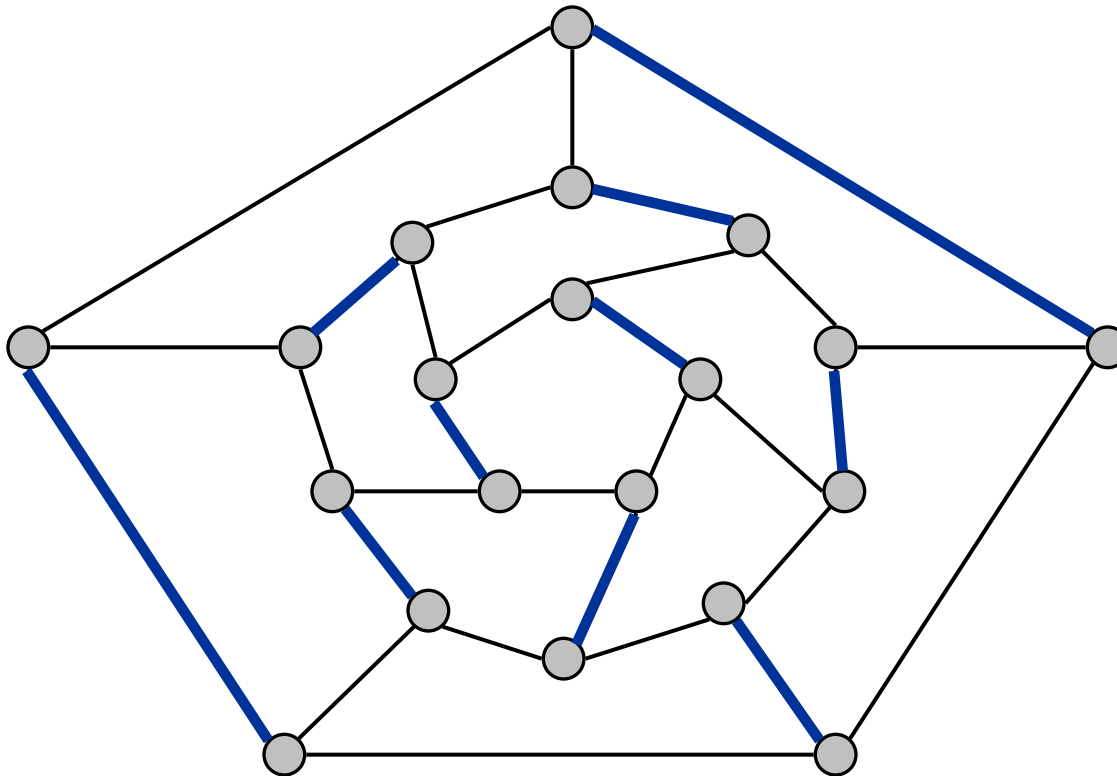
## 7.5 Bipartite Matching

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# Matching

## Matching.

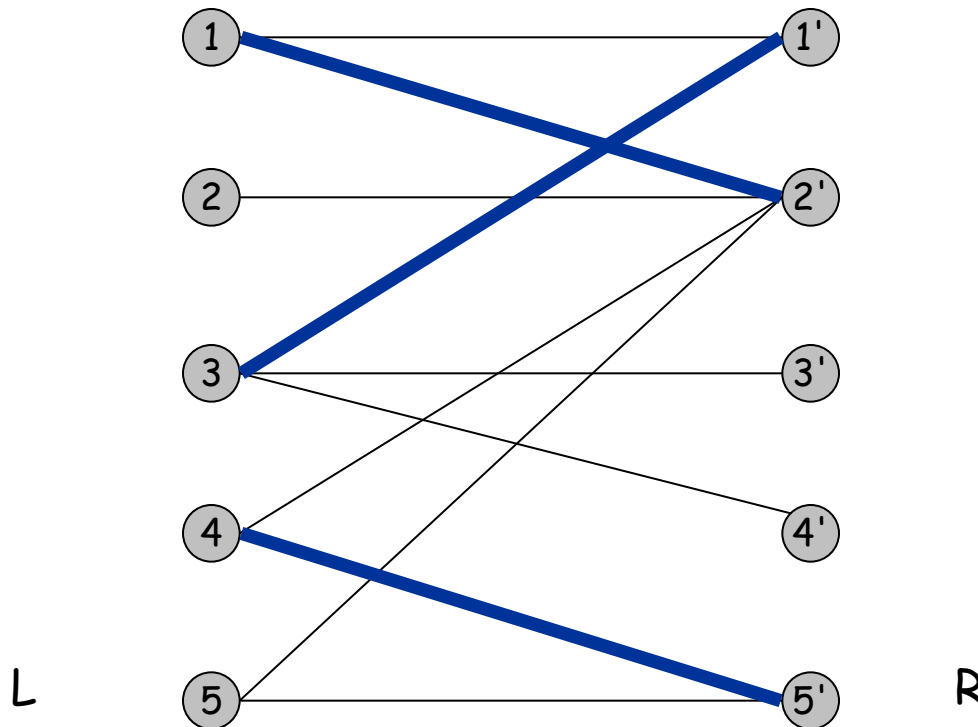
- Input: undirected graph  $G = (V, E)$ .
- $M \subseteq E$  is a **matching** if each node appears in at most edge in  $M$ .
- Max matching: find a max cardinality matching.



