

Network Flow

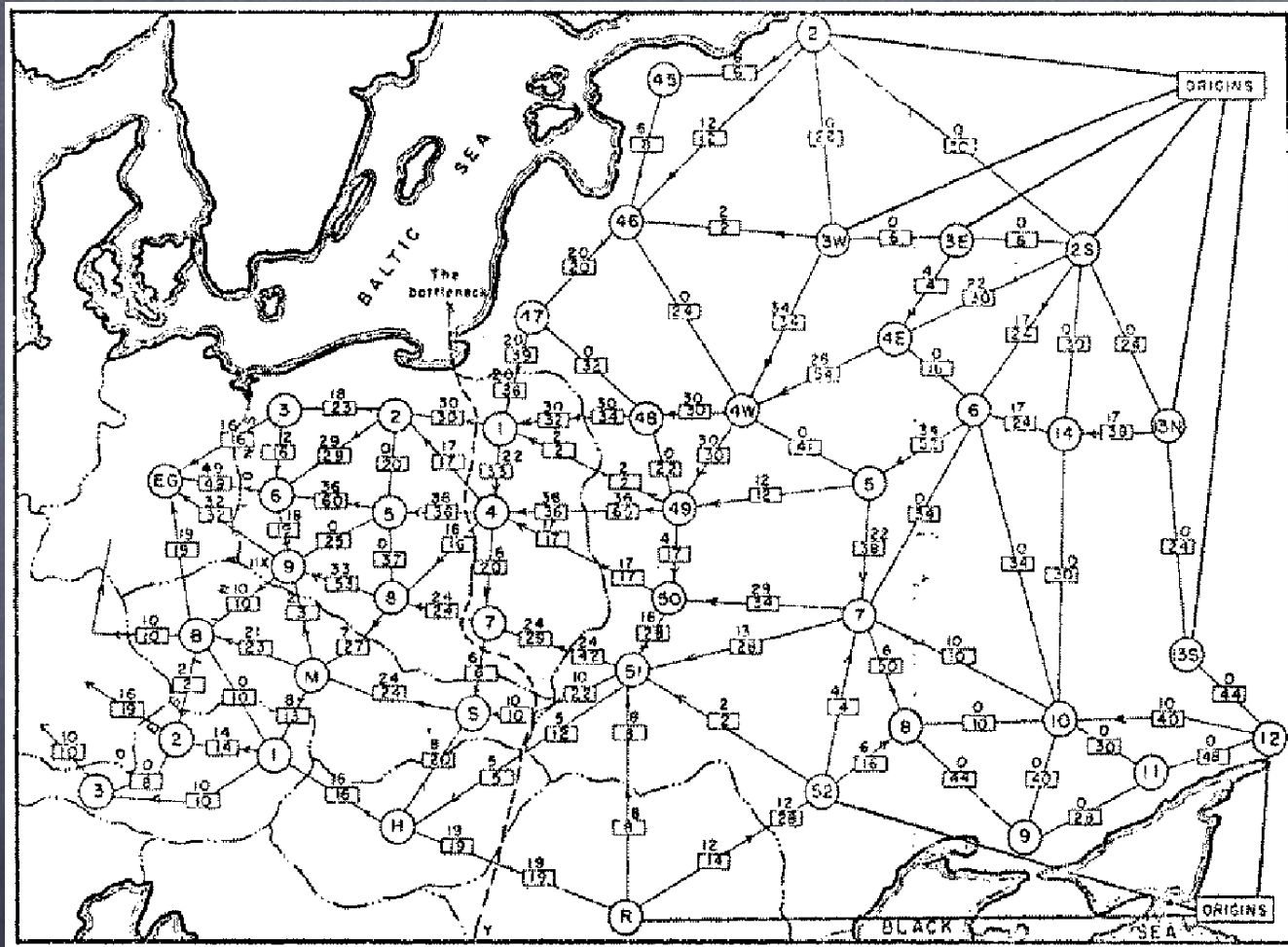
Algorithm Design

	Greedy	Divide and Conquer	Dynamic Programming
Formulate problem	?	?	?
Design algorithm	less work	more work	more work
Prove correctness	more work	less work	less work
Analyze running time	less work	more work	less work

