



# Design and Analysis of Algorithms

## Network Flow

**Si Wu**

School of CSE, SCUT

[cswusi@scut.edu.cn](mailto:cswusi@scut.edu.cn)

TA: [1684350406@qq.com](mailto:1684350406@qq.com)



# Topics

- **Max-Flow and Min-Cut Problems**
- **Ford-Fulkerson Algorithm**
- **Max-Flow Min-Cut Theorem**
- **Capacity-Scaling Algorithm**
- **Shortest Augmenting Paths**
- **Blocking-Flow Algorithm**

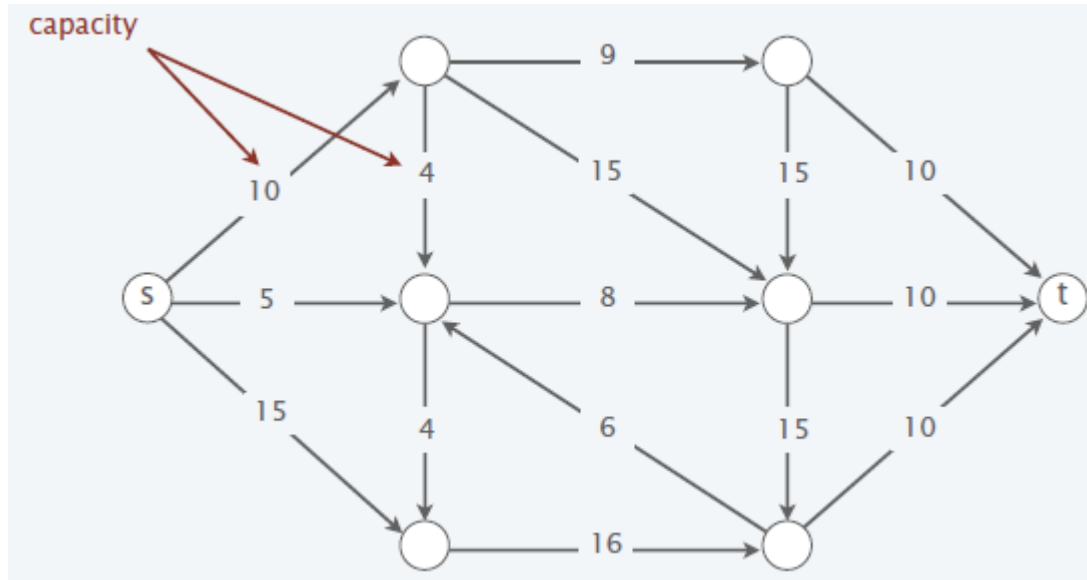


# Flow Network

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

- Digraph  $(V, E)$  with source  $s \in V$  and sink  $t \in V$ .
- Non-negative capacity  $c(e)$  for each  $e \in E$ .

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



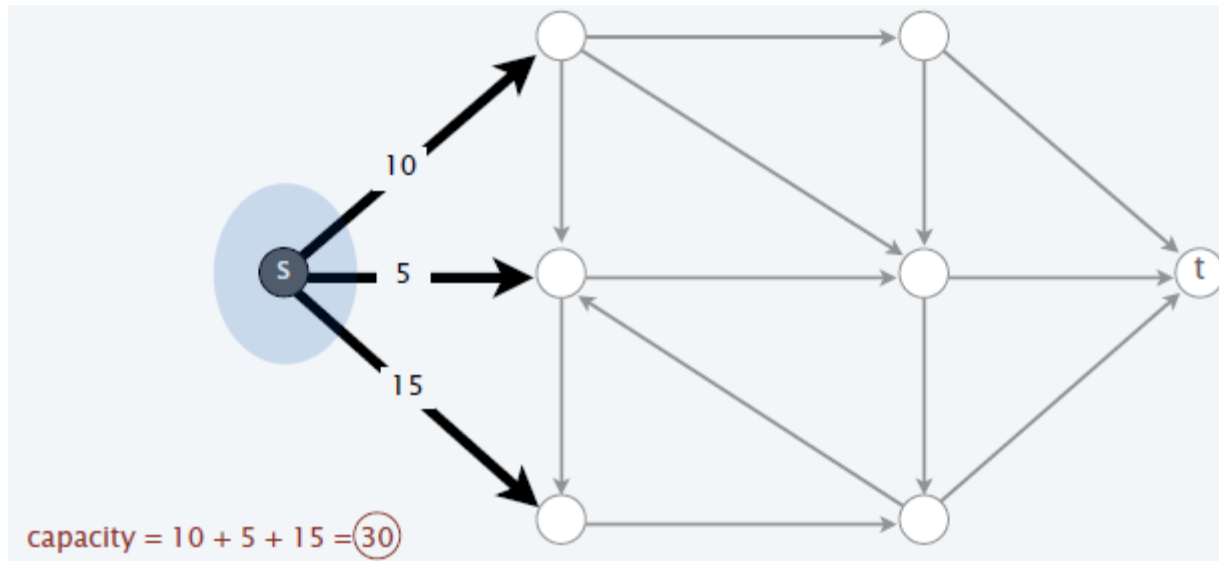


# Minimum-Cut Problem

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

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

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



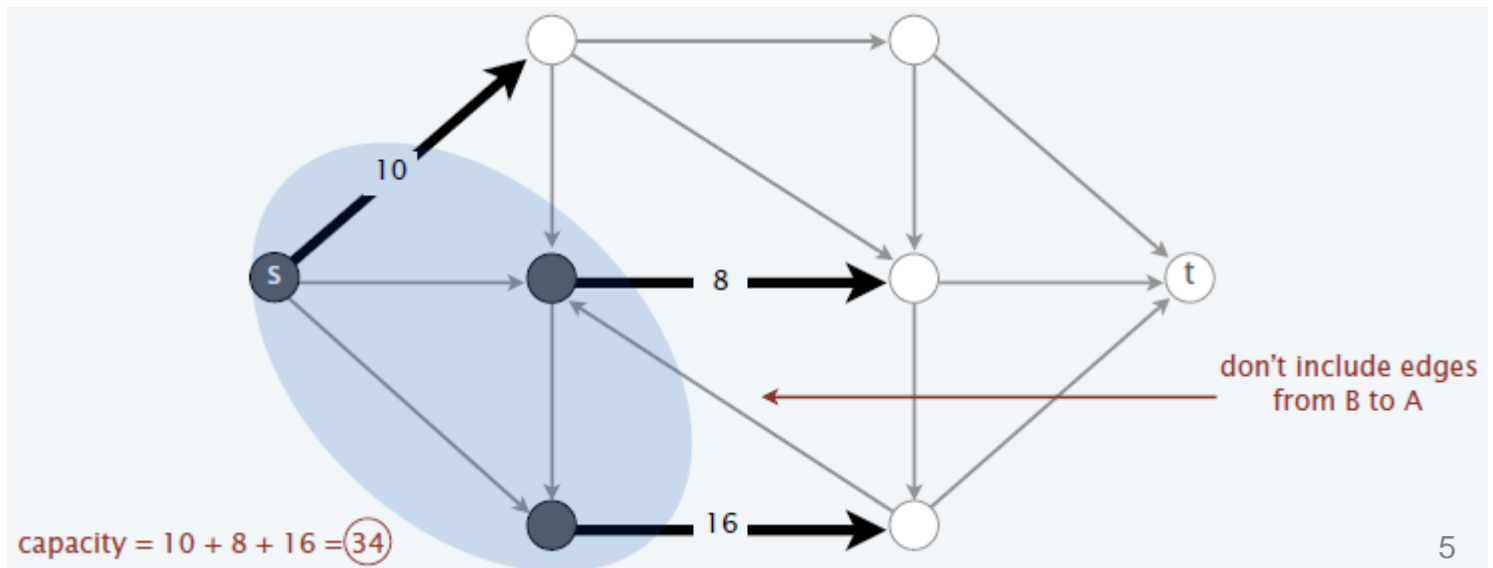


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# Minimum-Cut Problem

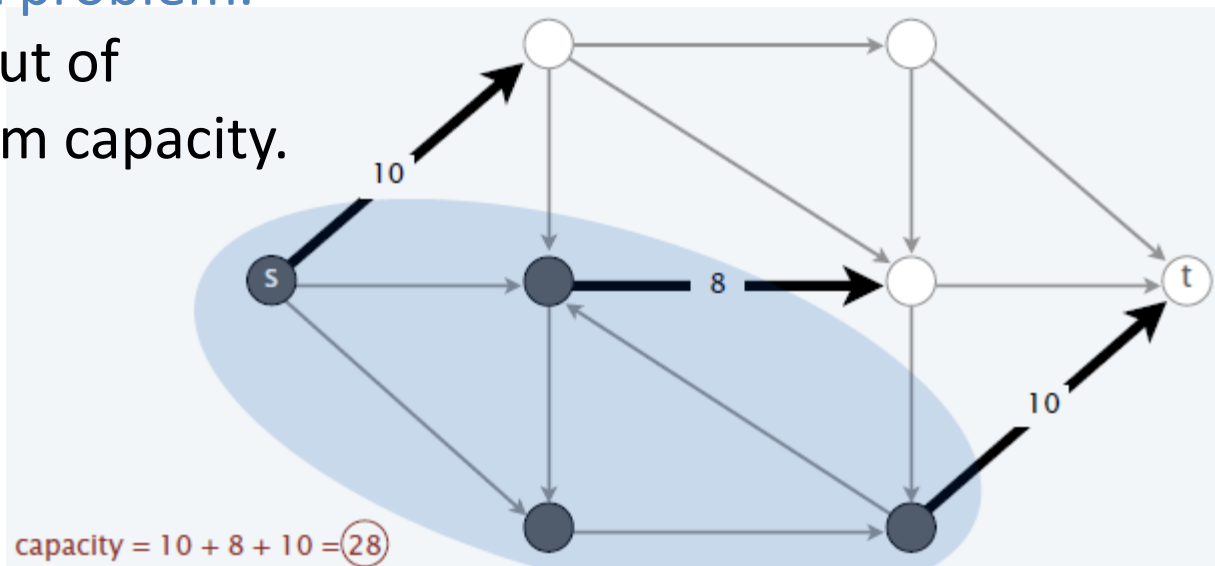
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**Min-cut problem.**

Find a cut of minimum capacity.



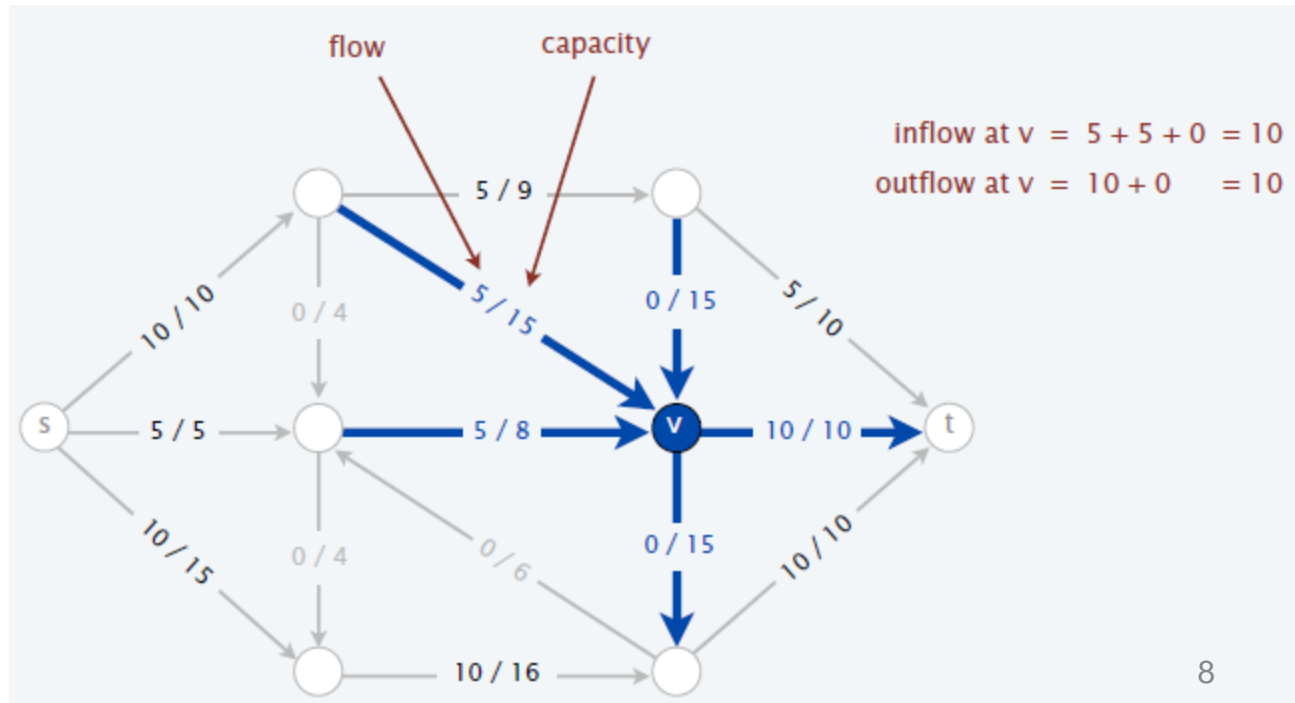

$$cap(A, B) = 45 \text{ (} 20 + 25 \text{)}$$




# 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  $v \in V - \{s, t\}$ :  $\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$  [flow conservation]





