

Objectives

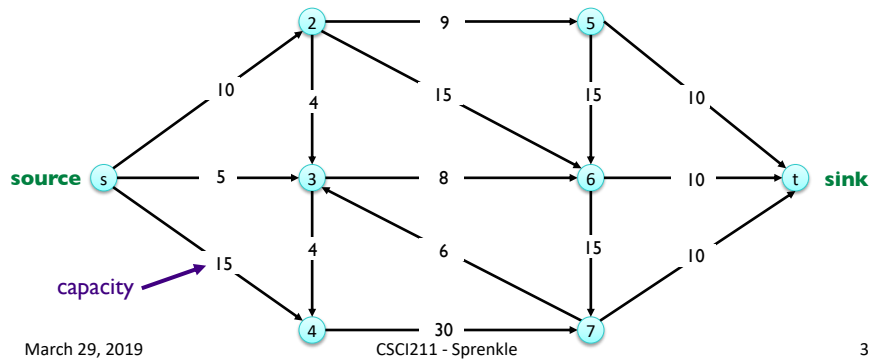
- Network Flow
 - Max flow
 - Min cut

Motivating Flow Network Problems

- Modeling *transportation* networks
 - Edges: carry traffic
 - Nodes: pass traffic between edges
- Can represent many different types of problems
 - Instead of looking at all possibilities, formulate as a flow problem

Flow Network

- $G = (V, E)$ = directed graph, no parallel edges
- Two distinguished nodes: s = source, t = sink
- $c(e)$ = capacity of edge e , > 0



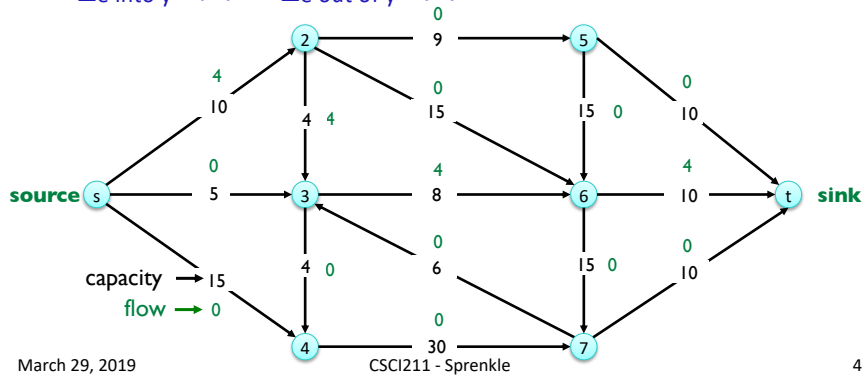
Flows: Definitions

- An **s-t flow** is a function that satisfies
 - **Capacity condition:** For each $e \in E$: $0 \leq f(e) \leq c(e)$
 - **Conservation condition:** For each $v \in V - \{s, t\}$:

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

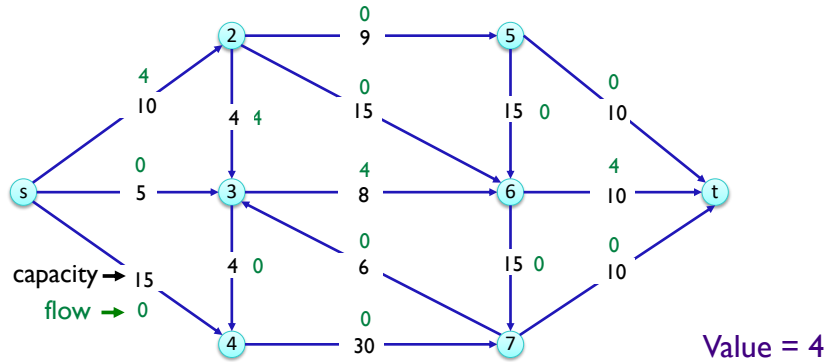
Flow can't exceed capacity

Flow in == Flow out



Flows: Definitions

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



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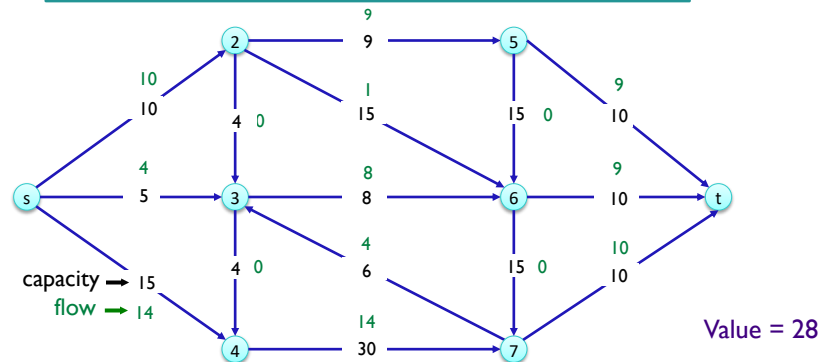
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Maximum Flow Problem