# 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 edge in  $M$ .
- Max matching: find a max cardinality matching.

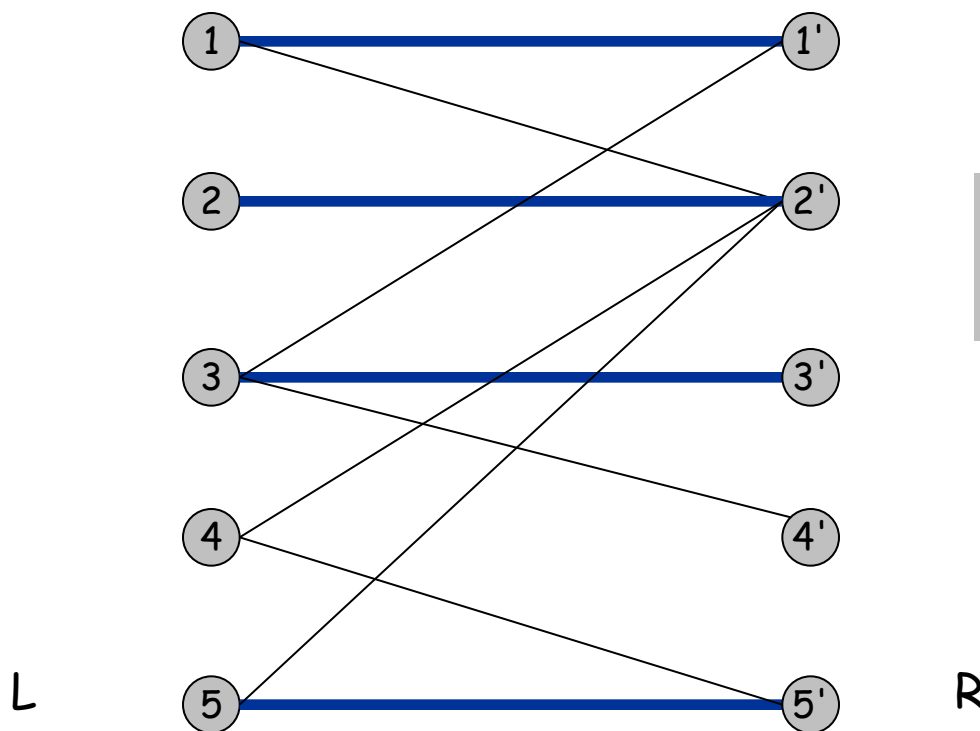


matching  
1-2', 3-1', 4-5'

# Bipartite Matching

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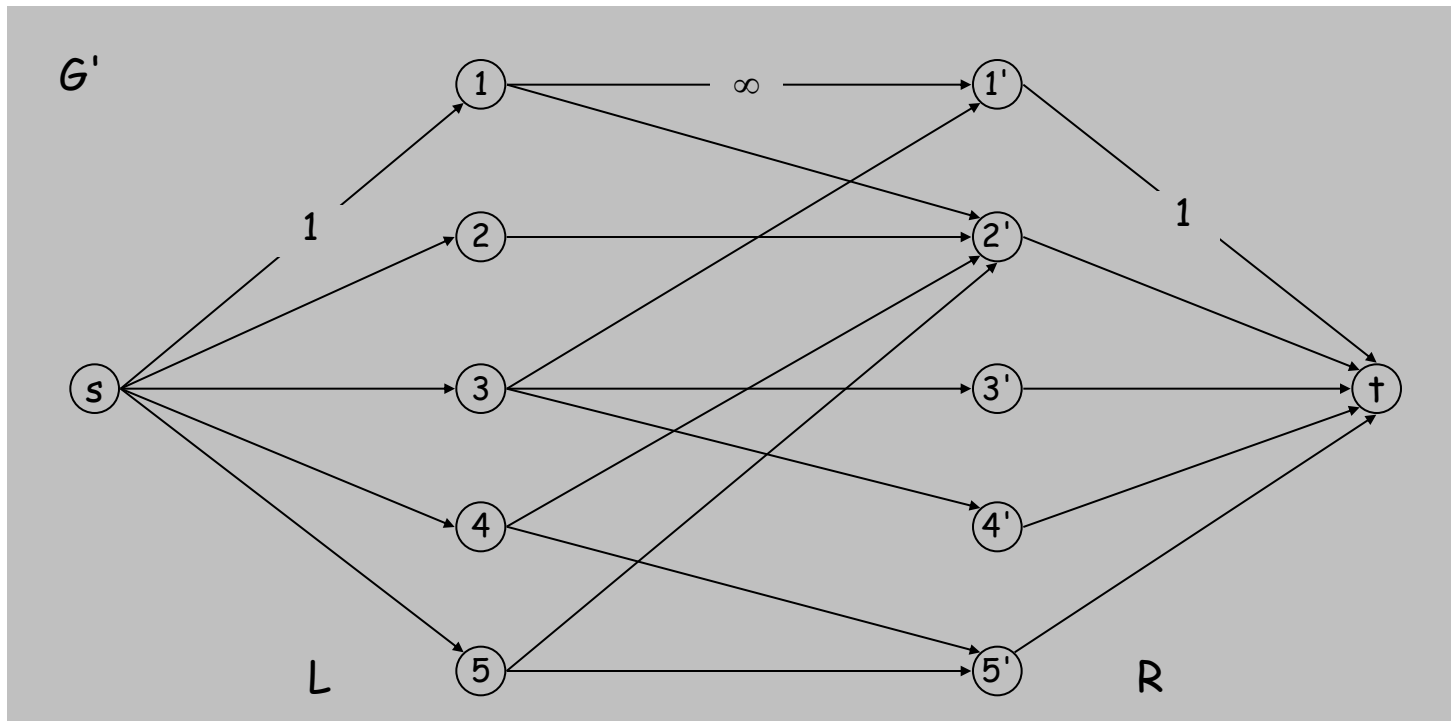


max matching  
1-1', 2-2', 3-3' 4-4'