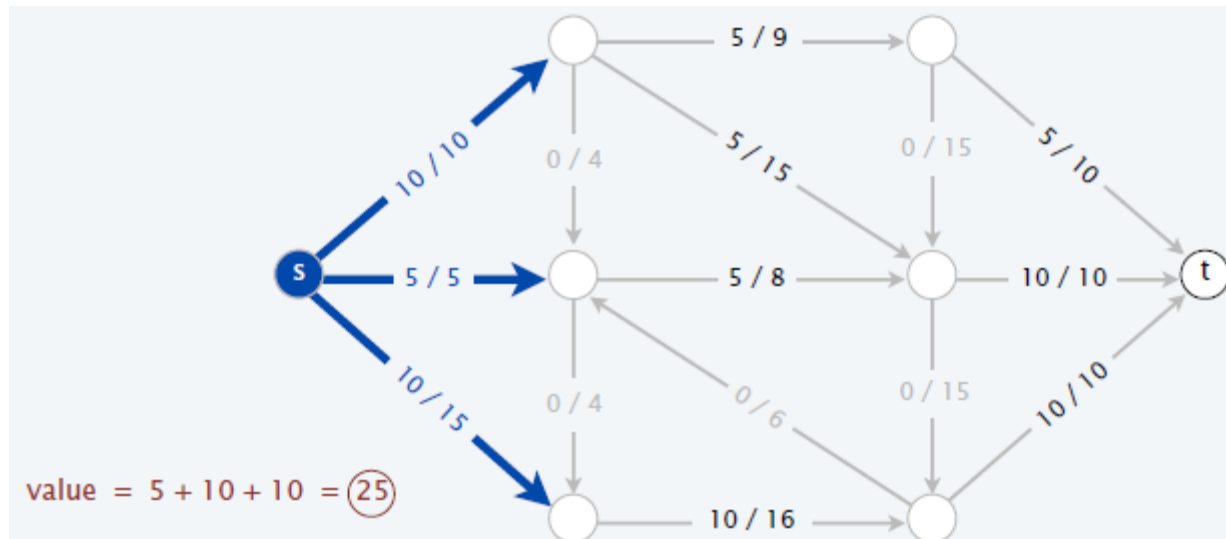


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**Def.** The value of a flow  $f$  is:  $val(f) = \sum_{e \text{ out of } s} f(e) - \sum_{e \text{ into } s} f(e)$





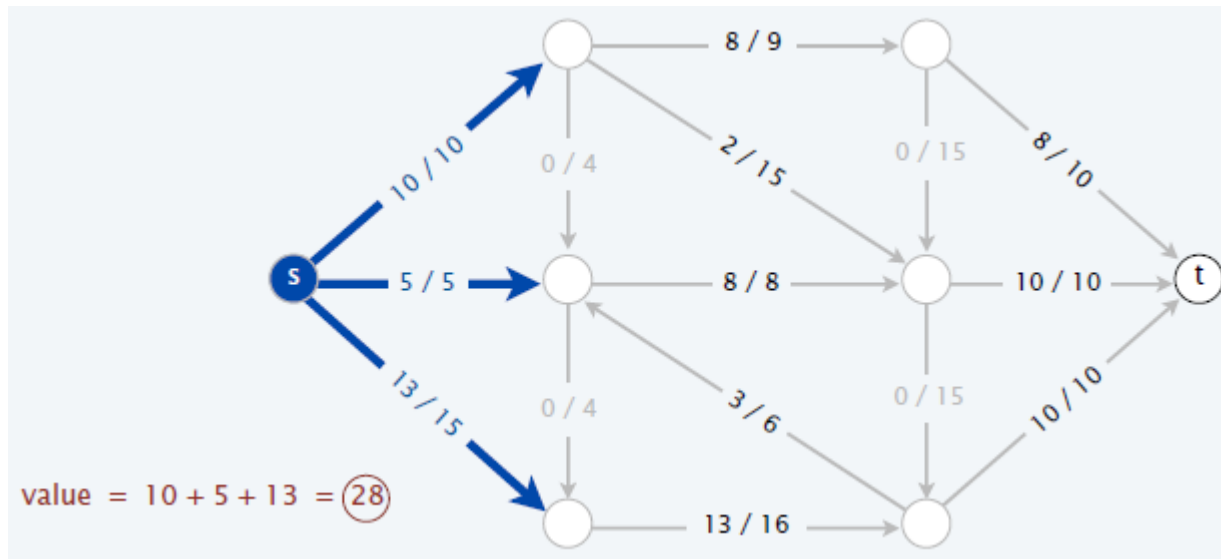
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**Max-flow problem.** Find a flow of maximum value?

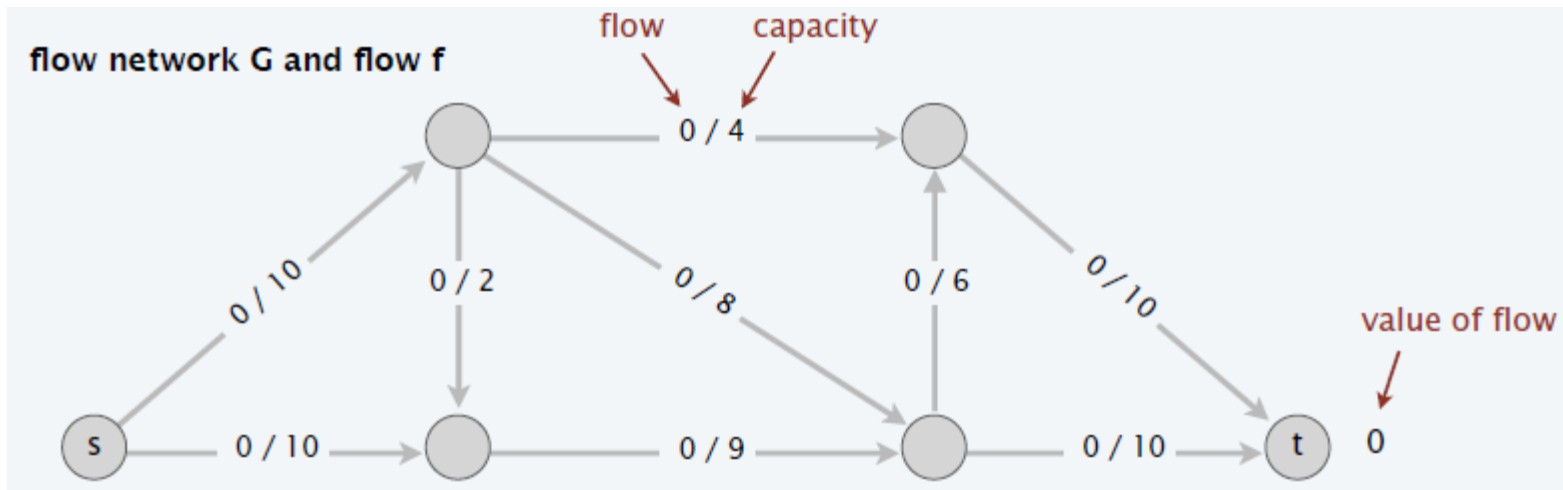




# Towards a Max-Flow Algorithm

## Greedy algorithm.

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



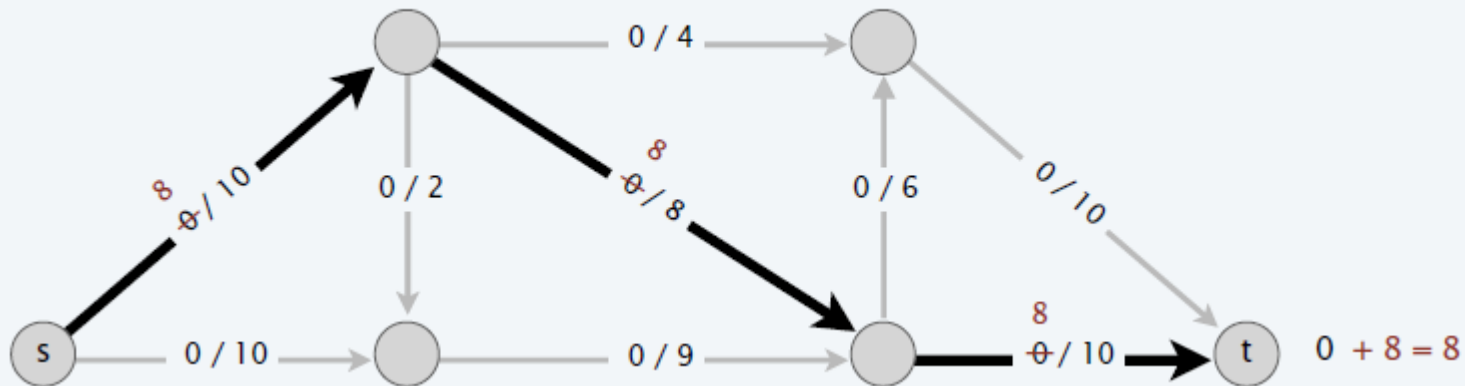


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flow network  $G$  and flow  $f$



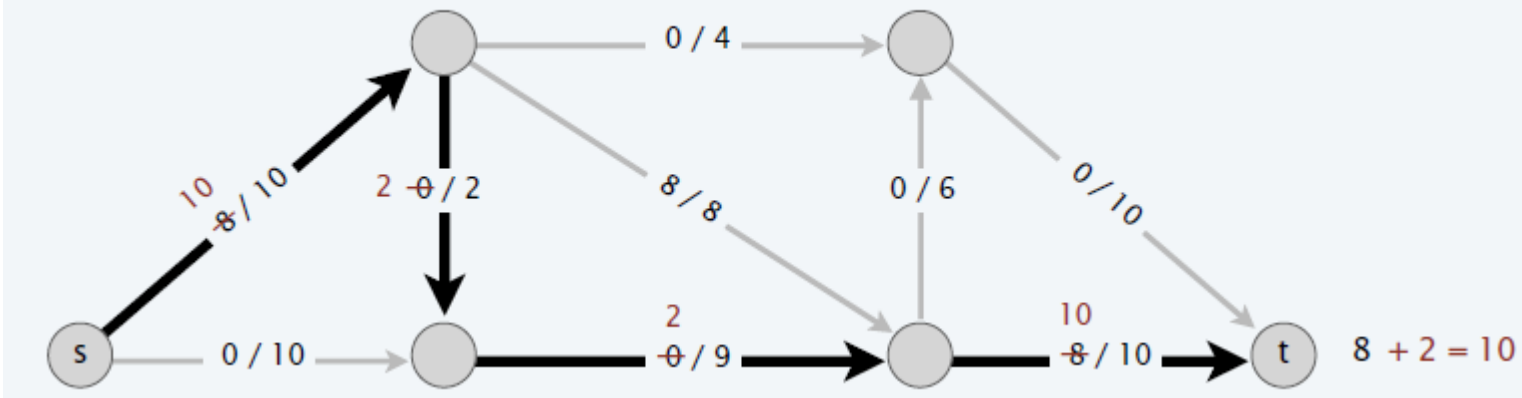


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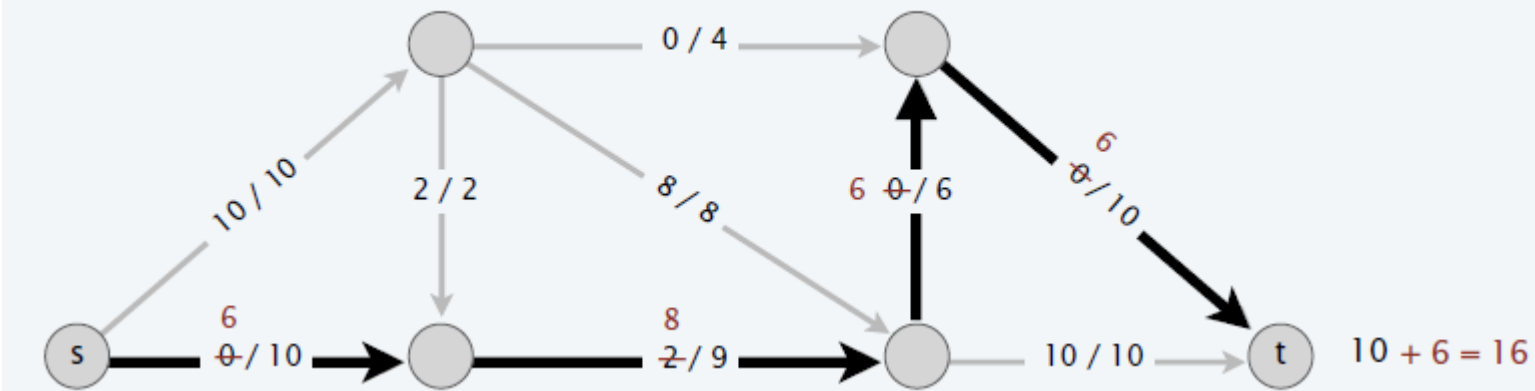


