



<http://algs4.cs.princeton.edu>

## 6.4 MAXIMUM FLOW

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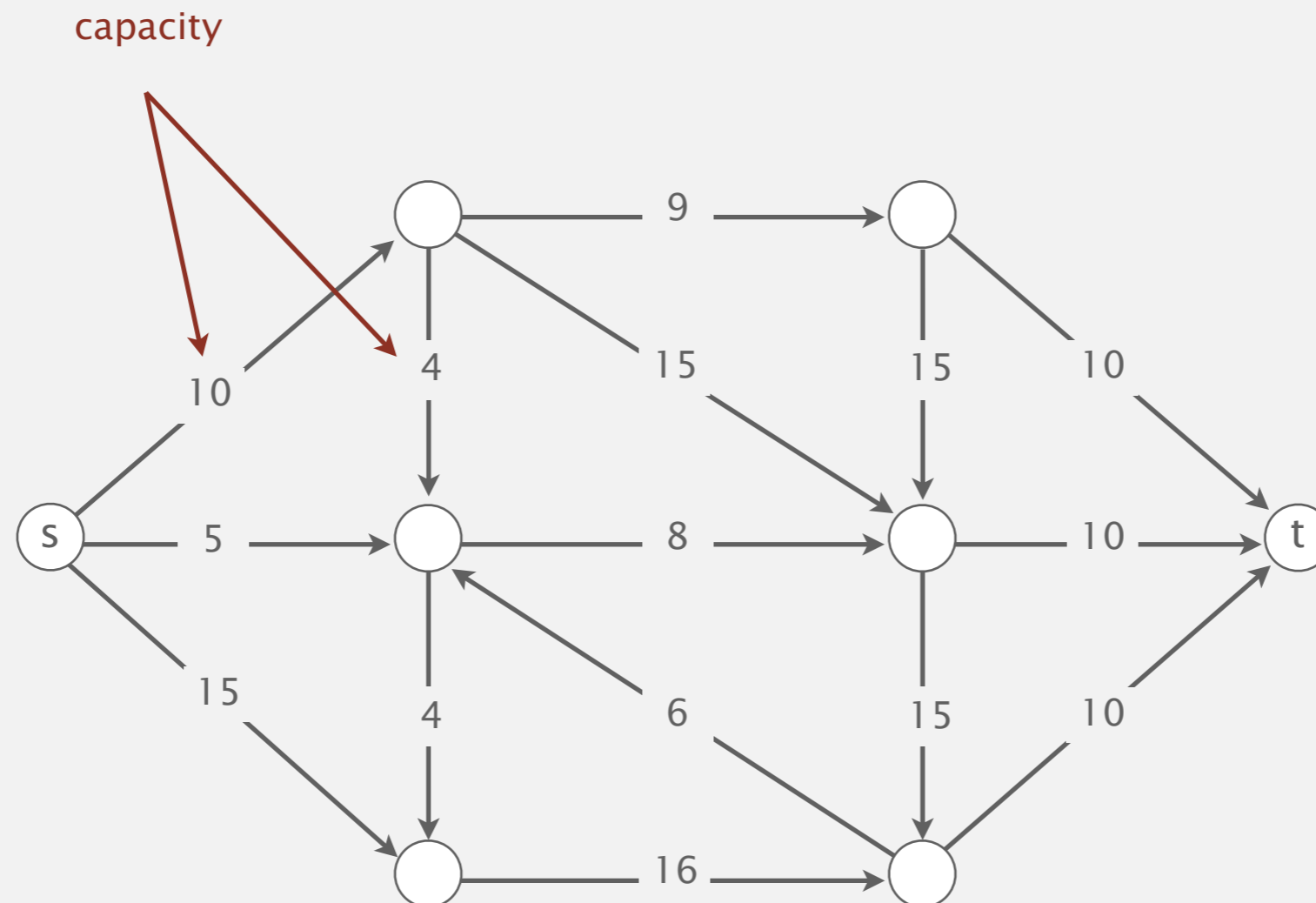
- *introduction*
- *Ford-Fulkerson algorithm*
- *maxflow-mincut theorem*
- *analysis of running time*
- *Java implementation*
- *applications*

# Mincut problem

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**Input.** An edge-weighted digraph, source vertex  $s$ , and target vertex  $t$ .

each edge has a  
positive capacity

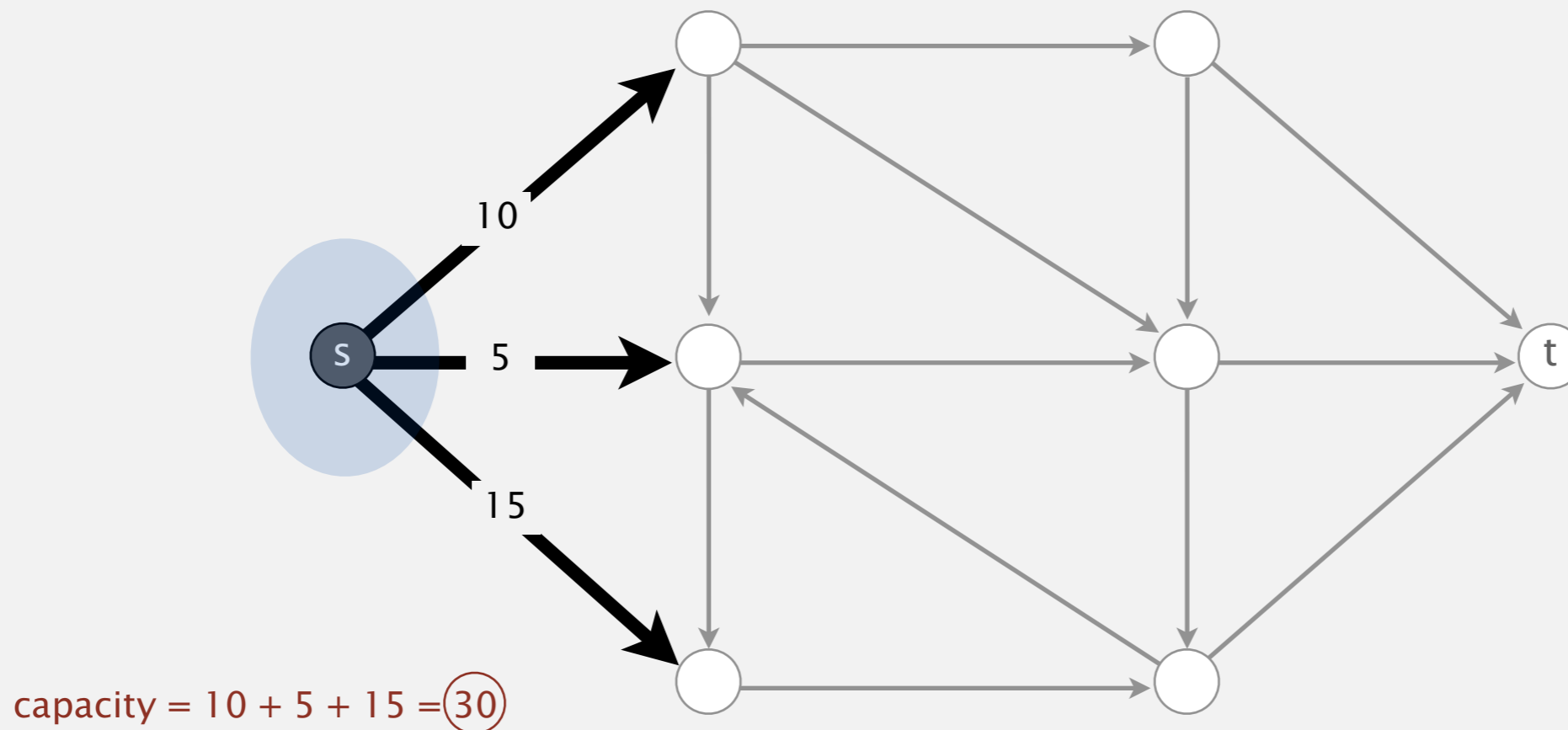


# Mincut problem

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**Def.** A *st-cut (cut)* is a partition of the vertices into two disjoint sets, with  $s$  in one set  $A$  and  $t$  in the other set  $B$ .

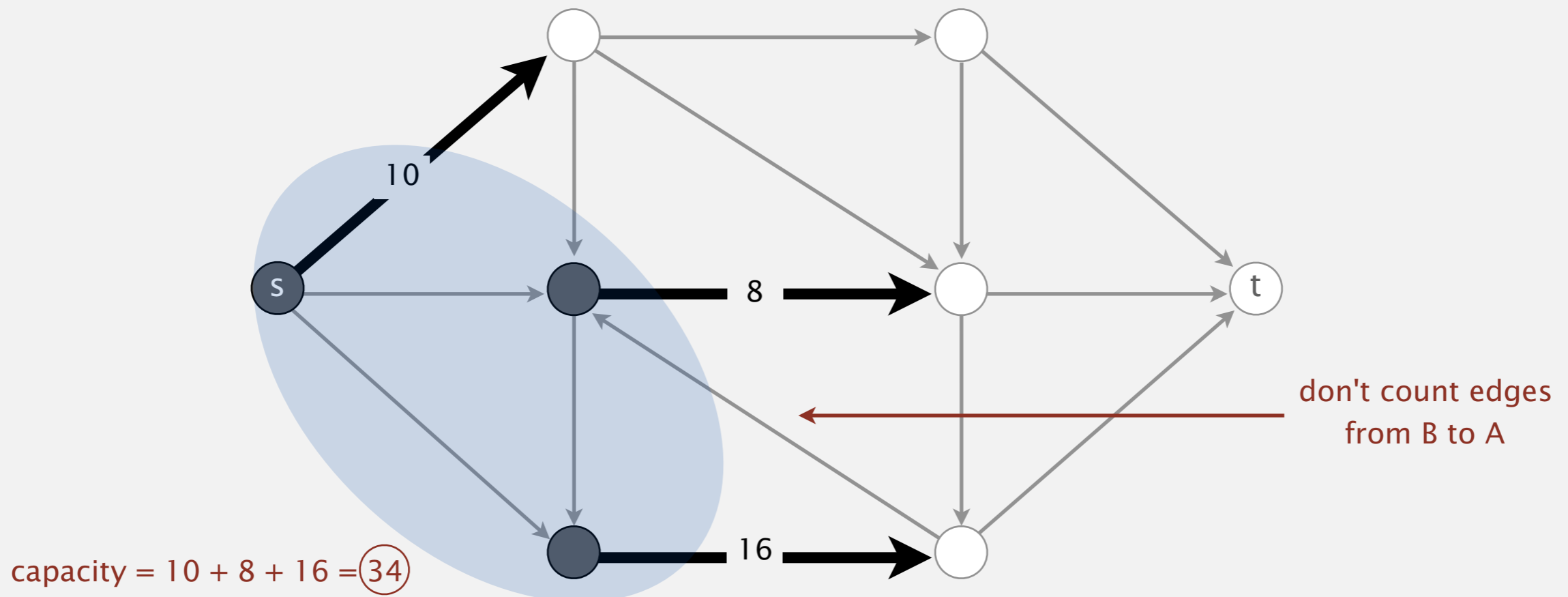
**Def.** Its *capacity* is the sum of the capacities of the edges from  $A$  to  $B$ .



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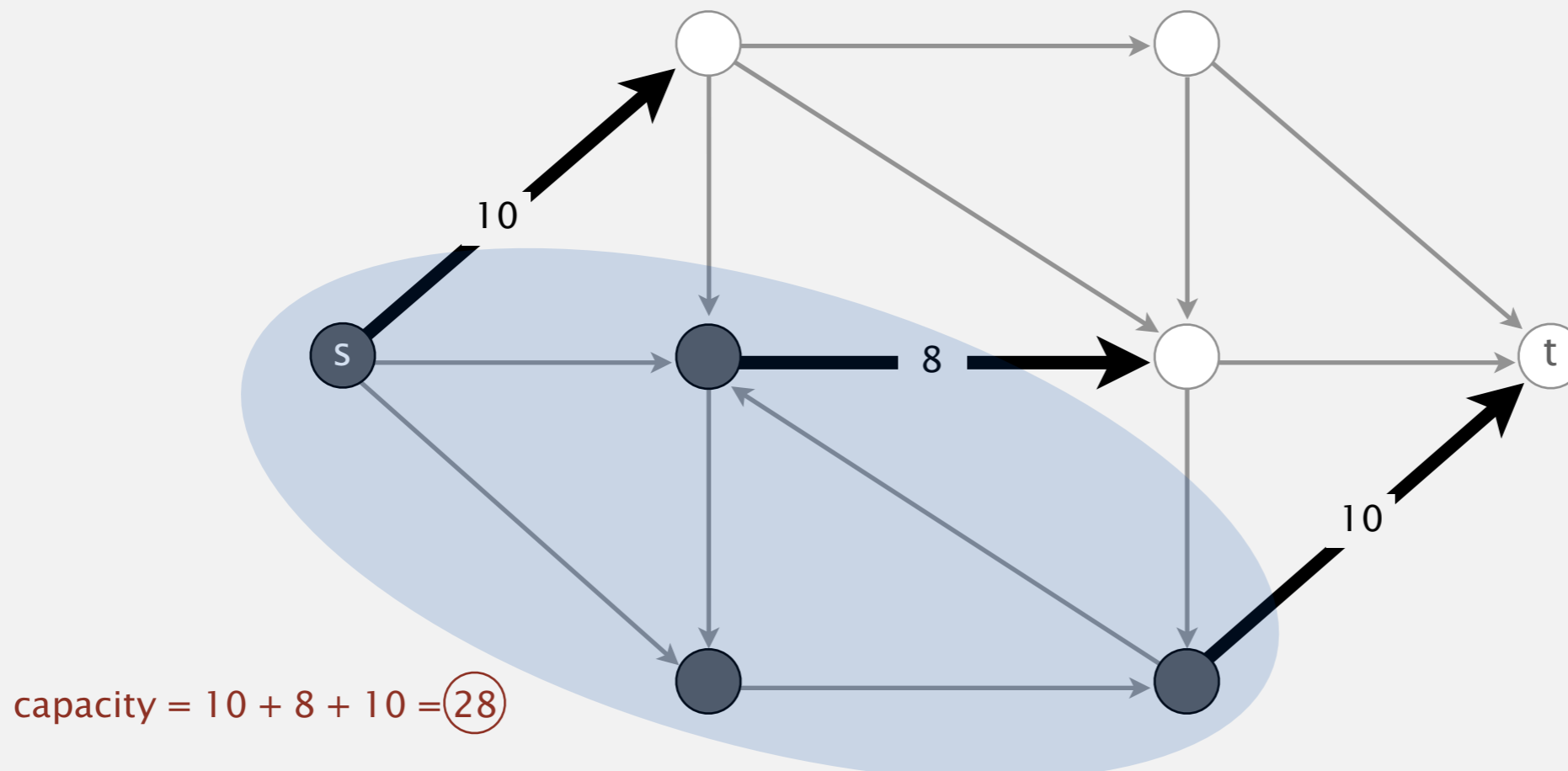
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**Minimum st-cut (mincut) problem.** Find a cut of minimum capacity.

