



# **Chapter 6**

# **Graph Algorithms**

**Algorithm Theory**  
**WS 2018/19**

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# Strongly Polynomial Algorithm

- Time of regular Ford-Fulkerson algorithm with integer capacities:

$$\underline{\underline{O(mC)}}$$

- Time of algorithm with scaling parameter:

$$\underline{\underline{O(m^2 \log C)}}$$

- $O(\log C)$  is polynomial in the size of the input, but not in  $n$
- Can we get an algorithm that runs in time polynomial in  $n$ ?
- Always picking a **shortest augmenting path** leads to running time

$$\underline{\underline{O(m^2 n)}}$$

- also works for arbitrary real-valued weights

# Other Algorithms

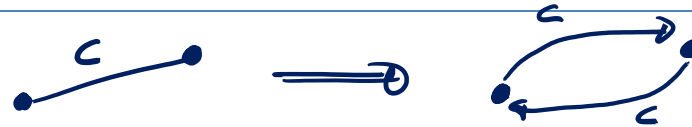
- There are many other algorithms to solve the maximum flow problem, for example:
- **Preflow-push algorithm:**
  - Maintains a preflow ( $\forall$  nodes: inflow  $\geq$  outflow)
  - Alg. guarantees: As soon as we have a flow, it is optimal
  - Detailed discussion in 2012/13 lecture
  - Running time of basic algorithm:  $O(m \cdot n^2)$
  - Doing steps in the “right” order:  $O(n^3)$
- **Current best known complexity:  $O(m \cdot n)$** 
  - For graphs with  $m \geq n^{1+\epsilon}$  [King,Rao,Tarjan 1992/1994]  
(for every constant  $\epsilon > 0$ )
  - For sparse graphs with  $m \leq n^{16/15-\delta}$  [Orlin, 2013]

# Maximum Flow Applications

- Maximum flow has many applications
- Reducing a problem to a max flow problem can even be seen as an important algorithmic technique
- Examples:
  - related network flow problems
  - computation of small cuts
  - computation of matchings
  - computing disjoint paths
  - scheduling problems
  - assignment problems with some side constraints
  - ...

# Undirected Edges and Vertex Capacities

## Undirected Edges:

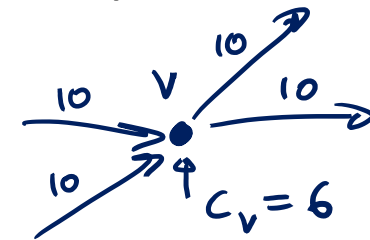


- Undirected edge  $\{u, v\}$ : add edges  $(u, v)$  and  $(v, u)$  to network

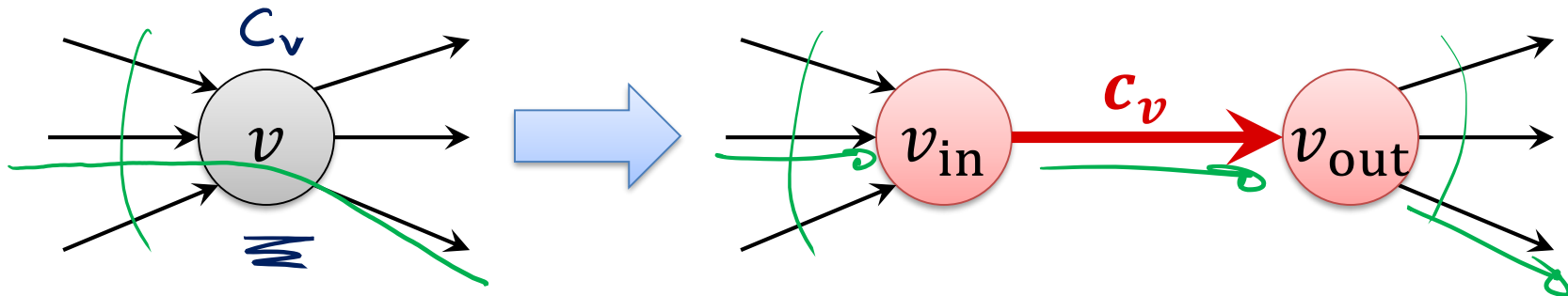
## Vertex Capacities:

- Not only edges, but also (or only) nodes have capacities
- Capacity  $c_v$  of node  $v \notin \{s, t\}$ :

$$\underline{f^{\text{in}}(v)} = \underline{f^{\text{out}}(v)} \leq \underline{c_v}$$



- Replace node  $v$  by edge  $e_v = \{v_{\text{in}}, v_{\text{out}}\}$ :



# Minimum $s$ - $t$ Cut

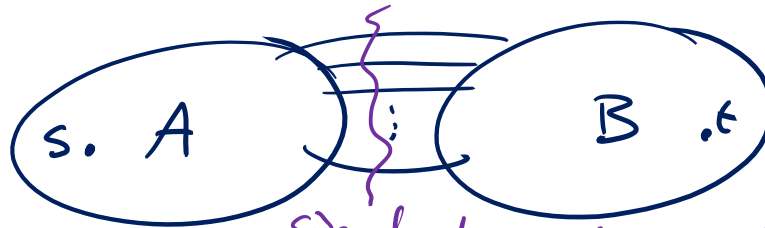
max flow min cut theorem



**Given:** undirected graph  $G = (V, E)$ , nodes  $s$ ,  $t$   $\in V$   
*(Handwritten:  $n$  and  $m$  with arrows pointing to  $s$  and  $t$  respectively)*