- Make network most efficient
 - Use most of available capacity

Goal: Find $s-t$ flow of maximum value



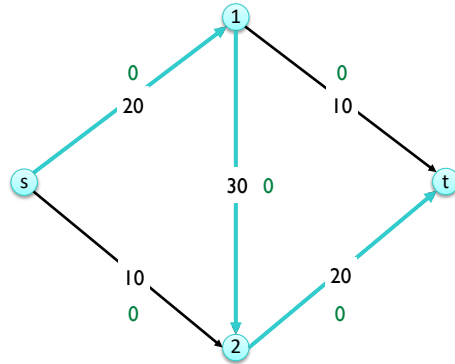
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Towards a Max Flow Algorithm

- Greedy algorithm
 - Start all edges $e \in E$ at $f(e) = 0$
 - Find an $s-t$ path P with the most capacity: $f(e) < c(e)$
 - Augment flow along path P
 - Repeat until you get stuck



Flow value = 0

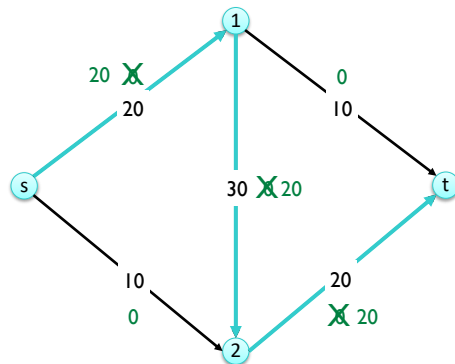
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Towards a Max Flow Algorithm

- Greedy algorithm
 - Start all edges $e \in E$ at $f(e) = 0$
 - Find an $s-t$ path P with the most capacity: $f(e) < c(e)$
 - Augment flow along path P
 - Repeat until you get stuck



Is this optimal?

Flow value = 20

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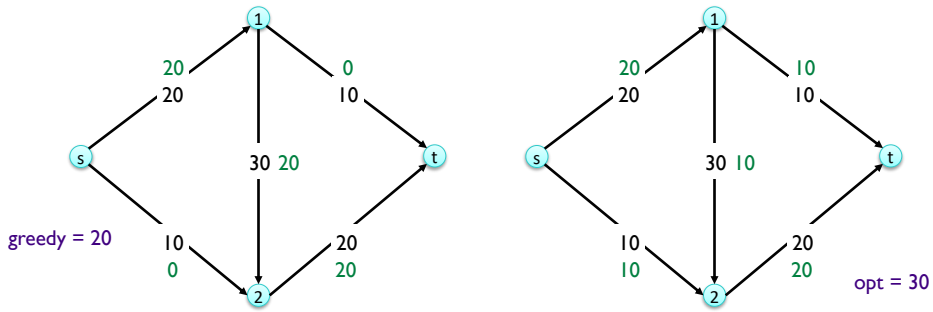
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Towards a Max Flow Algorithm

- Greedy algorithm
 - Start all edges $e \in E$ at $f(e) = 0$
 - Find an $s-t$ path P with the most capacity: $f(e) < c(e)$
 - Augment flow along path P
 - Repeat until you get **stuck**

locally optimality does not \Rightarrow global optimality



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Towards a solution...

RESIDUAL GRAPHS

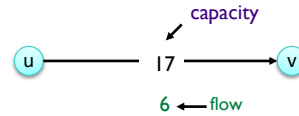
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Towards a Residual Graph

- Original edge: $e = (u, v) \in E$
 - Flow $f(e)$, capacity $c(e)$



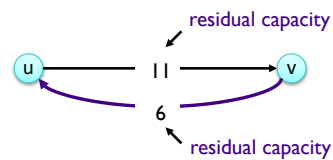
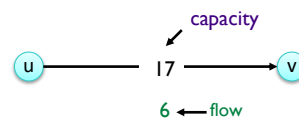
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Towards a Residual Graph

- Original edge: $e = (u, v) \in E$
 - Flow $f(e)$, capacity $c(e)$
- Residual edge
 - $e = (u, v)$ w/ capacity $c(e) - f(e)$
 - $e^R = (v, u)$ with capacity $f(e)$
 - To undo flow



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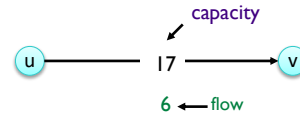
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Residual Graph: G_f

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

➤ Flow $f(e)$, capacity $c(e)$

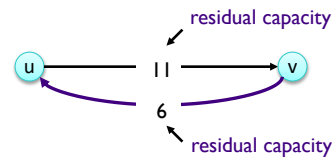


- **Residual edge**

➤ $e = (u, v)$ w/ capacity $c(e) - f(e)$

➤ $e^R = (v, u)$ with capacity $f(e)$

- To undo flow



- **Residual graph:** $G_f = (V, E_f)$

➤ Residual edges with *positive* residual capacity

➤ $E_f = \{e : f(e) < c(e)\} \cup \{e^R : f(e) > 0\}$

Forward edges

Backward edges

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Applying Residual Graph

- Used to find the maximum flow
 - Use similar idea to greedy algorithm

- **Residual path:** simple s - t path in G_f
 - Also known as *augmenting path*

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Augmenting Path Algorithm

$c = \text{capacity}$

```

Ford-Fulkerson( $G, s, t, c$ ):
  foreach  $e \in E$   $f(e) = 0$  # initially no flow
   $G_f = \text{residual graph}$ 

  while there exists augmenting path  $P$ 
     $f = \text{Augment}(f, c, P)$  # change the flow
    update  $G_f$  # build a new residual graph

  return  $f$ 