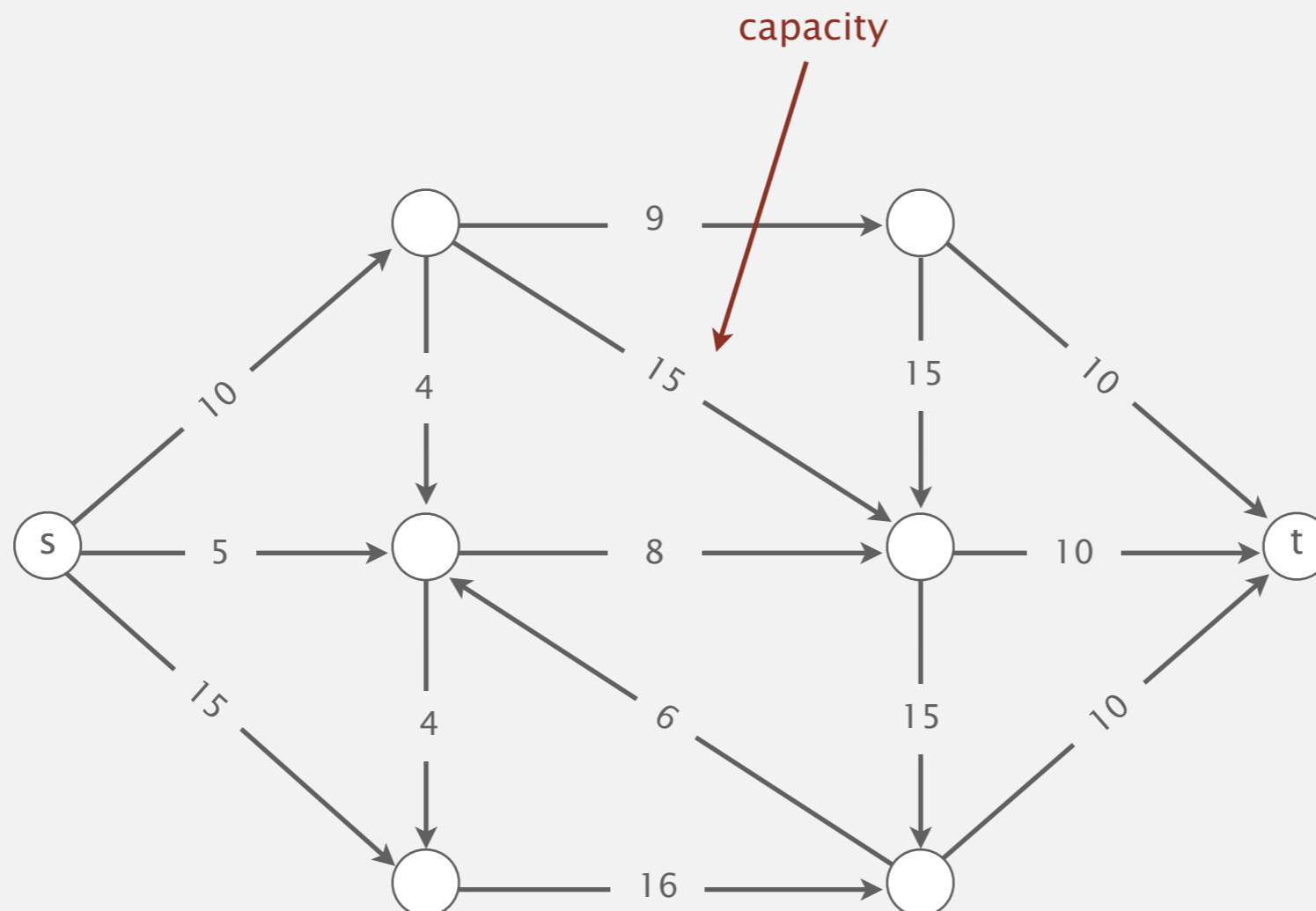


# Maxflow problem

---

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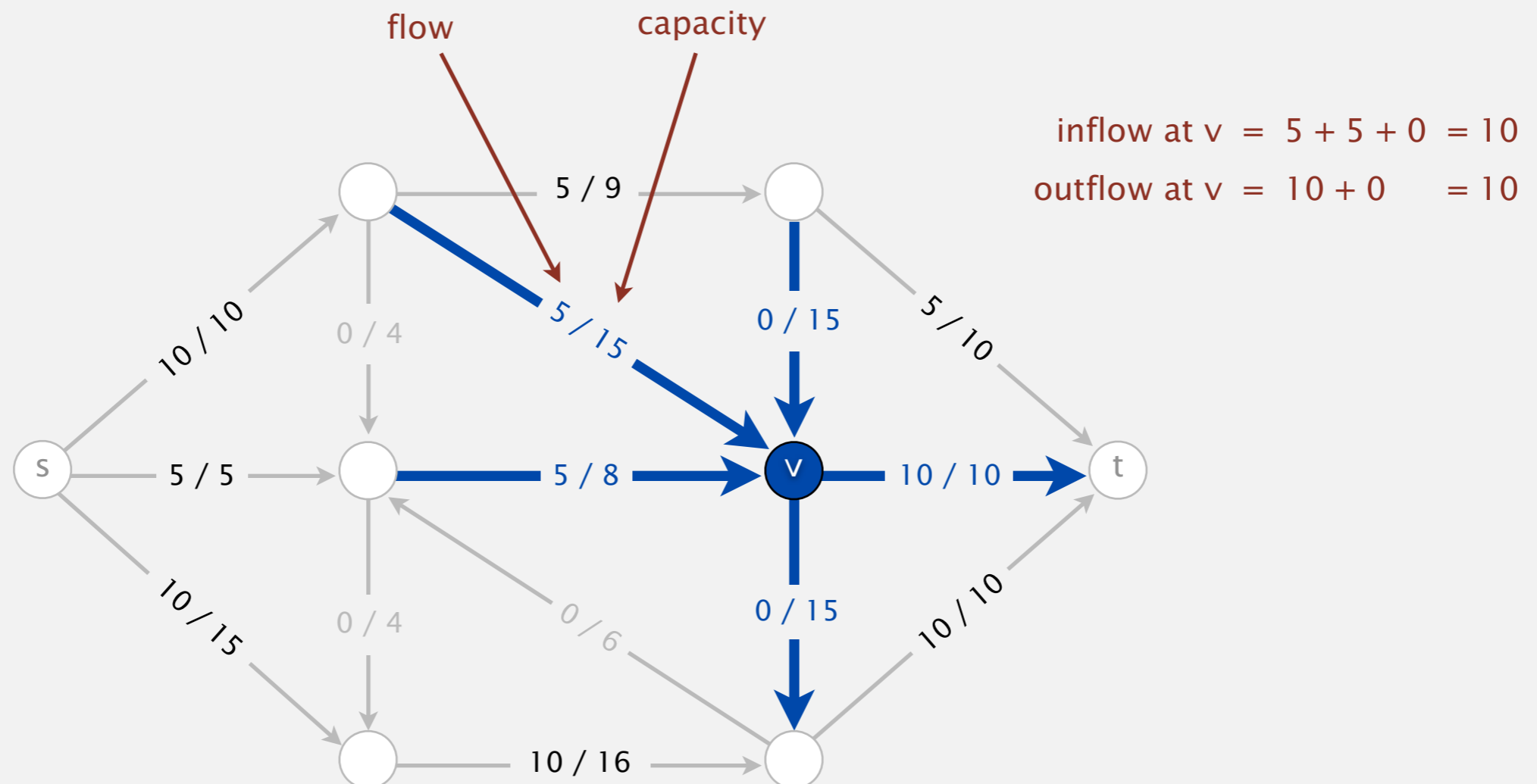
each edge has a  
positive capacity



# Maxflow problem

**Def.** An *st-flow* (flow) is an assignment of values to the edges such that:

- Capacity constraint:  $0 \leq \text{edge's flow} \leq \text{edge's capacity}$ .
- Local equilibrium: inflow = outflow at every vertex (except  $s$  and  $t$ ).



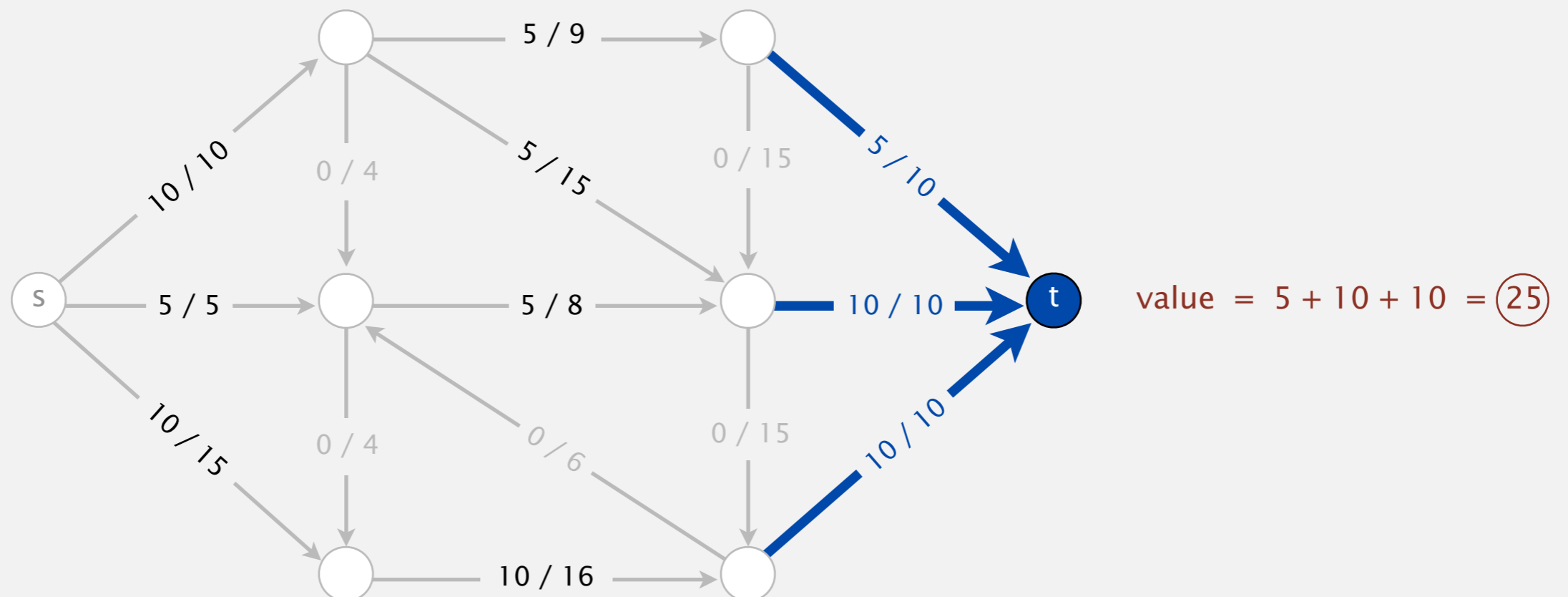
# Maxflow problem

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**Def.** The *value* of a flow is the inflow at  $t$ .

we assume no edges point to  $s$  or from  $t$



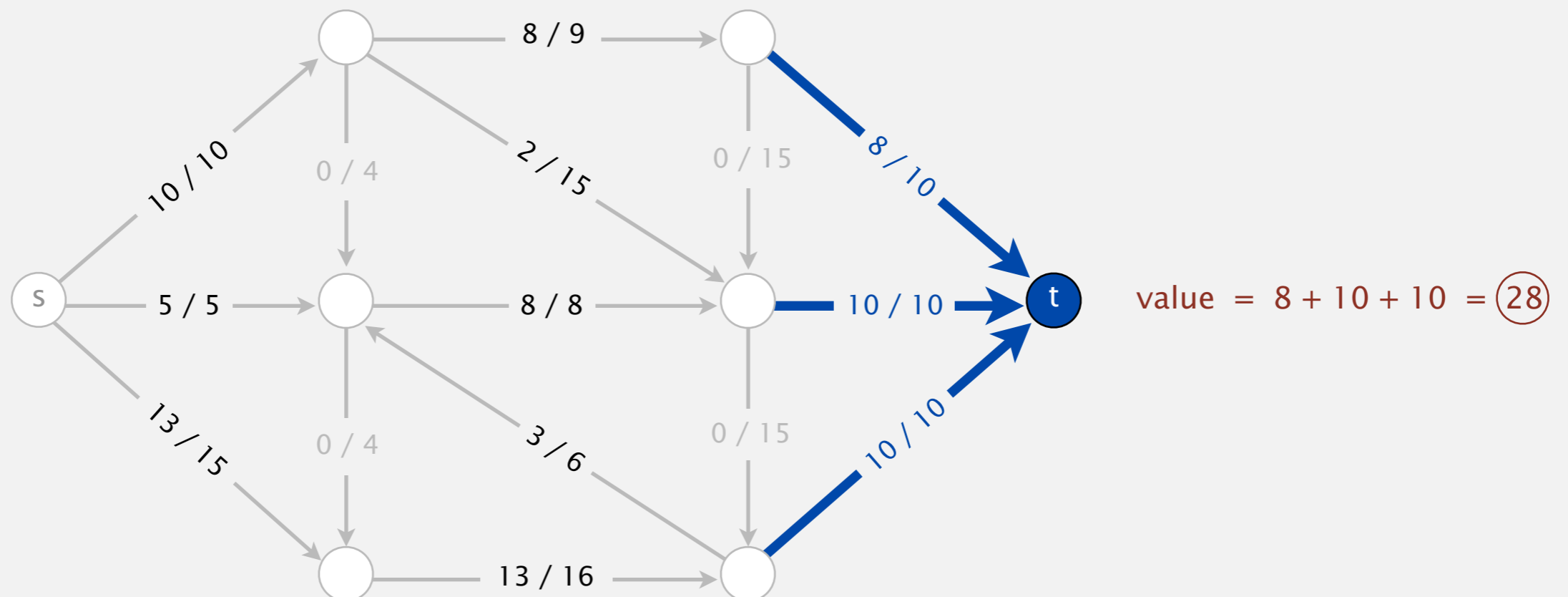
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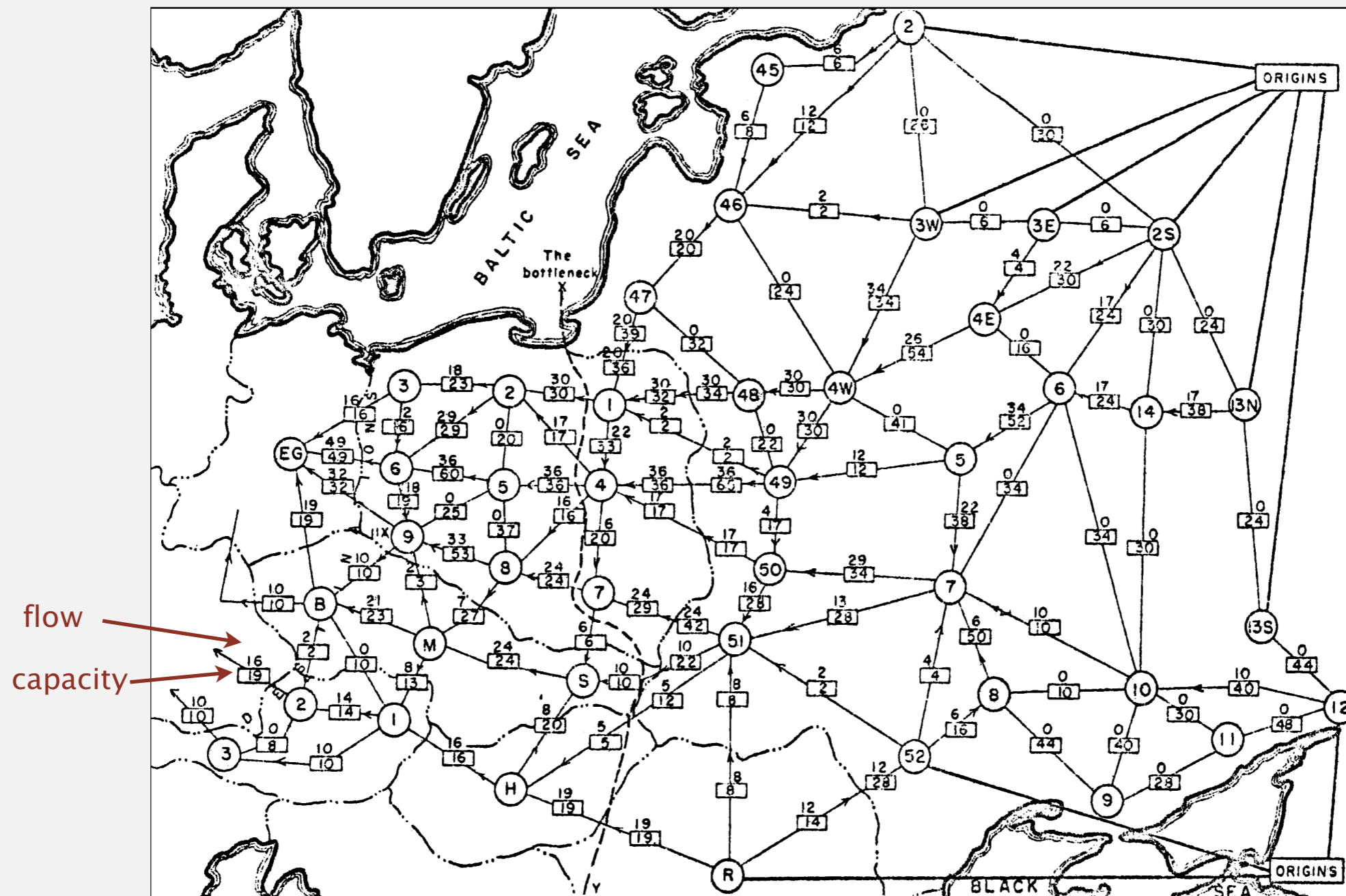
**Def.** The *value* of a flow is the inflow at  $t$ .

**Maximum st-flow (maxflow) problem.** Find a flow of maximum value.



# Maxflow application (Tolstoï 1930s)

Soviet Union goal. Maximize flow of supplies to Eastern Europe.



rail network connecting Soviet Union with Eastern European countries  
(map declassified by Pentagon in 1999)

# Potential maxflow application (2010s)

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"Free world" goal. Maximize flow of information to specified set of people.



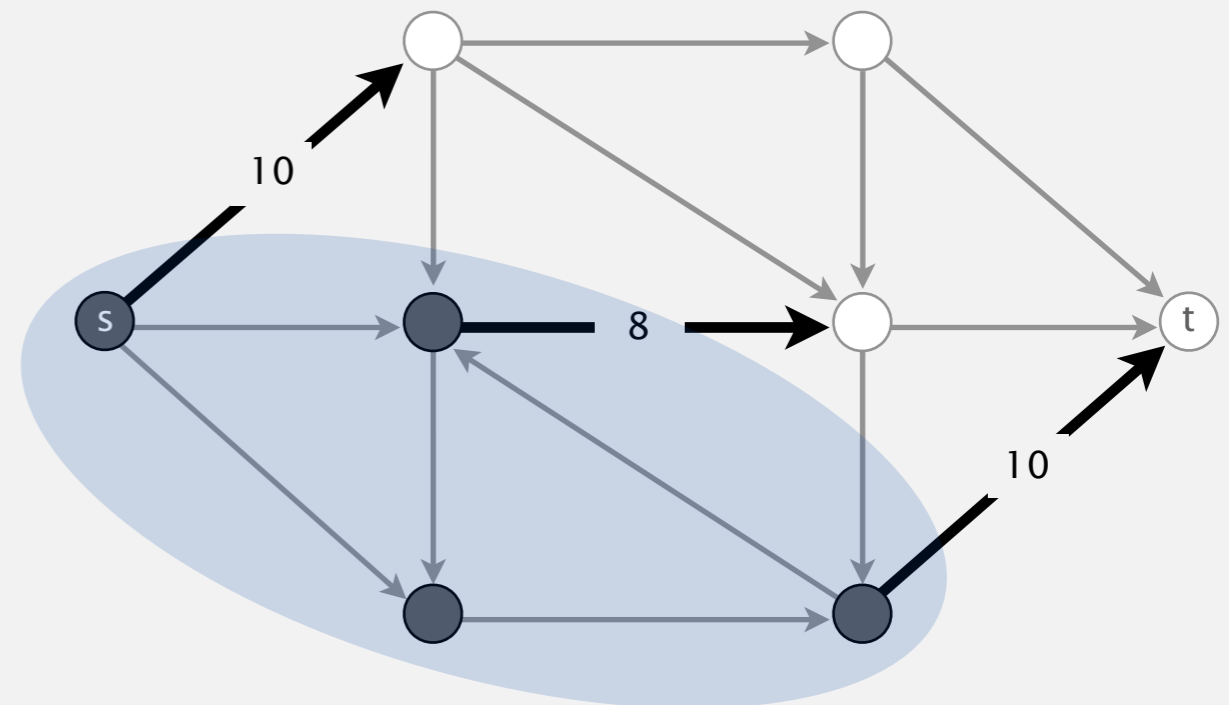
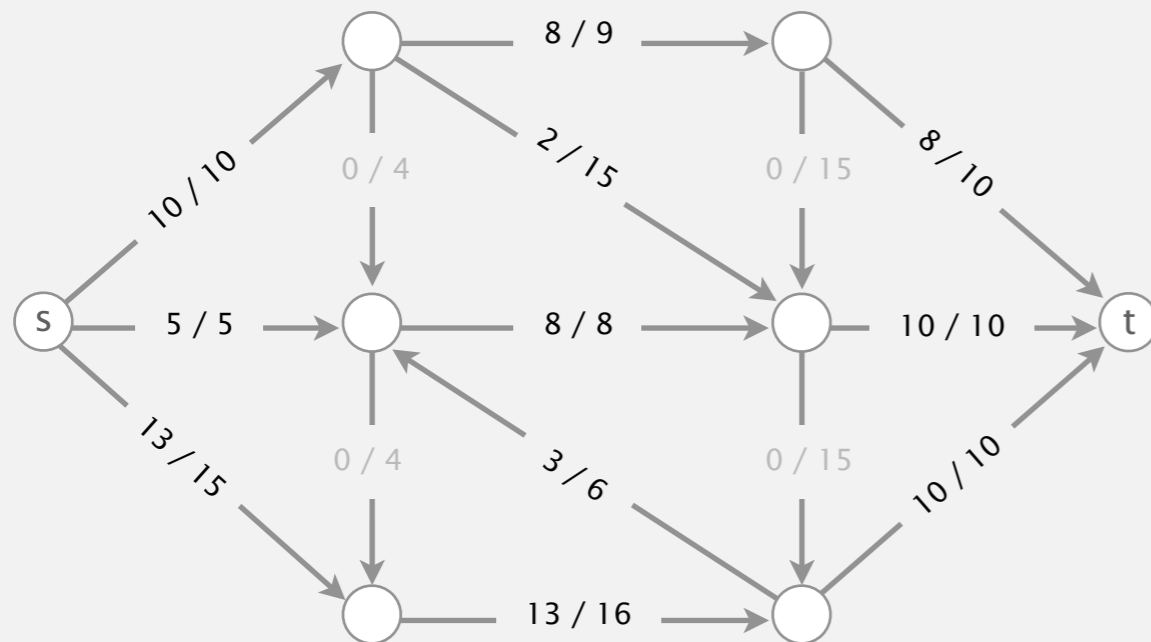
facebook graph