Network Flow

- Greedy, Divide-and-Conquer, and Dynamic Programming were **design techniques**
- Network flow → **a specific class of problems.**
 - Useful in many different applications!
(matching, transportation, network design, etc.)
- **Goal:** design and analyze algorithms for max-flow problem, then apply to solve other problems

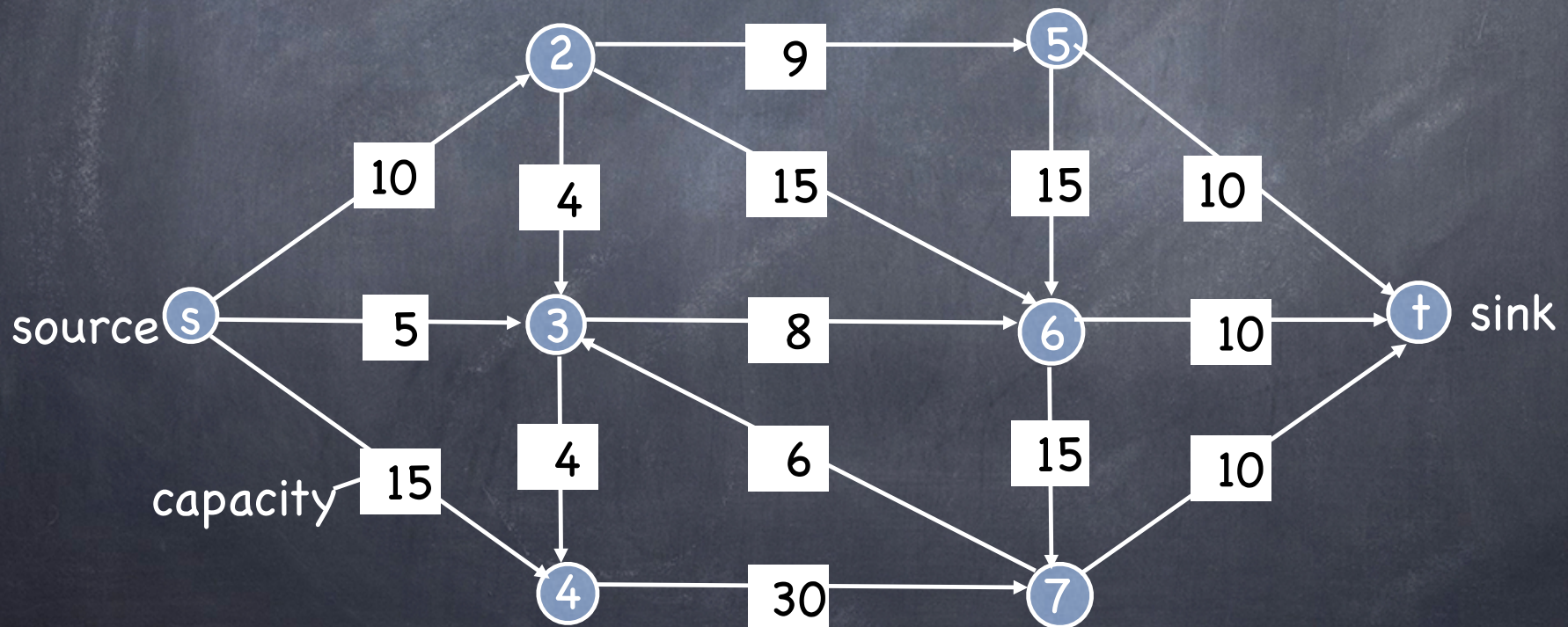
Soviet Rail Network, 1955



Reference: On the history of the transportation and maximum flow problems.
Alexander Schrijver in Math Programming, 91: 3, 2002.