```

```

Augment( $f, c, P$ ):
   $b = \text{bottleneck}(P)$  # edge on  $P$  with least capacity
  foreach  $e \in P$ 
    if ( $e \in E$ )  $f(e) = f(e) + b$  # forward edge,  $\uparrow$  flow
    else  $f(e^R) = f(e) - b$  # forward edge,  $\downarrow$  flow
  return  $f$ 

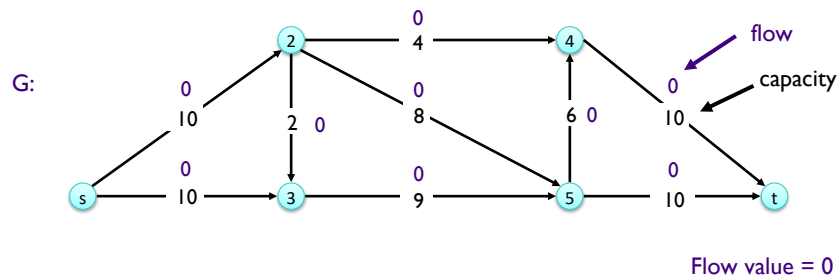
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Ford-Fulkerson Algorithm

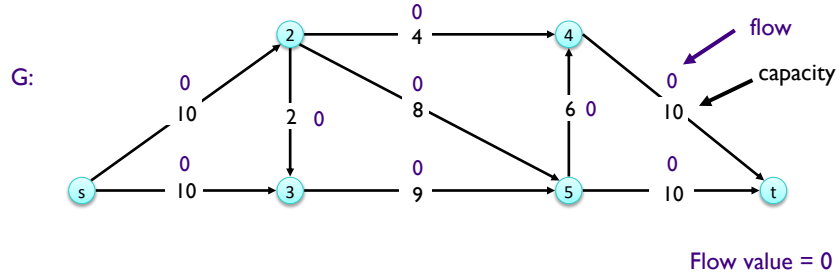


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Ford-Fulkerson Algorithm



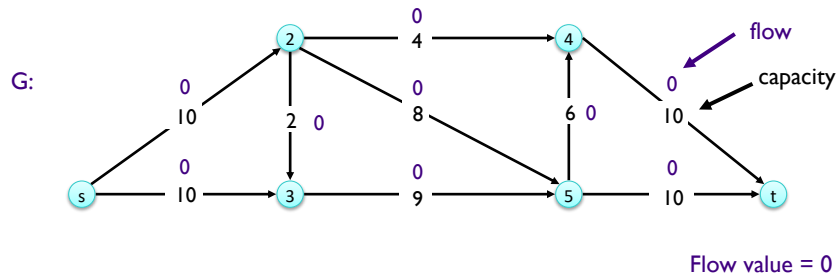
What does the residual graph look like?

