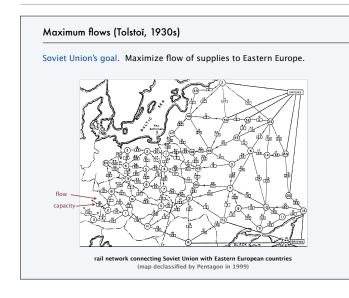
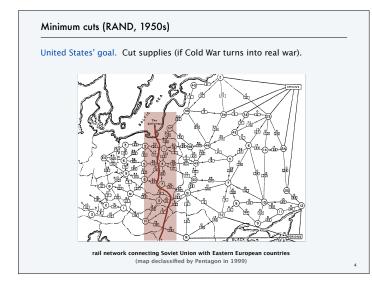
# CSCI 355: Algorithm Design and Analysis 9. Network Flow

- max-flow and min-cut problems
- ▶ Ford–Fulkerson algorithm
- max-flow min-cut theorem





#### Max-flow and min-cut

#### A widely applicable model.

- Data mining.
- Open-pit mining.
- Bipartite matching.
- Network reliability.
- Baseball elimination.
- Image segmentation.
- Network connectivity.
- Markov random fields.
- Distributed computing.
- Security of statistical data.
- Egalitarian stable matching.
- · Network intrusion detection.
- Multi-camera scene reconstruction.
- · Sensor placement for homeland security.
- Many, many, more.



liver and hepatic vascularization segmentati

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### CSCI 355: Algorithm Design and Analysis 9. Network Flow

# max-flow and min-cut problems

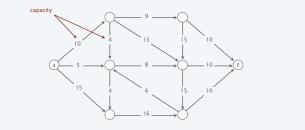
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#### Flow network

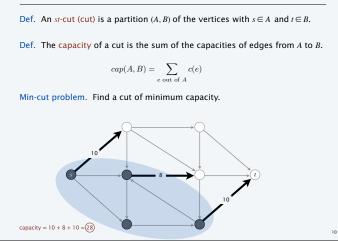
A flow network is a tuple G = (V, E, s, t, c).

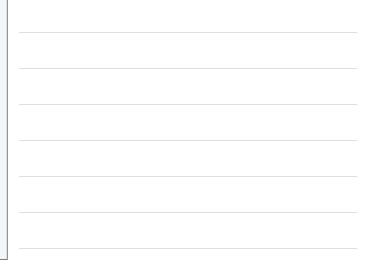
- Digraph (V, E) with source  $s \in V$  and sink  $t \in V$ .
- Capacity  $c(e) \ge 0$  for each  $e \in E$ . assume all vertices are reachable from s

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



#### Minimum-cut problem



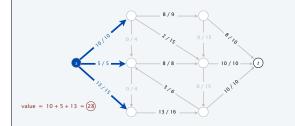


# Maximum-flow problem

Def. An <i>st</i> -flow (flow) <i>f</i> is a function that satisfies:		
• For each $e \in E$ :	$0 \leq f(e) \leq c(e)$	[capacity]
• For each $v \in V - \{s, t\}$ :	$\sum_{e \text{ in to } v} f(e) = \sum_{e \text{ out of } v} f(e)$	[flow conservation]
		_

Der. The value of a flow 
$$f$$
 is:  $val(f) = \sum_{e \text{ out of } s} f(e) - \sum_{e \text{ in to } s} f(e)$ 

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



# CSCI 355: Algorithm Design and Analysis 9. Network Flow

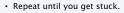
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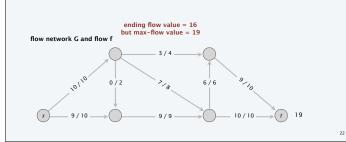
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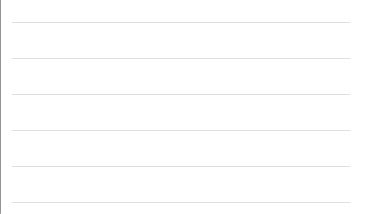
#### Toward a max-flow algorithm

#### 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.

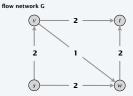






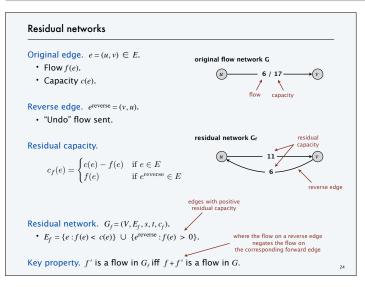
# Why the greedy algorithm fails

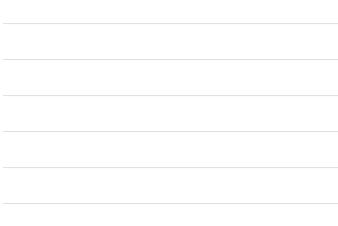
- Q. Why does the greedy algorithm fail?
- A. Once the greedy algorithm increases the flow on an edge, it never decreases it.
- Ex. Consider the flow network G.
- The unique max flow  $f^*$  has  $f^*(v, w) = 0$ .
- Greedy algorithm could choose  $s \rightarrow v \rightarrow w \rightarrow t$  as first path.



Bottom line. Need some mechanism to "undo" a bad decision.