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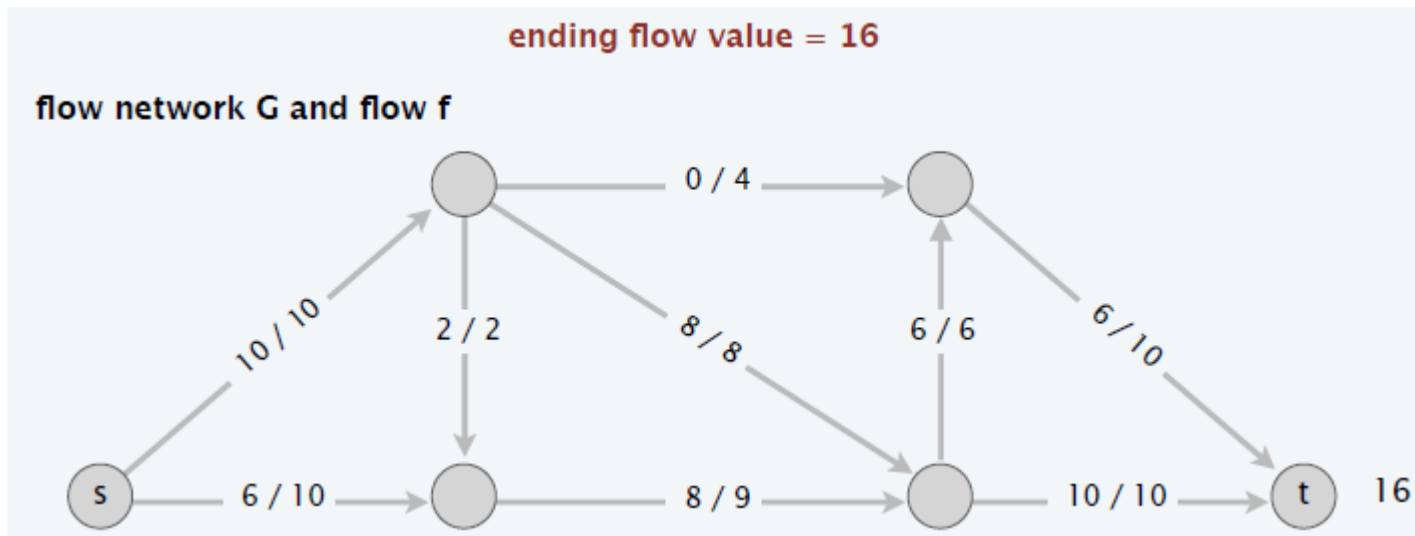




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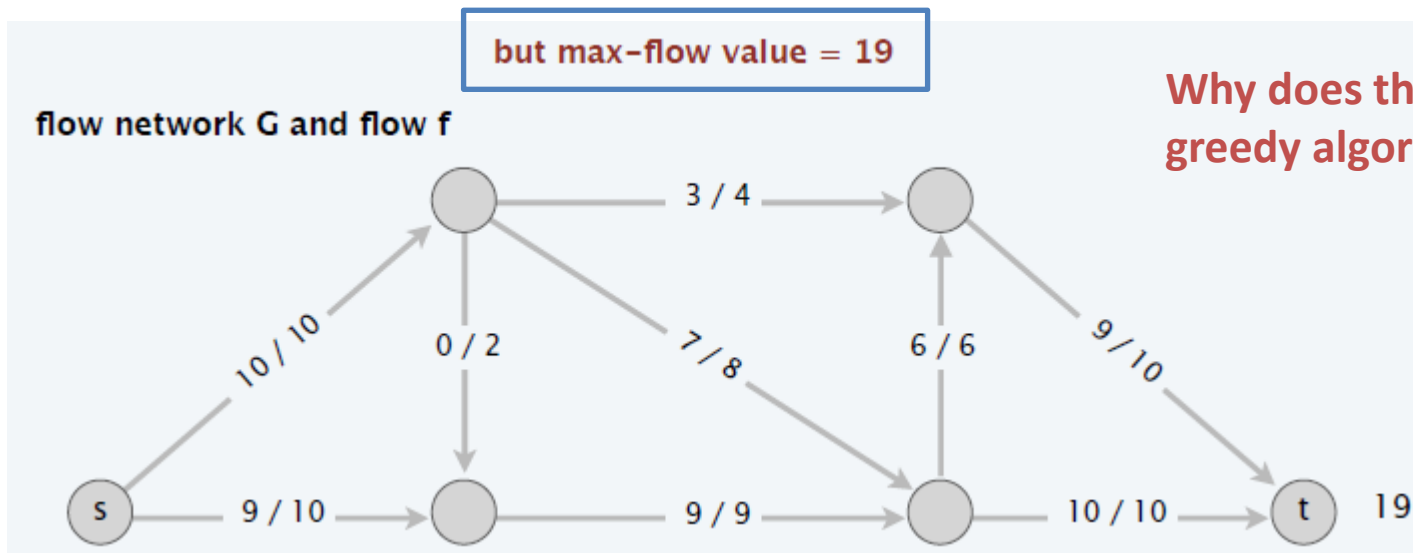




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- Repeat until get stuck.







# Why the Greedy Algorithm Fails

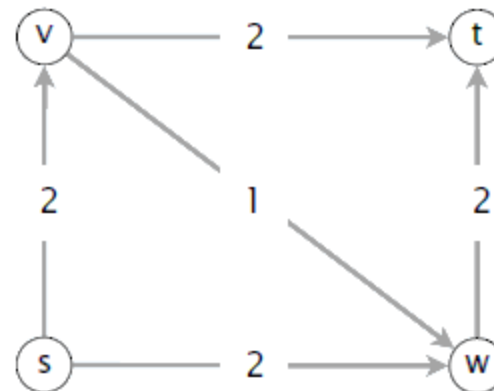
Q. Why does the greedy algorithm fail?

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

Ex.

- The max flow is unique; flow on edge  $(v, w)$  is zero.
- Greedy algorithm could choose  $s \rightarrow v \rightarrow w \rightarrow t$  for first augmenting path.

flow network G



Need some mechanism to “undo” bad decision.

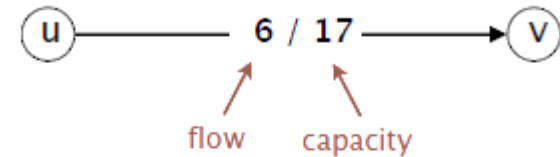


# Residual Network

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

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

original flow network  $G$



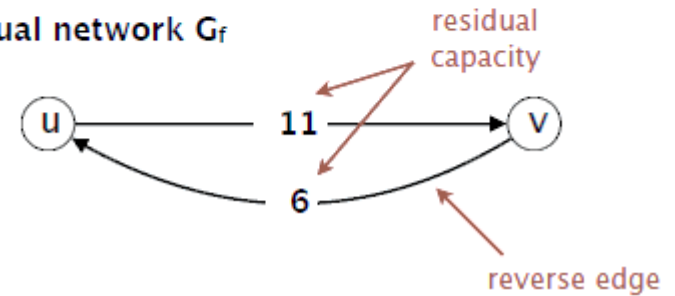
Reverse edge.  $e^{reverse} = (v, u)$ .

- “Undo” flow sent.

Residual capacity.

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

residual network  $G_f$



Edges with positive residual capacity

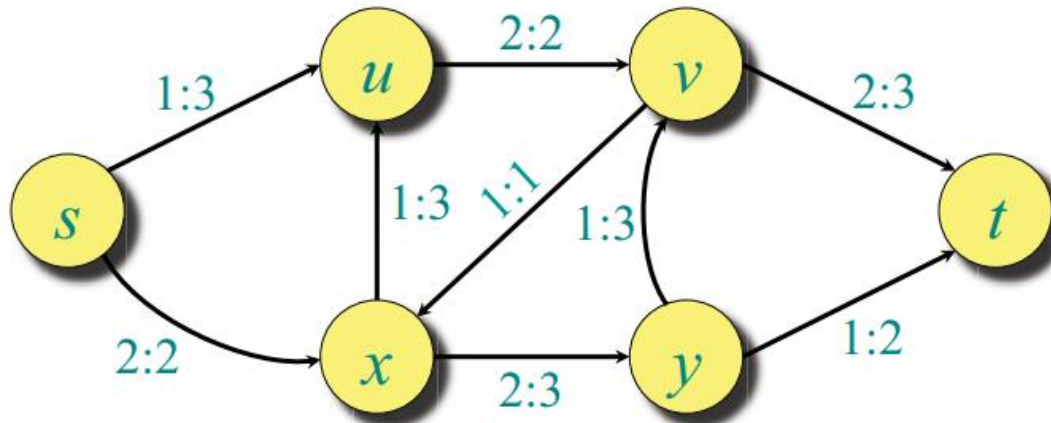
Residual network.  $G_f = (V, E_f, s, t, c_f)$ .

- $E_f = \{e: f(e) < c(e)\} \cup \{e^{reverse}: f(e) > 0\}$ .



# Flow and Residual Network

Let  $f$  be a flow on  $G = (V, E)$ :

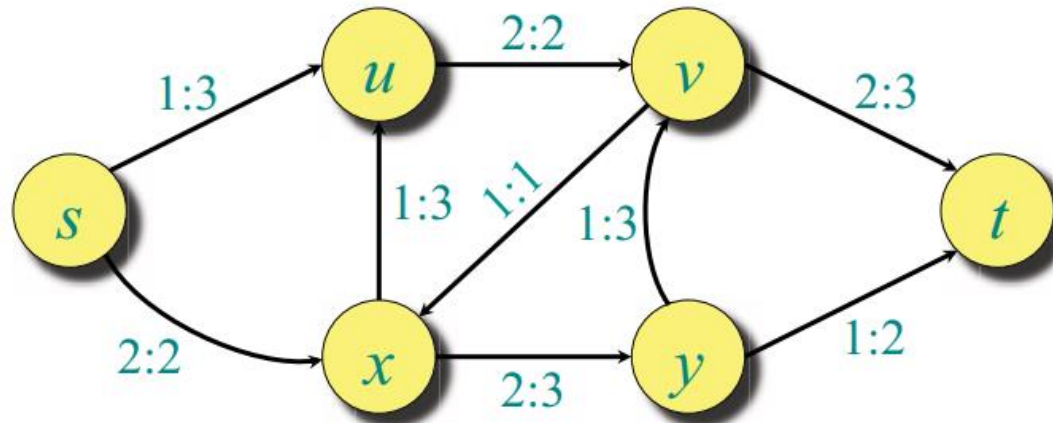


The residual network  $G_f(V, E_f)$ :

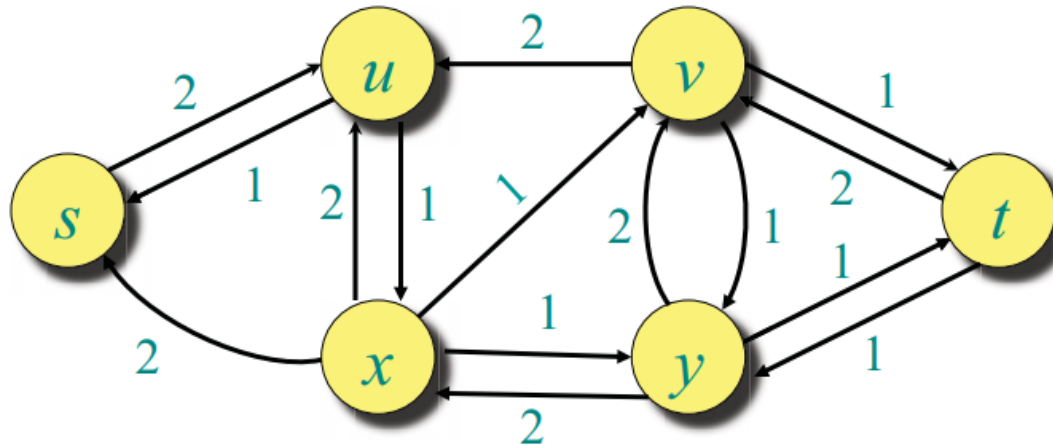


# Flow and Residual Network

Let  $f$  be a flow on  $G = (V, E)$ :



The residual network  $G_f(V, E_f)$ :





# Augmenting Path

**Def.** An augmenting path is a simple  $s \rightarrow t$  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 call **Augment**, the resulting  $f'$  is a flow and  $val(f') = val(f) + bottleneck(G_f, P)$ .