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Ford-Fulkerson Algorithm

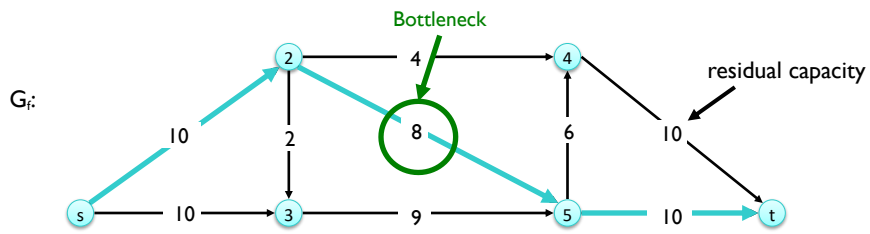
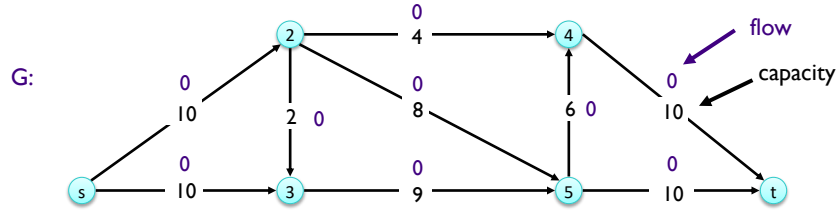


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Ford-Fulkerson Algorithm

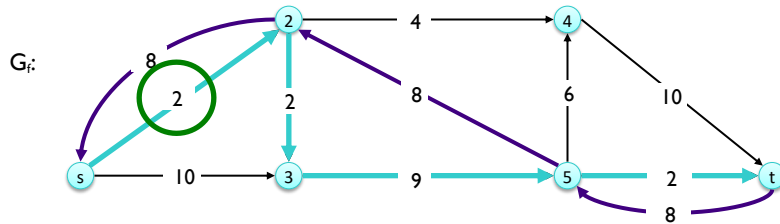
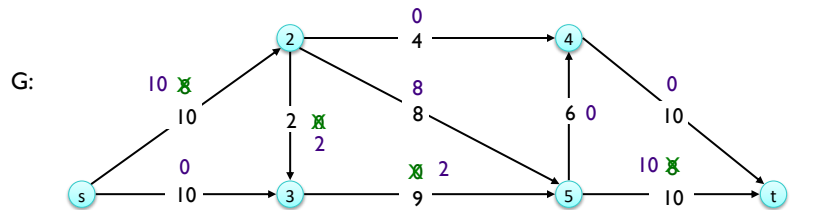


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Ford-Fulkerson Algorithm

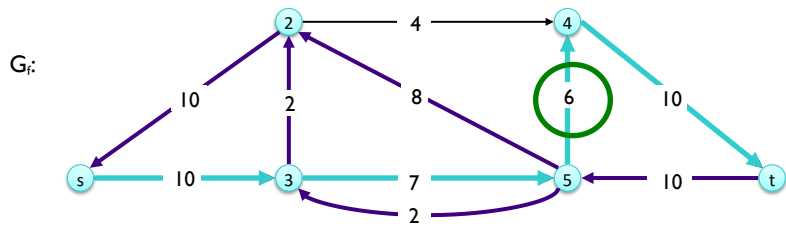
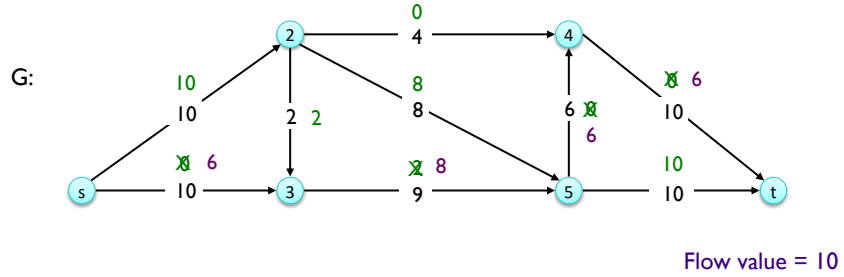


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Ford-Fulkerson Algorithm

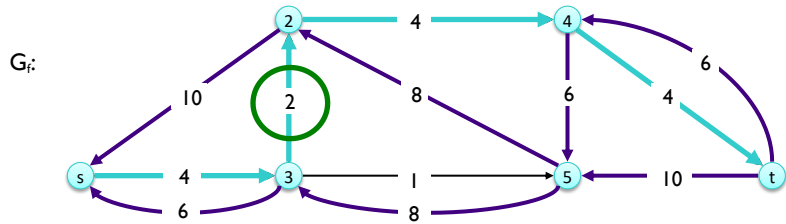
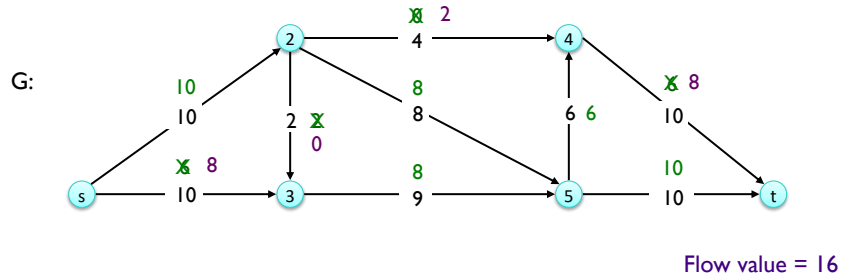