**$s$ - $t$  cut:** Partition ( $A$ ,  $B$ ) of  $V$  such that  $s$   $\in$   $A$ ,  $t$   $\in$   $B$

**Size of cut ( $A, B$ ):** number of edges crossing the cut



running time:

$$O(m^2)$$

*Size of cut = #edges crossing the cut*

**Objective:** find  $s$ - $t$  cut of minimum size

create flow network

1) make edges directed



2) edge cap. = 1

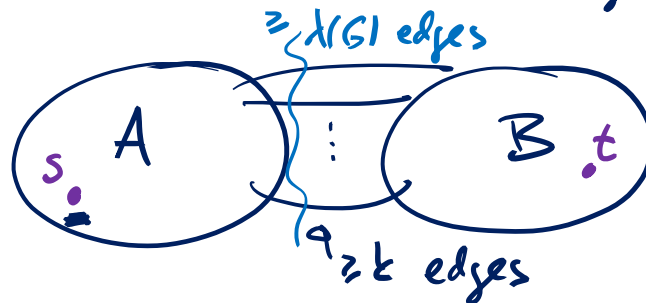
size of cut in  $G$  = cap. cut in flow network

# Edge Connectivity

**Definition:** A graph  $G = (V, E)$  is  **$k$ -edge connected** for an integer  $k \geq 1$  if the graph  $G_X = (V, E \setminus X)$  is **connected** for every edge set

$$X \subseteq E, |X| \leq k - 1.$$

need to remove  $\geq k$  edges to disconnect  $G$



edge connectivity  $\lambda(G)$ :  
 max.  $k$  s.t.  $G$  is  
 $k$ -edge connected

**Goal:** Compute edge connectivity  $\lambda(G)$  of  $G$   
 (and edge set  $X$  of size  $\lambda(G)$  that divides  $G$  into  $\geq 2$  parts)

- minimum set  $X$  is a minimum  $s$ - $t$  cut for some  $s, t \in V$ 
  - Actually for all  $s, t$  in different components of  $G_X = (V, E \setminus X)$

running time  $O(n^2 \cdot m)$

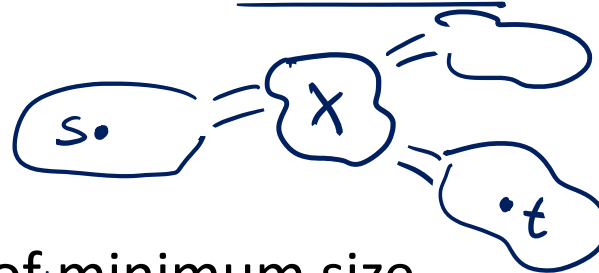
- Possible algorithm: fix  $s$  and find min  $s$ - $t$  cut for all  $t \neq s$

# Minimum $s$ - $t$ Vertex-Cut

**Given:** undirected graph  $G = (V, E)$ , nodes  $s, t \in V$

**$s$ - $t$  vertex cut:** Set  $X \subset V$  such that  $s, t \notin X$  and  $s$  and  $t$  are in different components of the sub-graph  $G[V \setminus X]$  induced by  $V \setminus X$

**Size of vertex cut:**  $|X|$

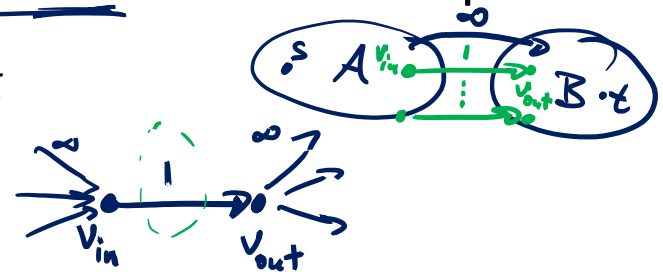


**Objective:** find  $s$ - $t$  vertex-cut of minimum size,

- Replace undirected edge  $\{u, v\}$  by  $(u, v)$  and  $(v, u)$
- Compute max  $s$ - $t$  flow for edge capacities  $\infty$  and node capacities

$$\underline{c_v = 1 \text{ for } v \neq s, t}$$

- Replace each node  $v$  by  $v_{in}$  and  $v_{out}$ :



- Min edge cut corresponds to min vertex cut in  $G$



# Vertex Connectivity

**Definition:** A graph  $G = (V, E)$  is  $k$ -vertex connected for an integer  $k \geq 1$  if the sub-graph  $G[V \setminus X]$  induced by  $V \setminus X$  is connected for every edge set

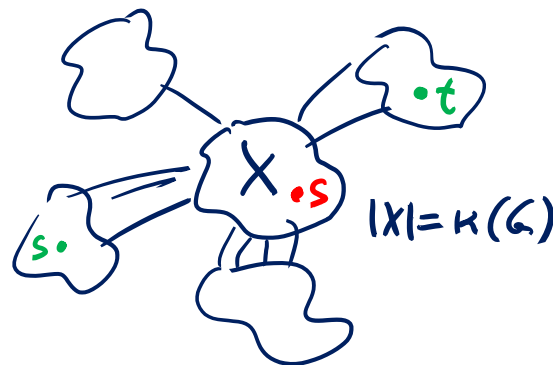
$$X \subseteq V, |X| \leq k - 1.$$

need to remove at least  $k$  nodes to make  $G$  disconnected

vertex conn. :  $\kappa(G)$

max.  $k$  s.t.