# Augmenting Path

**Key property.** Let  $f$  be a flow and let  $P$  be an augmenting path in  $G_f$ . Then, after call **Augment**, the resulting  $f'$  is a flow and  $val(f') = val(f) + bottleneck(G_f, P)$ .

**Augment** ( $f, c, P$ )

---

$b \leftarrow$  bottleneck capacity of path  $P$ .

**For each** edge  $e \in P$

**If** ( $e \in E$ )

$f[e] \leftarrow f[e] + b$ .

**Else**

$f[e^{reverse}] \leftarrow f[e^{reverse}] - b$ .

**Return**  $f$ .



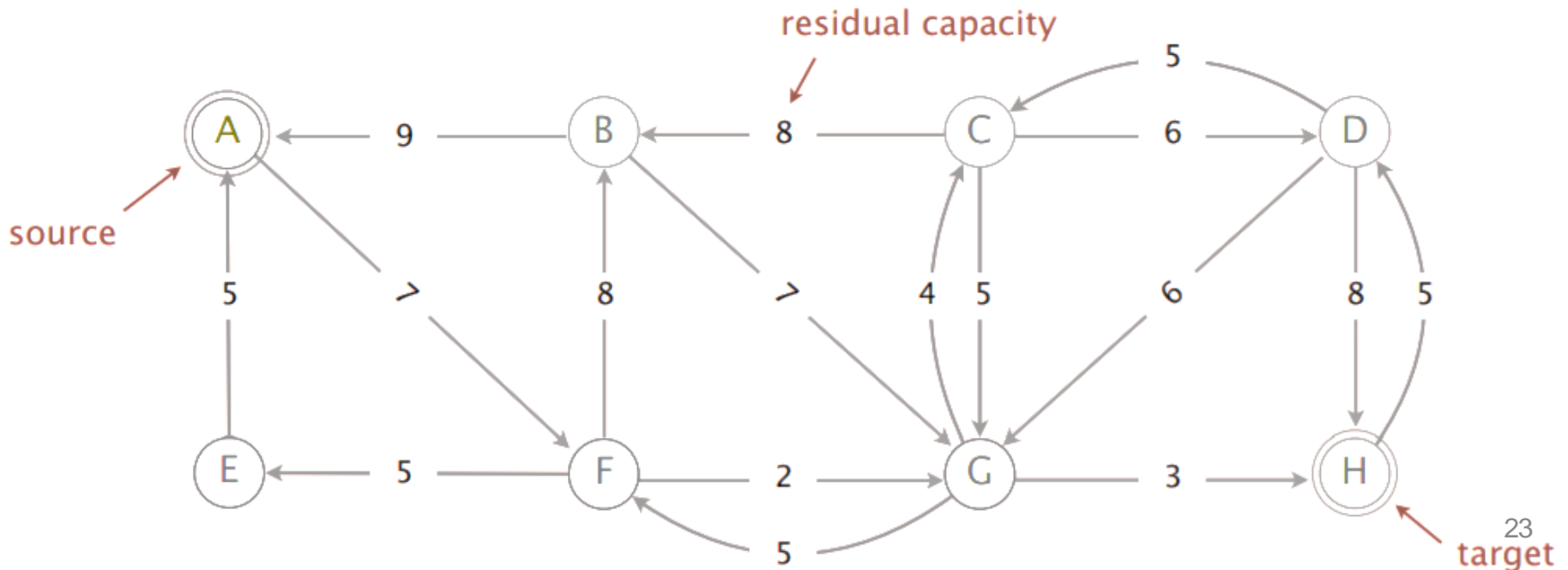
# Augmenting Path

Which is the augmenting path of highest bottleneck capacity?

A.  $A \rightarrow F \rightarrow G \rightarrow H$

B.  $A \rightarrow F \rightarrow B \rightarrow G \rightarrow H$

C.  $A \rightarrow F \rightarrow B \rightarrow G \rightarrow C \rightarrow D \rightarrow H$





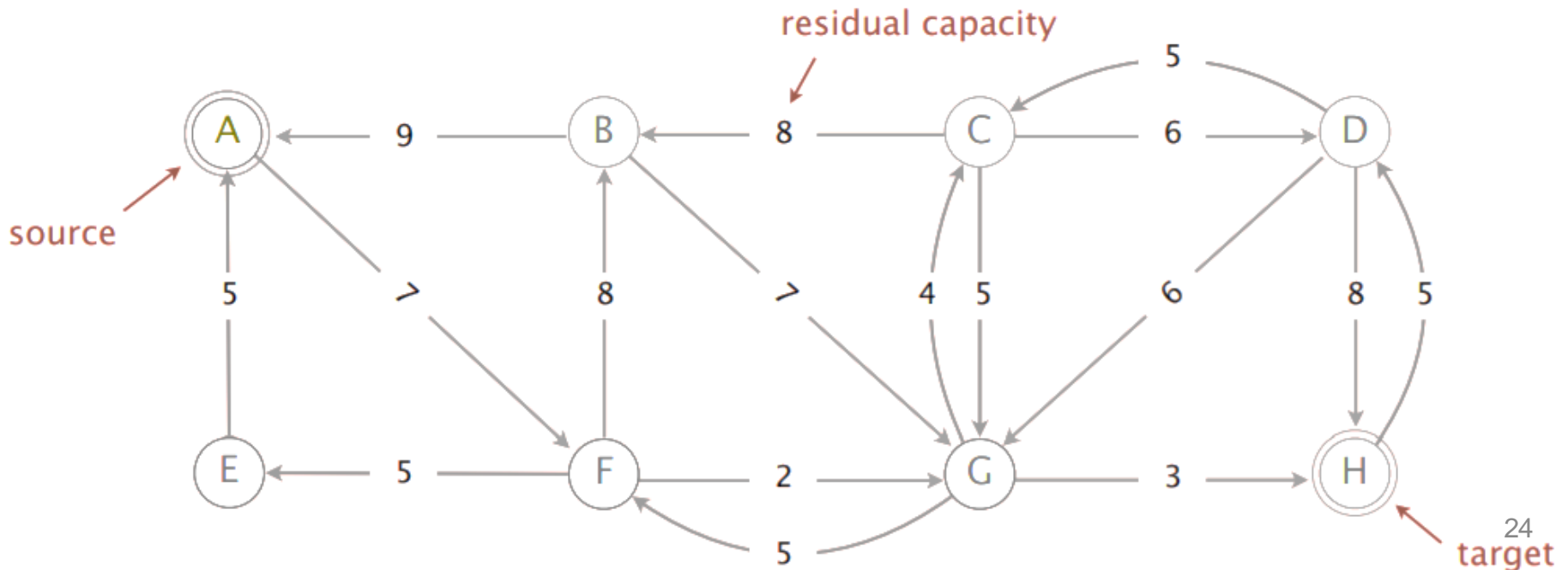
# Augmenting Path

Which is the augmenting path of highest bottleneck capacity?

A.  $A \rightarrow F \rightarrow G \rightarrow H$  (2)

B.  $A \rightarrow F \rightarrow B \rightarrow G \rightarrow H$  (3)

C.  $A \rightarrow F \rightarrow B \rightarrow G \rightarrow C \rightarrow D \rightarrow H$  (4)







# Ford-Fulkerson Algorithm

Ford-Fulkerson augmenting path algorithm.

- Start with  $f(e) = 0$  for each edge  $e \in E$ .
- Find an  $s \rightarrow t$  path  $P$  in the residual network  $G_f$ .
- Augment flow along path  $P$ .
- Repeat until you get stuck.

Ford-Fulkerson ( $G$ )

---

For each edge  $e \in E: f[e] \leftarrow 0$ .

$G_f \leftarrow$  residual network of  $G$  with respect to  $f$ .

While (there exists an  $s \rightarrow t$  augmenting path  $P$  in  $G_f$

$f \leftarrow \text{Augment}(f, c, P)$ .

    Update  $G_f$ .

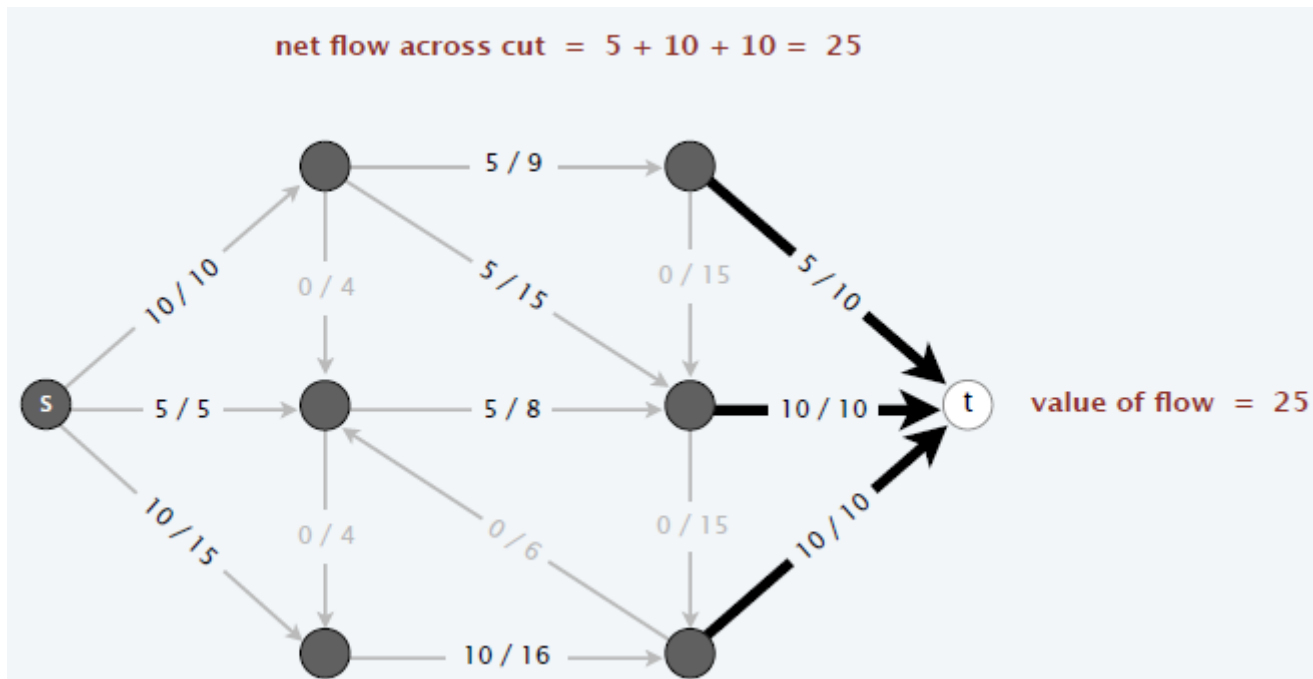
Return  $f$ .



# Relationship Between Flows and Cuts

**Flow value lemma.** Let  $f$  be any flow and let  $(A, B)$  be any cut. Then, the value of the flow  $f$  equals **the net flow across the cut  $(A, B)$** .