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Ford-Fulkerson Algorithm

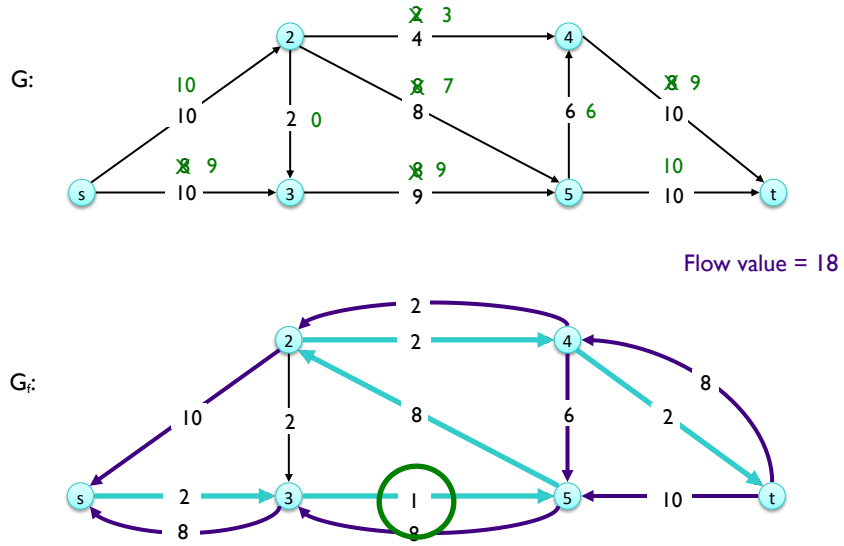


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Ford-Fulkerson Algorithm

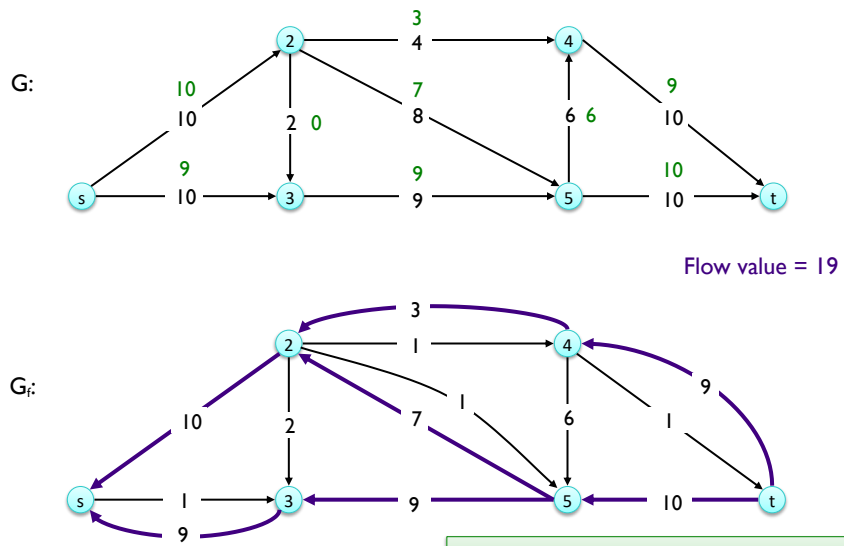


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Ford-Fulkerson Algorithm

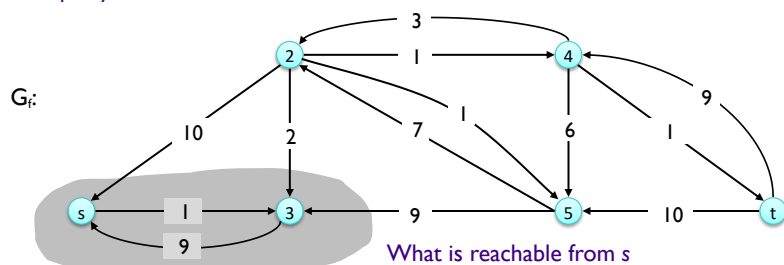
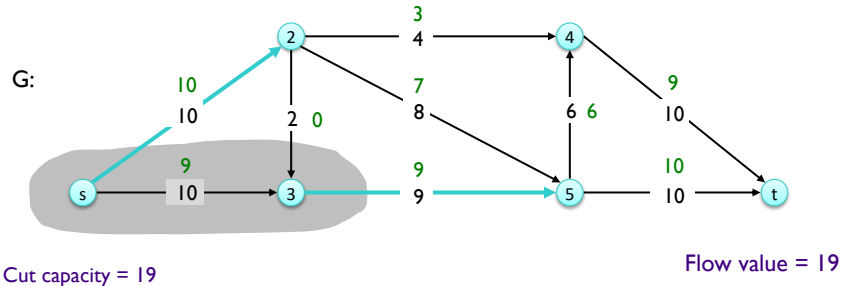


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How do we know we're done?

Ford-Fulkerson Algorithm



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Analyzing Augmenting Path Algorithm