Flow Networks

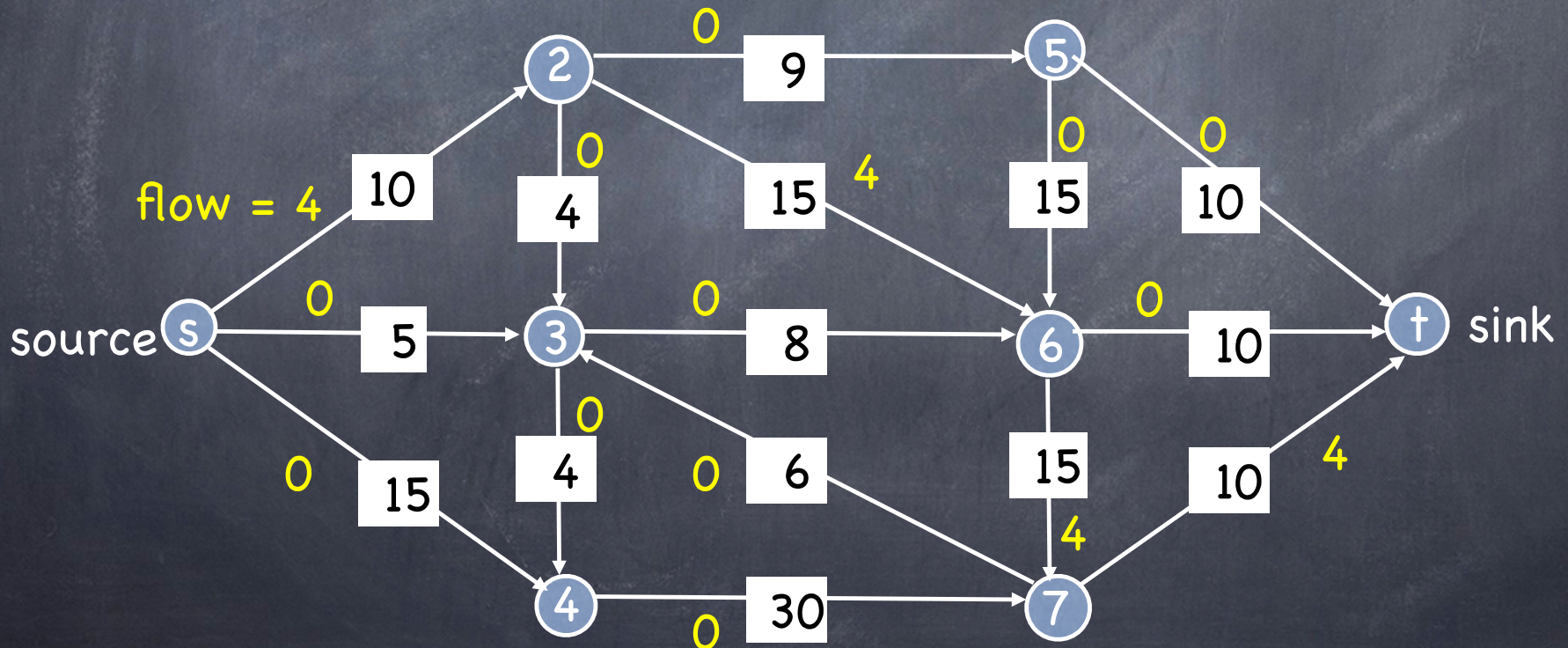
- Flow network.
 - Abstraction for material **flowing** through the edges.
 - $G = (V, E)$ = directed graph
 - Two distinguished nodes: s = **source**, t = **sink**.
 - $c(e)$ = **capacity** of edge e .