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

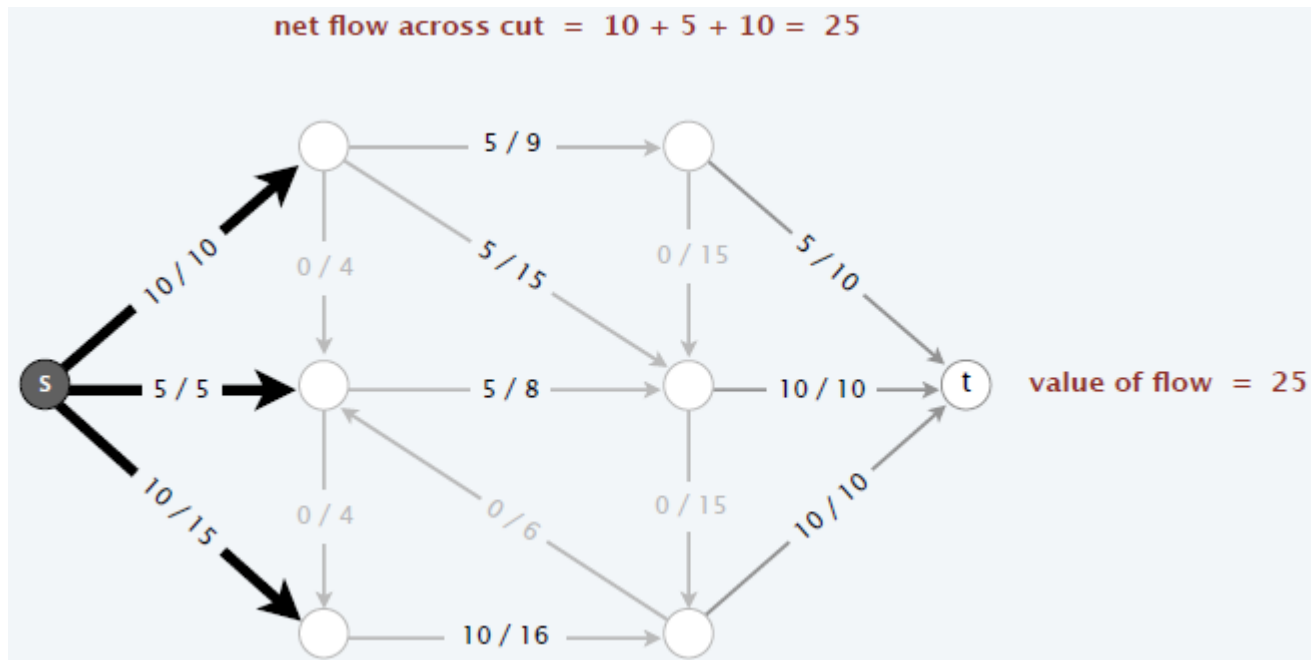




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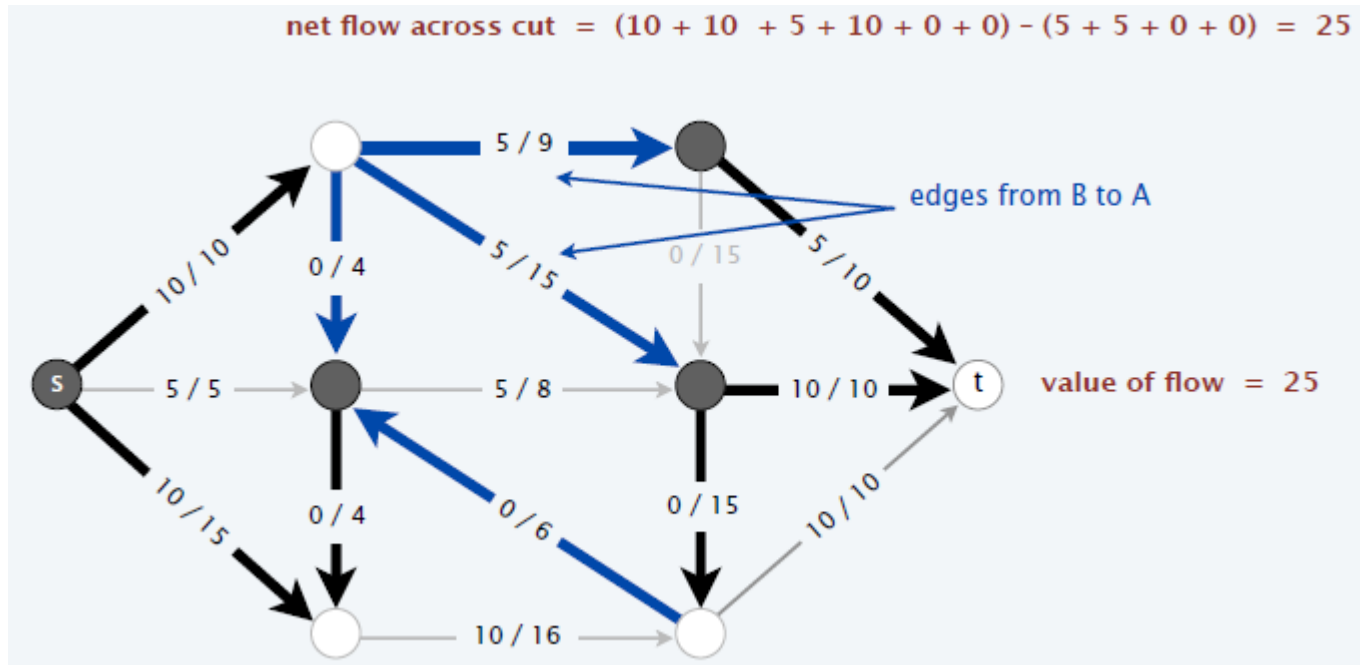




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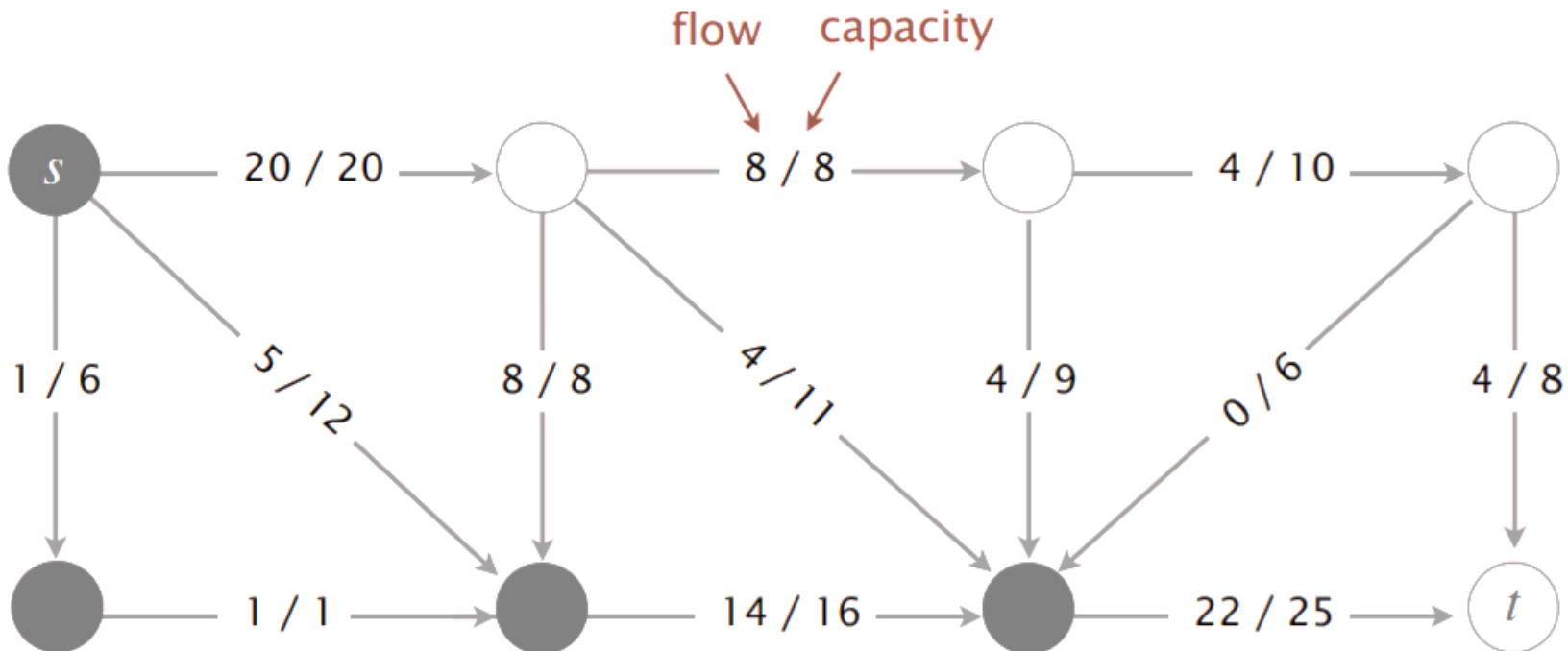




# Relationship Between Flows and Cuts

What is the net flow across the given *st*-cut?

$$val(f) = 26 \quad (20 + 22 - 8 - 4 - 4)$$





# Relationship Between Flows and Cuts

**Flow value lemma.** Let  $f$  be any flow and let  $(A, B)$  be any cut. Then, the value of the flow  $f$  equals the net flow across the cut  $(A, B)$ .

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

**Pf.**

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

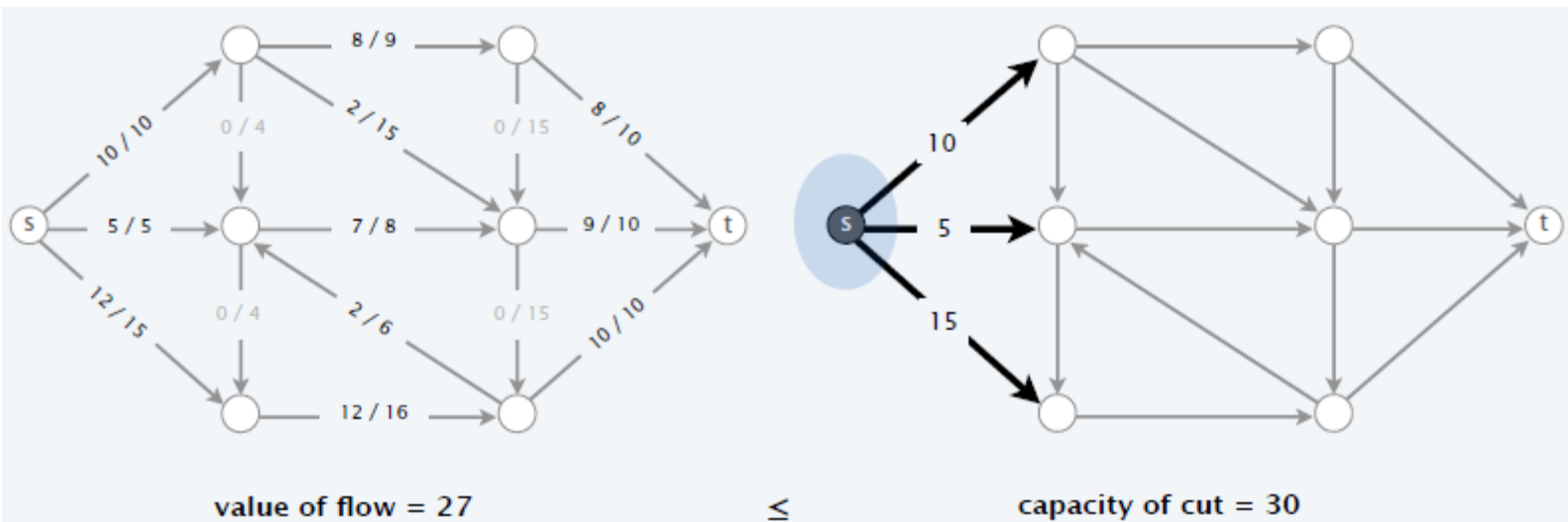
By flow conservation, all terms except for  $s$  are 0



# Relationship Between Flows and Cuts

**Weak duality.** Let  $f$  be any flow and  $(A, B)$  be any cut. Then,  $val(f) \leq cap(A, B)$ .

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



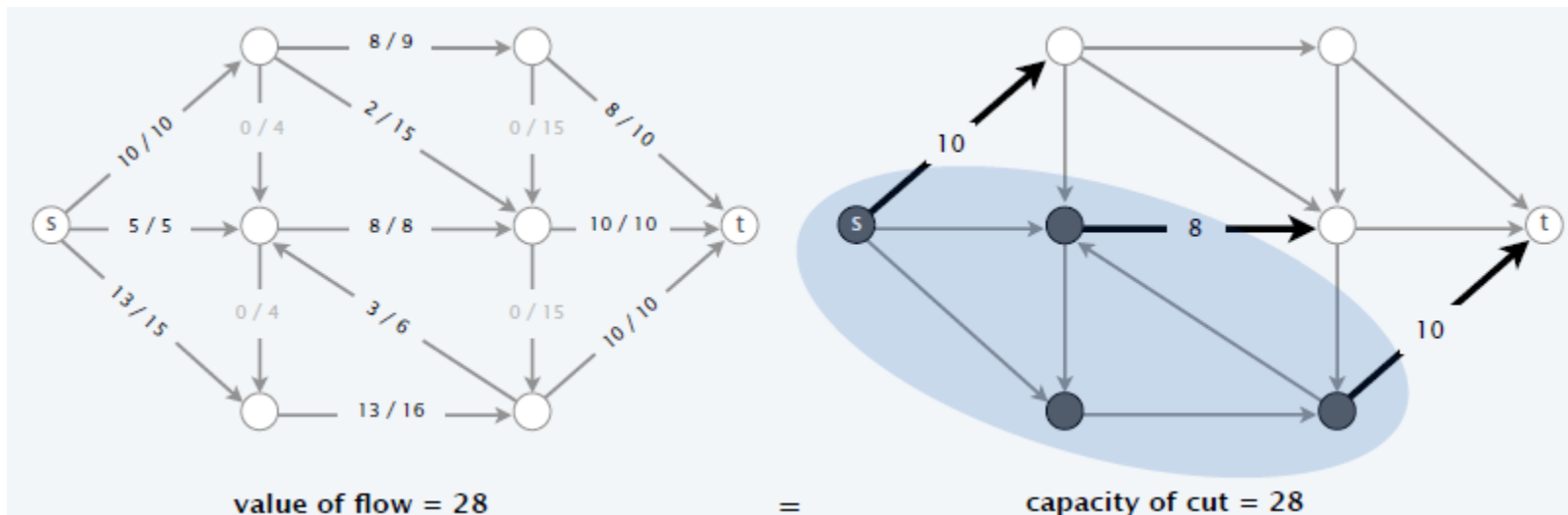


# Certificate of Optimality

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

**Pf.**

- For any flow  $f'$ :  $val(f') \leq cap(A, B) = val(f)$ .
- For any cut  $(A', B')$ :  $cap(A', B') \geq val(f) = cap(A, B)$ .