```

Ford-Fulkerson(G, s, t, c)
  foreach e ∈ E f(e) = 0 # initially no flow
  Gf = residual graph

  while there exists augmenting path P
    f = Augment(f, c, P) # change the flow
    update Gf # build a new residual graph

  return f

```

```

Augment(f, c, P)
  b = bottleneck(P) # edge on P with least capacity
  foreach e ∈ P
    if (e ∈ E) f(e) = f(e) + b # forward edge, ↑ flow
    else f(eR) = f(e) - b # forward edge, ↓ flow
  return f

```

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Why does alg work? What is happening at each iteration?
 What is the running time? Need more analysis ...

MINIMUM CUTS

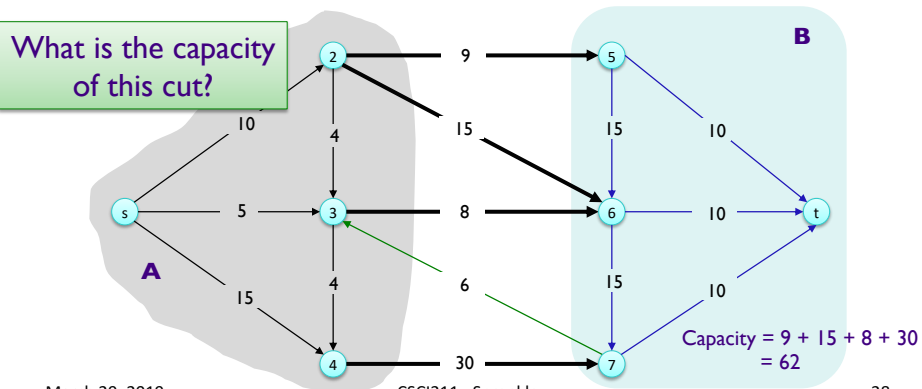
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Cuts

- An **s-t cut** is a partition (A, B) of V with $s \in A$ and $t \in B$
- The **capacity** of a cut (A, B) is $cap(A, B) = \sum_{e \text{ out of } A} c(e)$



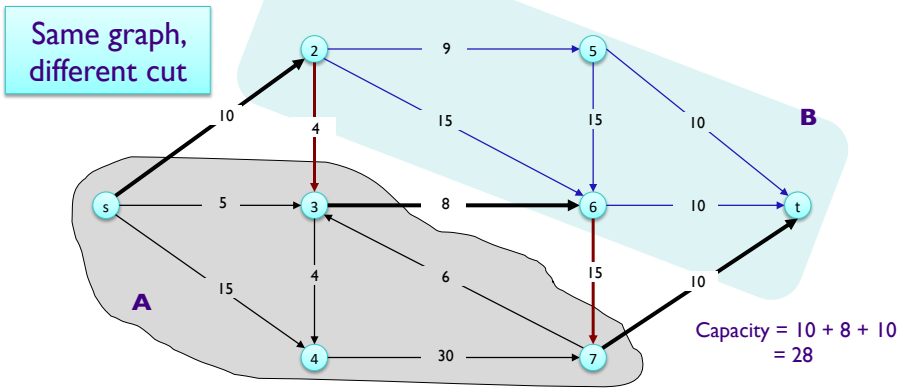
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Minimum Cut Problem

- Find an $s-t$ cut of *minimum* capacity
 - Puts *upperbound* on maximum flow



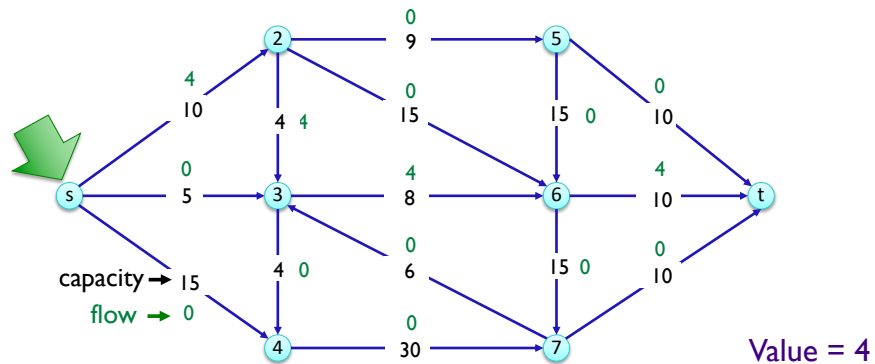
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Recall

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



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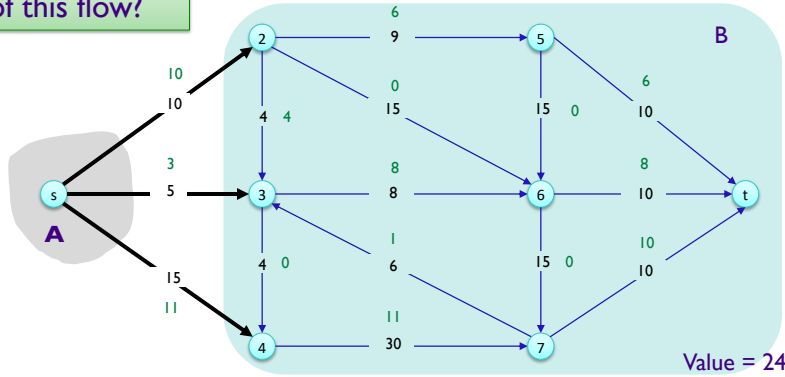
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Flow Value Lemma

- Let f be *any* flow, and let (A, B) be *any* s - t cut. Then, the **value of the flow** is $= f^{\text{out}}(A) - f^{\text{in}}(A)$.

What is the value of this flow?

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



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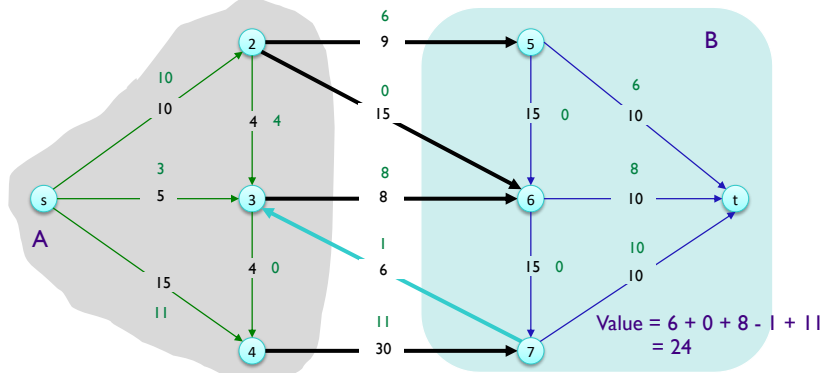
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Flow Value Lemma

- Let f be *any* flow, and let (A, B) be *any* s - t cut. Then, the **value of the flow** is $= f^{\text{out}}(A) - f^{\text{in}}(A)$.

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



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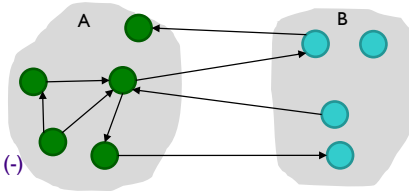