Flows

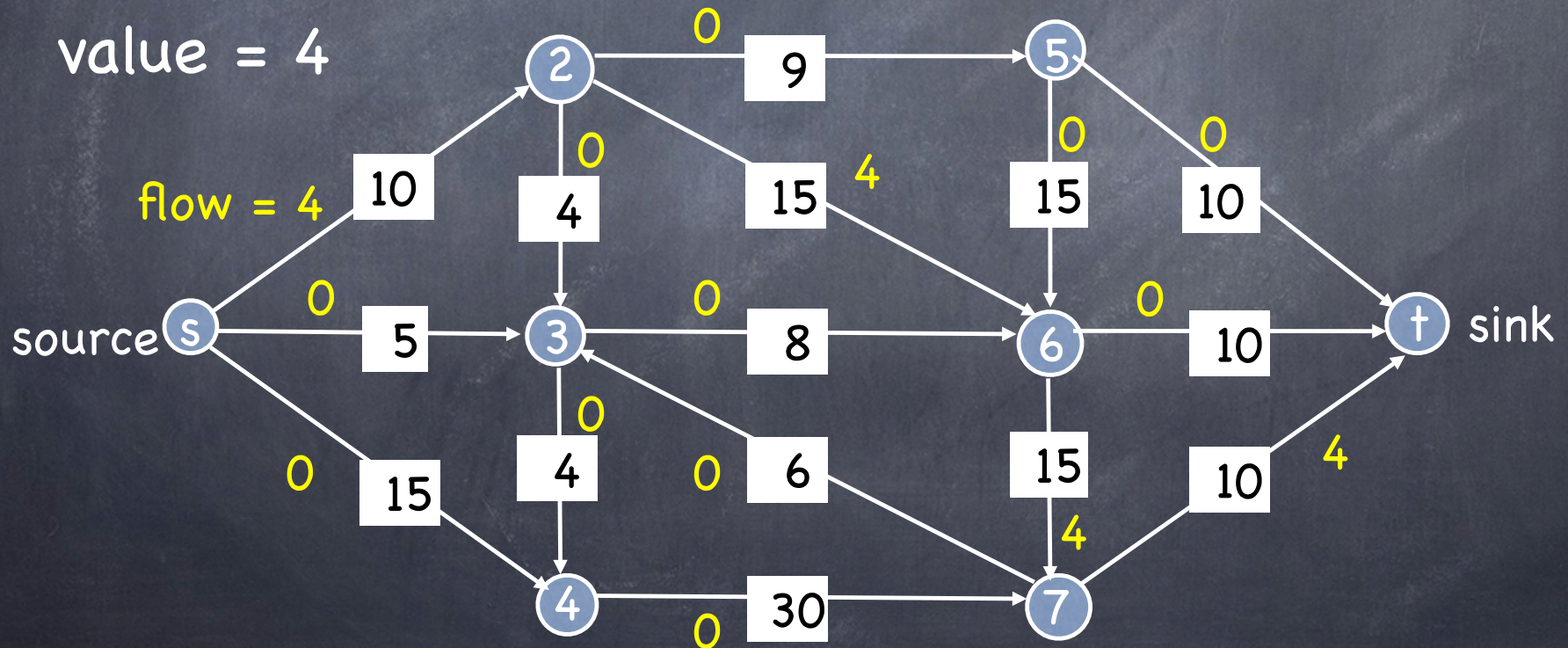
- An **s-t flow** is a function $f: E \rightarrow \mathbb{R}^+$ that satisfies:
 - Capacity condition:** For each $e \in E$: $0 \leq f(e) \leq c(e)$
 - Conservation condition:** For each $v \in V - \{s, t\}$:

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



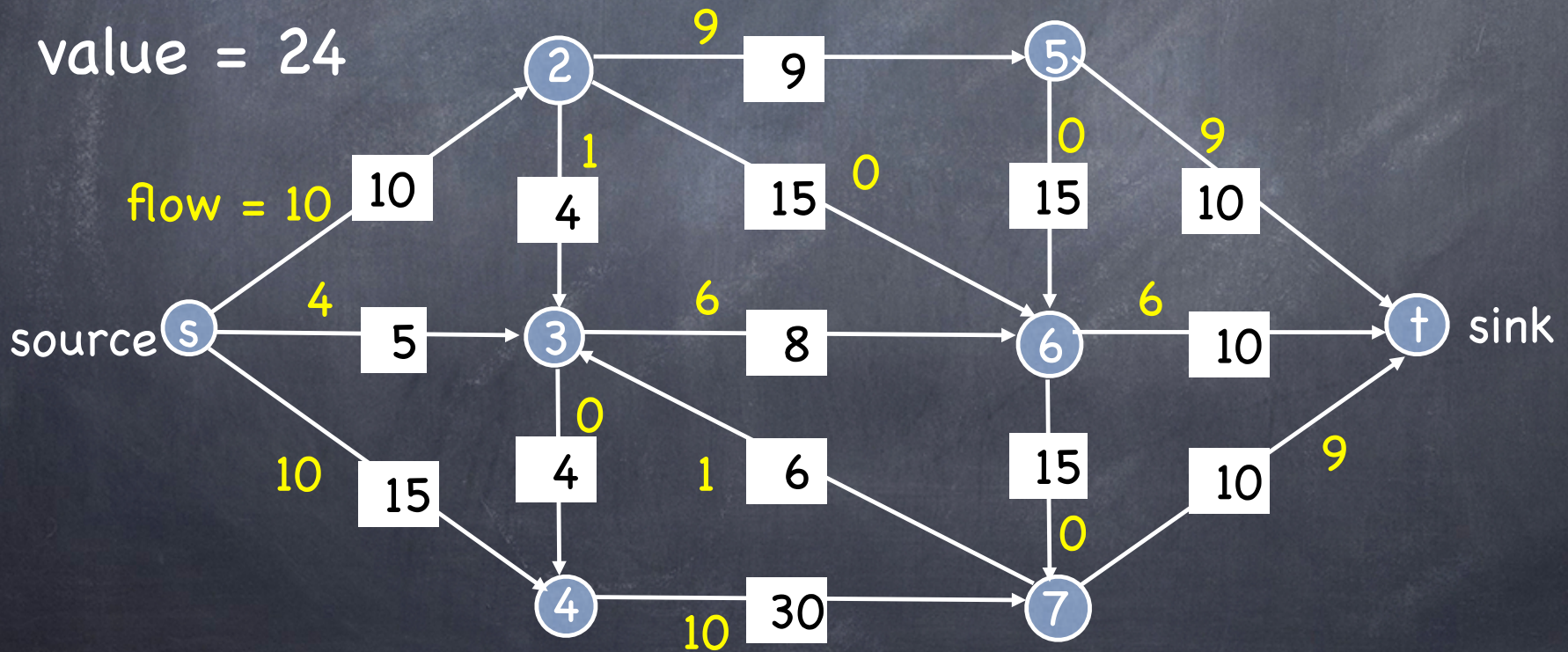
Flows

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