# 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  $t$ .

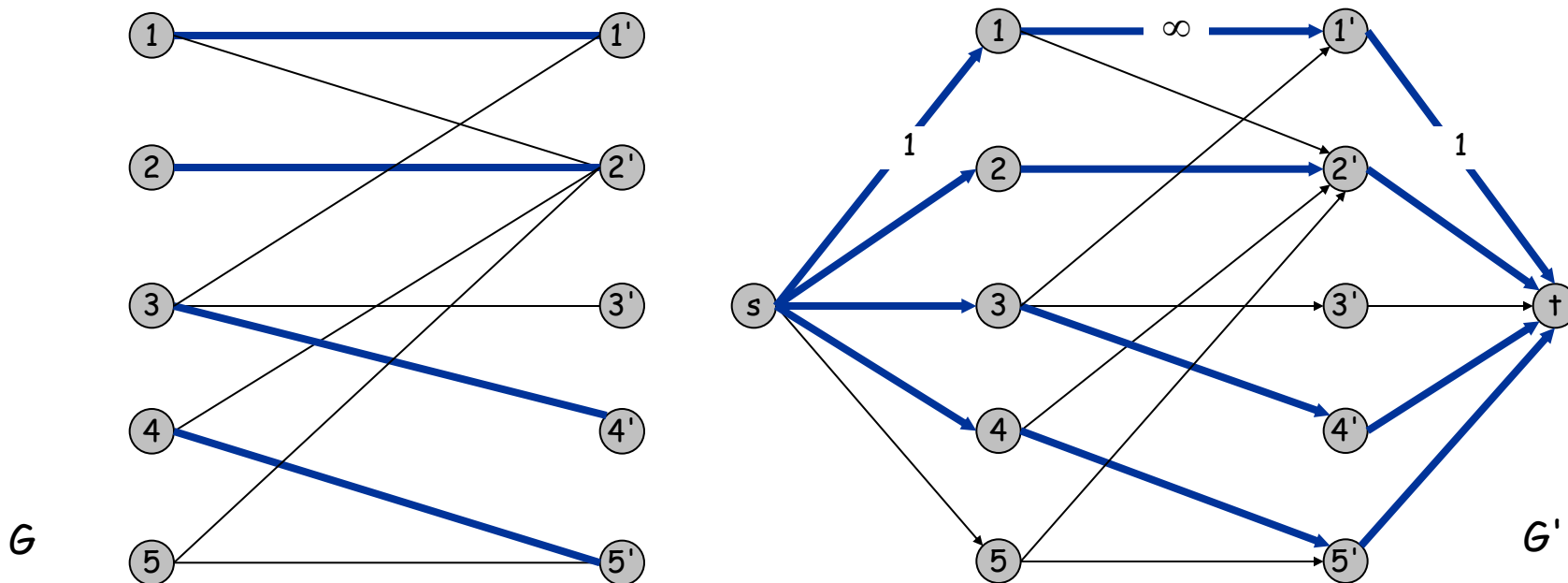


# Bipartite Matching: Proof of Correctness

**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  $k$ . ▪

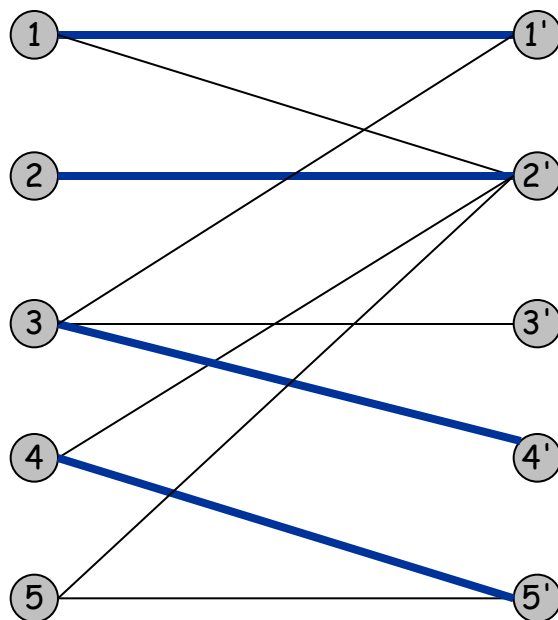
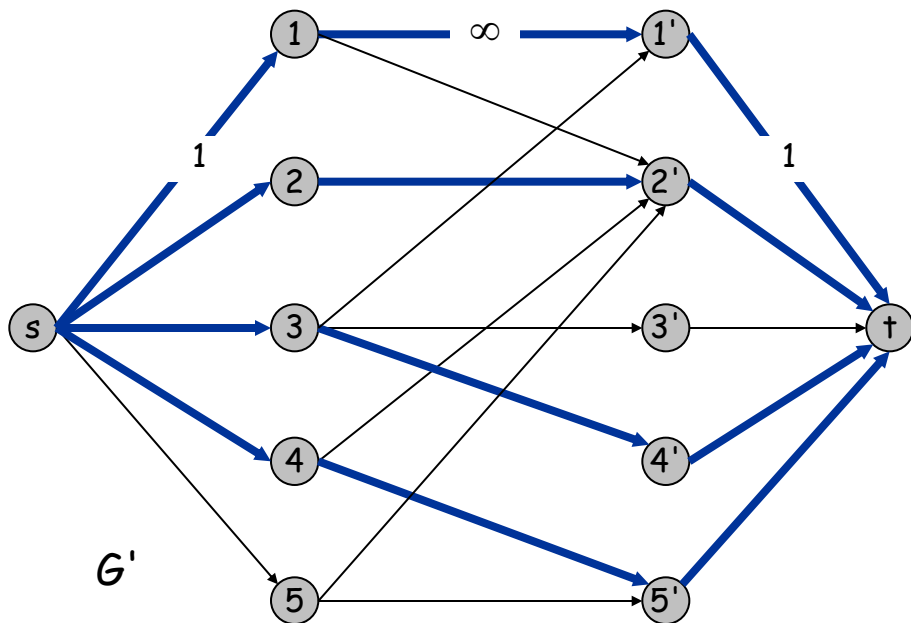