# Summary

**Input.** A weighted digraph, source vertex  $s$ , and target vertex  $t$ .

**Mincut problem.** Find a cut of minimum capacity.

**Maxflow problem.** Find a flow of maximum value.



**Remarkable fact.** These two problems are same!



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## 6.4 MAXIMUM FLOW

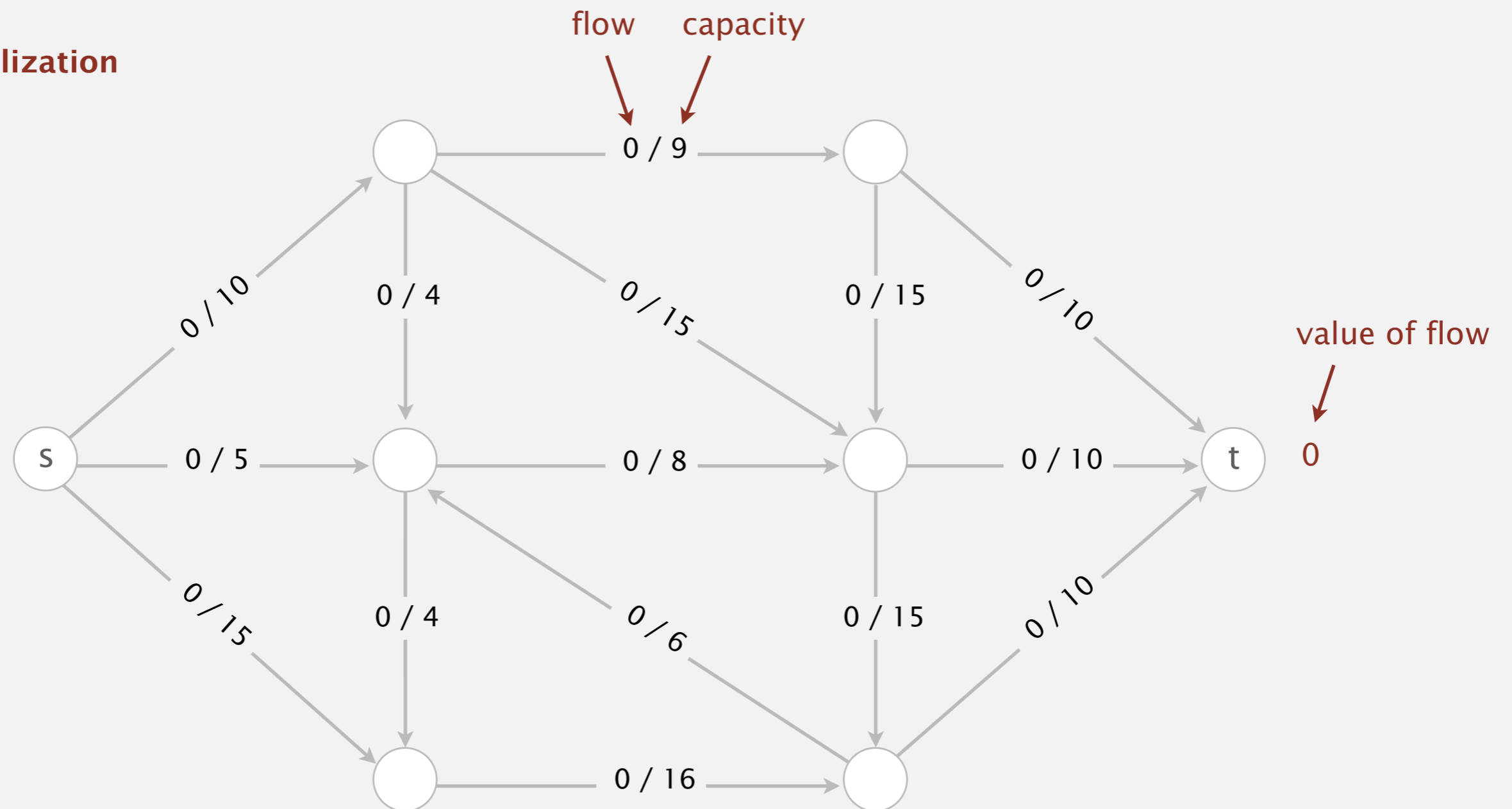
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- *introduction*
- *Ford-Fulkerson algorithm*
- *maxflow-mincut theorem*
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# Ford-Fulkerson algorithm

Initialization. Start with 0 flow.

initialization

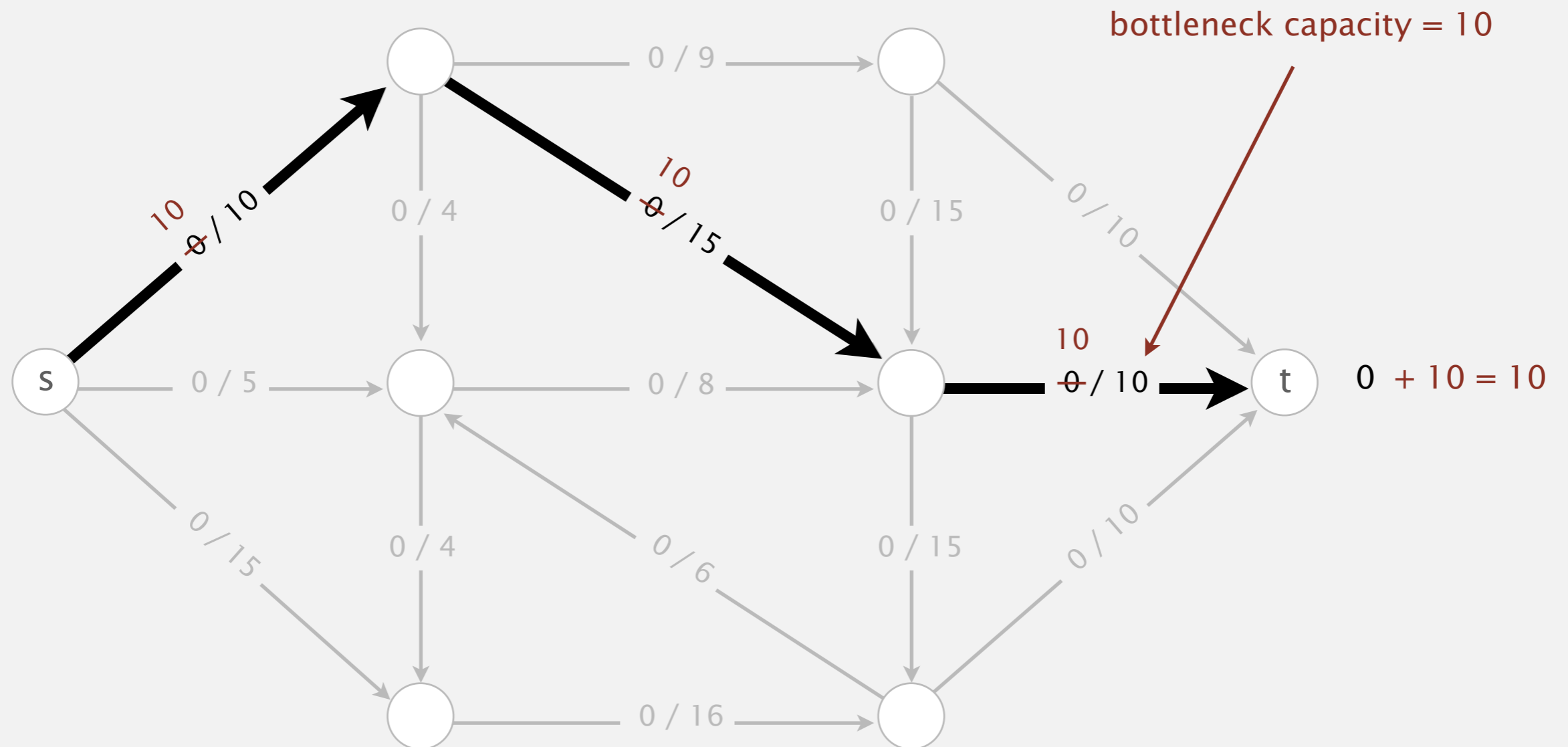


# Idea: increase flow along augmenting paths

**Augmenting path.** Find an undirected path from  $s$  to  $t$  such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

## 1<sup>st</sup> augmenting path

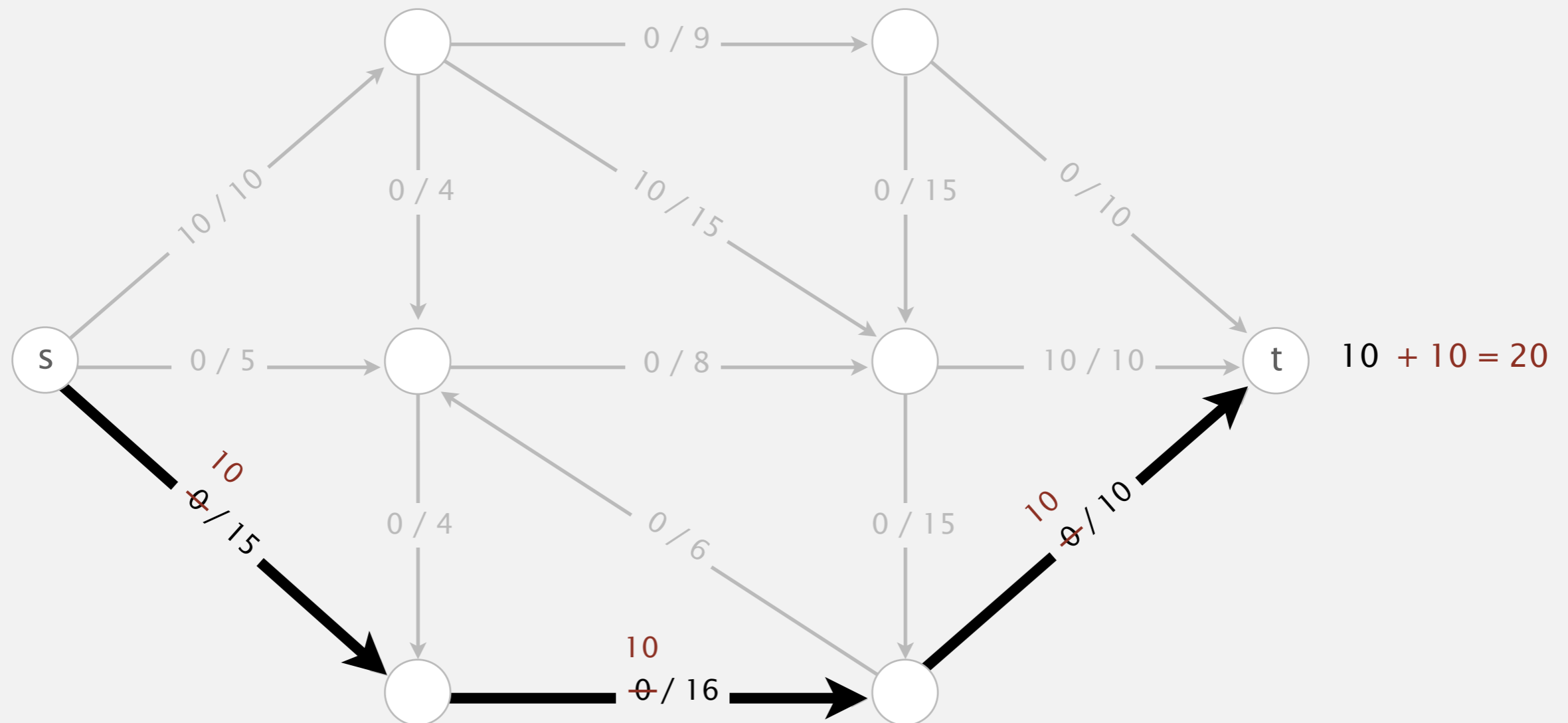


# Idea: increase flow along augmenting paths

**Augmenting path.** Find an undirected path from  $s$  to  $t$  such that:

- Can increase flow on forward edges (not full).
- Can decrease flow on backward edge (not empty).

## 2<sup>nd</sup> augmenting path

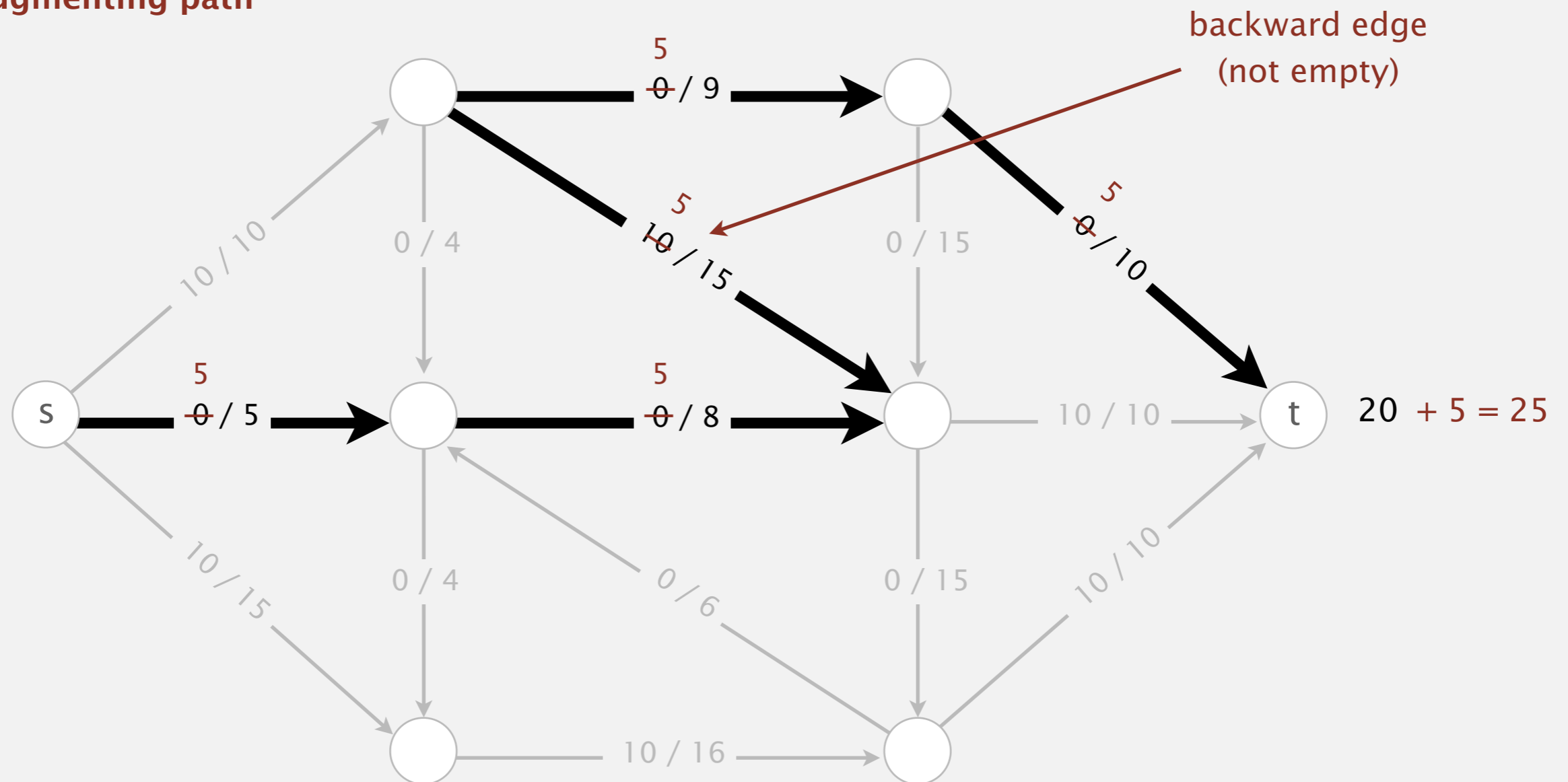


# Idea: increase flow along augmenting paths

**Augmenting path.** Find an undirected path from  $s$  to  $t$  such that:

- Can increase flow on forward edges (not full).
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**3<sup>rd</sup> augmenting path**

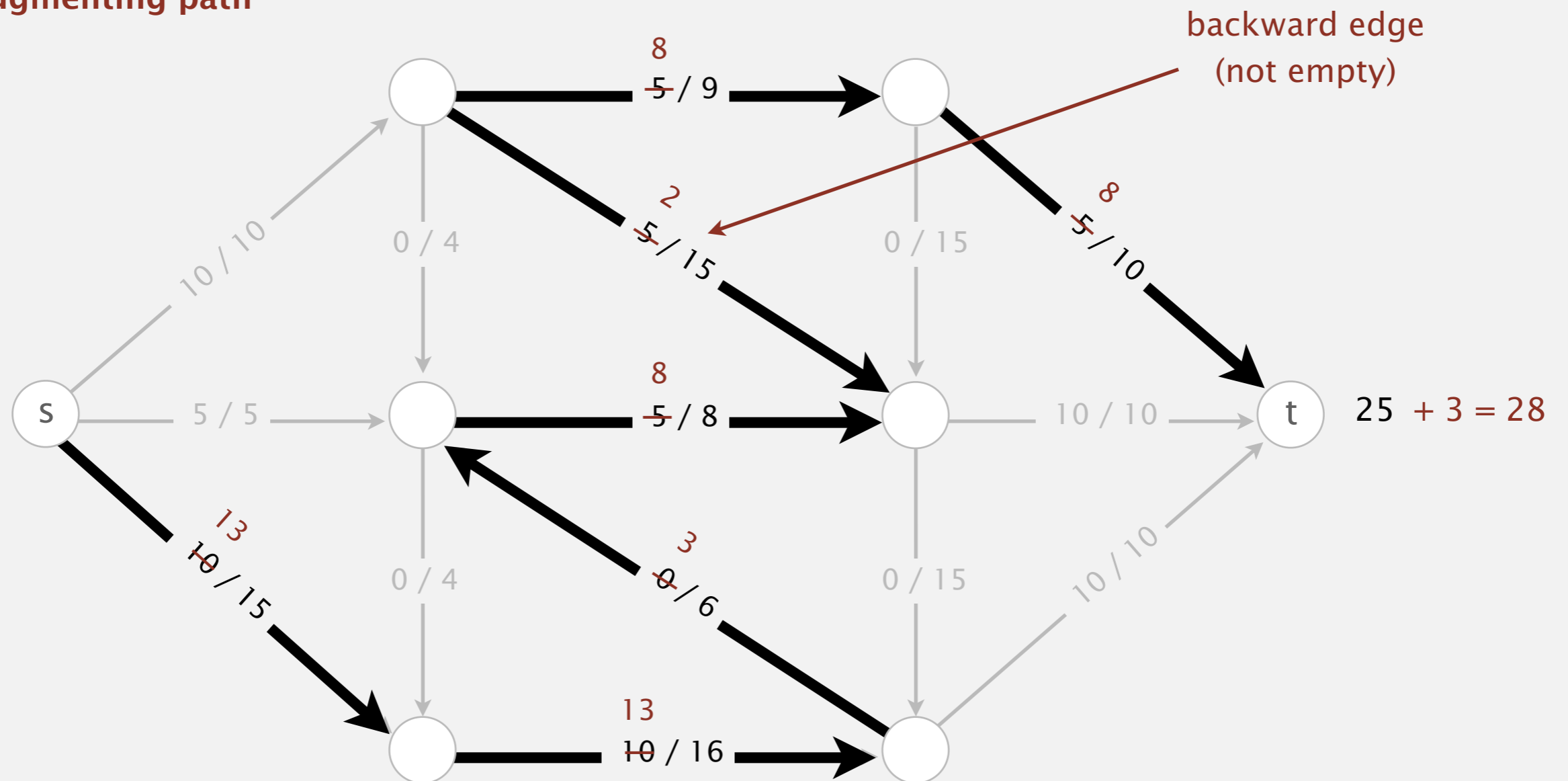


# Idea: increase flow along augmenting paths

**Augmenting path.** Find an undirected path from  $s$  to  $t$  such that:

- Can increase flow on forward edges (not full).
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4<sup>th</sup> augmenting path