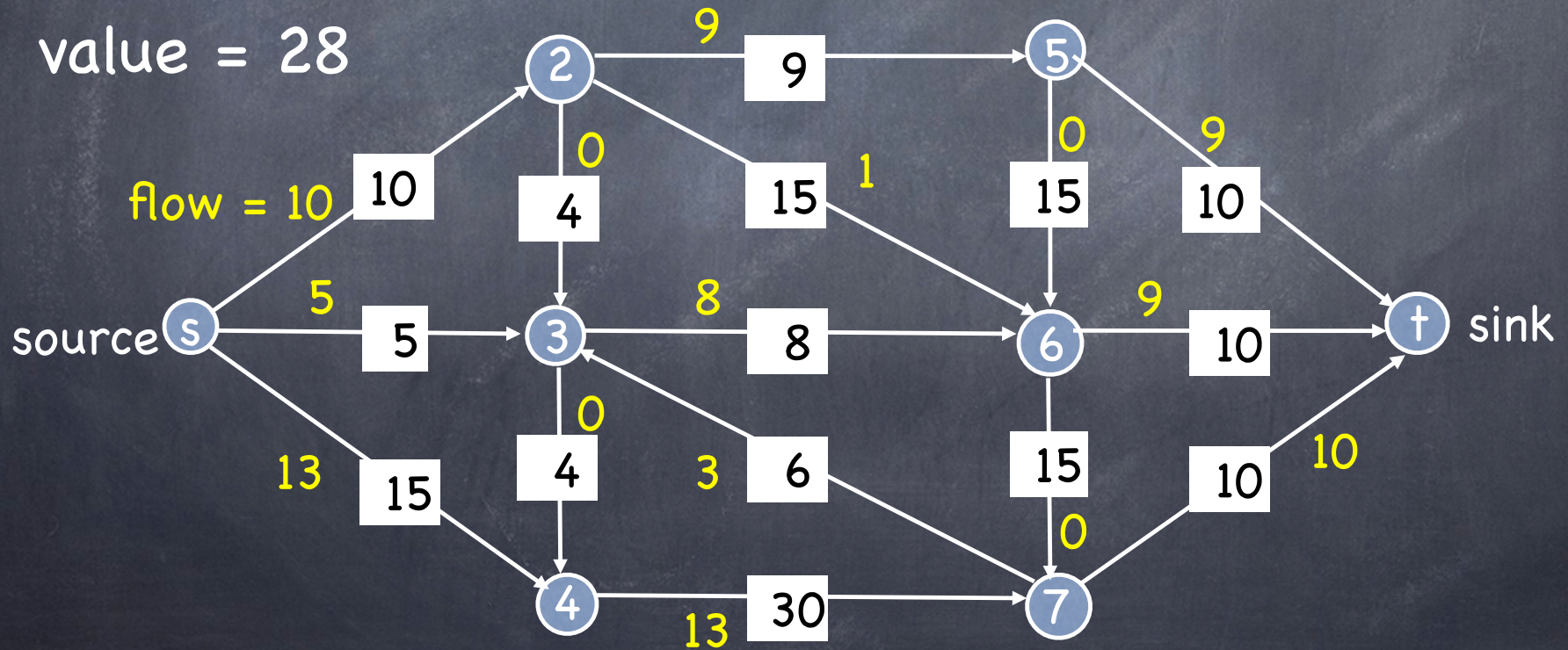
Flows

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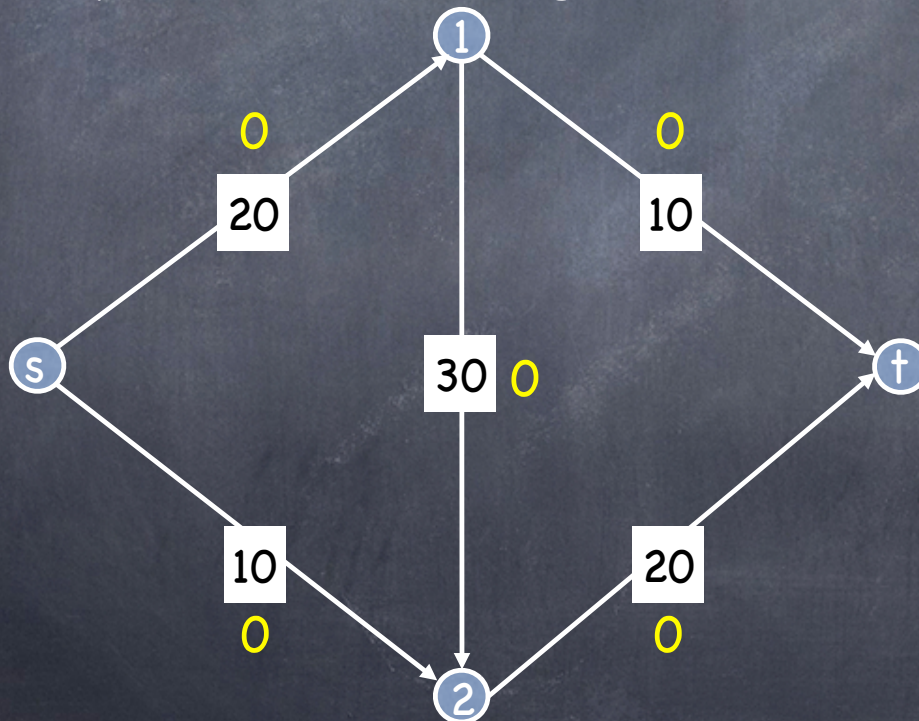
Maximum Flow Problem

Find s-t flow of maximum value.