# Bipartite Matching: Proof of Correctness

**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  $(L \cup s, R \cup t)$  .





## 7.6 Disjoint Paths

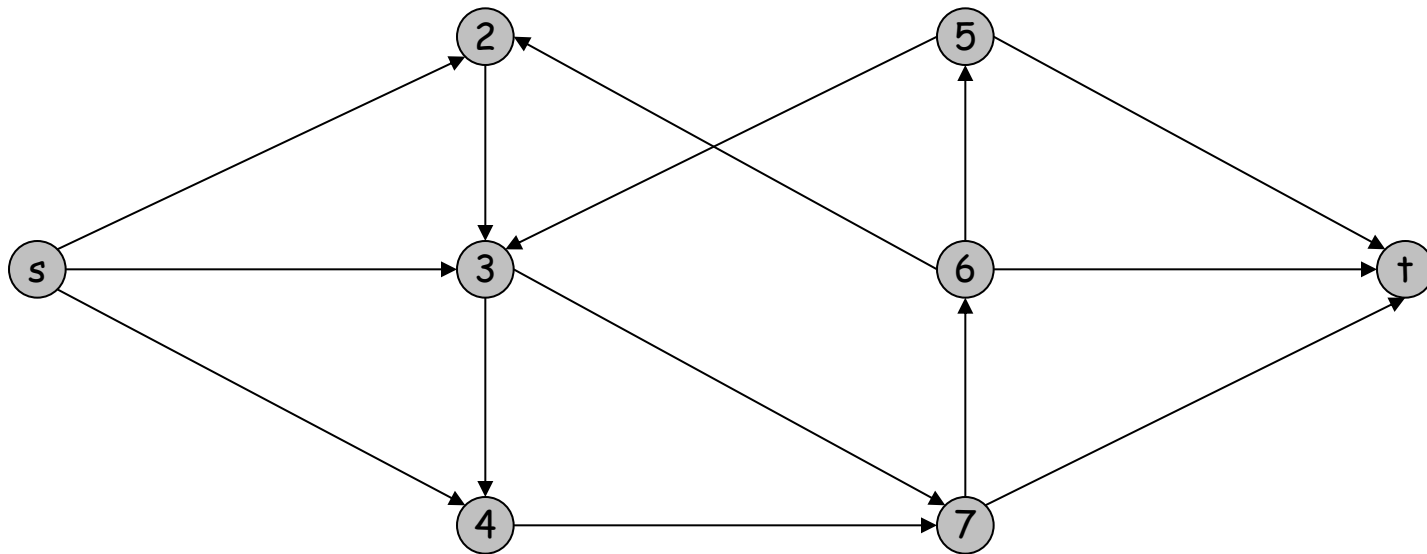
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# Edge Disjoint Paths

**Disjoint path problem.** Given a digraph  $G = (V, E)$  and two nodes  $s$  and  $t$ , find the max number of edge-disjoint  $s$ - $t$  paths.

**Def.** Two paths are **edge-disjoint** if they have no edge in common.

**Ex:** communication networks.

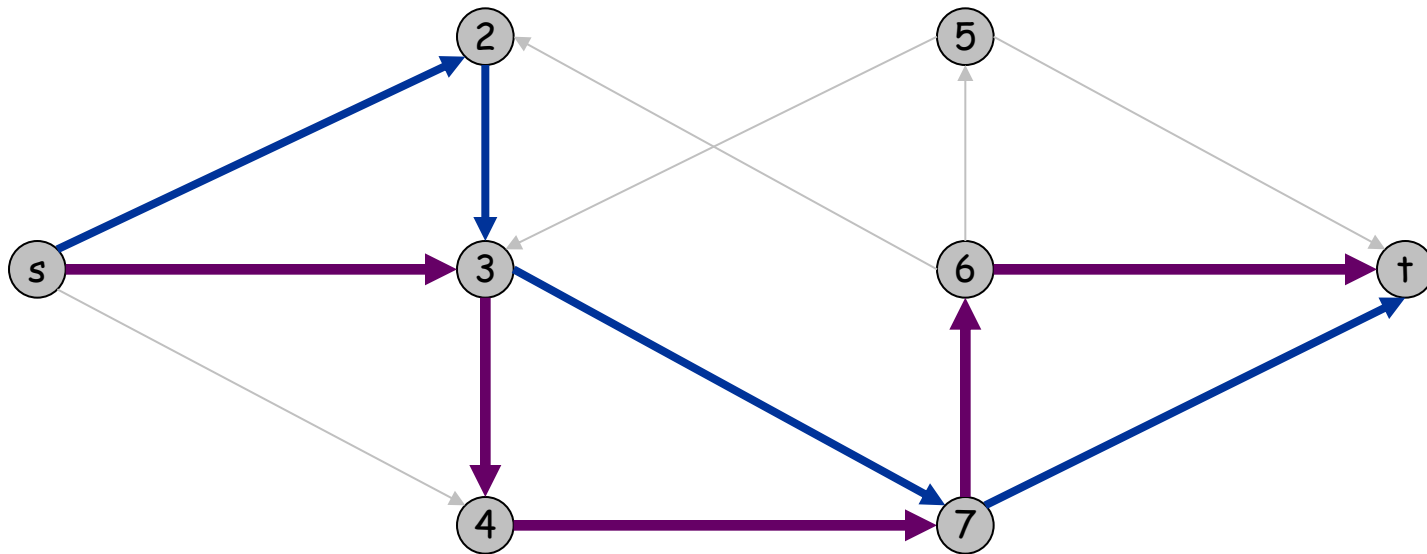


# Edge Disjoint Paths

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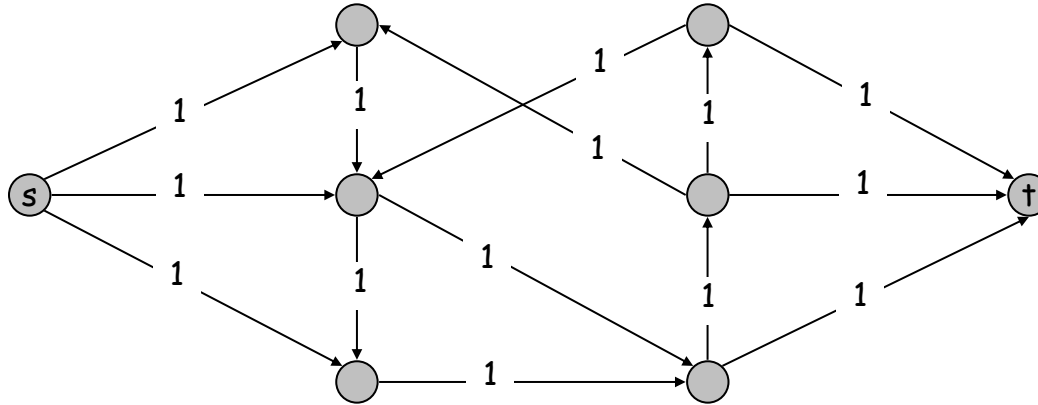
**Def.** Two paths are **edge-disjoint** if they have no edge in common.