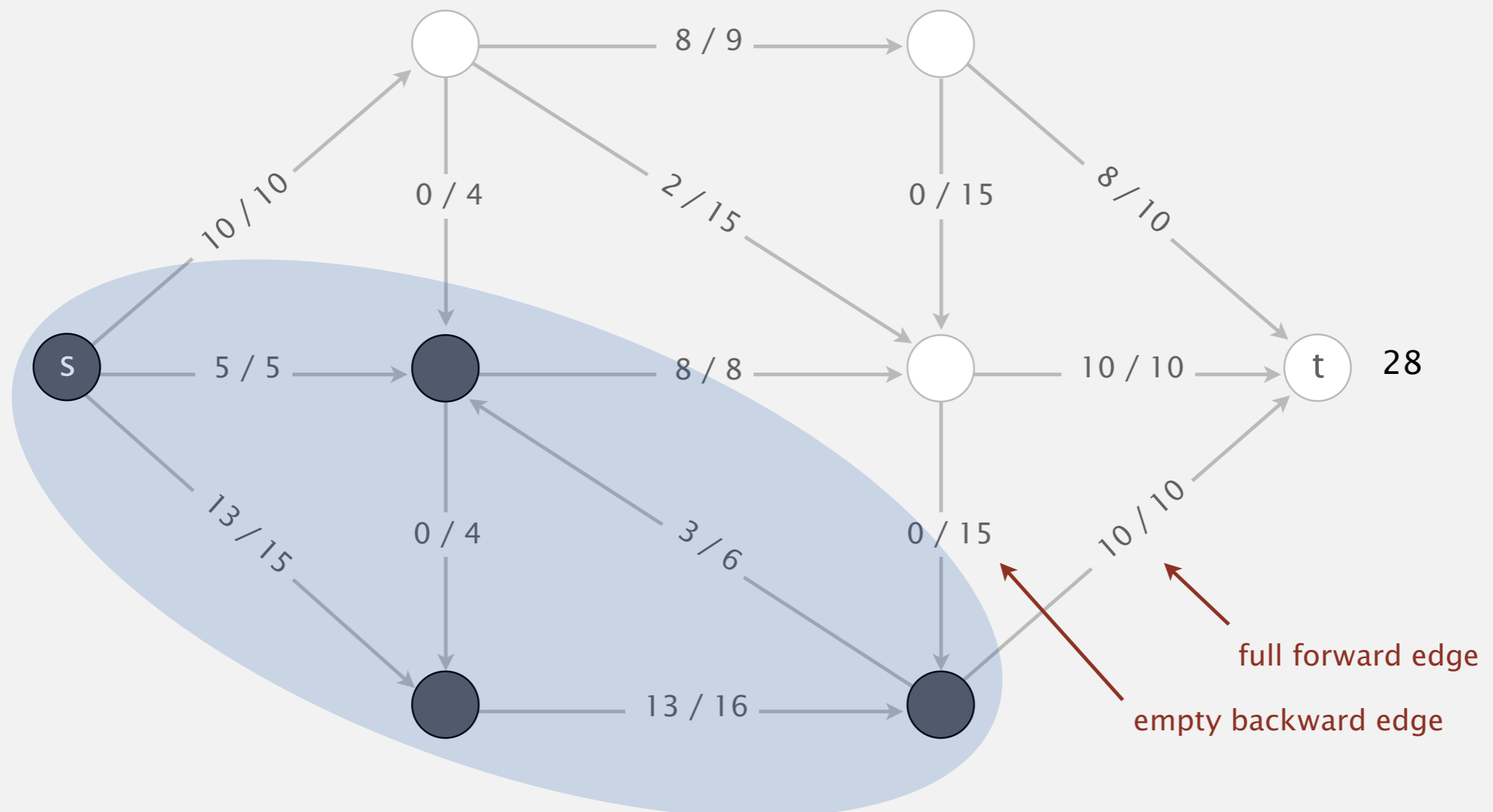


# Idea: increase flow along augmenting paths

**Termination.** All paths from  $s$  to  $t$  are blocked by either a

- Full forward edge.
- Empty backward edge.

no more augmenting paths



# Ford-Fulkerson algorithm

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## **Ford-Fulkerson algorithm**

---

**Start with 0 flow.**

**While there exists an augmenting path:**

- **find an augmenting path**
  - **compute bottleneck capacity**
  - **increase flow on that path by bottleneck capacity**
- 

## **Fundamental questions.**

- How to compute a mincut?
- How to find an augmenting path?
- If FF terminates, does it always compute a maxflow?
- Does FF always terminate? If so, after how many augmentations?



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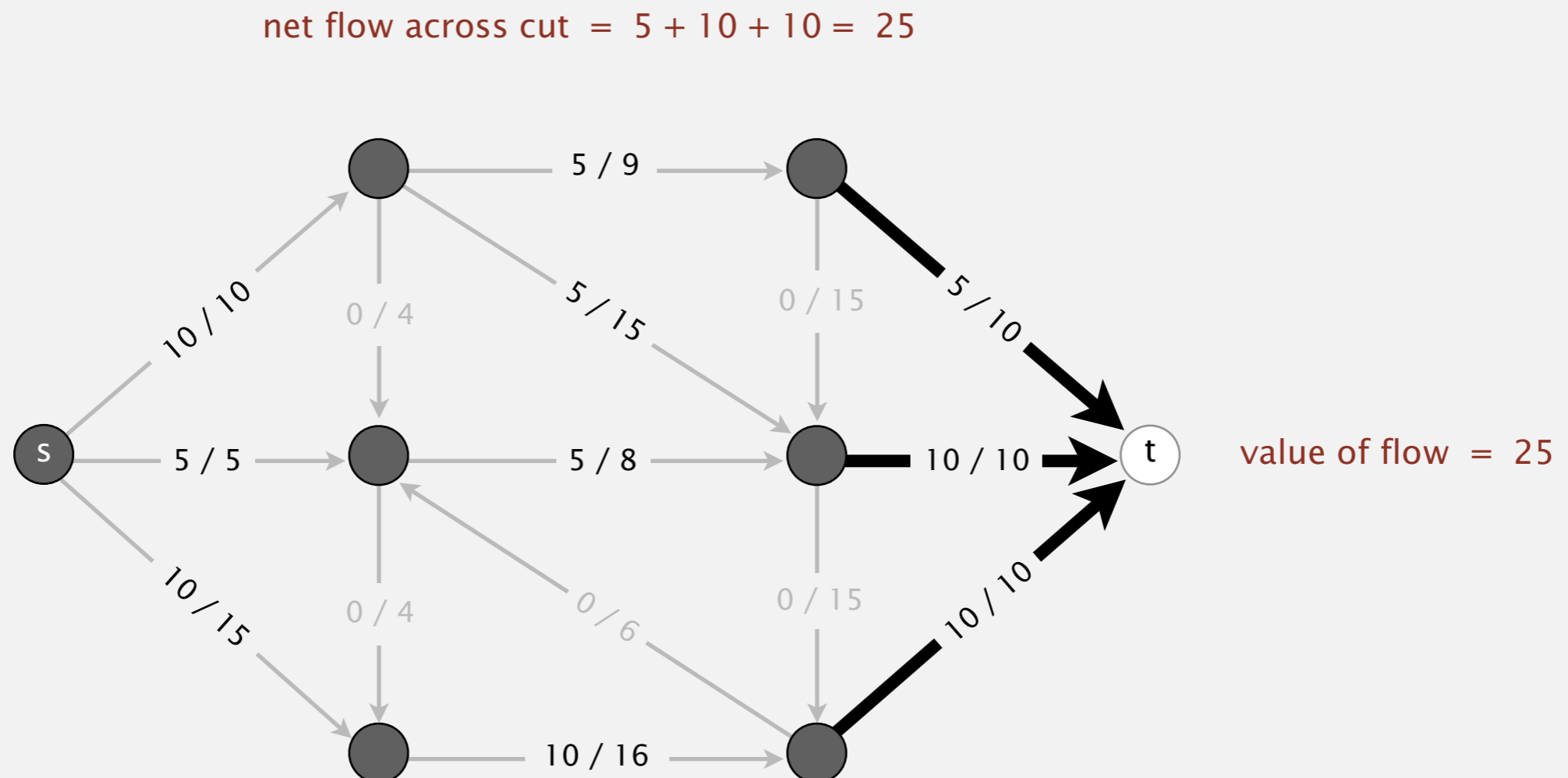
## 6.4 MAXIMUM FLOW

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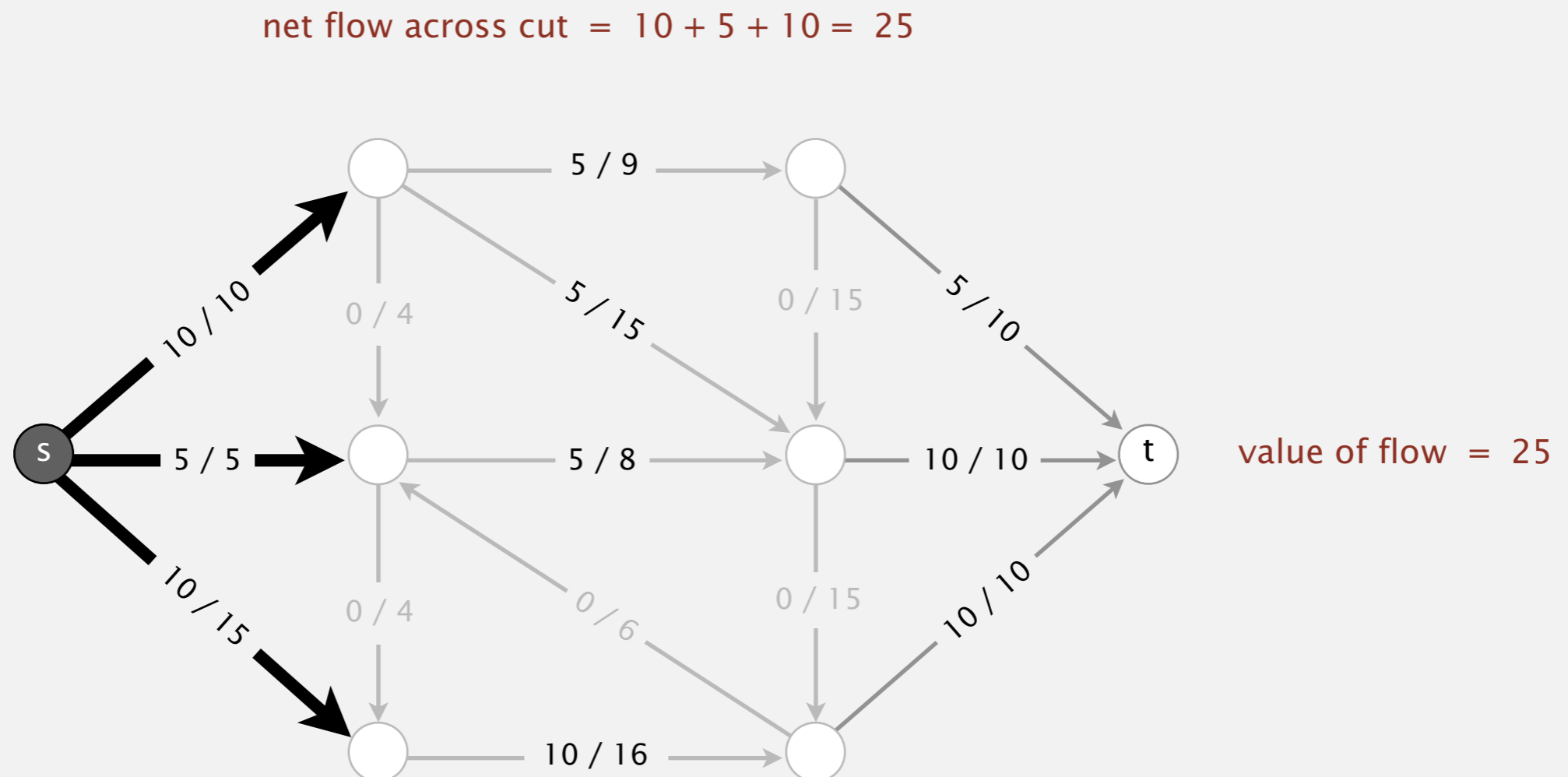
# Relationship between flows and cuts

**Def.** The **net flow across** a cut  $(A, B)$  is the sum of the flows on its edges from  $A$  to  $B$  minus the sum of the flows on its edges from  $B$  to  $A$ .



# Relationship between flows and cuts

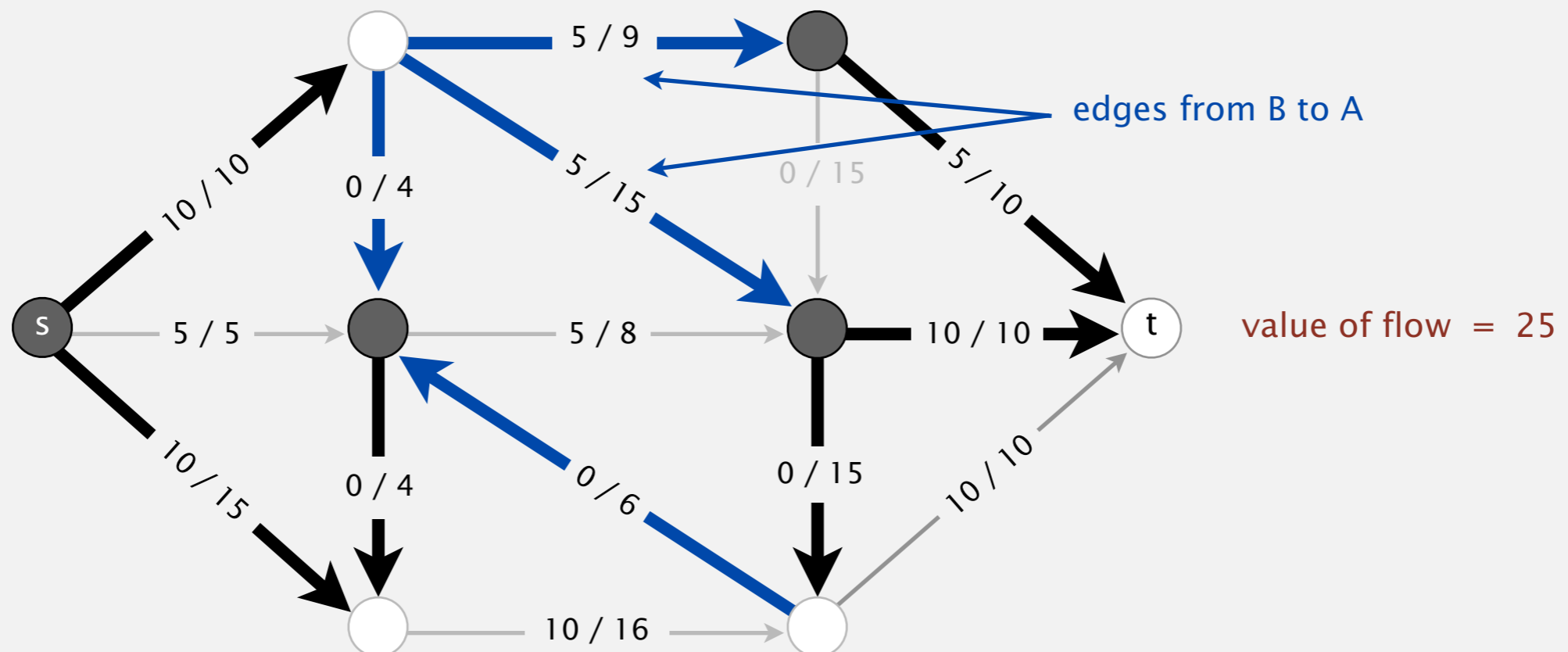
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# Relationship between flows and cuts

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$$\text{net flow across cut} = (10 + 10 + 5 + 10 + 0 + 0) - (5 + 5 + 0 + 0) = 25$$



# Relationship between flows and cuts

---

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

**Intuition.** Conservation of flow.

**Pf.** By induction on the size of  $B$ .

- Base case:  $B = \{ t \}$ .
- Induction step: remains true by local equilibrium when moving any vertex from  $A$  to  $B$ .

**Key Idea.** Outflow from  $s$  = inflow to  $t$  = value of flow.

# Relationship between flows and cuts

Let  $f$  be any flow and let  $(A, B)$  be any cut.

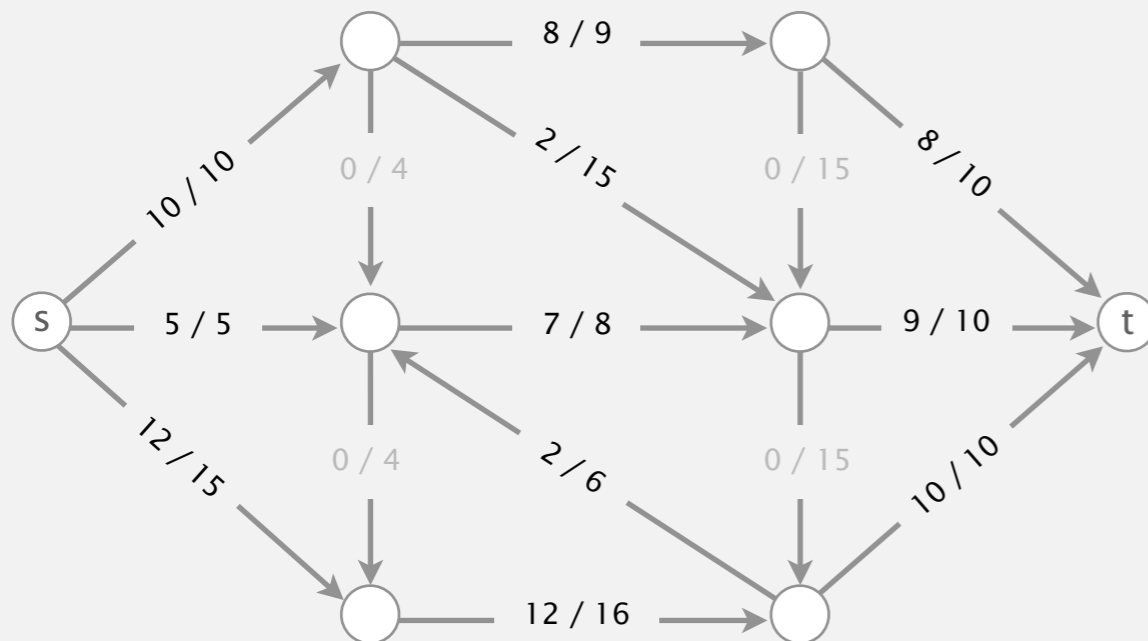
Then, the value of the flow  $\leq$  the capacity of the cut.

Value of flow  $f$  = net flow across cut  $(A, B) \leq$  capacity of cut  $(A, B)$ .