Flow Value Lemma (FVL)

- Let f be any flow, and let (A, B) be any s - t cut.

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

- Pf.

$$\begin{aligned}
 v(f) &= \sum_{e \text{ out of } s} f(e) && \text{By definition} \\
 &= \sum_{e \text{ out of } s} f(e) + \sum_{v \in A \neq s} \left(\sum_{e \text{ out of } v} f(e) - \sum_{e \text{ into } v} f(e) \right) && \text{by flow conservation, all terms except } v = s \text{ are } 0 \\
 &= \sum_{e \text{ out of } s} f(e) - \sum_{e \text{ into } s} f(e) + \sum_{v \in A \neq s} \left(\sum_{e \text{ out of } v} f(e) - \sum_{e \text{ into } v} f(e) \right) \\
 &= \sum_{v \in A} \left(\sum_{e \text{ out of } v} f(e) - \sum_{e \text{ into } v} f(e) \right) \\
 &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e)
 \end{aligned}$$



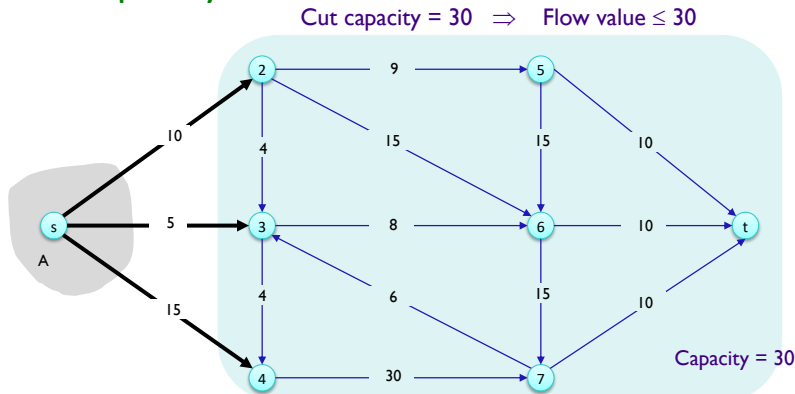
- Possibilities for edge e :
- Both ends in A (0)
 - Points out from A (+), Points in to A (-)

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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 cut's capacity



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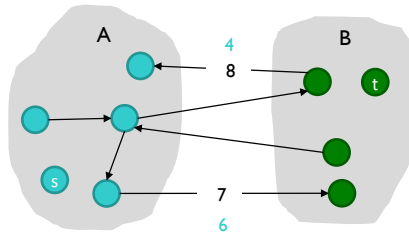
Weak Duality

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

- Pf.

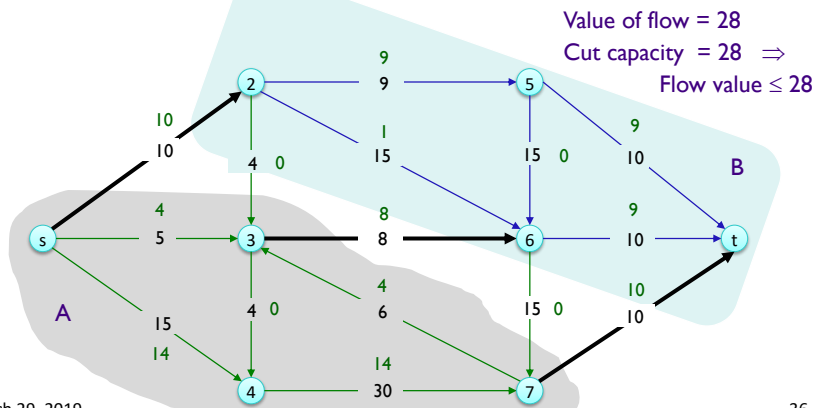
By FVL

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



Certificate of Optimality

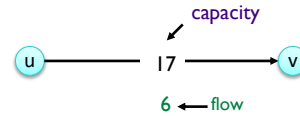
- Corollary.** Let f be any flow, and let (A, B) be any cut. If $v(f) = \text{cap}(A, B)$, then f is a **max flow** and (A, B) is a **min cut**.



Recall: Residual Graph G_f

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

➤ Flow $f(e)$, capacity $c(e)$

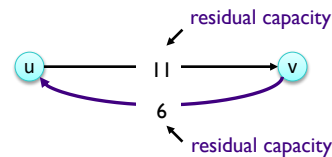


- Residual edge

➤ $e = (u, v)$ w/ capacity $c(e) - f(e)$

➤ $e^R = (v, u)$ with capacity $f(e)$

- To undo flow



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

➤ Residual edges with *positive* residual capacity

➤ $E_f = \{e : f(e) < c(e)\} \cup \{e^R : f(e) > 0\}$

Forward edges

Backward edges

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Recall: Augmenting Path Algorithm