**Ex:** communication networks.



# Edge Disjoint Paths

Max flow formulation: assign unit capacity to every edge.



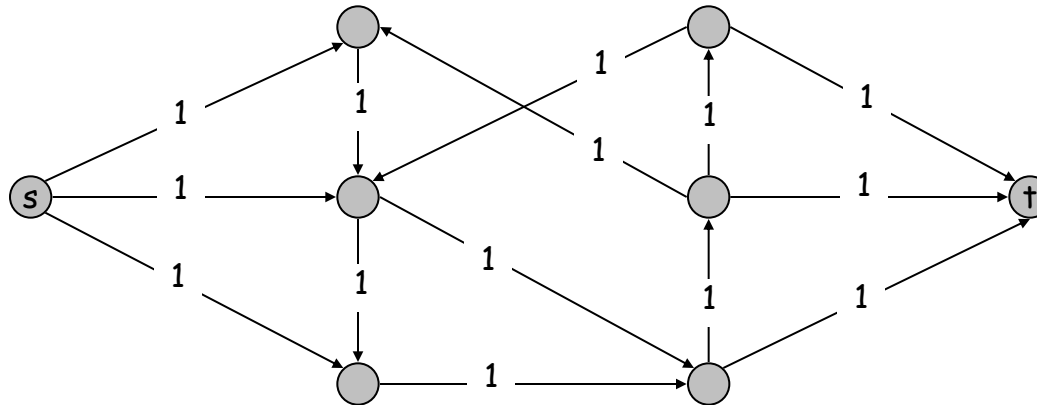
**Theorem.** Max number edge-disjoint  $s$ - $t$  paths equals max flow value.

**Pf.**  $\leq$

- Suppose there are  $k$  edge-disjoint paths  $P_1, \dots, P_k$ .
- Set  $f(e) = 1$  if  $e$  participates in some path  $P_i$ ; else set  $f(e) = 0$ .
- Since paths are edge-disjoint,  $f$  is a flow of value  $k$ . ▪

# Edge Disjoint Paths

Max flow formulation: assign unit capacity to every edge.



**Theorem.** Max number edge-disjoint  $s$ - $t$  paths equals max flow value.

**Pf.**  $\geq$

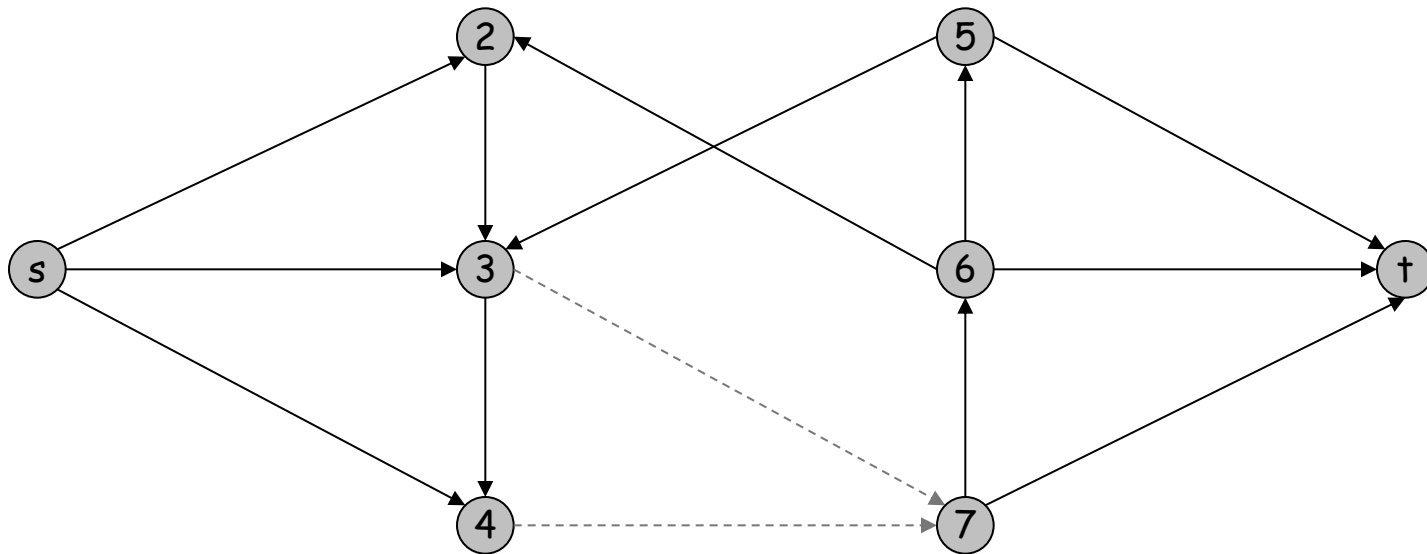
- Suppose max flow value is  $k$ .
- Integrality theorem  $\Rightarrow$  there exists 0-1 flow  $f$  of value  $k$ .
- Consider edge  $(s, u)$  with  $f(s, u) = 1$ .
  - by conservation, there exists an edge  $(u, v)$  with  $f(u, v) = 1$
  - continue until reach  $t$ , always choosing a new edge
- Produces  $k$  (not necessarily simple) edge-disjoint paths. •

can eliminate cycles to get simple paths if desired

# Network Connectivity

**Network connectivity.** Given a digraph  $G = (V, E)$  and two nodes  $s$  and  $t$ , find min number of edges whose removal disconnects  $t$  from  $s$ .