$G$  is  $k$ -vertex-connected



**Goal:** Compute vertex connectivity  $\kappa(G)$  of  $G$

(and node set  $X$  of size  $\kappa(G)$  that divides  $G$  into  $\geq 2$  parts)

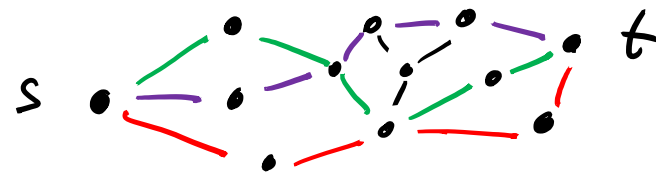
- Compute minimum  $s$ - $t$  vertex cut for all  $s$  and all  $t \neq s$

running time:  $O(m \cdot n^3) \Rightarrow O(m \cdot n \cdot \kappa^2(G))$

# Edge-Disjoint Paths

**Given:** Graph  $G = (V, E)$  with nodes  $s, t \in V$

**Goal:** Find as many edge-disjoint  $s$ - $t$  paths as possible



**Solution:**

*integer*

- Find max  $s$ - $t$  flow in  $G$  with **edge capacities  $c_e = 1$**  for all  $e \in E$



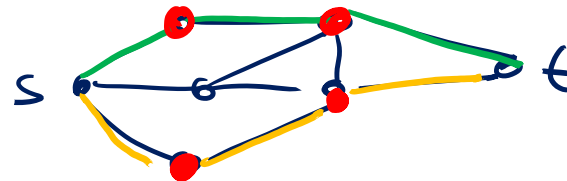
Flow  $f$  induces  $|f|$  **edge-disjoint paths**:

- Integral capacities  $\rightarrow$  can compute integral max flow  $f$
- Get  $|f|$  edge-disjoint paths by greedily picking them
- Correctness follows from flow conservation  $f^{\text{in}}(v) = f^{\text{out}}(v)$

# Vertex-Disjoint Paths

**Given:** Graph  $G = (V, E)$  with nodes  $s, t \in V$

**Goal:** Find as many internally vertex-disjoint  $s$ - $t$  paths as possible



**Solution:**

- Find max  $s$ - $t$  flow in  $G$  with node capacities  $c_v = 1$  for all  $v \in V$

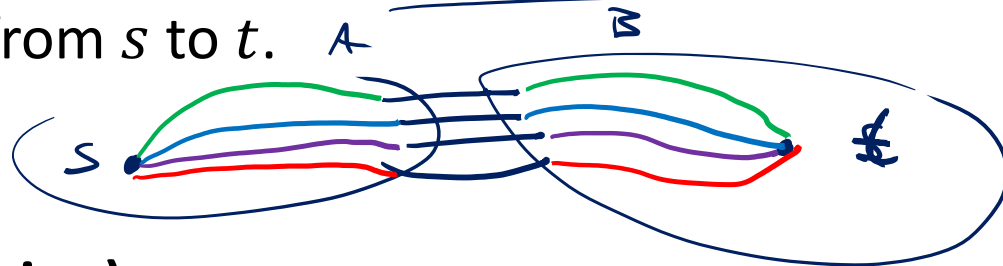
Flow  $f$  induces  $|f|$  **vertex-disjoint paths**:

- Integral capacities  $\rightarrow$  can compute integral max flow  $f$
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# Menger's Theorem

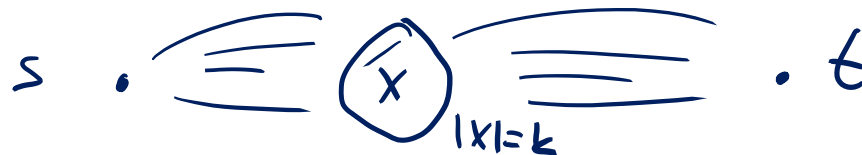
## Theorem: (edge version)

For every graph  $G = (V, E)$  with nodes  $s, t \in V$ , the size of the minimum  $s-t$  (edge) cut equals the maximum number of pairwise edge-disjoint paths from  $s$  to  $t$ .



## Theorem: (node version)

For every graph  $G = (V, E)$  with nodes  $s, t \in V$ , the size of the minimum  $s-t$  vertex cut equals the maximum number of pairwise internally vertex-disjoint paths from  $s$  to  $t$ .



- Both versions can be seen as a special case of the max flow min cut theorem

# Baseball Elimination

Team $i$	Wins $w_i$	Losses $\ell_i$	To Play $r_i$	Against = $r_{ij}$				
				NY	Balt.	T. Bay	Tor.	Bost.
New York	<u>81</u>	69	<u>12</u>	-	<u>2</u>	<u>5</u>	2	3
Baltimore	<u>79</u>	77	<u>6</u>	<u>2</u>	-	2	1	1
Tampa Bay	<u>79</u>	74	<u>9</u>	<u>5</u>	2	-	1	1
Toronto	<u>76</u>	80	<u>6</u>	2	1	1	-	2
Boston	<u>71</u>	84	<u>7</u>	3	1	1	2	-

- Only wins/losses possible (no ties), winner: team with most wins
- Which teams can still win (as least as many wins as top team)?
- Boston is eliminated (cannot win):
  - Boston can get at most 78 wins, New York already has 81 wins
- If for some  $i, j$ :  $\underline{w_i} + \underline{r_i} < \underline{w_j} \rightarrow$  team  $i$  is eliminated
- **Sufficient** condition, **but not** a **necessary** one!