# Computing a Minimum Cut from a Maximum Flow

**Theorem.** Given any max flow  $f$ , can compute a min cut  $(A, B)$  in  $O(m)$  time.

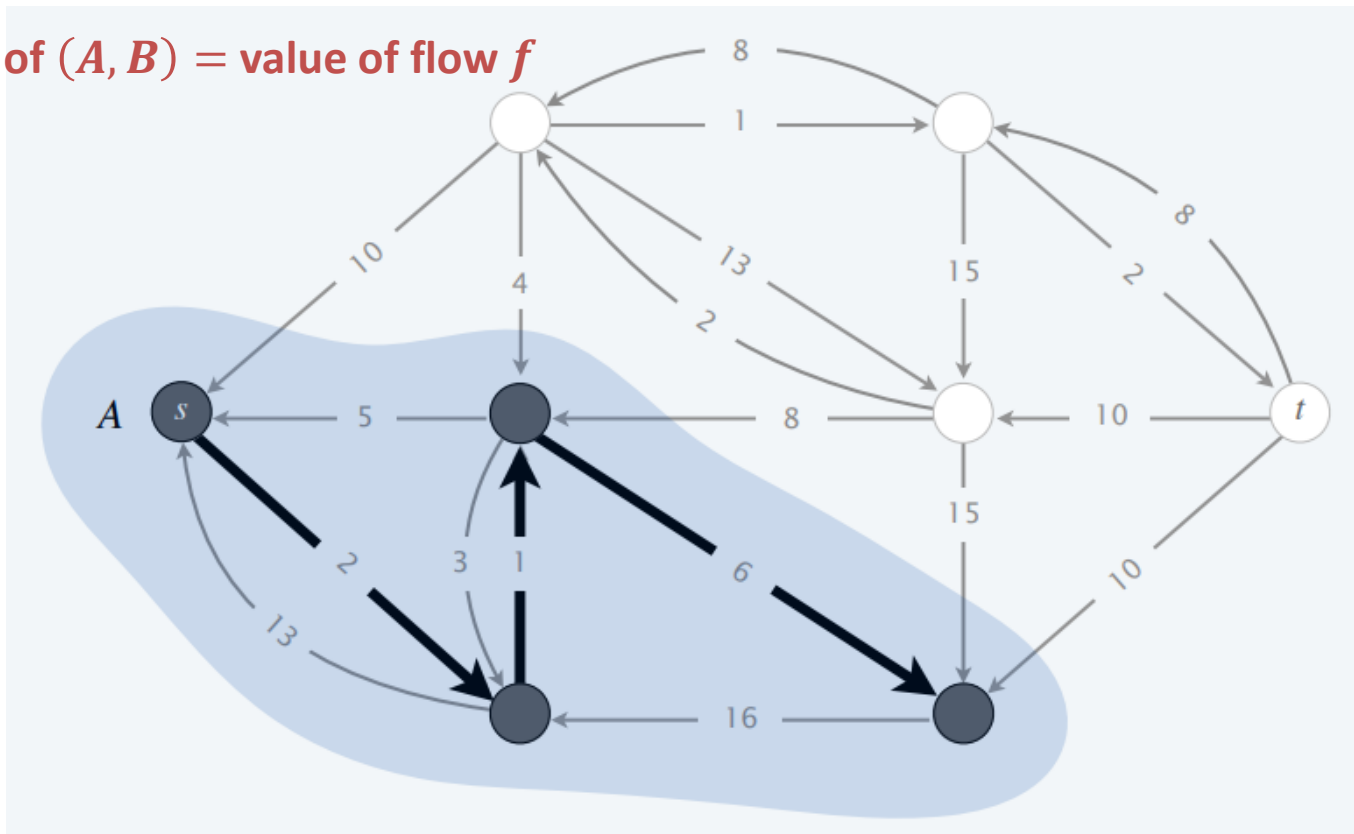


# Computing a Minimum Cut from a Maximum Flow

**Theorem.** Given any max flow  $f$ , can compute a min cut  $(A, B)$  in  $O(m)$  time.

**Pf.** Let  $A$  = set of nodes reachable from  $s$  in residual network  $G_f$ .

**Capacity of  $(A, B)$  = value of flow  $f$**





# Max-Flow Min-Cut Theorem

**Augmenting path theorem.** A flow  $f$  is a max flow iff no augmenting paths.

**Max-flow min-cut theorem.** Value of a max flow = capacity of a min cut.

**Pf.** The following three conditions are equivalent for any flow  $f$ :

- I. There exists a cut  $(A, B)$  such that  $cap(A, B) = val(f)$ .
- II.  $f$  is a max flow.
- III. There is no augmenting path with respect to  $f$ .

[I  $\Rightarrow$  II]

- Suppose that  $(A, B)$  is a cut such that  $cap(A, B) = val(f)$ .
- Then, for any flow  $f'$ :  $val(f') \leq cap(A, B) = val(f)$ .
- Thus,  $f$  is a max flow.



# Max-Flow Min-Cut Theorem

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- II.  $f$  is a max flow.
- III. There is no augmenting path with respect to  $f$ .

[II  $\Rightarrow$  III] We prove contrapositive:  $\sim$ III  $\Rightarrow$   $\sim$ II.

- Suppose that there is an augmenting path with respect to  $f$ .
- Can improve flow  $f$  by sending flow along this path.
- Thus,  $f$  is not a max flow.

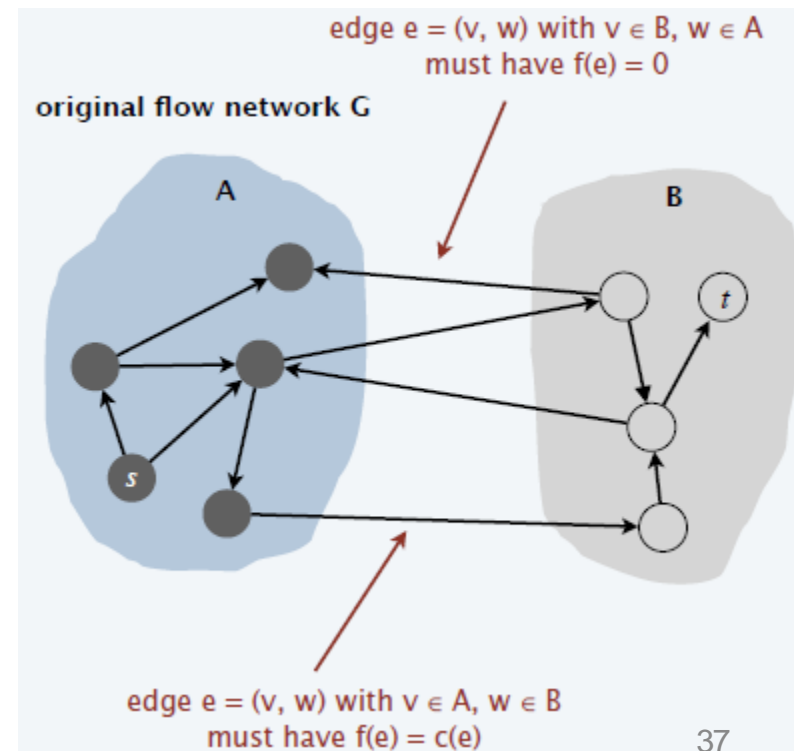


# Max-Flow Min-Cut Theorem

[III  $\Rightarrow$  I]

- Let  $f$  be a flow with no augmenting paths.
- Let  $A$  be set of nodes reachable from  $s$  in residual network  $G_f$ .
- By definition of cut  $A: s \in A$ .
- By definition of flow  $f: t \notin A$ .

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



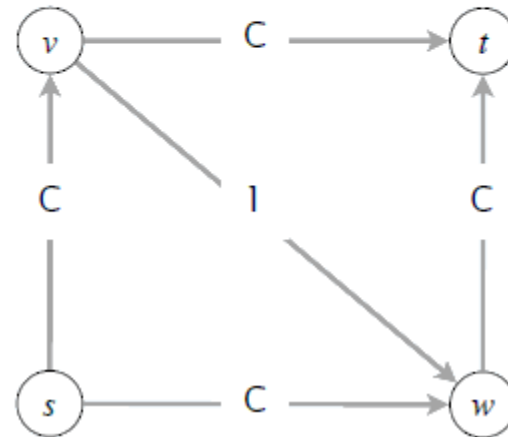


# Bad Case for Ford-Fulkerson

If max capacity is  $C$ , then algorithm can take  $\geq C$  iterations.

- $s \rightarrow v \rightarrow w \rightarrow t$
- $s \rightarrow w \rightarrow v \rightarrow t$
- $s \rightarrow v \rightarrow w \rightarrow t$
- $s \rightarrow w \rightarrow v \rightarrow t$
- ...
- $s \rightarrow v \rightarrow w \rightarrow t$
- $s \rightarrow w \rightarrow v \rightarrow t$

← each augmenting path  
sends only 1 unit of flow  
(# augmenting paths =  $2C$ )





# Choosing Good Augmenting Paths

Use care when selecting augmenting paths.

- Some choices lead to exponential algorithms.
- Clever choices lead to polynomial algorithms.

**Pathology.** If capacities are irrational, algorithm does not guarantee to terminate (or converge to correct answer)!

**Goal.** Choose augmenting paths so that:

- Can find augmenting paths efficiently.
- Few iterations.

Choose augmenting paths with:

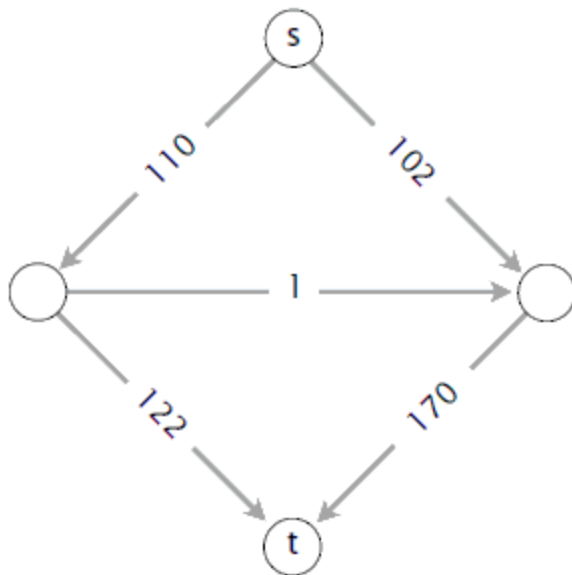
- Max bottleneck capacity (“fattest”).
- Sufficiency large bottleneck capacity.
- Fewest edges.



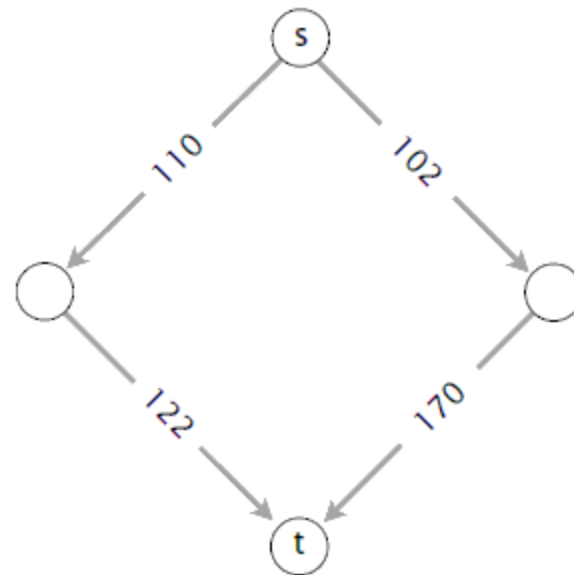
# Capacity-Scaling Algorithm

**Intuition.** Choose augmenting path with highest bottleneck capacity: it increases flow by max possible amount in given iteration.