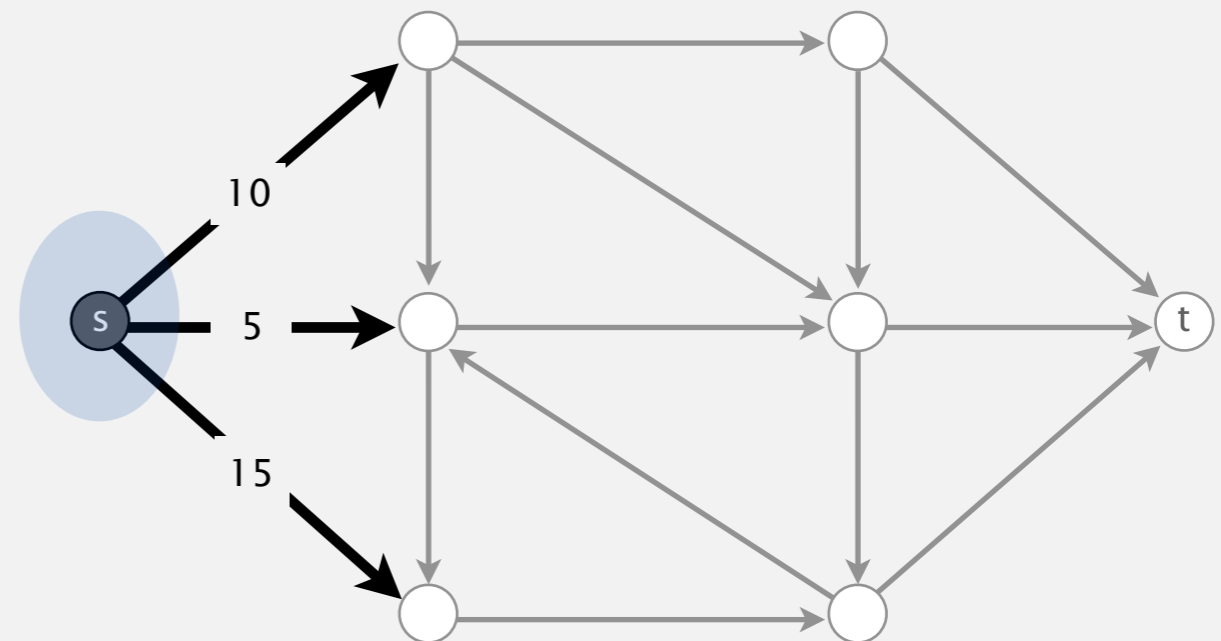
↑  
flow-value lemma

↑  
flow bounded by capacity

Think of the nodes collapsing on themselves.



value of flow = 27



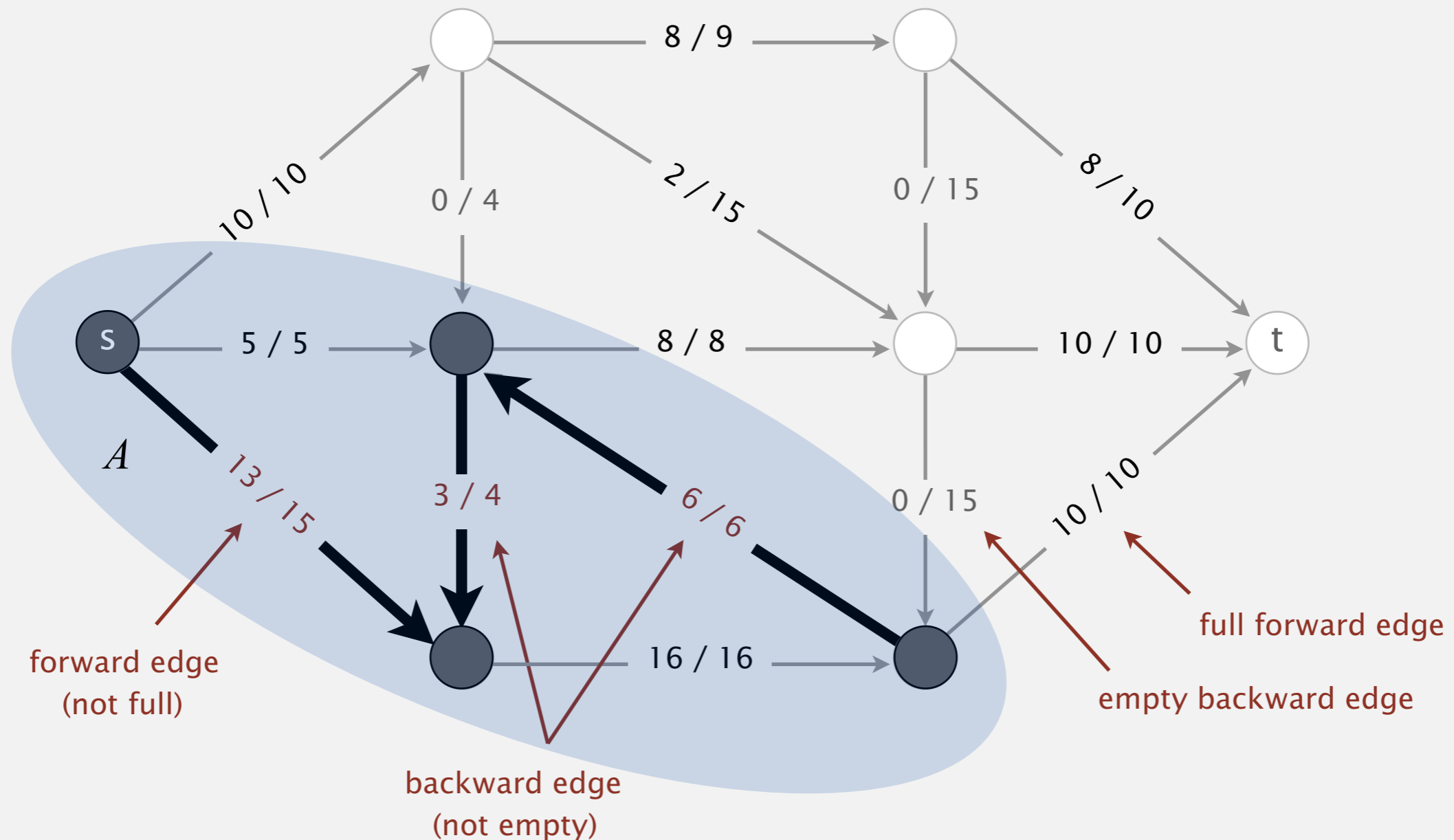
capacity of cut = 30

# Computing a mincut from a maxflow

To compute mincut  $(A, B)$  from maxflow  $f$  :

- Compute  $A$  = set of vertices connected to  $s$  by an undirected path with no full forward or empty backward edges.

Think of running DFS on the undirected graph that with full forward edges and empty backward edges removed





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# Ford-Fulkerson algorithm

---

## Ford-Fulkerson algorithm

---

Start with 0 flow.

While there exists an augmenting path:

- find an augmenting path
  - compute bottleneck capacity
  - increase flow on that path by bottleneck capacity
- 

## Fundamental questions.

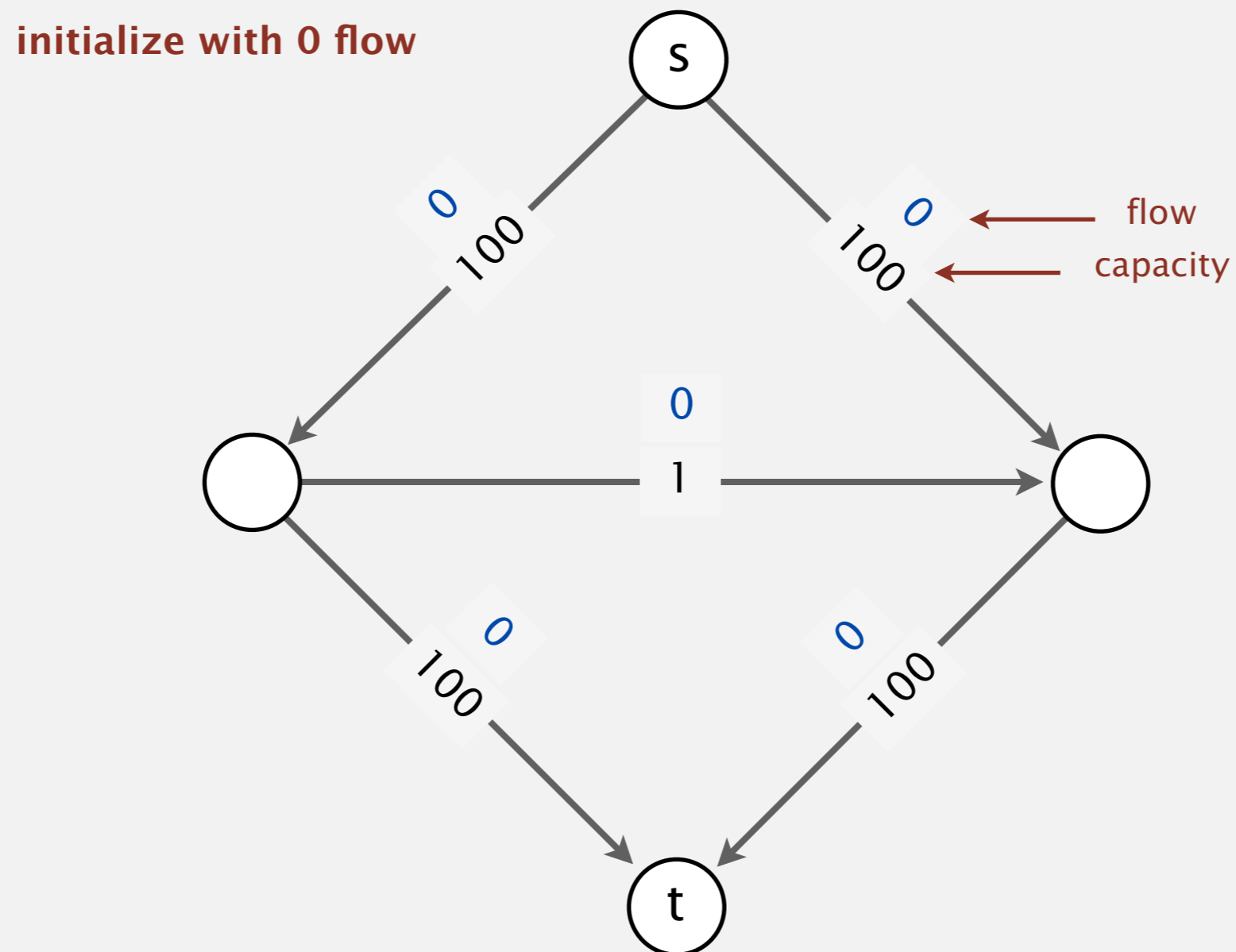
- How to compute a mincut? **Easy. ✓**
- How to find an augmenting path? **BFS works well.**
- If FF terminates, does it always compute a maxflow? **Yes. ✓**
- Does FF always terminate? If so, after how many augmentations?

yes, provided edge capacities are integers  
(or augmenting paths are chosen carefully)

requires clever analysis

# Bad case for Ford-Fulkerson

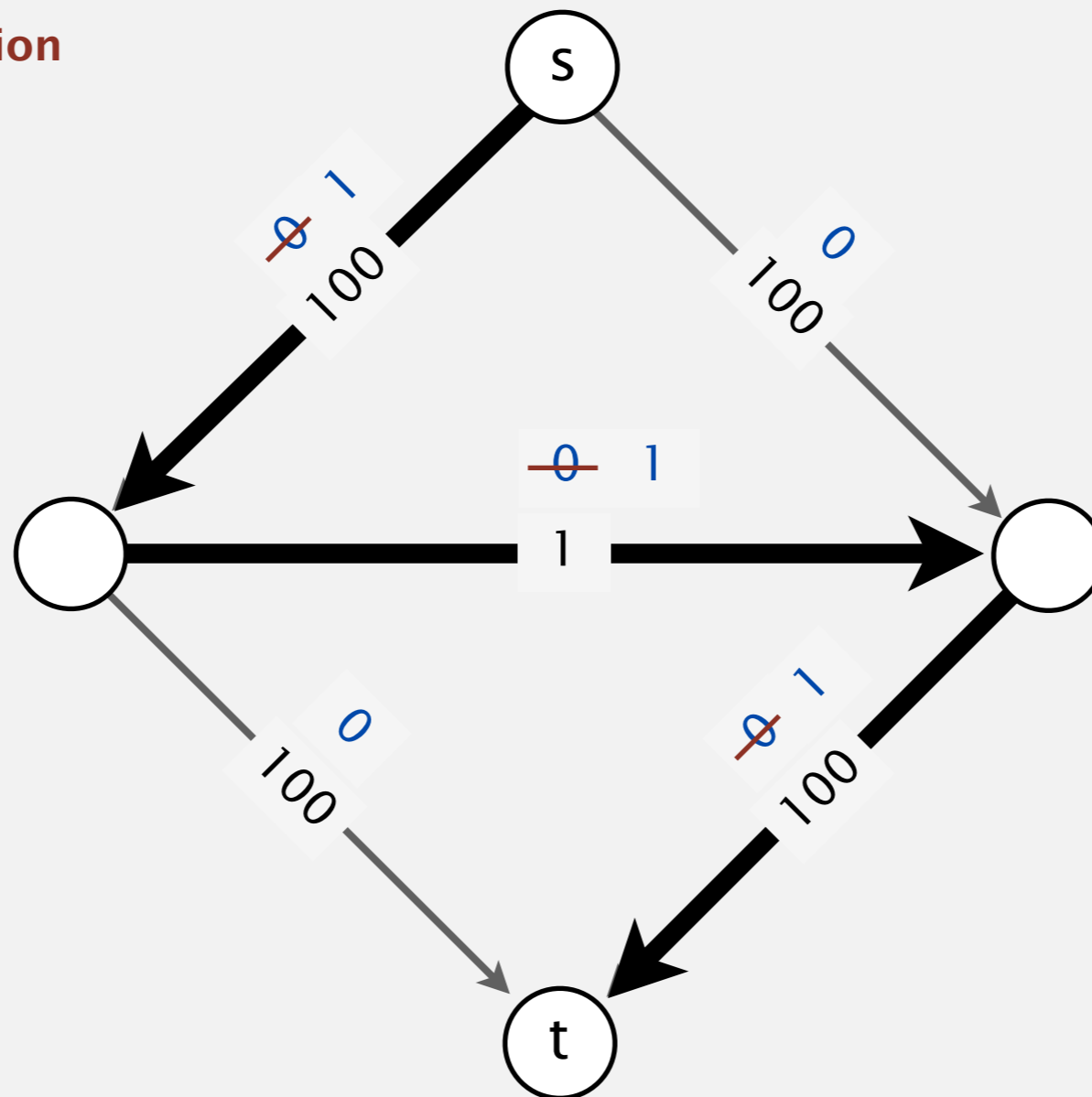
**Bad news.** Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.



## Bad case for Ford-Fulkerson

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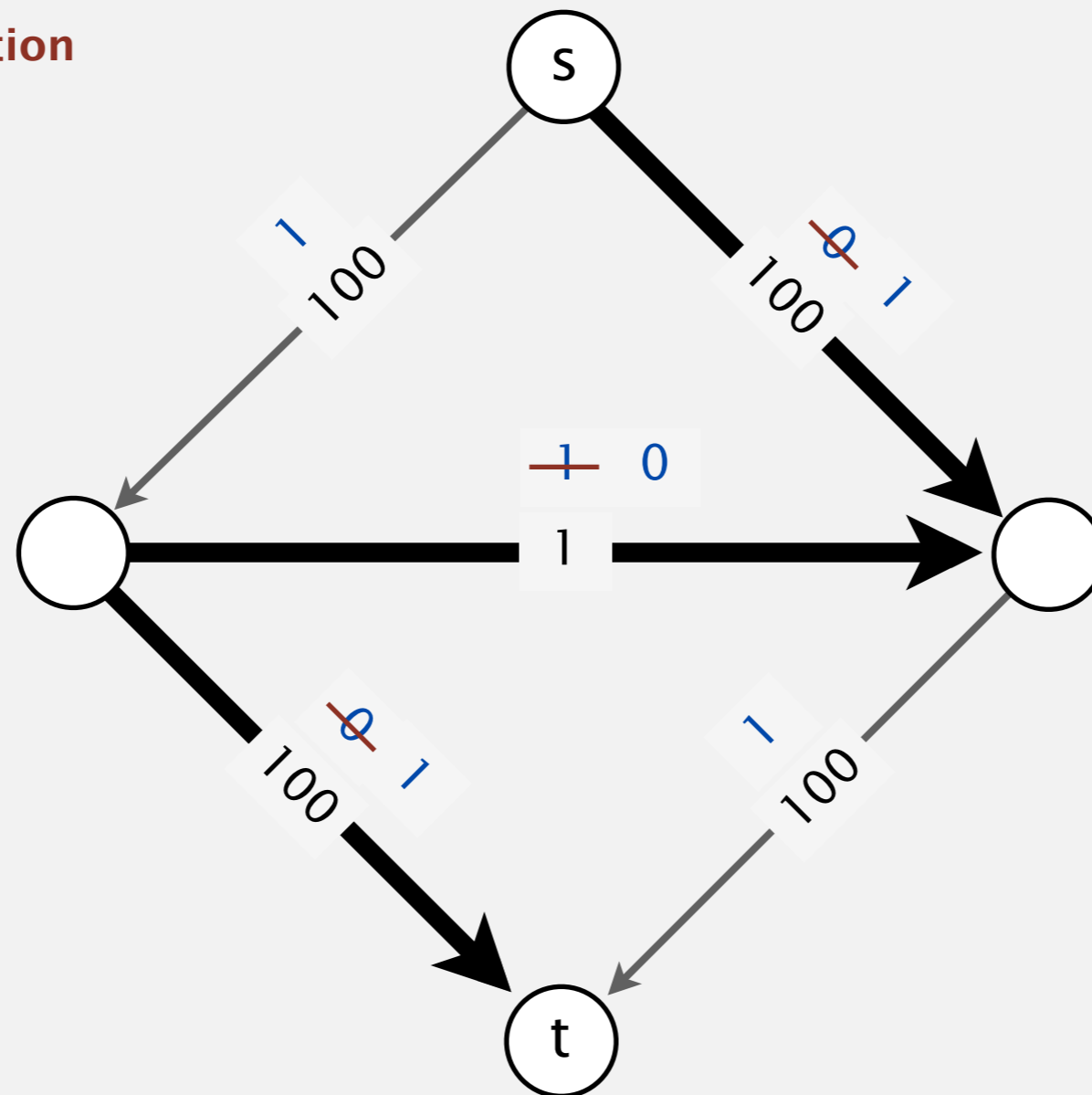
1<sup>st</sup> iteration



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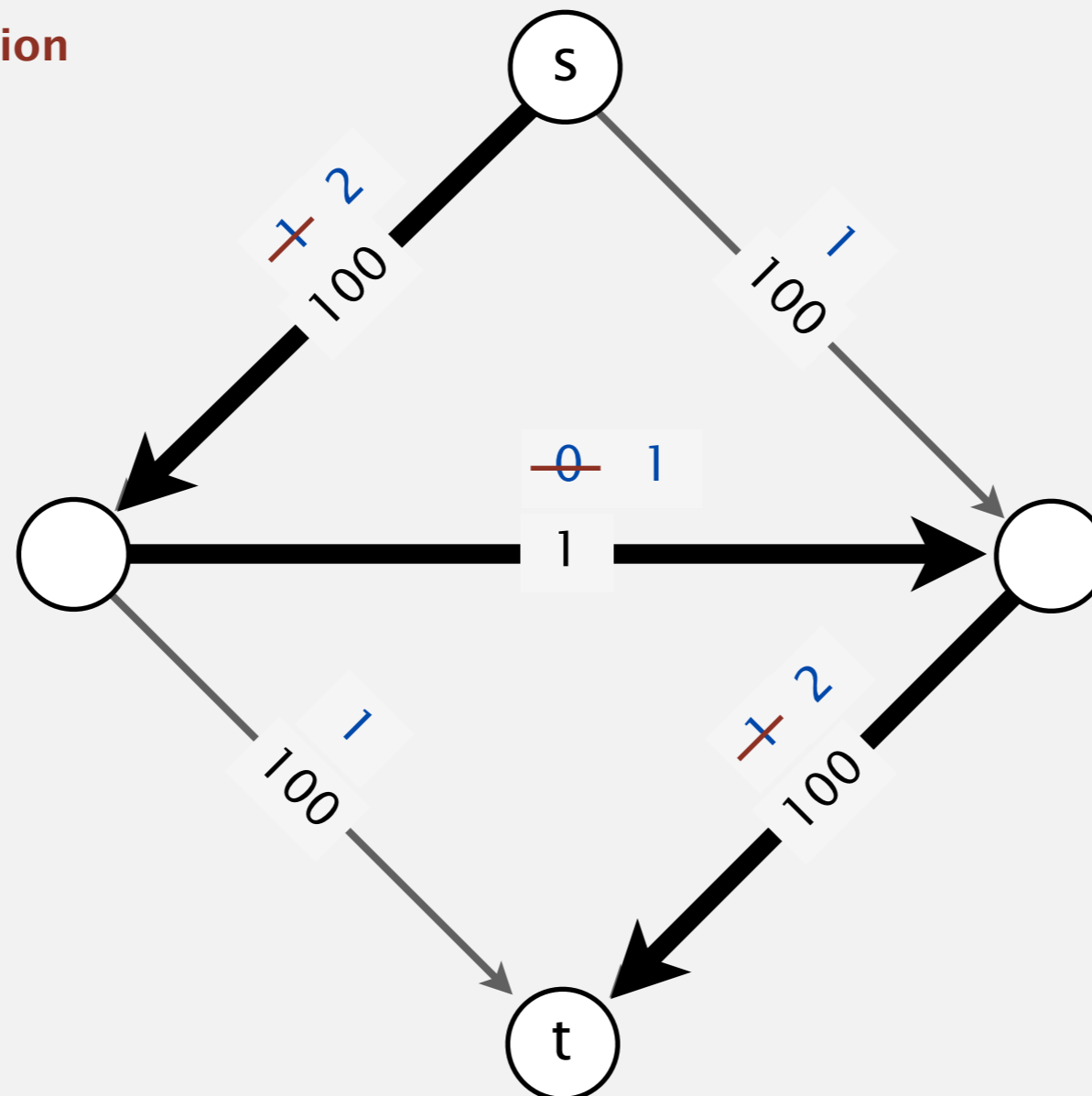
2<sup>nd</sup> iteration