# Baseball Elimination

Team $i$	Wins $w_i$	Losses $\ell_i$	To Play $r_i$	Against = $r_{ij}$				
				NY	Balt.	T. Bay	Tor.	Bost.
<u>New York</u>	81	69	12	-	2	<u>5</u>	2	3
Baltimore	79	77	6	2	-	2	1	1
<u>Tampa Bay</u>	79	74	9	<u>5</u>	2	-	1	1
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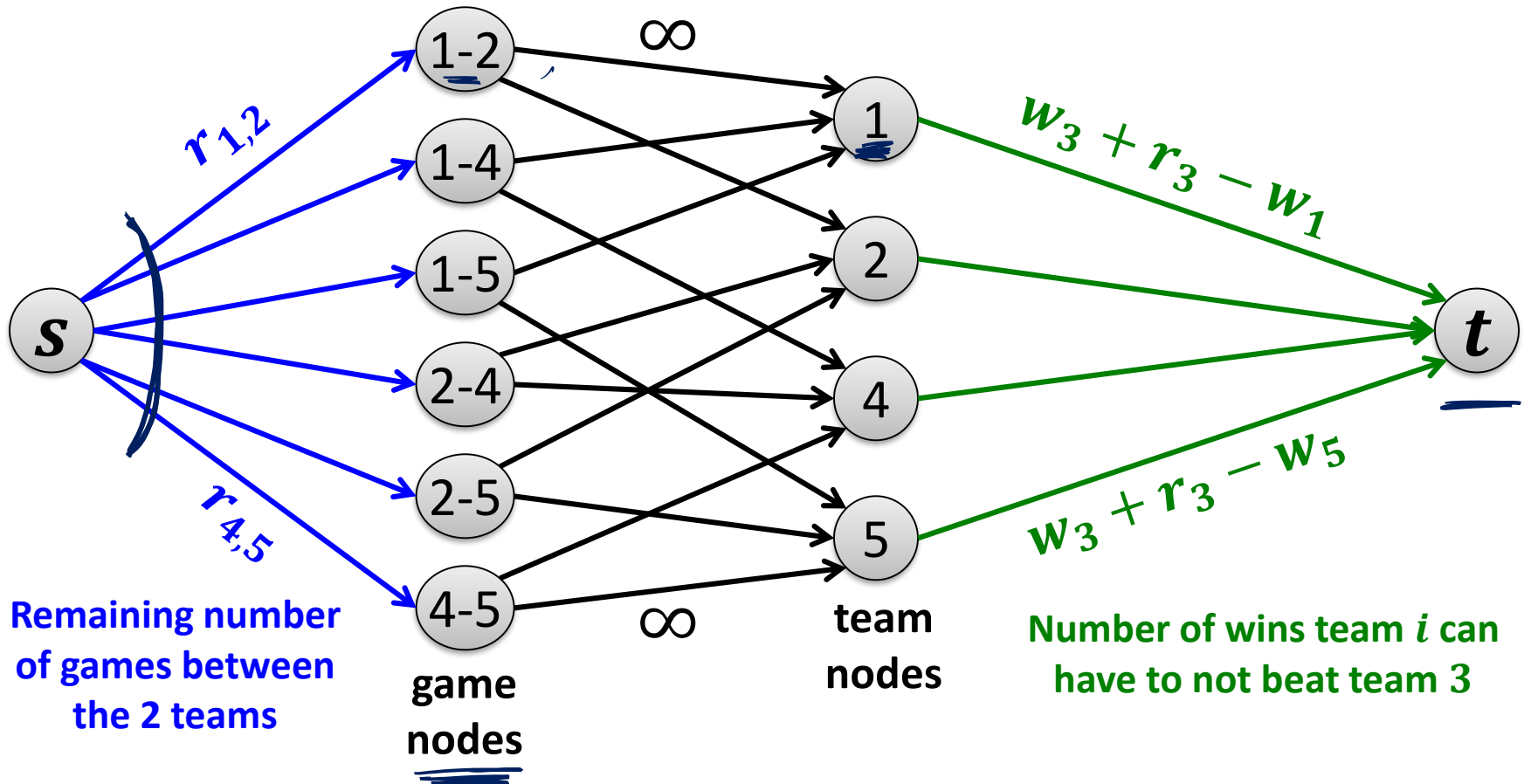
- Can Toronto still finish first?
- Toronto can get  $82 > 81$  wins, but:  
 NY and Tampa have to play 5 more times against each other  
 → if NY wins two, it gets 83 wins, otherwise, Tampa has 83 wins
- Hence: Toronto cannot finish first
- How about the others? How can we solve this in general?

# Max Flow Formulation

Team 3 will have  $\leq w_3 + r_3$  wins  
 $w_i$ : #wins of team  $i$  so far  
 $r_i$ : #rem. games of team  $i$



- Can team 3 finish with most wins?