**Def.** A set of edges  $F \subseteq E$  **disconnects  $t$  from  $s$**  if all  $s$ - $t$  paths uses at least on edge in  $F$ .

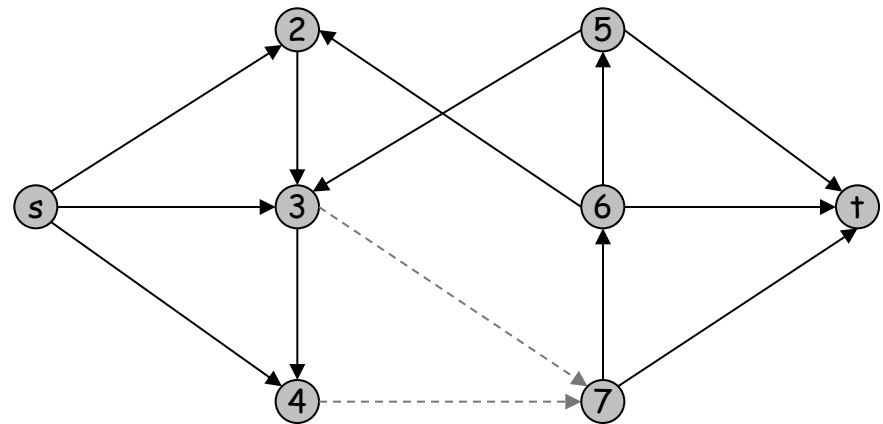
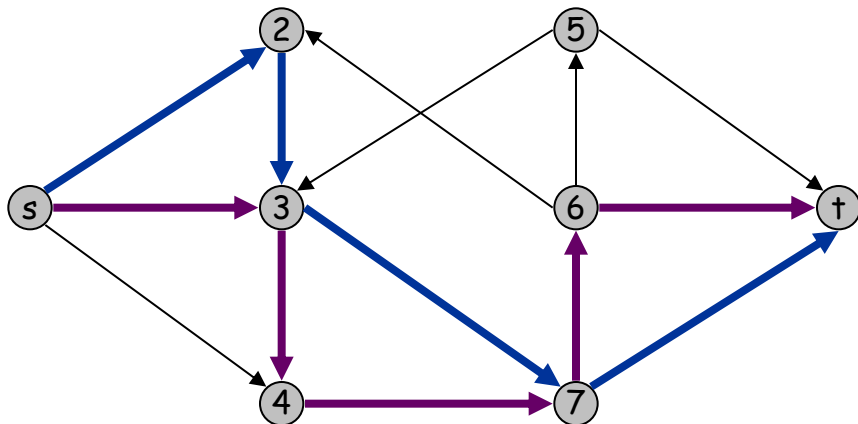


# Edge Disjoint Paths and Network Connectivity

**Theorem.** [Menger 1927] The max number of edge-disjoint  $s$ - $t$  paths is equal to the min number of edges whose removal disconnects  $t$  from  $s$ .

Pf.  $\leq$

- Suppose the removal of  $F \subseteq E$  disconnects  $t$  from  $s$ , and  $|F| = k$ .
- All  $s$ - $t$  paths use at least one edge of  $F$ . Hence, the number of edge-disjoint paths is at most  $k$ . ▪

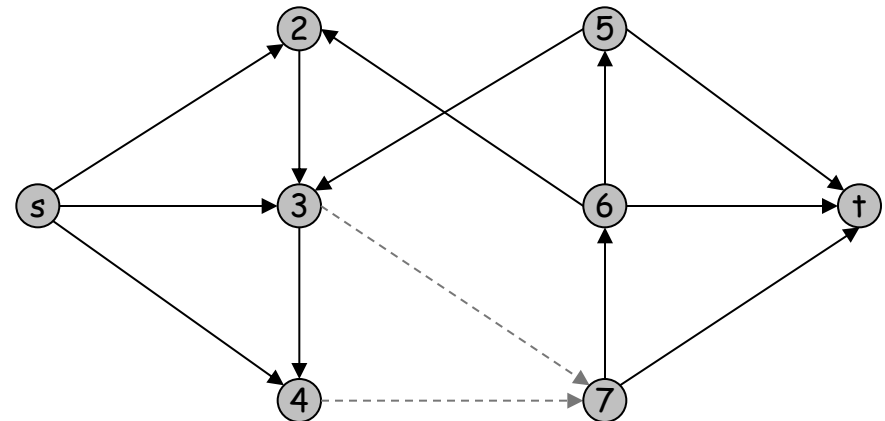
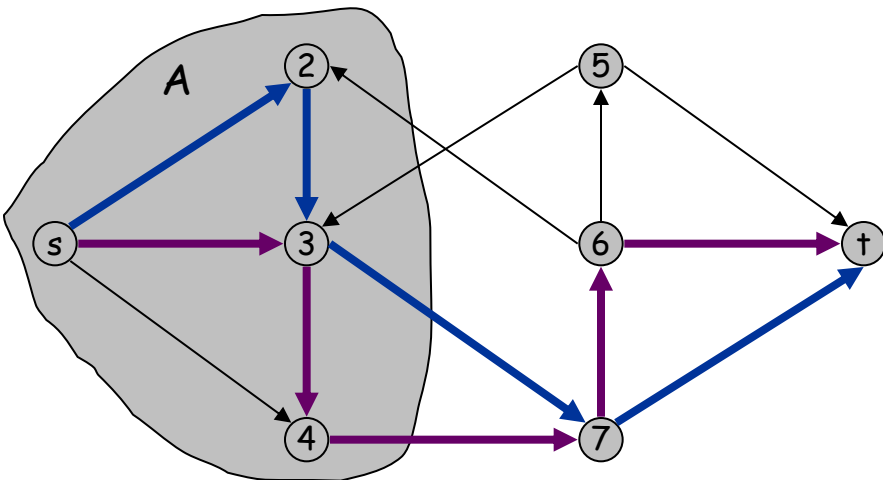


# Disjoint Paths and Network Connectivity

**Theorem.** [Menger 1927] The max number of edge-disjoint  $s$ - $t$  paths is equal to the min number of edges whose removal disconnects  $t$  from  $s$ .