# Bad case for Ford-Fulkerson

**Bad news.** Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

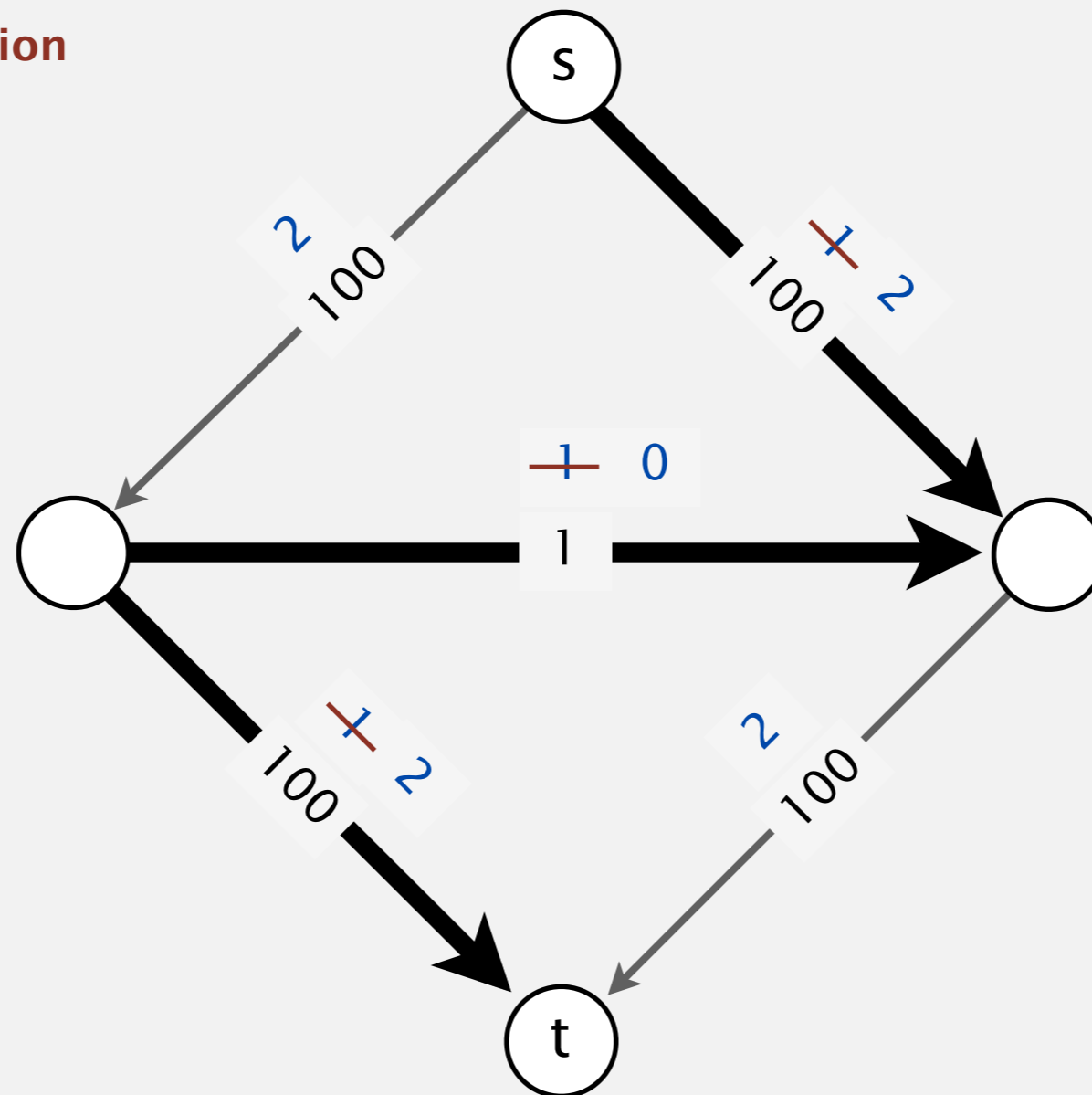
3<sup>rd</sup> iteration



# Bad case for Ford-Fulkerson

**Bad news.** Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

4<sup>th</sup> iteration



## Bad case for Ford-Fulkerson

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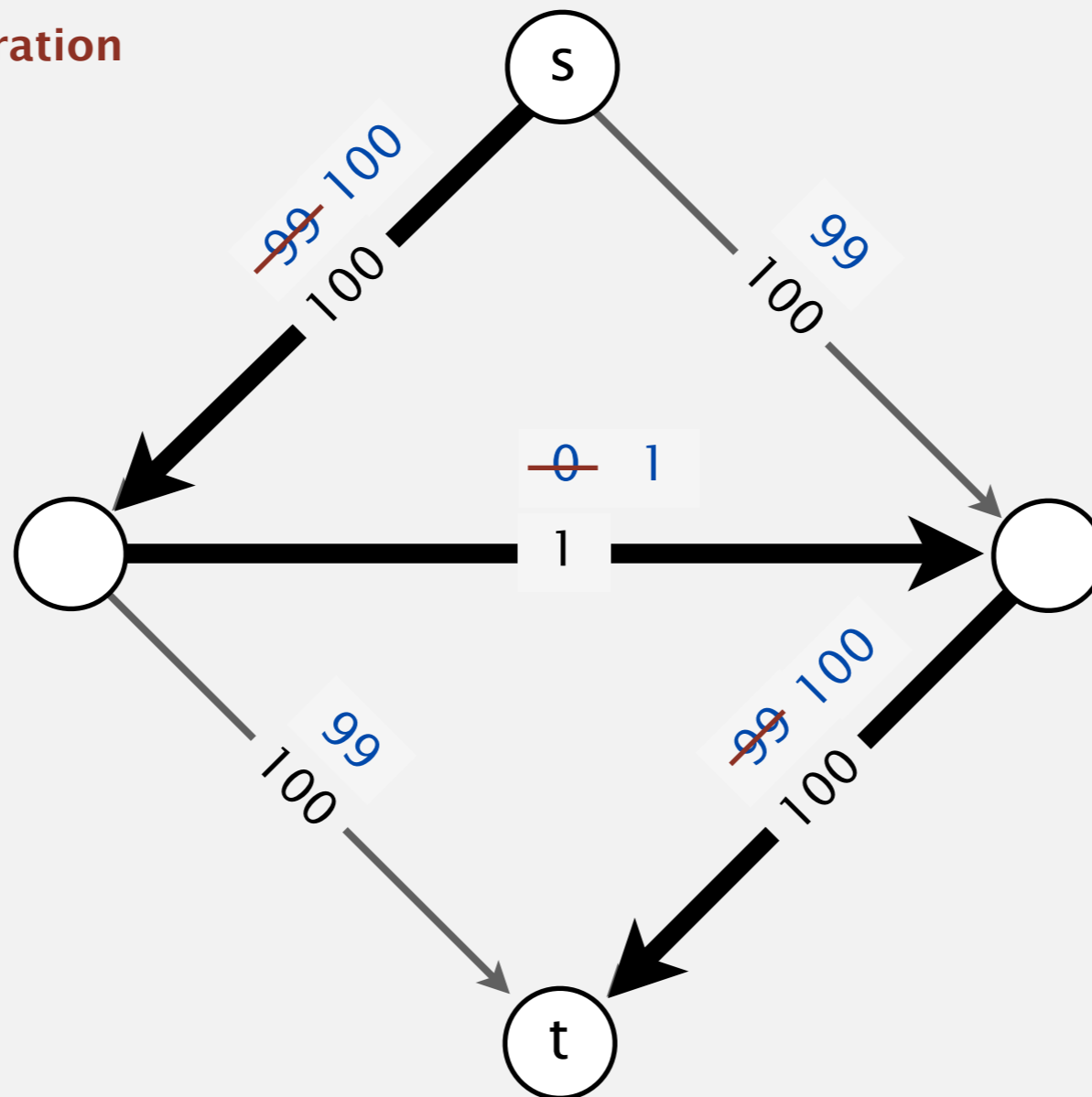
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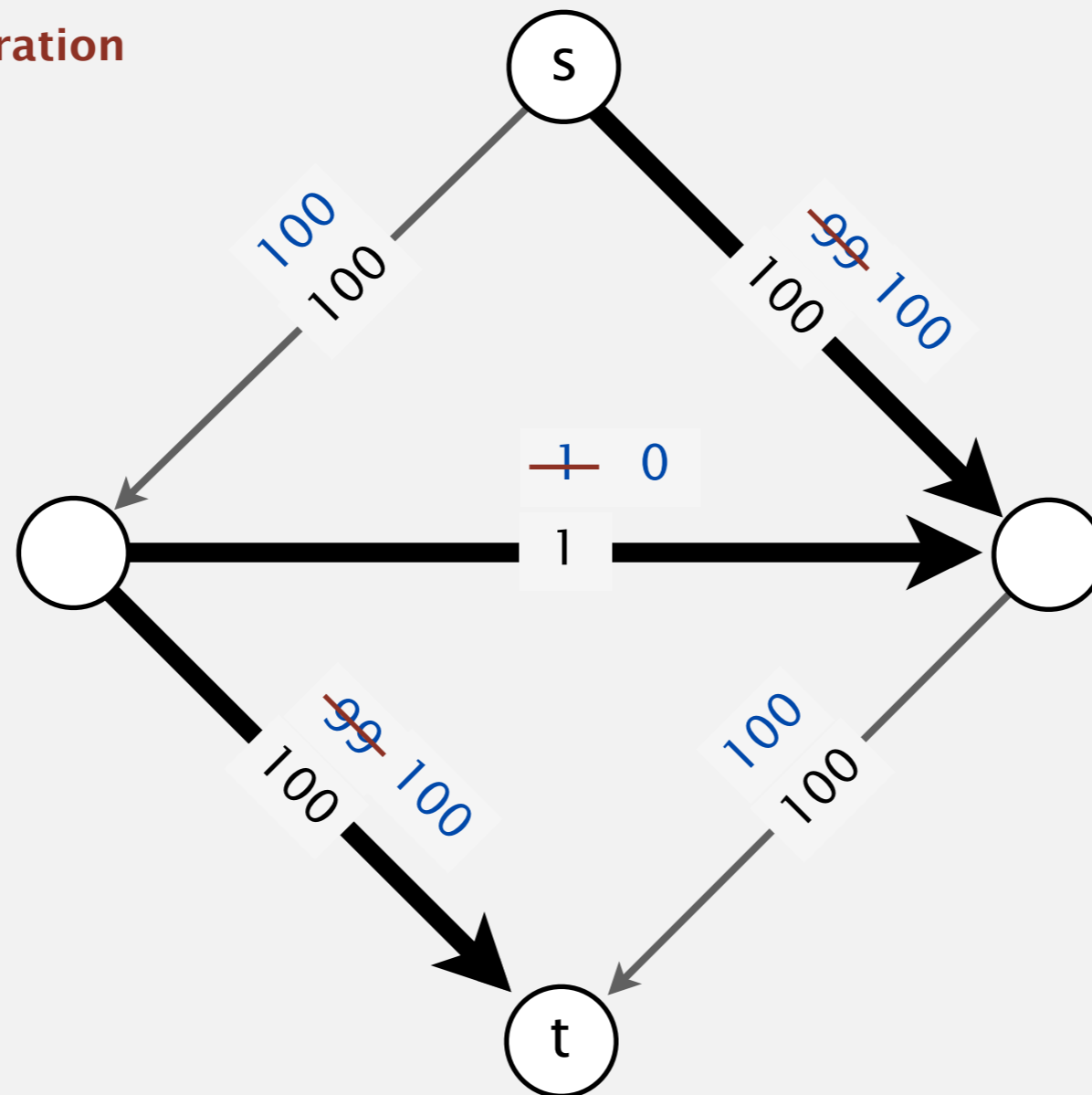
199<sup>th</sup> iteration



## Bad case for Ford-Fulkerson

**Bad news.** Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

200<sup>th</sup> iteration



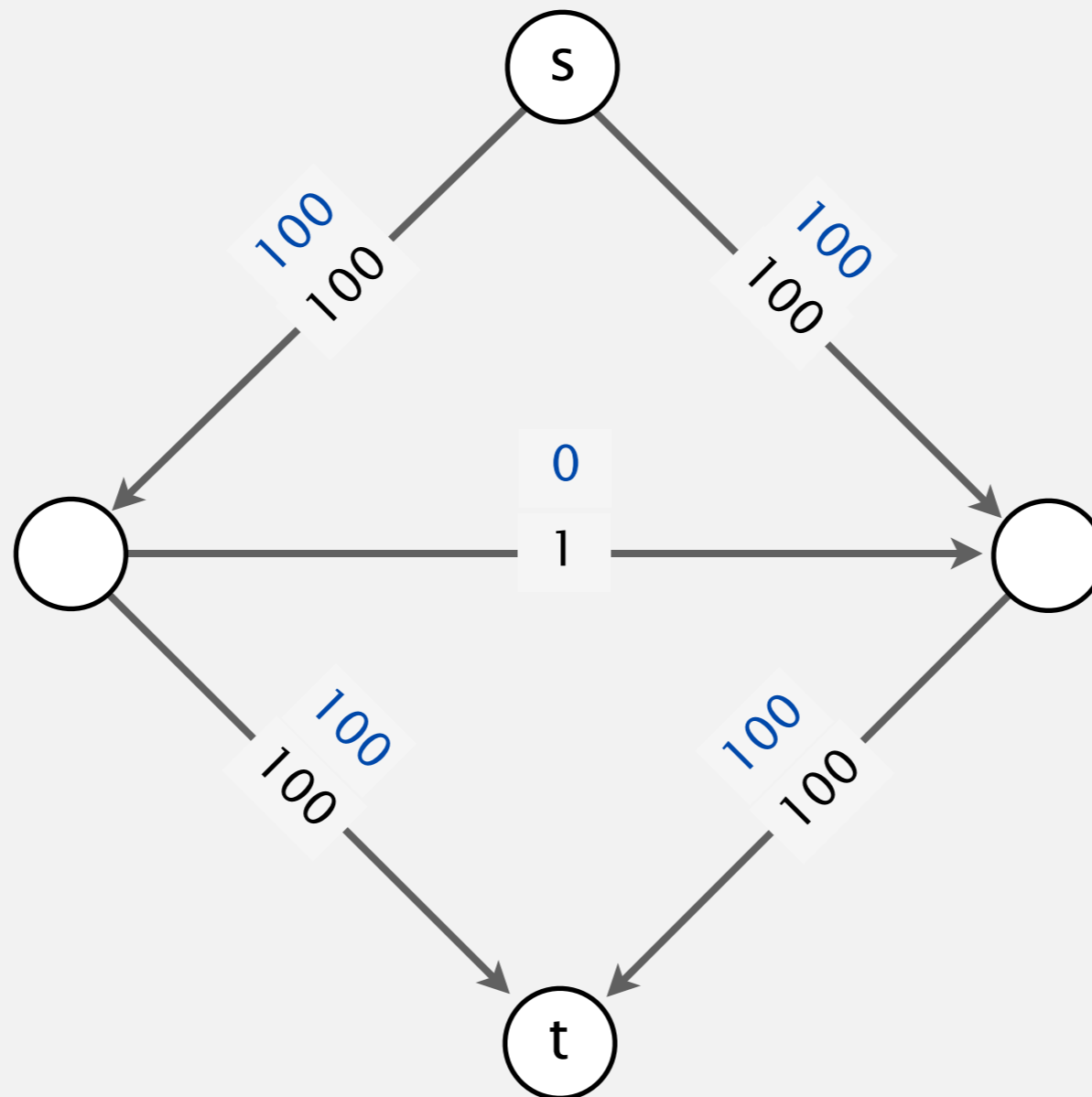
# Bad case for Ford-Fulkerson

---

**Bad news.** Even when edge capacities are integers, number of augmenting paths could be equal to the value of the maxflow.

↖ can be exponential in input size

**Good news.** This case is easily avoided. [ use shortest/fattest path ]



# How to choose augmenting paths?

---

## Choose augmenting paths with:

- Shortest path: fewest number of edges.
- Fattest path: max bottleneck capacity.

### **Theoretical Improvements in Algorithmic Efficiency for Network Flow Problems**

JACK EDMONDS

*University of Waterloo, Waterloo, Ontario, Canada*

AND

RICHARD M. KARP

*University of California, Berkeley, California*

**ABSTRACT.** This paper presents new algorithms for the maximum flow problem, the Hitchcock transportation problem, and the general minimum-cost flow problem. Upper bounds on the numbers of steps in these algorithms are derived, and are shown to compare favorably with upper bounds on the numbers of steps required by earlier algorithms.

**Edmonds–Karp 1972 (USA)**

Dokl. Akad. Nauk SSSR  
Tom 194 (1970), No. 4

Soviet Math. Dokl.  
Vol. 11 (1970), No. 5

### **ALGORITHM FOR SOLUTION OF A PROBLEM OF MAXIMUM FLOW IN A NETWORK WITH POWER ESTIMATION**

UDC 518.5

E. A. DINIC

Different variants of the formulation of the problem of maximal stationary flow in a network and its many applications are given in [1]. There also is given an algorithm solving the problem in the case where the initial data are integers (or, what is equivalent, commensurable). In the general case this algorithm requires preliminary rounding off of the initial data, i.e. only an approximate solution of the problem is possible. In this connection the rapidity of convergence of the algorithm is inversely proportional to the relative precision.

**Dinic 1970 (Soviet Union)**

# How to choose augmenting paths?

---

Use care when selecting augmenting paths.