- Team 3 can finish first iff all source-game edges are saturated

# Reason for Elimination

AL East: Aug 30, 1996

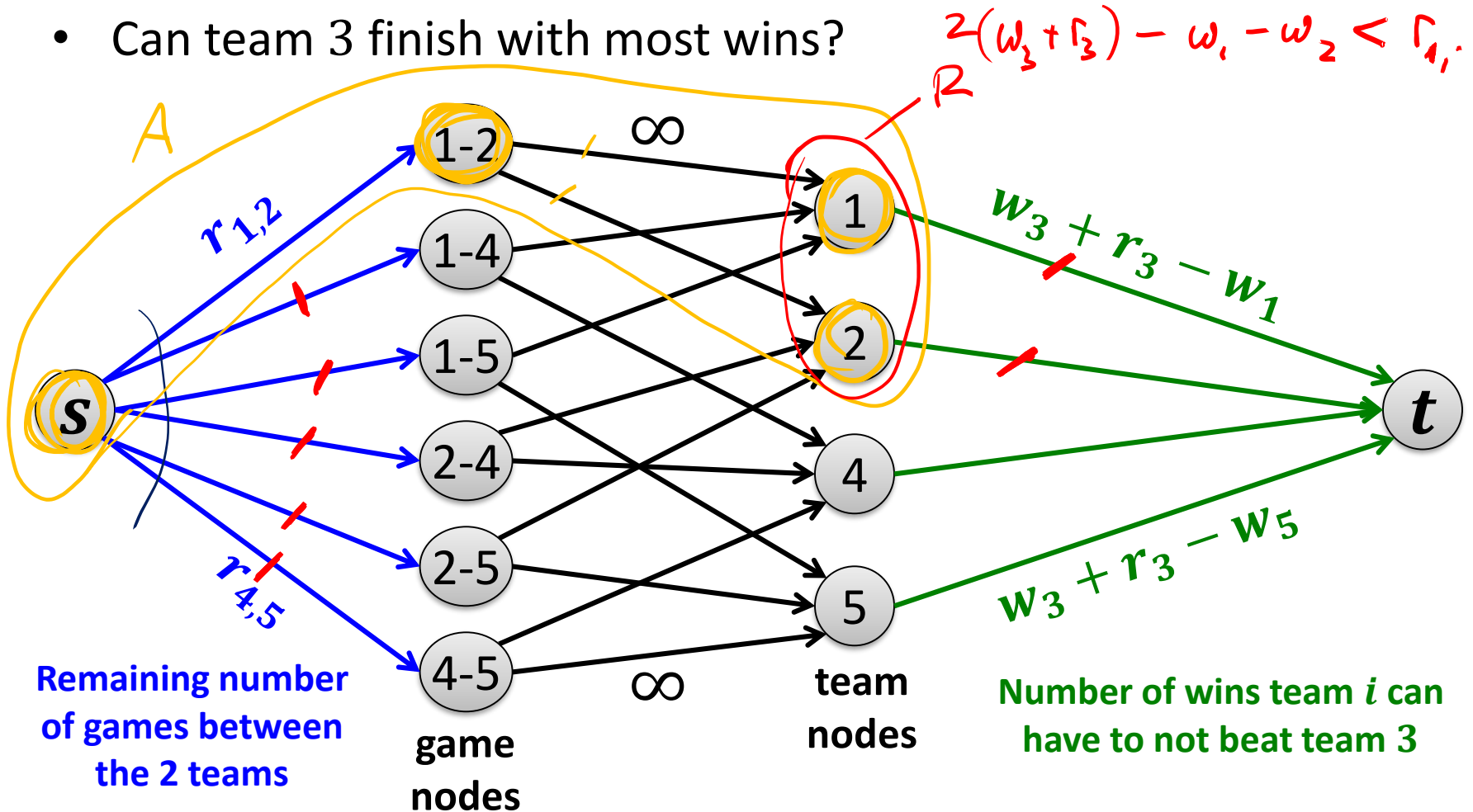
Team $i$	Wins $w_i$	Losses $\ell_i$	To Play $r_i$	Against = $r_{ij}$				
				NY	Balt.	Bost.	Tor.	Detr.
New York	75	59	28	-	3	8	7	3
Baltimore	71	63	28	3	-	2	7	4
Boston	69	66	27	8	2	-	0	0
Toronto	63	72	27	7	7	0	-	0
<b><u>Detroit</u></b>	49	86	27	3	4	0	0	-

- Detroit could finish with  $49 + 27 = \underline{76}$  wins
- Consider  $R = \{\text{NY, Bal, Bos, Tor}\}$ 
  - Have together already won  $w(R) = \underline{278}$  games
  - Must together win at least  $r(R) = \underline{27}$  more games
- On average, teams in  $R$  win  $\frac{278+27}{4} = \underline{76.25}$  games



# Reason for Elimination

- Can team 3 finish with most wins?



- Team 3 cannot finish first  $\Leftrightarrow$  min cut of size  $<$  “all blue edges”

# Reason for Elimination

Certificate of elimination:

$$\underline{R} \subseteq X, \quad w(R) := \sum_{i \in R} w_i, \quad r(R) := \sum_{i, j \in R} r_{i, j}$$

#wins of nodes in  $R$ 
#remaining games among nodes in  $R$

Team  $x \in X$  is eliminated by  $R$  if

$$\frac{w(R) + r(R)}{|R|} > w_x + r_x.$$

# Reason for Elimination

**Theorem:** Team  $x$  is eliminated if and only if there exists a subset  $R \subseteq X$  of the teams  $X$  such that  $x$  is eliminated by  $R$ .

## Proof Idea:

- Minimum cut gives a certificate...
- If  $x$  is eliminated, max flow solution does not saturate all outgoing edges of the source.
- Team nodes of unsaturated source-game edges are saturated
- Source side of min cut contains all teams of saturated team-dest. edges of unsaturated source-game edges
- Set of team nodes in source-side of min cut give a certificate  $R$

# Circulations with Demands

**Given:** Directed network with positive edge capacities

**Sources & Sinks:** Instead of one source and one destination, several sources that generate flow and several sinks that absorb flow.