Pf.  $\geq$

- Suppose max number of edge-disjoint paths is  $k$ .
- Then max flow value is  $k$ .
- Max-flow min-cut  $\Rightarrow$  cut  $(A, B)$  of capacity  $k$ .
- Let  $F$  be set of edges going from  $A$  to  $B$ .
- $|F| = k$  and disconnects  $t$  from  $s$ . ▪





## 7.11 Project Selection

---

# Project Selection

can be positive or negative



## Projects with prerequisites.

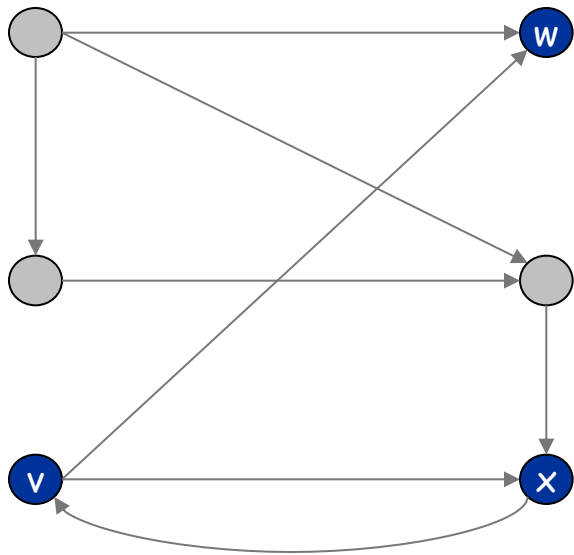
- Set  $P$  of possible projects. Project  $v$  has associated revenue  $p_v$ .
  - some projects generate money: create interactive e-commerce interface, redesign web page
  - others cost money: upgrade computers, get site license
- Set of prerequisites  $E$ . If  $(v, w) \in E$ , can't do project  $v$  and unless also do project  $w$ .
- A subset of projects  $A \subseteq P$  is **feasible** if the prerequisite of every project in  $A$  also belongs to  $A$ .

**Project selection.** Choose a feasible subset of projects to maximize revenue.

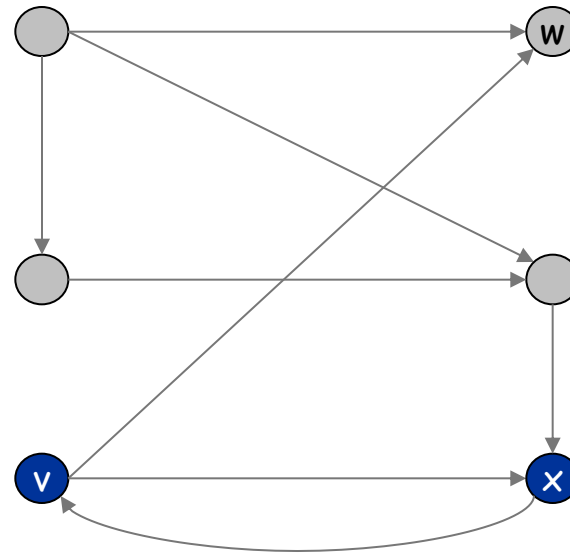
# Project Selection: Prerequisite Graph

## Prerequisite graph.

- Include an edge from  $v$  to  $w$  if can't do  $v$  without also doing  $w$ .
- $\{v, w, x\}$  is feasible subset of projects.
- $\{v, x\}$  is infeasible subset of projects.



feasible

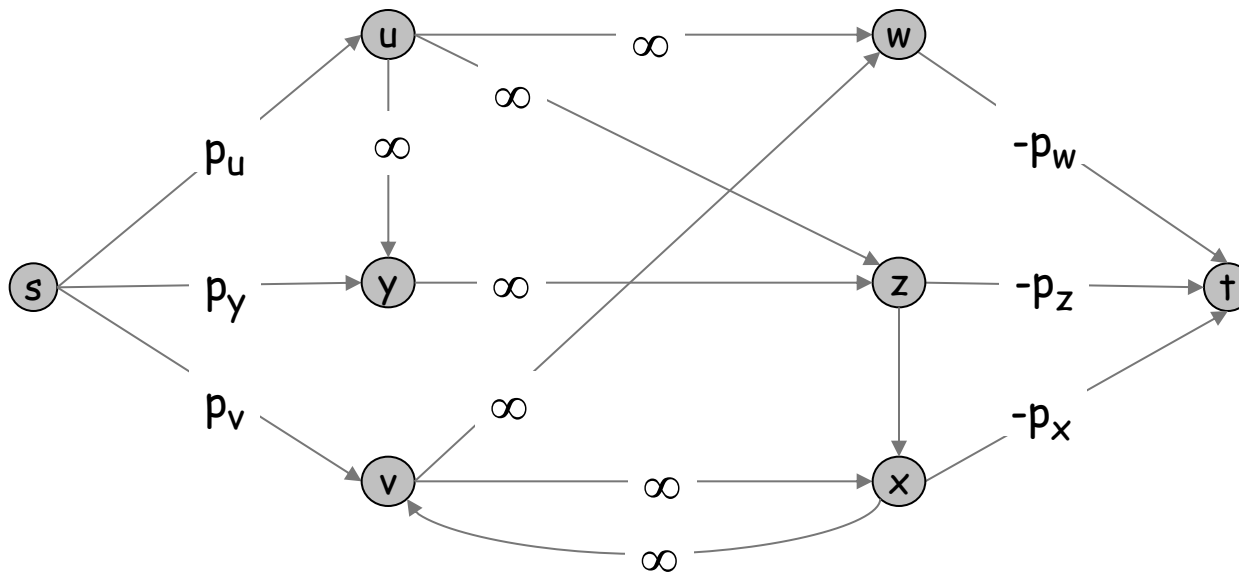


infeasible

# Project Selection: Min Cut Formulation

## Min cut formulation.

- Assign capacity  $\infty$  to all prerequisite edge.
- Add edge  $(s, v)$  with capacity  $p_v$  if  $p_v > 0$ .
- Add edge  $(v, t)$  with capacity  $-p_v$  if  $p_v < 0$ .
- For notational convenience, define  $p_s = p_t = 0$ .