Towards a Max Flow Algorithm

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



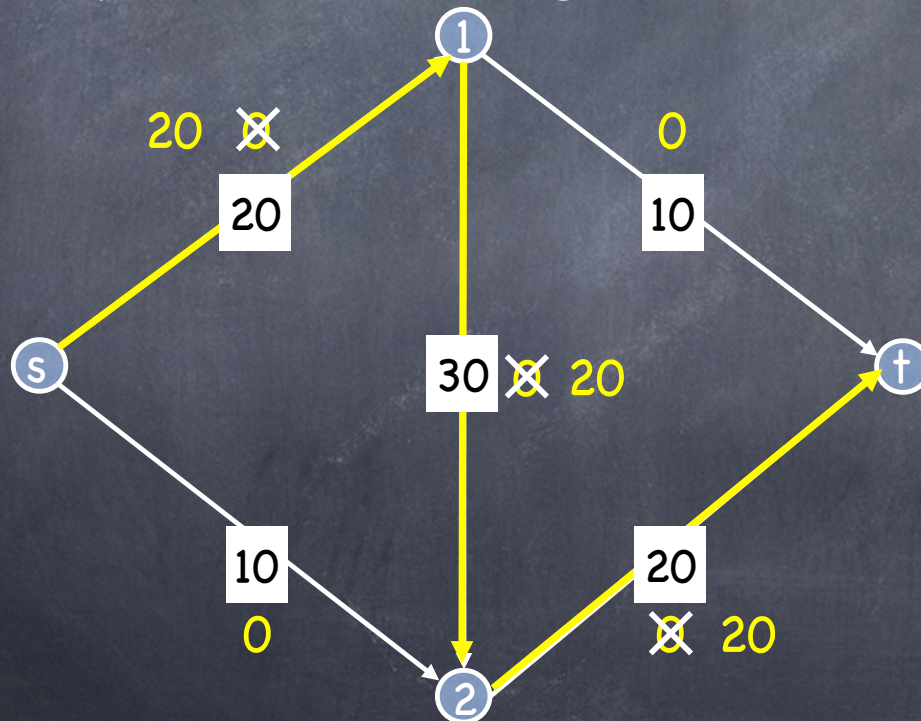
Flow value = 0

Towards a Max-Flow Algorithm

Key idea: repeatedly choose paths and “augment” the amount of flow on those paths as much as possible until capacities are met

Towards a Max Flow Algorithm

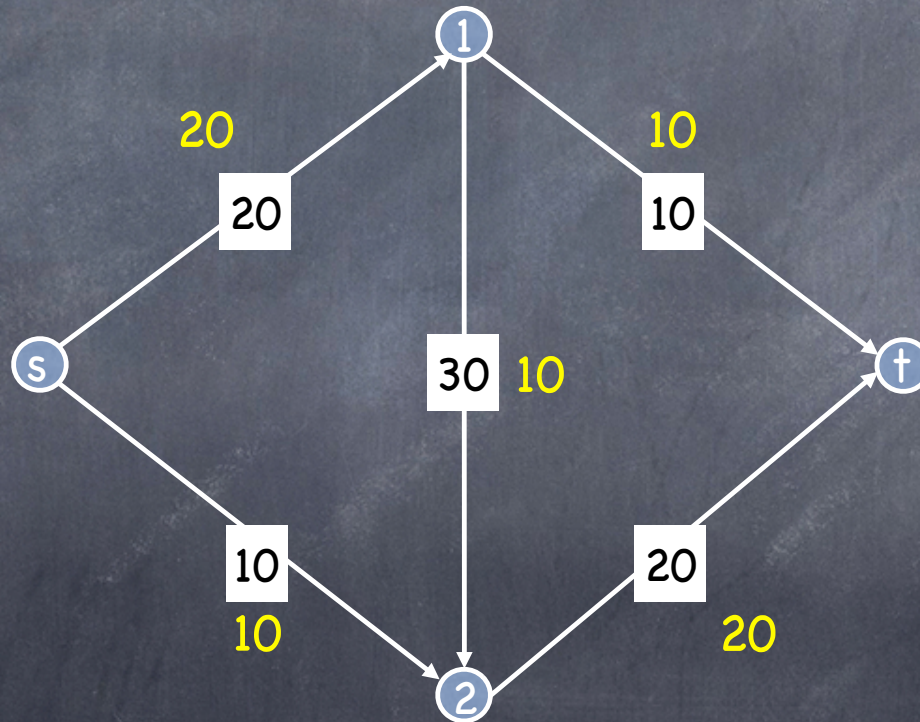
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Flow value = ~~20~~