**Supply & Demand:** sources have supply values, sinks demand values

**Goal:** Compute a flow such that source supplies and sink demands are exactly satisfied

- The circulation problem is a feasibility rather than a maximization problem

# Circulations with Demands: Formally

**Given:** Directed network  $G = (V, E)$  with

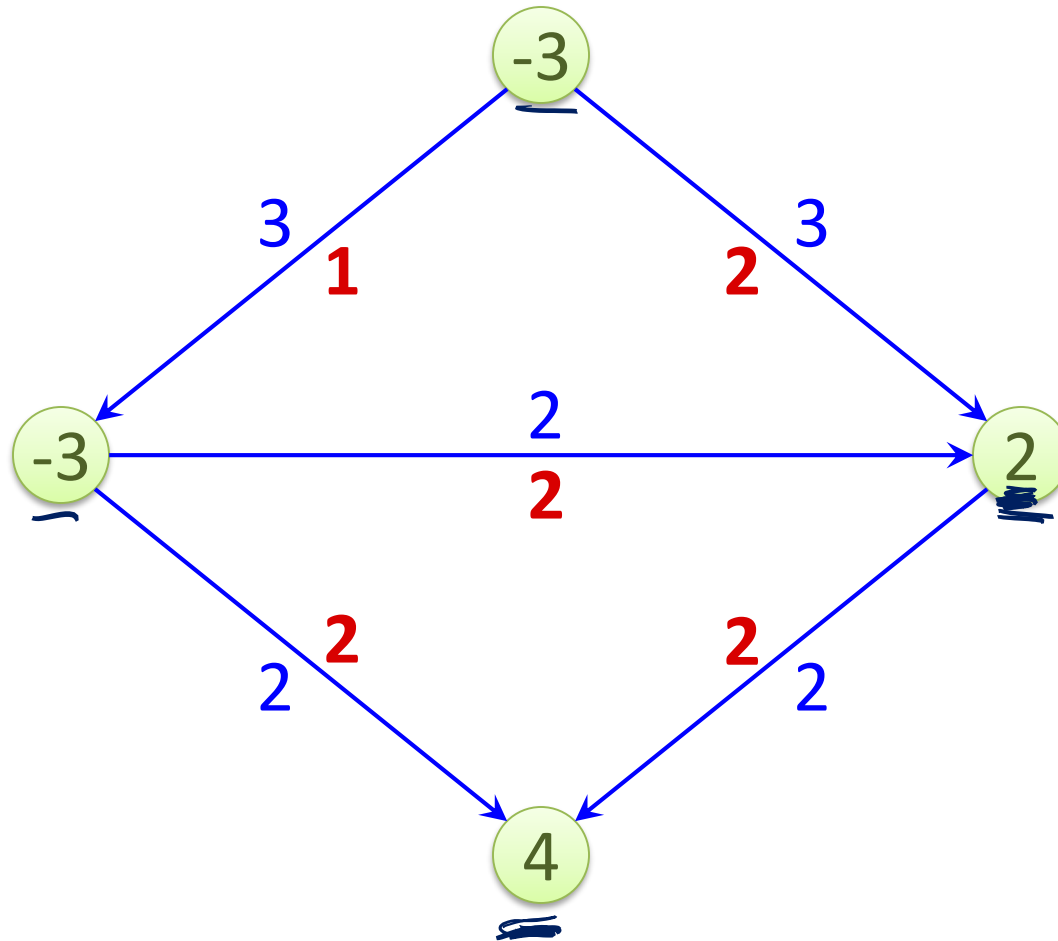
- Edge capacities  $c_e > 0$  for all  $e \in E$
- Node demands  $\underline{d}_v \in \mathbb{R}$  for all  $v \in V$ 
  - $\underline{d}_v > 0$ : node needs flow and therefore is a sink
  - $\underline{d}_v < 0$ : node has a supply of  $-\underline{d}_v$  and is therefore a source
  - $\underline{d}_v = 0$ : node is neither a source nor a sink

**Flow:** Function  $f: E \rightarrow \mathbb{R}_{\geq 0}$  satisfying

- *Capacity Conditions:*  $\forall e \in E: 0 \leq f(e) \leq c_e$
- *Demand Conditions:*  $\forall v \in V: \underline{f^{\text{in}}}(v) - \underline{f^{\text{out}}}(v) = \underline{d}_v$

**Objective:** Does a flow  $f$  satisfying all conditions exist?  
If yes, find such a flow  $f$ .

# Example



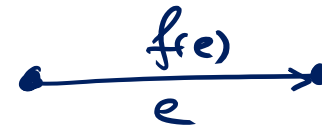
# Condition on Demands

**Claim:** If there exists a feasible circulation with demands  $d_v$  for  $v \in V$ , then

$$\sum_{v \in V} d_v = 0. \quad d_v = f^{\text{in}}(v) - f^{\text{out}}(v)$$

**Proof:**

- $\sum_v d_v = \sum_v (f^{\text{in}}(v) - f^{\text{out}}(v)) = \sum_v f^{\text{in}}(v) - \sum_v f^{\text{out}}(v) = 0$
- $f(e)$  of each edge  $e$  appears twice in the above sum with different signs  $\rightarrow$  overall sum is 0