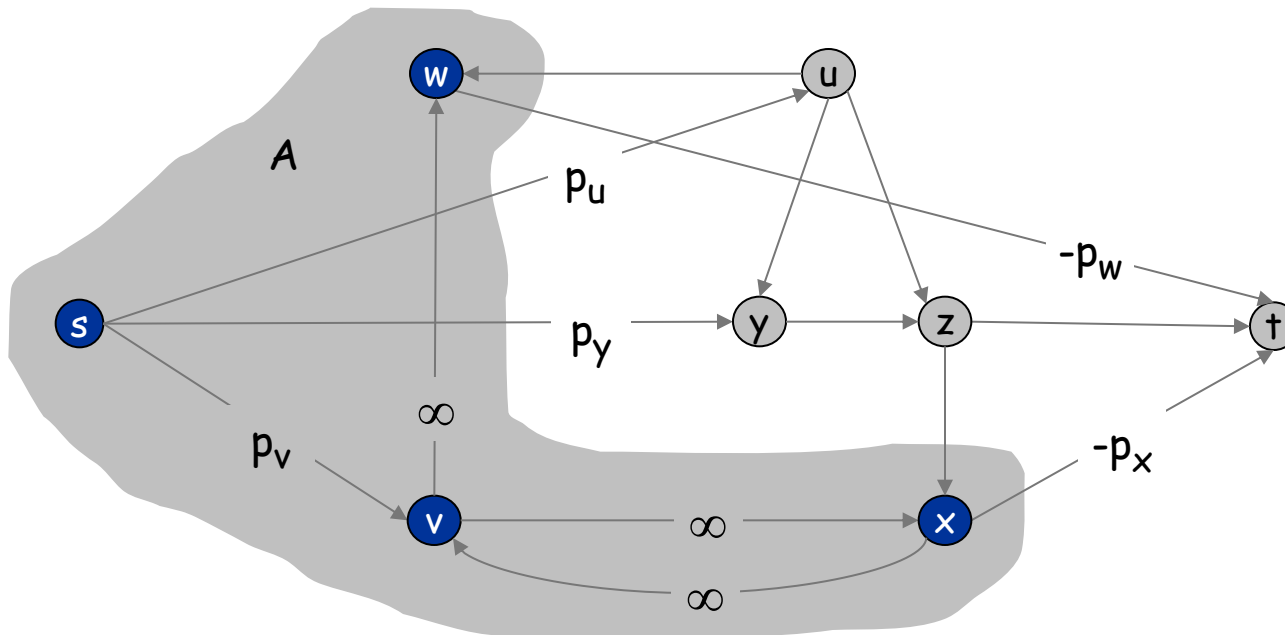


# Project Selection: Min Cut Formulation

**Claim.**  $(A, B)$  is min cut iff  $A - \{s\}$  is optimal set of projects.

- Infinite capacity edges ensure  $A - \{s\}$  is feasible.

- Max revenue because:
- $$\begin{aligned} \text{cap}(A, B) &= \sum_{v \in B: p_v > 0} p_v + \sum_{v \in A: p_v < 0} (-p_v) \\ &= \underbrace{\sum_{v: p_v > 0} p_v}_{\text{constant}} - \sum_{v \in A} p_v \end{aligned}$$



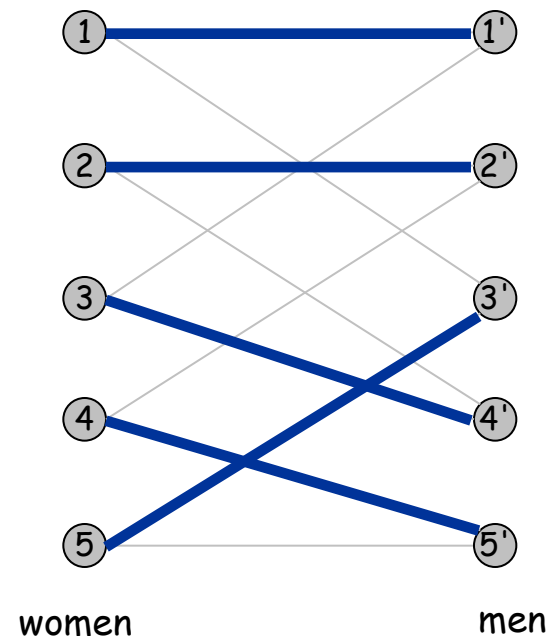
# k-Regular Bipartite Graphs

## Dancing problem.

- Exclusive Ivy league party attended by  $n$  men and  $n$  women.
- Each man knows exactly  $k$  women; each woman knows exactly  $k$  men.
- Acquaintances are mutual.
- Is it possible to arrange a dance so that each woman dances with a man that she knows?

**Mathematical reformulation.** Does every  $k$ -regular bipartite graph have a perfect matching?

**Ex.** Boolean hypercube.



# k-Regular Bipartite Graphs Have Perfect Matchings

**Theorem.** [König 1916, Frobenius 1917] Every k-regular bipartite graph has a perfect matching.

**Pf.** Size of max matching = value of max flow in  $G'$ . Consider flow:

$$f(u, v) = \begin{cases} 1/k & \text{if } (u, v) \in E \\ 1 & \text{if } u = s \text{ or } v = t \\ 0 & \text{otherwise} \end{cases}$$

- $f$  is a flow and its value =  $n \Rightarrow$  perfect matching. ▪