Optimal Solution

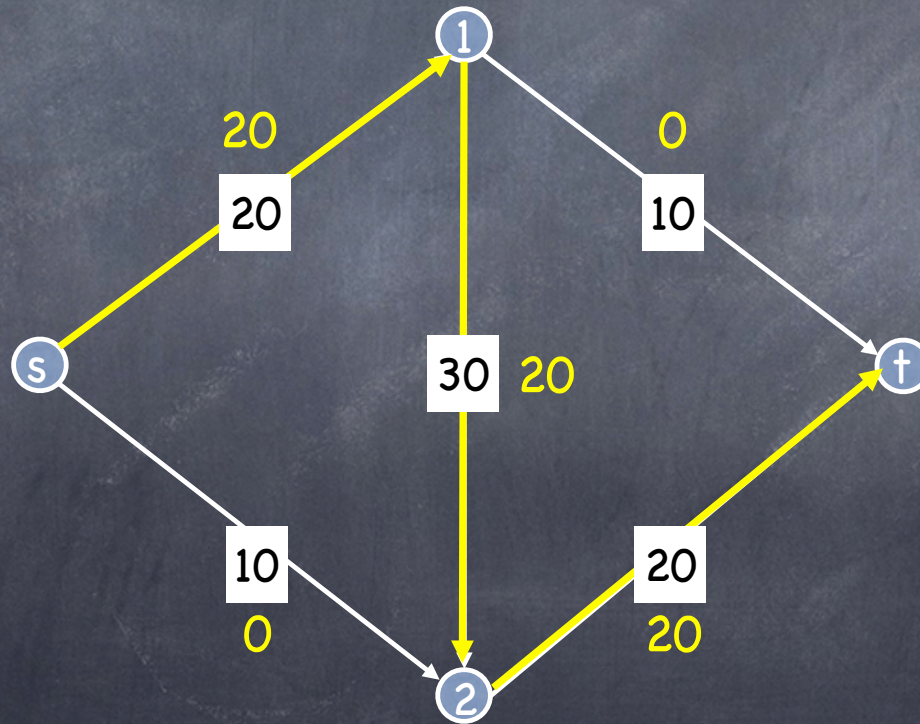
Flow value = 30



Problem

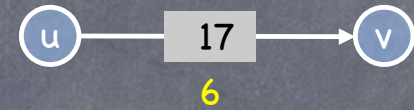
To fix the greedy algorithm, we need a way to track:

- (1) how much more flow can we send on any edge?
- (2) how much flow can we "undo" on each edge?



Residual Graph

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



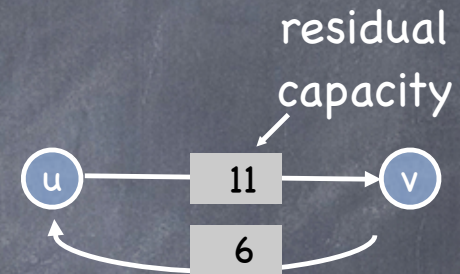
- Create two residual edges

- "Forward edge"

$e = (u, v)$ with capacity $c(e) - f(e)$

- "Backward edge"

$e' = (v, u)$ with capacity $f(e)$



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

- E_f = edges with positive residual capacity

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

Augmenting Path

Use path P in G_f to update flow in G

```
Augment(f, P) {  
    b = bottleneck(P)           // edge on P with least residual capacity  
    foreach e = (u,v) ∈ P {  
        if e is a forward edge  
            f(e) = f(e) + b     // forward edge: increase flow  
        else  
            let e' = (v, u)  
            f(e') = f(e') - b   // backward edge: decrease flow  
    }  
    return f  
}
```

Example on board

Ford-Fulkerson Algorithm

Repat: find an augmenting path, and augment!

```
Ford-Fulkerson(G, s, t) {  
    foreach e ∈ E  f(e) = 0 // initially, no flow  
    Gf = copy of G      // residual graph = original graph  
  
    while (there exists an s-t path P in Gf) {  
        f = Augment(f, P) // change the flow  
        update Gf       // build a new residual graph  
    }  
    return f  
}
```


Next Time

- Termination and running time (easy)
- **Correctness:** Max-Flow Min-Cut Theorem