```

Ford-Fulkerson( $G, s, t, c$ )
  foreach  $e \in E$   $f(e) = 0$  # initially no flow
   $G_f =$  residual graph

  while there exists augmenting path  $P$ 
     $f =$  Augment( $f, c, P$ ) # change the flow
    update  $G_f$  # build a new residual graph

  return  $f$ 

```

```

Augment( $f, c, P$ )
   $b =$  bottleneck( $P$ ) # edge on  $P$  with least capacity
  foreach  $e \in P$ 
    if ( $e \in E$ )  $f(e) = f(e) + b$  # forward edge, ↑ flow
    else  $f(e^R) = f(e) - b$  # backward edge, ↓ flow
  return  $f$ 

```

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Intuition Behind Correctness of F-F Algorithm

- Let A be set of vertices *reachable* from s in residual graph at end of F-F alg execution
- By definition of A , $s \in A$
- By definition of the F-F algorithm's resulting flow, $t \notin A$

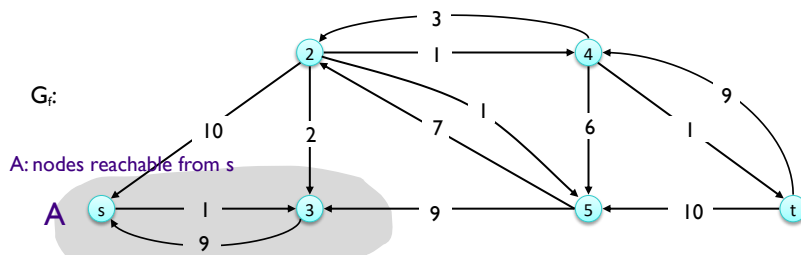
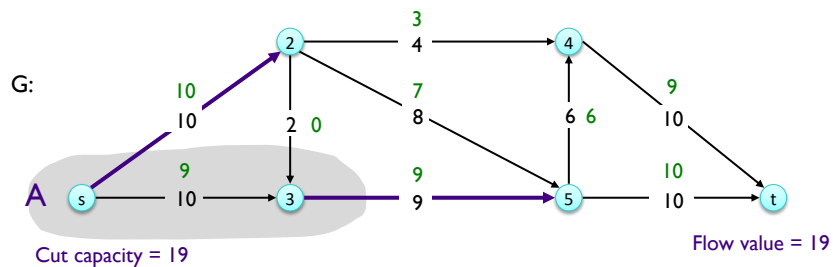
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Ford-Fulkerson

- What do we know about the flow out of A ?
- What do we know about the flow into A ?



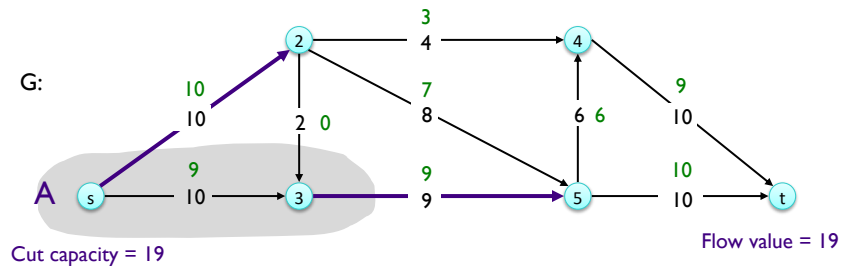
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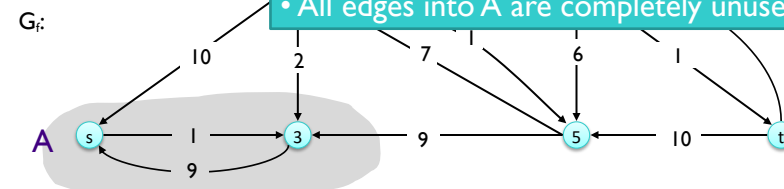
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Ford-Fulkerson

- What do we know about the flow out of A?
- What do we know about the flow into A?



- All edges out of A are completely saturated
- All edges into A are completely unused



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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. [Ford-Fulkerson 1956]

The value of the max flow is equal to the value of the min cut.

- **Proof strategy.** Prove both simultaneously by showing the following are equivalent:
 - There exists a cut (A, B) such that $v(f) = \text{cap}(A, B)$.
 - Flow f is a max flow.
 - There is no augmenting path relative to f .

See formal proof in book

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Analyzing Augmenting Path Algorithm

```

Ford-Fulkerson(G, s, t, c)
  foreach e ∈ E f(e) = 0 # initially no flow
  Gf = residual graph

  while there exists augmenting path P
    f = Augment(f, c, P) # change the flow
    update Gf # build a new residual graph

  return f

```

```

Augment(f, c, P)
  b = bottleneck(P) # edge on P with least capacity
  foreach e ∈ P
    if (e ∈ E) f(e) = f(e) + b # forward edge, ↑ flow
    else f(eR) = f(e) - b # forward edge, ↓ flow
  return f

```

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Analyzing Augmenting Path Algorithm

```

Ford-Fulkerson(G, s, t, c)
O(m)  foreach e ∈ E f(e) = 0 # initially no flow
O(m)  Gf = residual graph
Find path: O(m); Iterations: O(F) iterations, where F = max flow
  while there exists augmenting path P
O(m)  f = Augment(f, c, P) # change the flow
O(m)  update Gf # build a new residual graph

  return f

```

Total: O(Fm)

```

Augment(f, c, P)
O(n)  b = bottleneck(P) # edge on P with least capacity
O(n)  foreach e ∈ P
O(l)  if (e ∈ E) f(e) = f(e) + b # forward edge, ↑ flow
O(l)  else f(eR) = f(e) - b # forward edge, ↓ flow
  return f

```

Total: O(n) → O(m), since n ≤ 2m

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Running Time

- **Assumption.** All capacities are integers between 1 and F .
- **Invariant.** Every flow value $f(e)$ and every residual capacity's $c_f(e)$ remains an integer throughout algorithm.
- **Theorem.** Algorithm terminates in at most $v(f^*) \leq nF$ iterations.
- **Pf.** Each augmentation increases value by at least 1.
- **Corollary.** If $F = 1$, 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.

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Looking Ahead

- PS 9 (last one!) due Friday
 - [See course schedule page for starter code](#)
- Wiki due Monday – Network flows focus

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