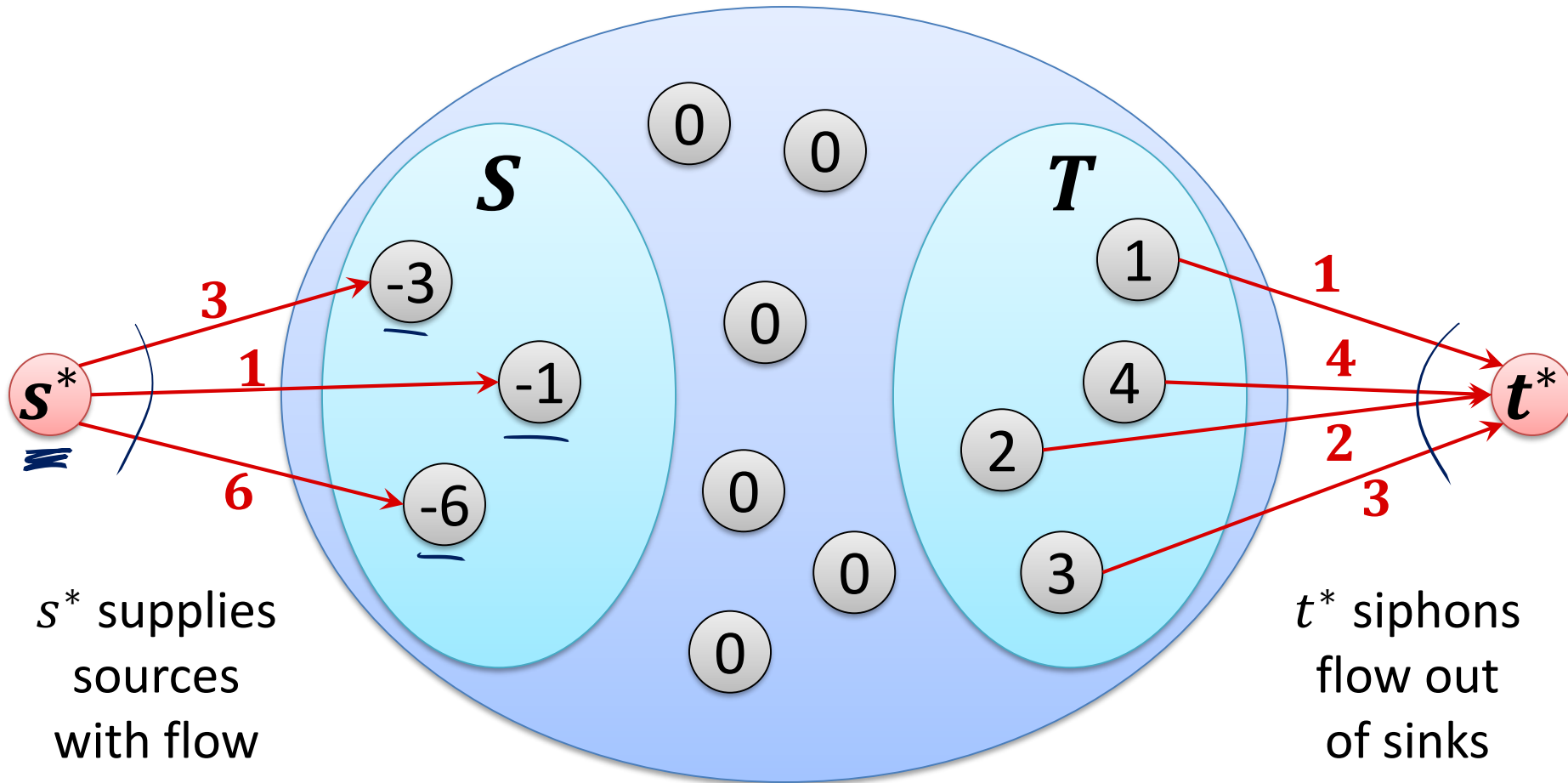


**Total supply = total demand:**

$$\text{Define } \underline{D} := \sum_{v: d_v > 0} d_v = \sum_{v: d_v < 0} -d_v$$