- Don't worry about finding exact highest bottleneck path.
- Maintain scaling parameter  $\Delta$ .
- Let  $G_f(\Delta)$  be the part of the residual network consisting of only those arcs with capacity  $\geq \Delta$ .



$G_f$



$G_f(\Delta), \Delta = 100$





# Capacity-Scaling Algorithm

## Capacity-Scaling ( $G$ )

---

For each edge  $e \in E$ :  $f[e] \leftarrow 0$ .

$\Delta \leftarrow$  largest power of 2  $\leq C$ .

While ( $\Delta \geq 1$ )

$G_f(\Delta) \leftarrow \Delta$ -residual network of  $G$  with respect to flow  $f$ .

While (there exists an  $s \rightarrow t$  path  $P$  in  $G_f(\Delta)$ )

$f \leftarrow \text{Augment}(f, c, P)$ .

Update  $G_f(\Delta)$ .

$\Delta \leftarrow \Delta/2$ .

Return  $f$ .



# Capacity-Scaling Algorithm: Proof of Correctness

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

**Integrality invariant.** All flows and residual capacities are integral.

**Theorem.** If capacity-scaling algorithm terminates, then  $f$  is a max flow.

**Pf.**

- By integrality invariant, when  $\Delta = 1 \implies G_f(\Delta) = G_f$ .
- Upon termination of  $\Delta = 1$  phase, there are no augmenting paths.



# Shortest Augmenting Path

Q. Which augmenting path?

A. The one with the fewest edges (can find via **Breadth-First-Search**).

## Shortest-Augmenting-Path ( $G$ )

---

For each edge  $e \in E: f[e] \leftarrow 0$ .

$G_f \leftarrow$  residual network of  $G$  with respect to flow  $f$ .

**While** (there exists an  $s \rightarrow t$  path in  $G_f$ )

$P \leftarrow$  **Breadth-First-Search** ( $G_f$ ).

$f \leftarrow$  **Augment** ( $f, c, P$ )

Update  $G_f$ .

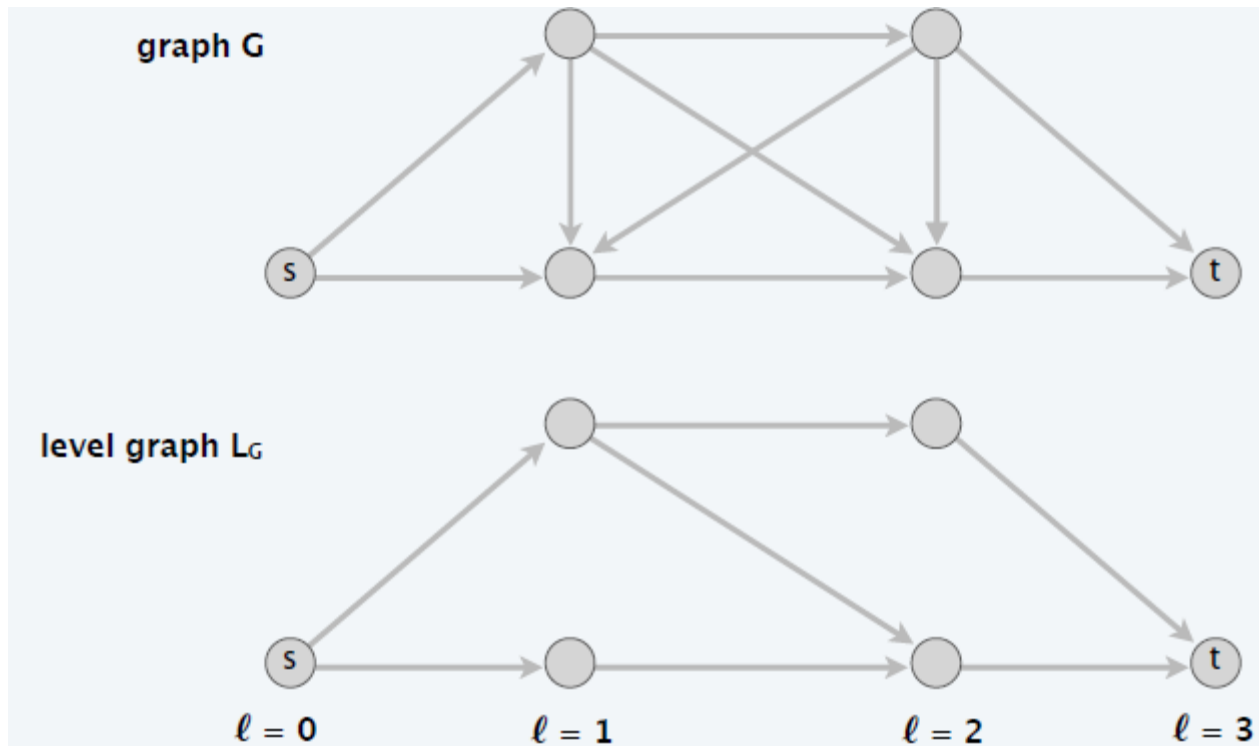
**Return**  $f$ .



# Shortest Augmenting Path: Analysis

**Def.** Given a digraph  $G = (V, E)$  with source  $s$ , its **level graph** is defined by:

- $l(v)$  = number of edges in shortest path from  $s$  to  $v$ .
- $L_G = (V, E_G)$  is the subgraph of  $G$  that contains only those edge  $(v, w) \in E$  with  $l(w) = l(v) + 1$ .



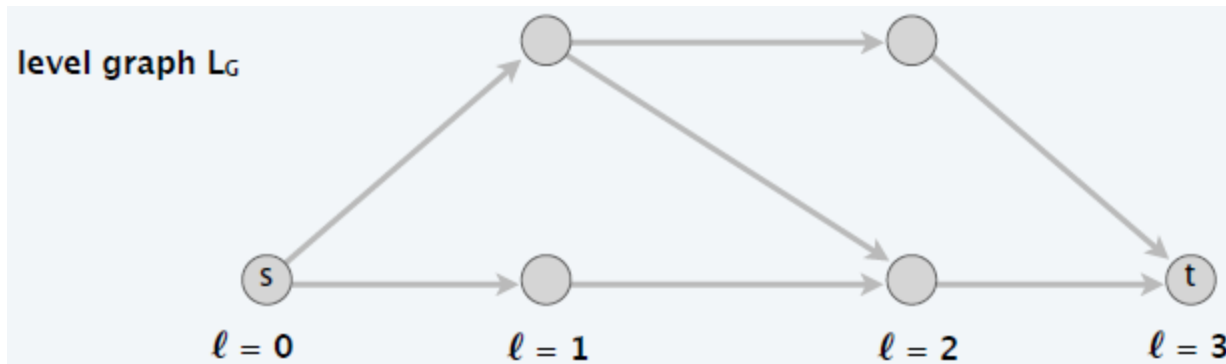


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**Key property.**  $P$  is a shortest path  $s \rightarrow v$  path in  $G$  iff  $P$  is an  $s \rightarrow v$  path in  $L_G$ .



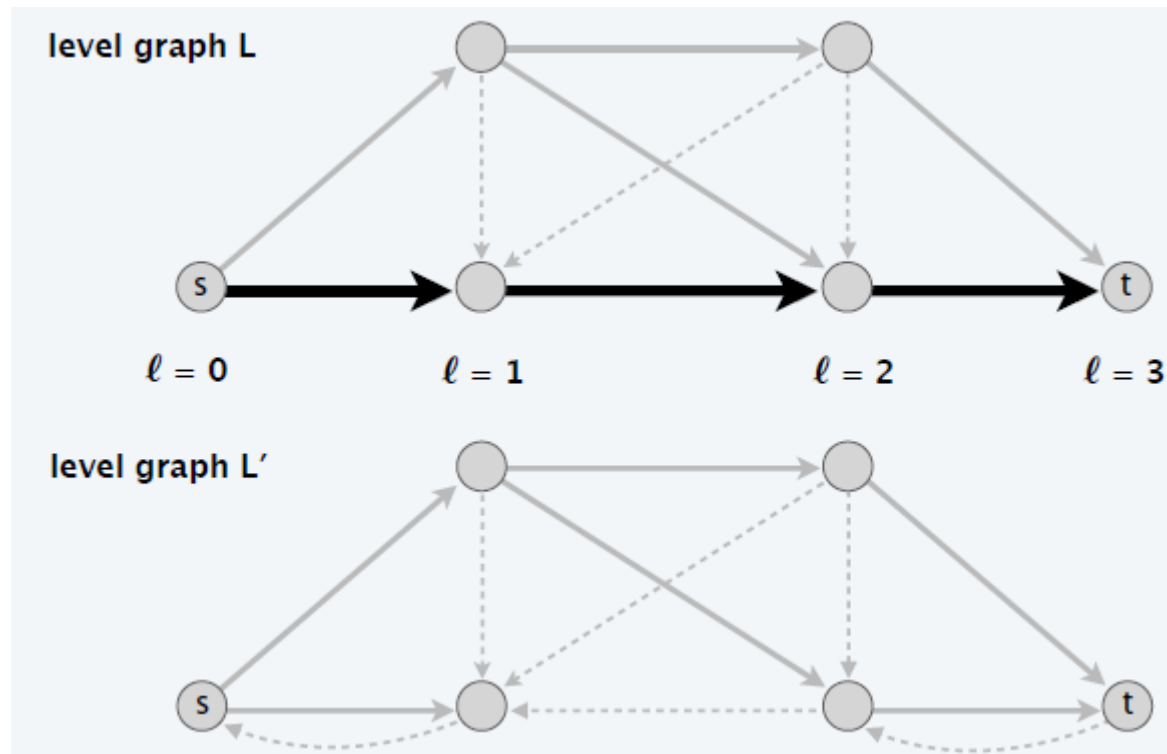


# Shortest Augmenting Path: Analysis

**Lemma 1.** The length of a shortest augmenting path never decreases.

- Let  $f$  and  $f'$  be flow before and after a shortest-path augmentation.
- Let  $L$  and  $L'$  be level graphs of  $G_f$  and  $G_{f'}$ .
- Only back edges added to  $G_{f'}$ .

(any path with a back edge is longer than previous length)





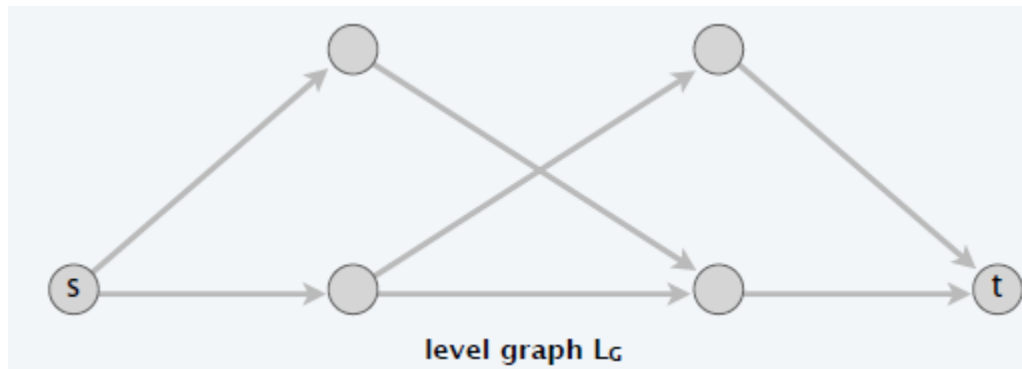
# Blocking-Flow Algorithm

## Two types of augmentations.

- Normal: length of shortest path does not change.
- Special: length of shortest path strictly increases.