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#### Augmenting paths

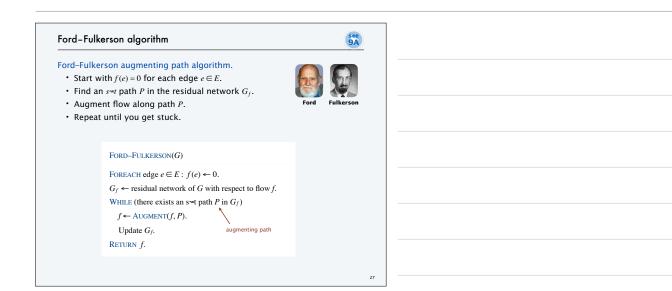
Def. An augmenting path is a simple  $s \sim t$  path in the residual network  $G_f$ .

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 calling  $f' \leftarrow AUGMENT(f, P)$ , the resulting f' is a flow and  $val(f') = val(f) + bottleneck(G_f, P)$ .

AUGMENT(f, P)

$$\begin{split} \delta &\leftarrow \text{bottleneck capacity of augmenting path } P.\\ \text{FOREACH edge } e &\in P:\\ \text{IF } (e \in E) \ f(e) \leftarrow f(e) + \delta.\\ \text{ELSE} \qquad f(e^{\text{revense}}) \leftarrow f(e^{\text{revense}}) - \delta.\\ \text{RETURN } f. \end{split}$$



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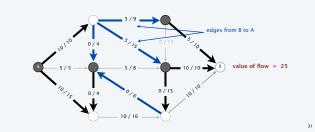
# CSCI 355: Algorithm Design and Analysis 9. Network Flow

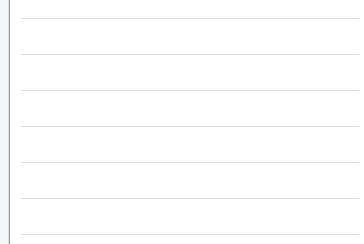
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#### Relationship between flows and cuts

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).

$$val(f) = \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e)$$
  
net flow across cut = (10 + 10 + 5 + 10 + 0 + 0) - (5 + 5 + 0 + 0) = 25



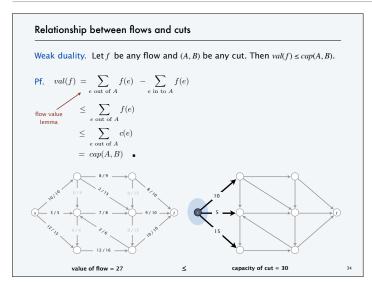


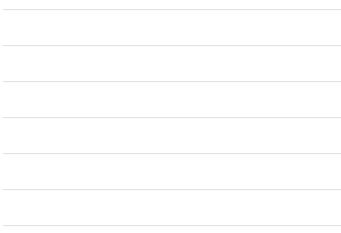
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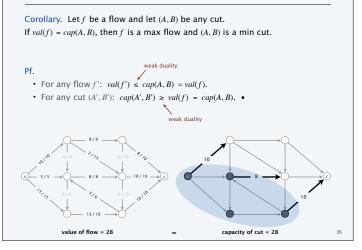
 $val(f) = \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e)$ Pf.  $val(f) = \sum_{e \text{ out of } s} f(e) - \sum_{e \text{ in to } s} f(e)$ by flow conservation, all terms  $\longrightarrow = \sum_{v \in A} \left( \sum_{e \text{ out of } v} f(e) - \sum_{e \text{ in to } v} f(e) \right)$   $= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e) \quad \bullet$ 



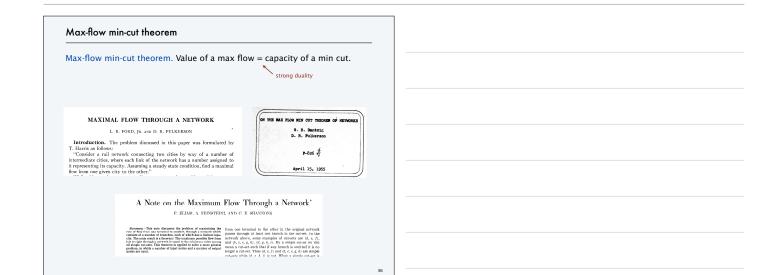




#### Certificate of optimality







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#### Max-flow min-cut theorem

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

Augmenting path theorem. A flow f is a max flow iff there are no augmenting paths.

- **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.

 $[ \ i \rightarrow ii \ ]$ 

This is the weak duality corollary.

#### Max-flow min-cut theorem

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

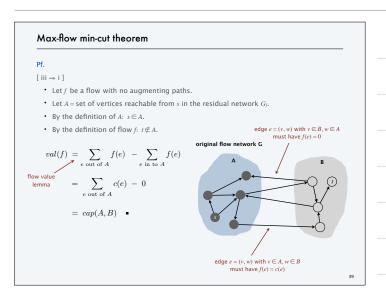
# Augmenting path theorem. A flow f is a max flow iff there are no augmenting paths.

**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  $f\!.$

#### [ ii $\Rightarrow$ iii ] We prove the contrapositive: $\neg$ iii $\Rightarrow$ $\neg$ ii.

- Suppose that there is an augmenting path with respect to *f*.
- We can improve the flow f by sending the flow along this path.
- Thus, f is not a max flow.  $\hfill \hfill \h$



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