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

augmenting path	number of paths	implementation
random path	$\leq E U$	randomized queue
shortest path	$\leq \frac{1}{2} E V$	queue (BFS)
fattest path	$\leq E \ln(E U)$	priority queue

digraph with  $V$  vertices,  $E$  edges, and integer capacities between 1 and  $U$



<http://algs4.cs.princeton.edu>

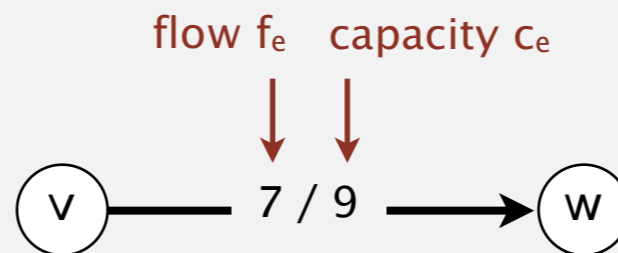
## 6.4 MAXIMUM FLOW

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# Flow network representation

**Flow edge data type.** Associate flow  $f_e$  and capacity  $c_e$  with edge  $e = v \rightarrow w$ .



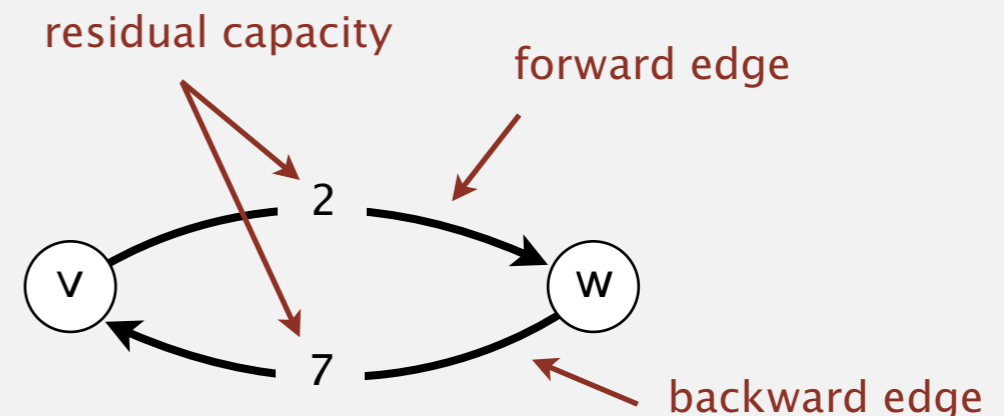
**Flow network data type.** Must be able to process edge  $e = v \rightarrow w$  in either direction: include  $e$  in adjacency lists of both  $v$  and  $w$ .

**Residual (spare) capacity.**

- Forward edge: residual capacity  $= c_e - f_e$ .
- Backward edge: residual capacity  $= f_e$ .

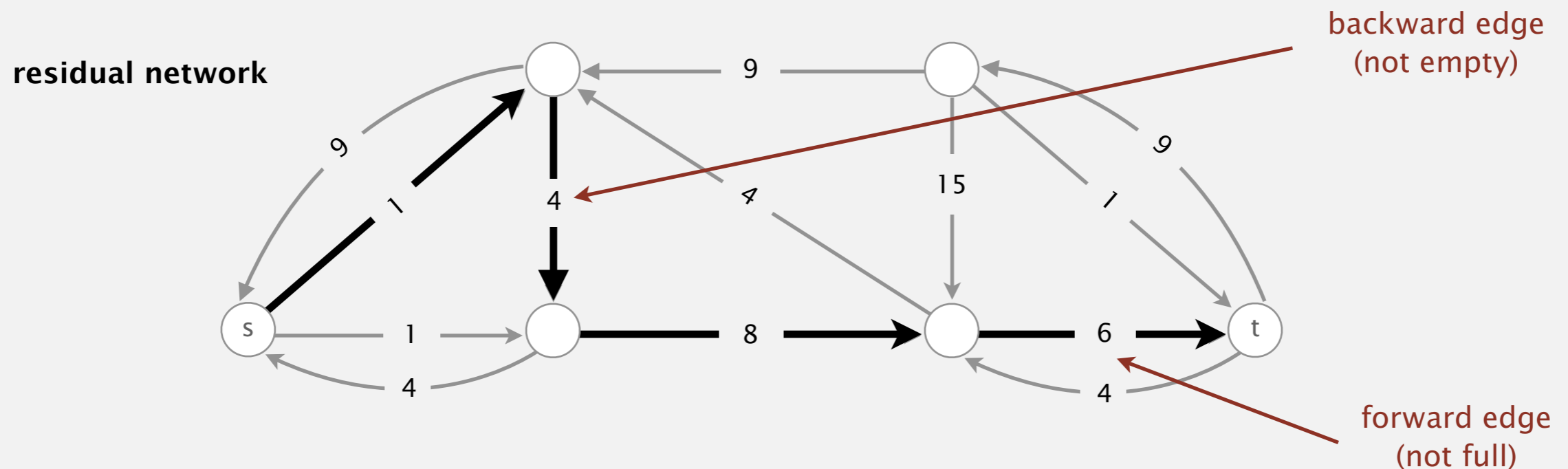
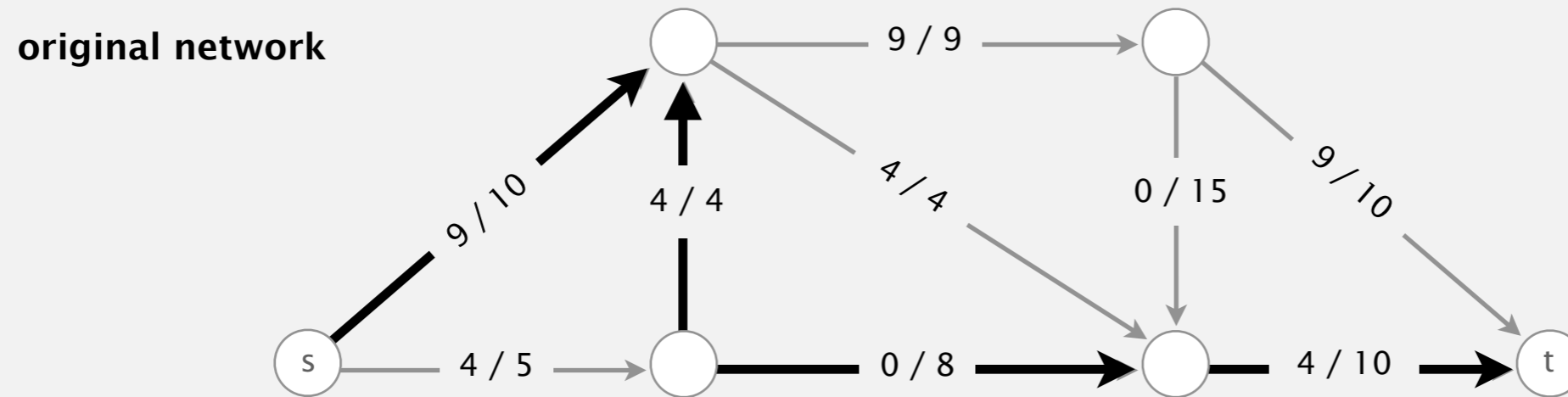
**Augment flow.**

- Forward edge: add  $\Delta$ .
- Backward edge: subtract  $\Delta$ .



# Flow network representation

**Residual network.** A useful view of a flow network. ← includes all edges with positive residual capacity

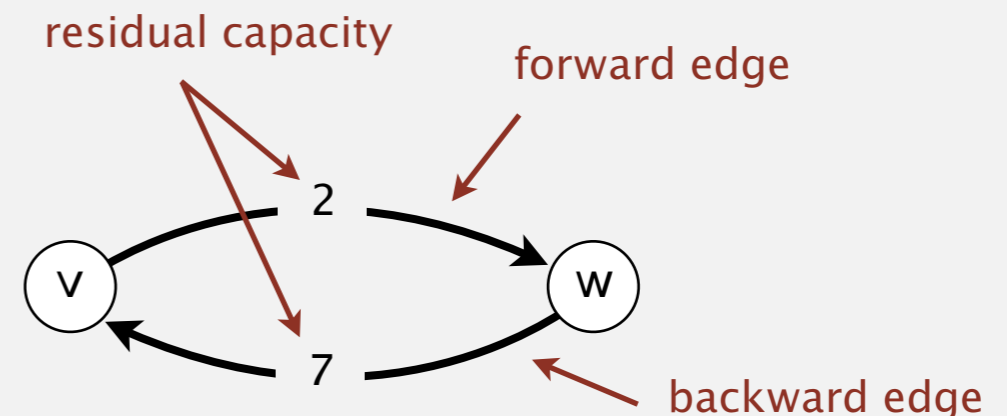
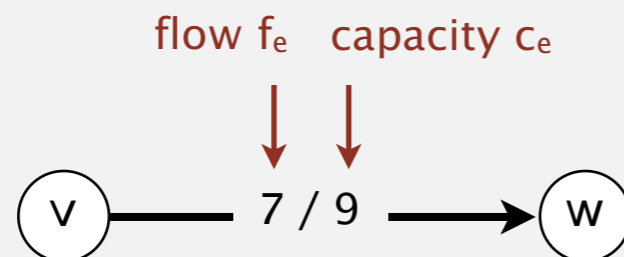


**Key point.** Augmenting paths in original network are in 1-1 correspondence with directed paths in residual network.

# Flow edge API

```
public class FlowEdge
```

<code>FlowEdge(int v, int w, double capacity)</code>	<i>create a flow edge <math>v \rightarrow w</math></i>
<code>int from()</code>	<i>vertex this edge points from</i>
<code>int to()</code>	<i>vertex this edge points to</i>
<code>int other(int v)</code>	<i>other endpoint</i>
<code>double capacity()</code>	<i>capacity of this edge</i>
<code>double flow()</code>	<i>flow in this edge</i>
<code>double residualCapacityTo(int v)</code>	<i>residual capacity toward v</i>
<code>void addResidualFlowTo(int v, double delta)</code>	<i>add delta flow toward v</i>



# Flow edge: Java implementation

```
public class FlowEdge
{
    private final int v, w;           // from and to
    private final double capacity;    // capacity
    private double flow;              // flow

    public FlowEdge(int v, int w, double capacity)
    {
        this.v      = v;
        this.w      = w;
        this.capacity = capacity;
    }

    public int from()      { return v;      }
    public int to()        { return w;      }
    public double capacity() { return capacity; }
    public double flow()   { return flow;   }

    public int other(int vertex)
    {
        if (vertex == v) return w;
        else if (vertex == w) return v;
        else throw new IllegalArgumentException();
    }

    public double residualCapacityTo(int vertex) {...}
    public void addResidualFlowTo(int vertex, double delta) {...}
}
```

← flow variable (mutable)

← next slide

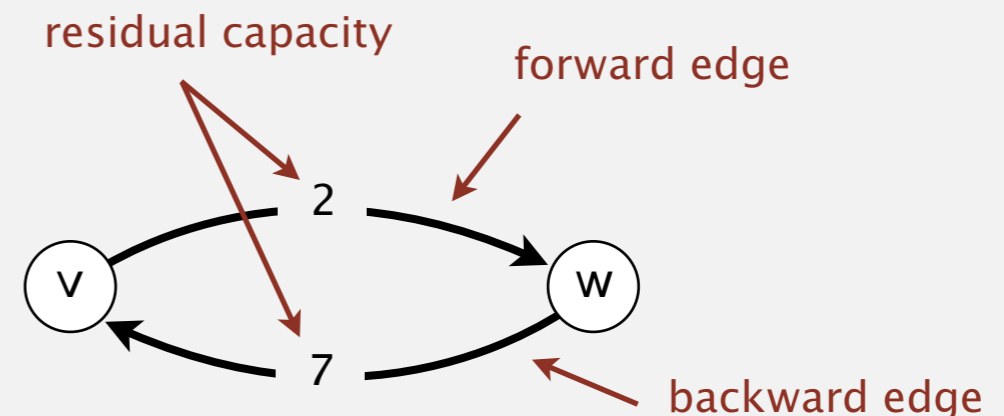
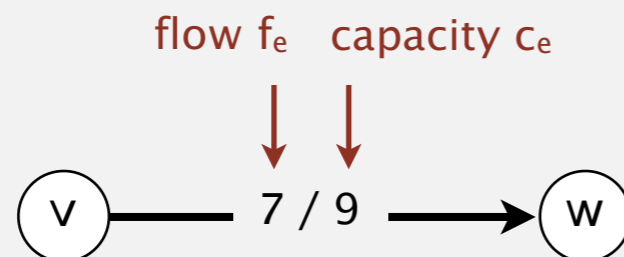
# Flow edge: Java implementation (continued)

```
public double residualCapacityTo(int vertex)
{
    if (vertex == v) return flow;
    else if (vertex == w) return capacity - flow;
    else throw new IllegalArgumentException();
}
```

← forward edge  
← backward edge

```
public void addResidualFlowTo(int vertex, double delta)
{
    if (vertex == v) flow -= delta;
    else if (vertex == w) flow += delta;
    else throw new IllegalArgumentException();
}
```

← forward edge  
← backward edge



# Flow network API

---

```
public class FlowNetwork
```

```
    FlowNetwork(int V)
```

*create an empty flow network with  $V$  vertices*

```
    FlowNetwork(In in)
```

*construct flow network input stream*

```
    void addEdge(FlowEdge e)
```

*add flow edge  $e$  to this flow network*

```
    Iterable<FlowEdge> adj(int v)
```

*forward and backward edges incident to  $v$*

```
    Iterable<FlowEdge> edges()
```

*all edges in this flow network*

```
    int V()
```

*number of vertices*

```
    int E()
```

*number of edges*

```
    String toString()
```

*string representation*

**Conventions.** Allow self-loops and parallel edges.

# Flow network: Java implementation

---

```
public class FlowNetwork
{
    private final int V;
    private Bag<FlowEdge>[] adj;
```

← same as EdgeWeightedGraph,  
but adjacency lists of  
FlowEdges instead of Edges

```
    public FlowNetwork(int V)
    {
        this.V = V;
        adj = (Bag<FlowEdge>[]) new Bag[V];
        for (int v = 0; v < V; v++)
            adj[v] = new Bag<FlowEdge>();
    }
```

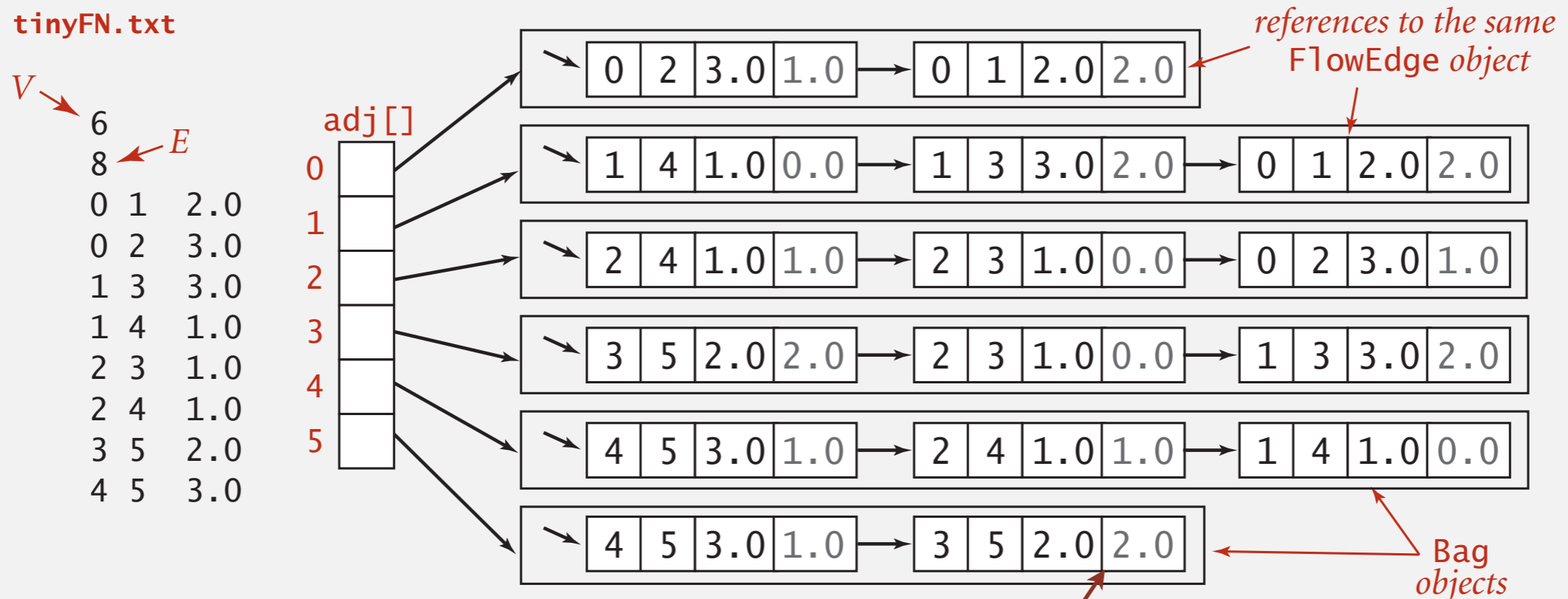
```
    public void addEdge(FlowEdge e)
    {
        int v = e.from();
        int w = e.to();
        adj[v].add(e);
        adj[w].add(e);
    }
```

← add forward edge  
← add backward edge

```
    public Iterable<FlowEdge> adj(int v)
    { return adj[v]; }
}
```

# Flow network: adjacency-lists representation

Maintain vertex-indexed array of FlowEdge lists (use Bag abstraction).



**Note.** Adjacency list includes edges with 0 residual capacity.  
(residual network is represented implicitly)

# Finding a shortest augmenting path (cf. breadth-first search)

```
private boolean hasAugmentingPath(FlowNetwork G, int s, int t)
{
    edgeTo = new FlowEdge[G.V()];
    marked = new boolean[G.V()];

    Queue<Integer> queue = new Queue<Integer>();
    queue.enqueue(s);
    marked[s] = true;
    while (!queue.isEmpty())
    {
        int v = queue.dequeue();

        for (FlowEdge e : G.adj(v))
        {
            int w = e.other(v);
            if (!marked[w] && (e.residualCapacityTo(w) > 0) )
            {
                edgeTo[w] = e;
                marked[w] = true;
                queue.enqueue(w);
            }
        }
    }

    return marked[t];
}
```

found path from s to w  
in the residual network?

save last edge on path to w;  
mark w;  
add w to the queue

is t reachable from s in residual network?

# Ford-Fulkerson: Java implementation

```
public class FordFulkerson
{
    private boolean[] marked;    // true if s->v path in residual network
    private FlowEdge[] edgeTo;  // last edge on s->v path
    private double value;       // value of flow

    public FordFulkerson(FlowNetwork G, int s, int t)
    {
        value = 0.0;
        while (hasAugmentingPath(G, s, t))
        {
            double bottle = Double.POSITIVE_INFINITY;
            for (int v = t; v != s; v = edgeTo[v].other(v))
                bottle = Math.min(bottle, edgeTo[v].residualCapacityTo(v));

            for (int v = t; v != s; v = edgeTo[v].other(v))
                edgeTo[v].addResidualFlowTo(v, bottle);

            value += bottle;
        }
    }

    private boolean hasAugmentingPath(FlowNetwork G, int s, int t)
    { /* See previous slide. */ }

    public double value()
    { return value; }

    public boolean inCut(int v)
    { return marked[v]; }
}
```

compute edgeTo[] and marked[]

compute bottleneck capacity

augment flow

is v reachable from s in residual network?



<http://algs4.cs.princeton.edu>

## 6.4 MAXIMUM FLOW

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- *introduction*
- *Ford-Fulkerson algorithm*
- *maxflow-mincut theorem*
- *analysis of running time*
- *Java implementation*
- ***applications***

# Bipartite matching problem

---

N students apply for N jobs.



Each gets several offers.



Is there a way to match all students to jobs?