## Phase of normal augmentations.

- Explicitly maintain level graph  $L_G$ .
- Start at  $s$ , advance along an edge in  $L_G$  until reach  $t$  or get stuck.
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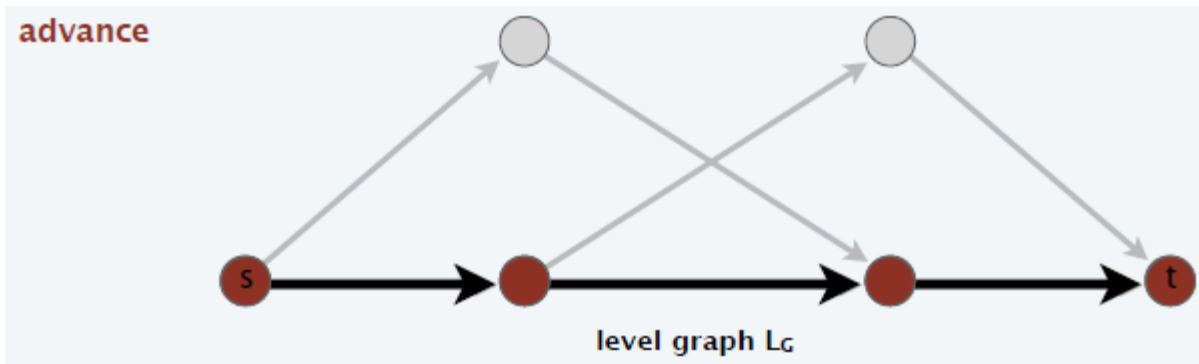
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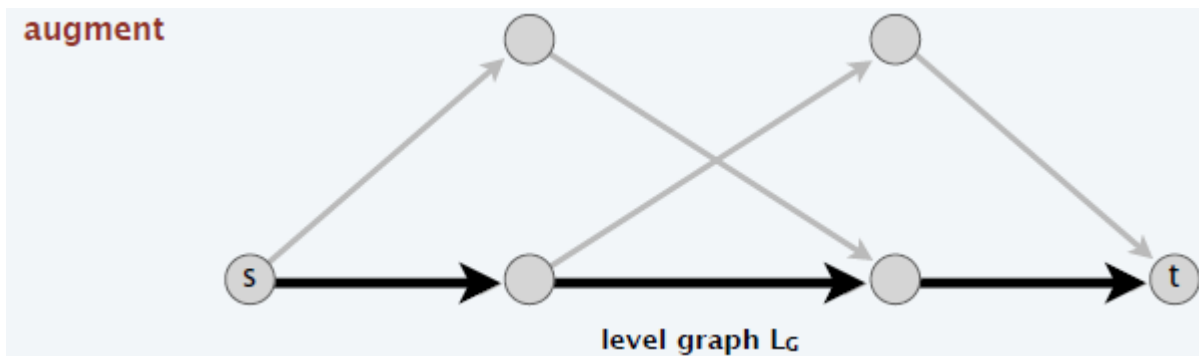
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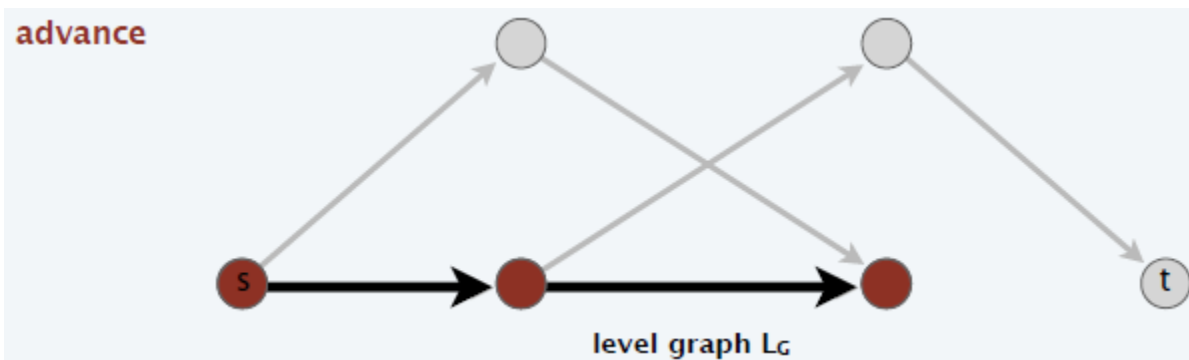
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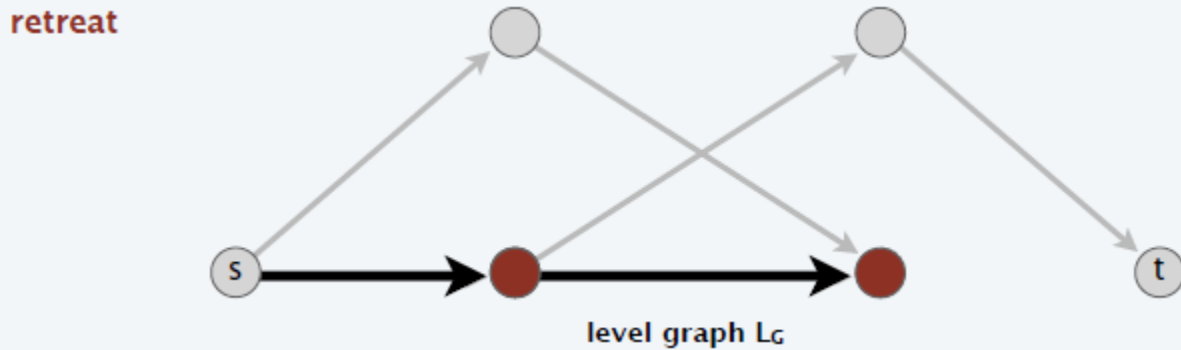
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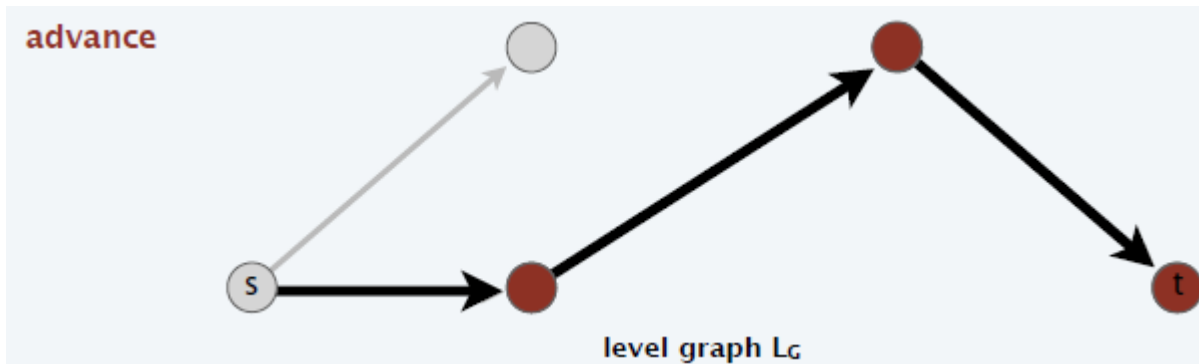
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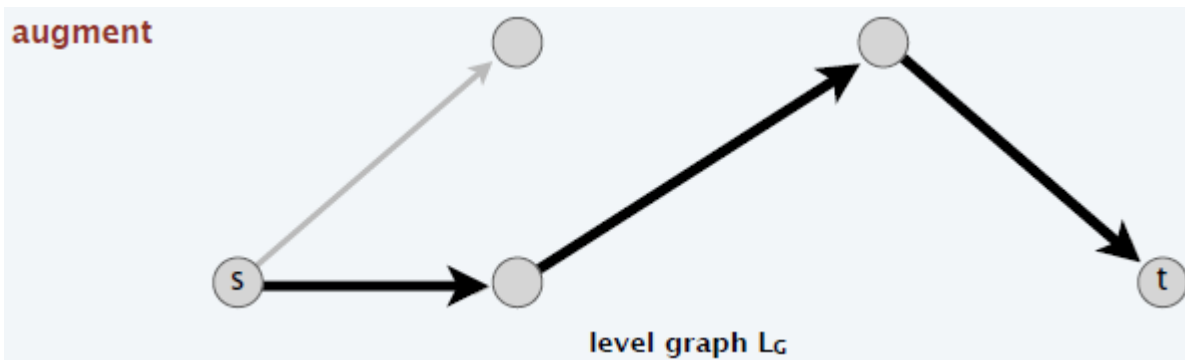
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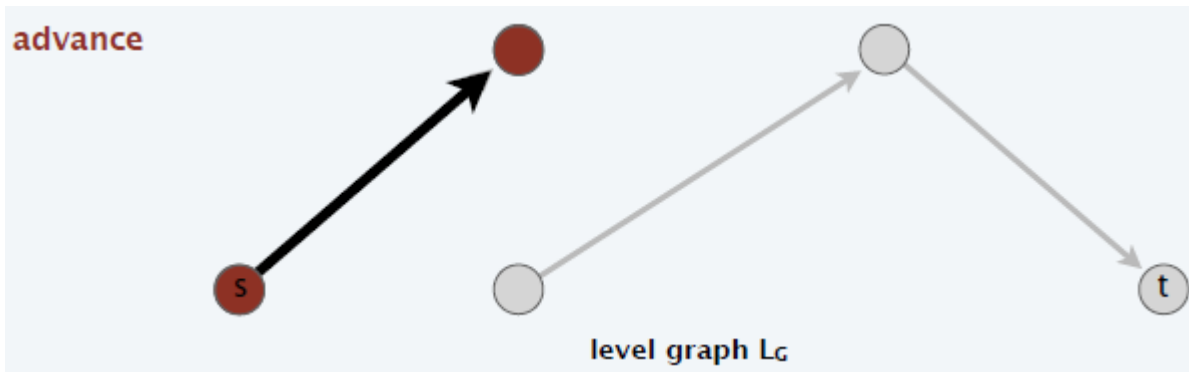
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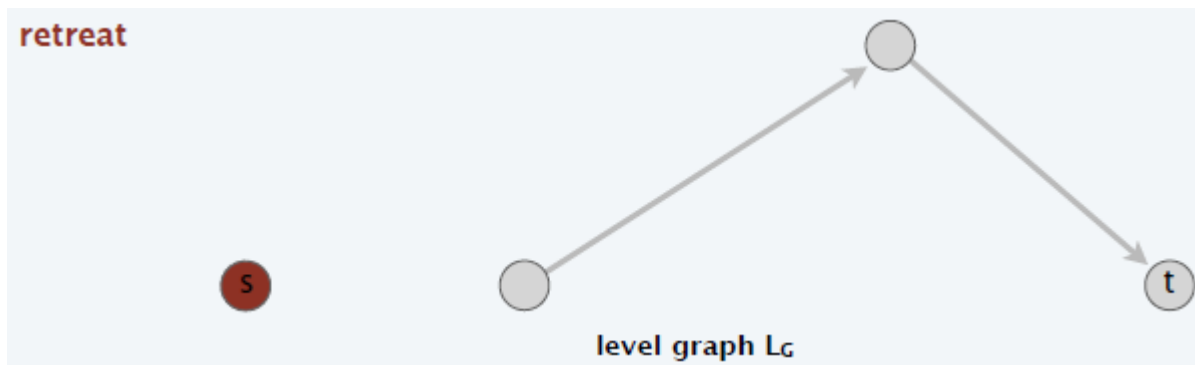
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# Blocking-Flow Algorithm

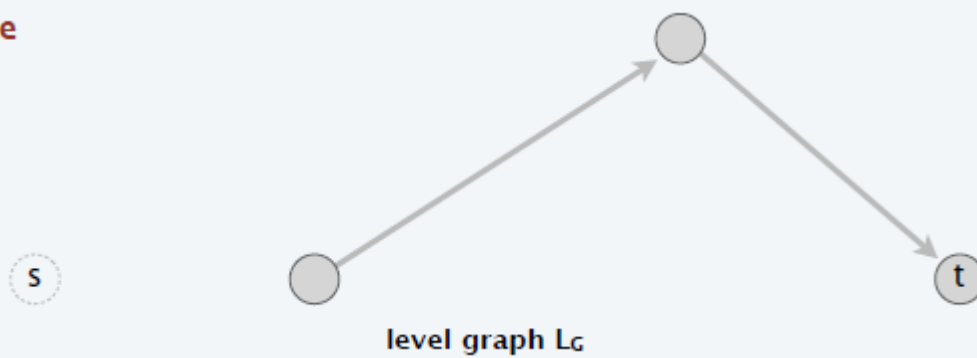
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end of phase







# Dinitz' Algorithm

Initialize ( $G, f$ )

---

$L_G \leftarrow$  level-graph of  $G_f$ .

$P \leftarrow \emptyset$ .

Goto Advance ( $s$ ).

Retreat ( $v$ )

---

If  $v = s$   
Stop.

Else

Delete  $v$  (and all incident edges) from  $L_G$ .

Remove last edge  $(u, v)$  from  $P$ .

Goto Advance ( $u$ )

Advance ( $v$ )

---

If  $v = t$

Augment ( $P$ ).

Remove saturated edge from  $L_G$ .

$P \leftarrow \emptyset$ .

Goto Advance ( $s$ )

If there exists edge  $(v, w) \in L_G$   
Add edge  $(v, w)$  to  $P$ .

Goto Advance ( $w$ ).

Else

Goto Retreat ( $v$ ).