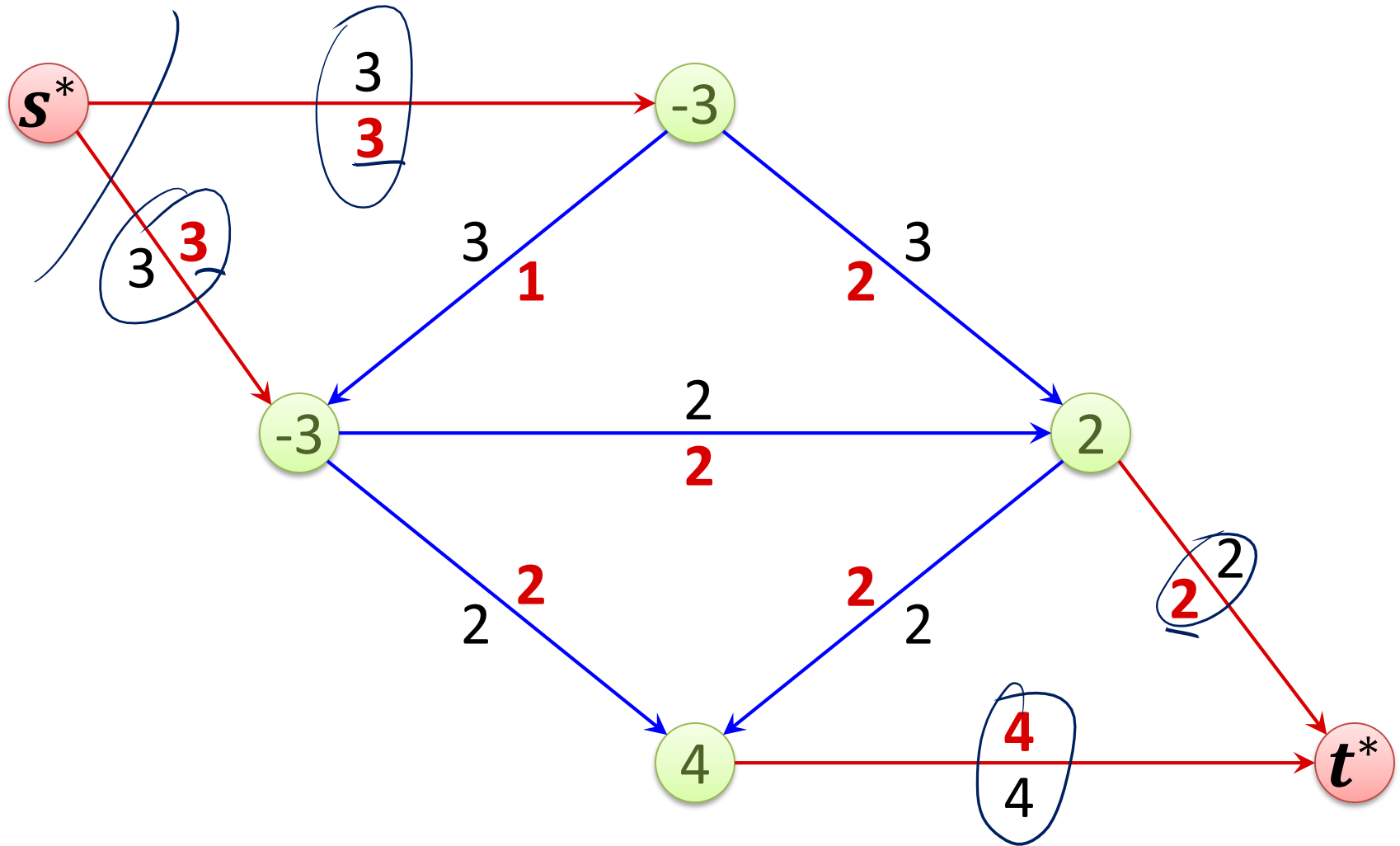
# Reduction to Maximum Flow

- Add “super-source”  $s^*$  and “super-sink”  $t^*$  to network





# Example



# Formally...

**Reduction:** Get graph  $G'$  from graph as follows

- Node set of  $G'$  is  $V \cup \{s^*, t^*\}$
- Edge set is  $E$  and edges
  - $(s^*, v)$  for all  $v$  with  $d_v < 0$ , capacity of edge is  $-d_v$
  - $(v, t^*)$  for all  $v$  with  $d_v > 0$ , capacity of edge is  $d_v$

## Observations:

- Capacity of min  $s^*$ - $t^*$  cut is at most  $D$  (e.g., the cut  $(s^*, V \cup \{t^*\})$ )
- A feasible circulation on  $G$  can be turned into a feasible flow of value  $D$  of  $G'$  by saturating all  $(s^*, v)$  and  $(v, t^*)$  edges.
- Any flow of  $G'$  of value  $D$  induces a feasible circulation on  $G$ 
  - $(s^*, v)$  and  $(v, t^*)$  edges are saturated
  - By removing these edges, we get exactly the demand constraints

# Circulation with Demands

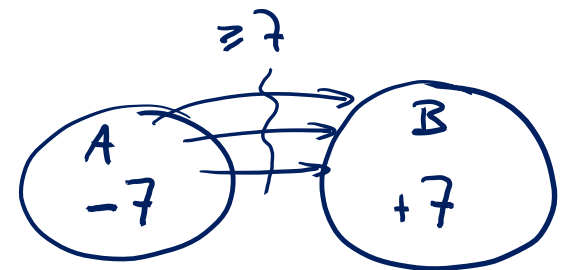
**Theorem:** There is a feasible circulation with demands  $d_v, v \in V$  on graph  $G$  if and only if there is a flow of value  $D$  on  $G'$ .

- If all capacities and demands are integers, there is an integer circulation

The **max flow min cut theorem** also implies the following:

**Theorem:** The graph  $G$  has a feasible circulation with demands  $d_v, v \in V$  if and only if for all cuts  $(A, B)$ ,

$$\sum_{v \in B} d_v \leq c(A, B).$$



**Given:** Directed network  $G = (V, E)$  with

- Edge capacities  $c_e > 0$  and **lower bounds  $0 \leq \ell_e \leq c_e$  for  $e \in E$**
- Node demands  $d_v \in \mathbb{R}$  for all  $v \in V$ 
  - $d_v > 0$ : node needs flow and therefore is a sink
  - $d_v < 0$ : node has a supply of  $-d_v$  and is therefore a source
  - $d_v = 0$ : node is neither a source nor a sink

**Flow:** Function  $f: E \rightarrow \mathbb{R}_{\geq 0}$  satisfying

- *Capacity Conditions:*  $\forall e \in E: \ell_e \leq f(e) \leq c_e$
- *Demand Conditions:*  $\forall v \in V: f^{\text{in}}(v) - f^{\text{out}}(v) = d_v$

**Objective:** Does a flow  $f$  satisfying all conditions exist?  
If yes, find such a flow  $f$ .