## bipartite matching problem

1	Alice	6	Adobe
	Adobe		Alice
	Amazon		Bob
	Google		Carol
2	Bob	7	Amazon
	Adobe		Alice
	Amazon		Bob
3	Carol		Dave
	Adobe		Eliza
	Facebook	8	Facebook
	Google		Carol
4	Dave	9	Google
	Amazon		Alice
	Yahoo		Carol
5	Eliza	10	Yahoo
	Amazon		Dave
	Yahoo		Eliza

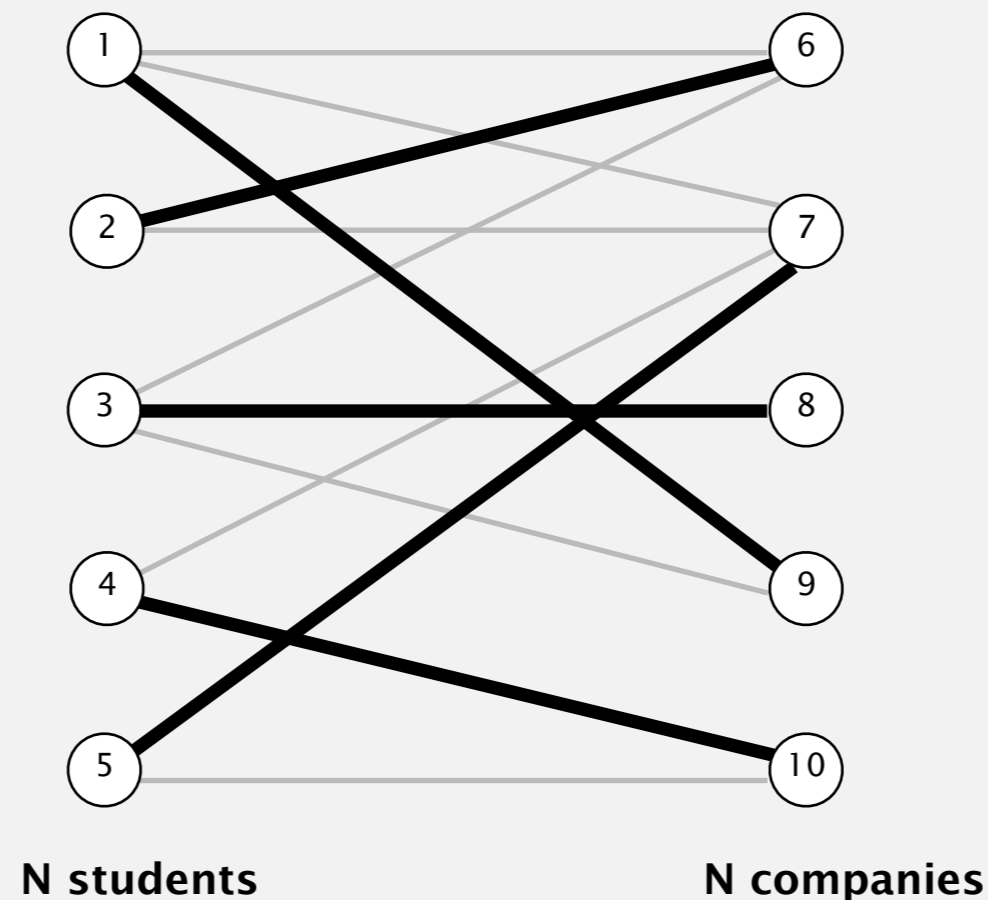
# Bipartite matching problem

Given a bipartite graph, find a perfect matching.

## perfect matching (solution)

Alice — Google  
Bob — Adobe  
Carol — Facebook  
Dave — Yahoo  
Eliza — Amazon

## bipartite graph

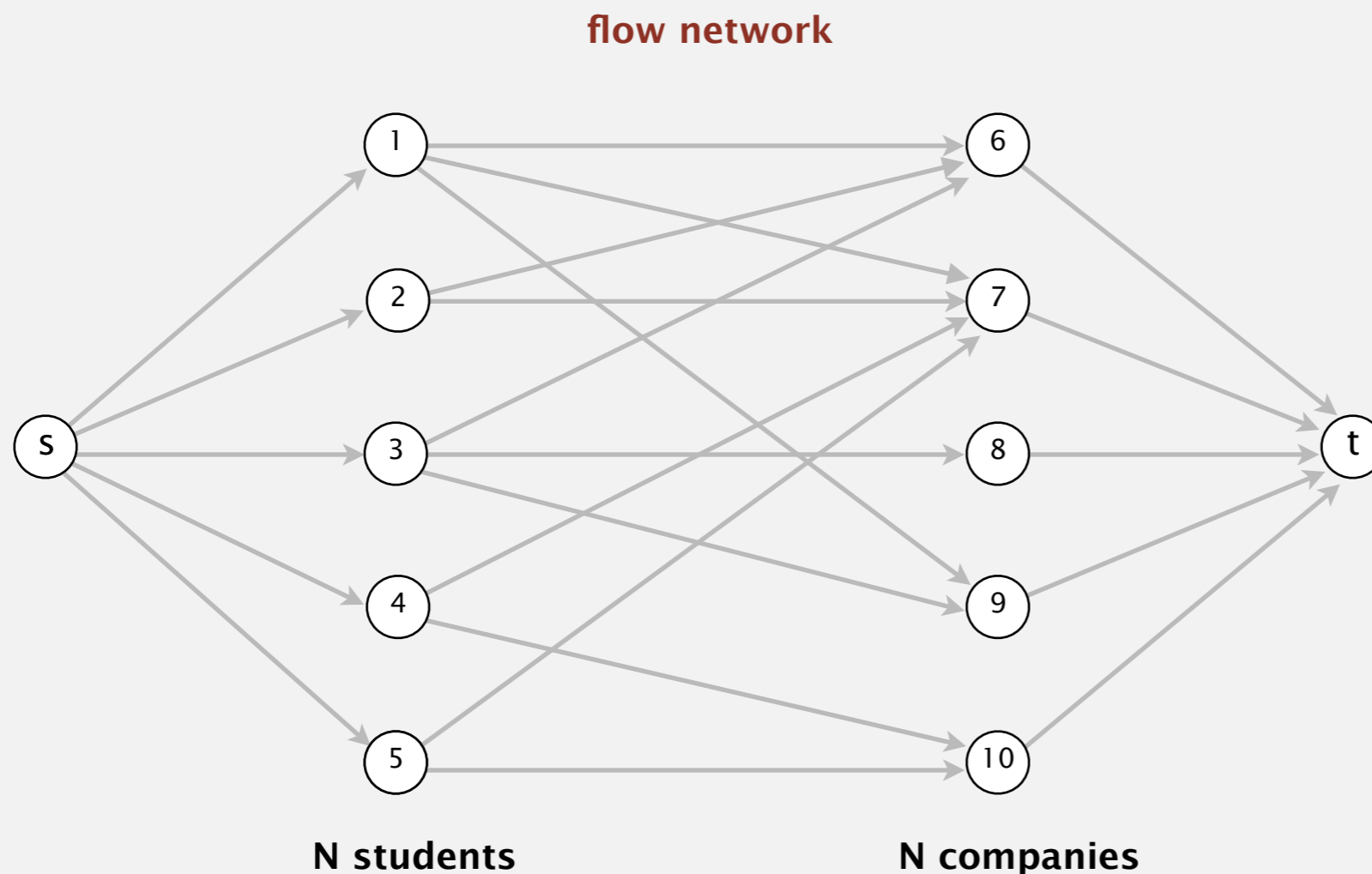


## bipartite matching problem

1	Alice	6	Adobe
	Adobe		Alice
	Amazon		Bob
	Google		Carol
2	Bob	7	Amazon
	Adobe		Alice
	Amazon		Bob
3	Carol		Dave
	Adobe		Eliza
	Facebook	8	Facebook
	Google		Carol
4	Dave	9	Google
	Amazon		Alice
	Yahoo		Carol
5	Eliza	10	Yahoo
	Amazon		Dave
	Yahoo		Eliza

# Network flow formulation of bipartite matching

- Create  $s$ ,  $t$ , one vertex for each student, and one vertex for each job.
- Add edge from  $s$  to each student (capacity 1).
- Add edge from each job to  $t$  (capacity 1).
- Add edge from student to each job offered (infinite capacity).

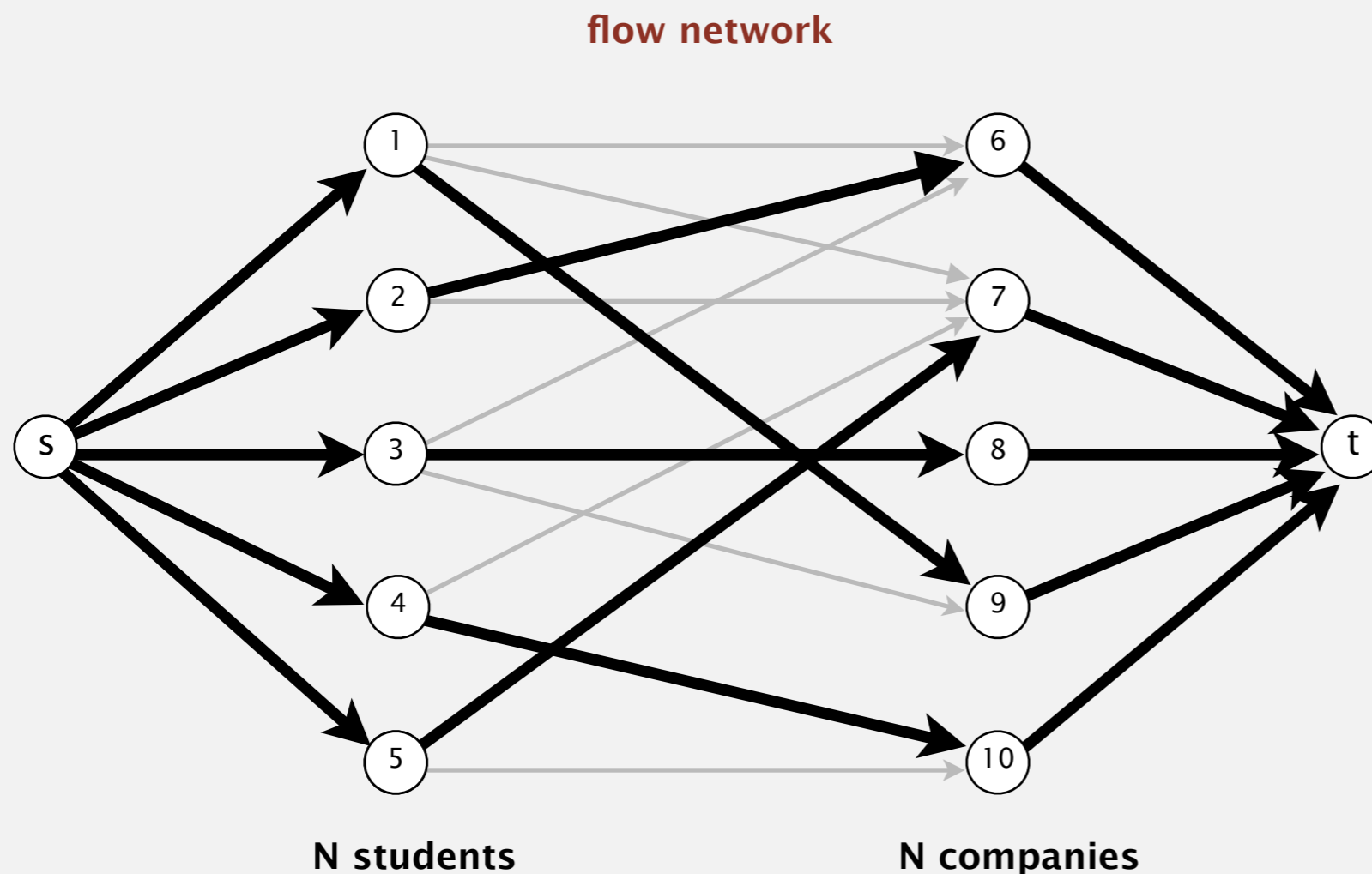


**bipartite matching problem**

1	Alice	6	Adobe
	Adobe		Alice
	Amazon		Bob
	Google		Carol
2	Bob	7	Amazon
	Adobe		Alice
	Amazon		Bob
3	Carol		Dave
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	Facebook	8	Facebook
	Google		Carol
4	Dave	9	Google
	Amazon		Alice
	Yahoo		Carol
5	Eliza	10	Yahoo
	Amazon		Dave
	Yahoo		Eliza

# Network flow formulation of bipartite matching

1-1 correspondence between perfect matchings in bipartite graph and **integer-valued** maxflows of value  $N$ .



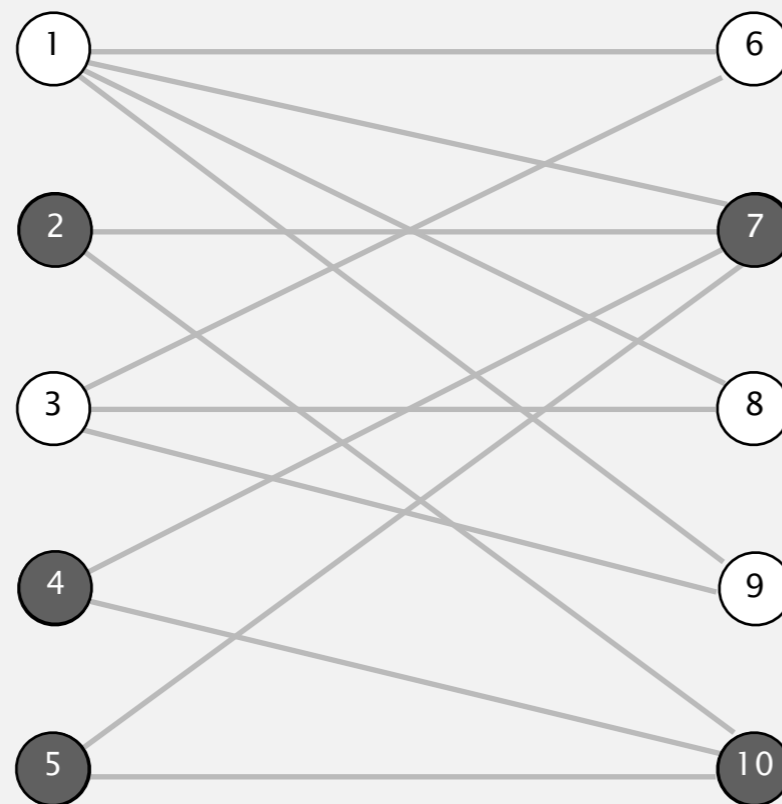
## bipartite matching problem

1	Alice	6	Adobe
	Adobe		Alice
	Amazon		Bob
	Google		Carol
2	Bob	7	Amazon
	Adobe		Alice
	Amazon		Bob
3	Carol		Dave
	Adobe		Eliza
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	Google		Carol
4	Dave	9	Google
	Amazon		Alice
	Yahoo		Carol
5	Eliza	10	Yahoo
	Amazon		Dave
	Yahoo		Eliza

# What the mincut tells us

---

**Goal.** When no perfect matching, explain why.



**no perfect matching exists**

$S = \{ 2, 4, 5 \}$

$T = \{ 7, 10 \}$

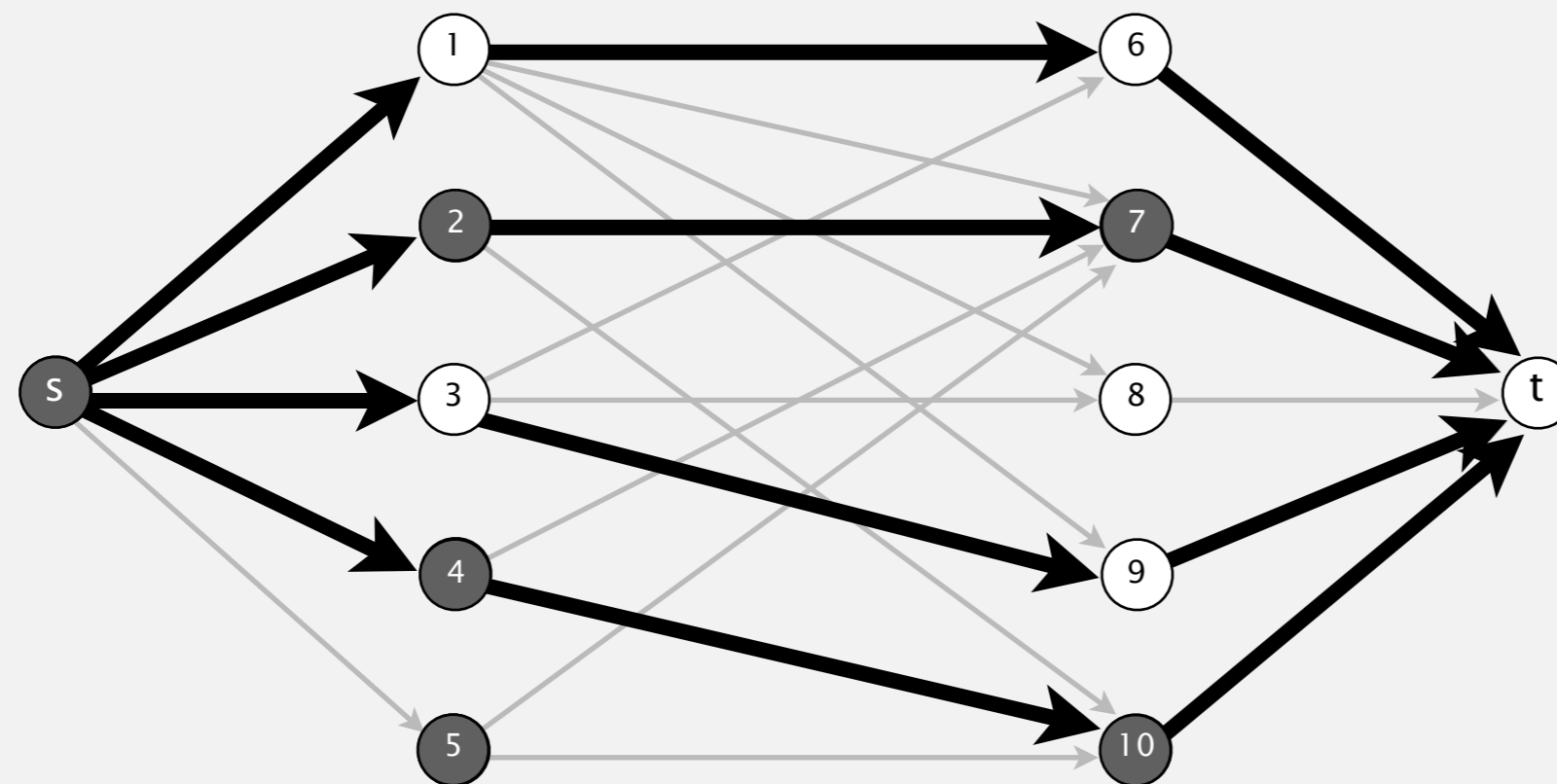
student in  $S$   
can be matched  
only to  
companies in  $T$

$|S| > |T|$

# What the mincut tells us

**Mincut.** Consider mincut  $(A, B)$ .

- Let  $S$  = students on  $s$  side of cut.
- Let  $T$  = companies on  $s$  side of cut.
- Fact:  $|S| > |T|$ ; students in  $S$  can be matched only to companies in  $T$ .



$S = \{ 2, 4, 5 \}$

$T = \{ 7, 10 \}$

student in  $S$   
can be matched  
only to  
companies in  $T$

$|S| > |T|$

no perfect matching exists

**Bottom line.** When no perfect matching, mincut explains why.

# Summary

---

**Mincut problem.** Find an  $st$ -cut of minimum capacity.

**Maxflow problem.** Find an  $st$ -flow of maximum value.

**Duality.** Value of the maxflow = capacity of mincut.

**Proven successful approaches.**

- Ford-Fulkerson (various augmenting-path strategies).
- Preflow-push (various versions).

**Open research challenges.**

- Practice: solve real-world maxflow/mincut problems in linear time.
- Theory: prove it for worst-case inputs.
- Still much to be learned!