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## 6.4 MAXIMUM FLOW

---

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



# Algorithms

ROBERT SEDGEWICK | KEVIN WAYNE

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

## 6.4 MAXIMUM FLOW

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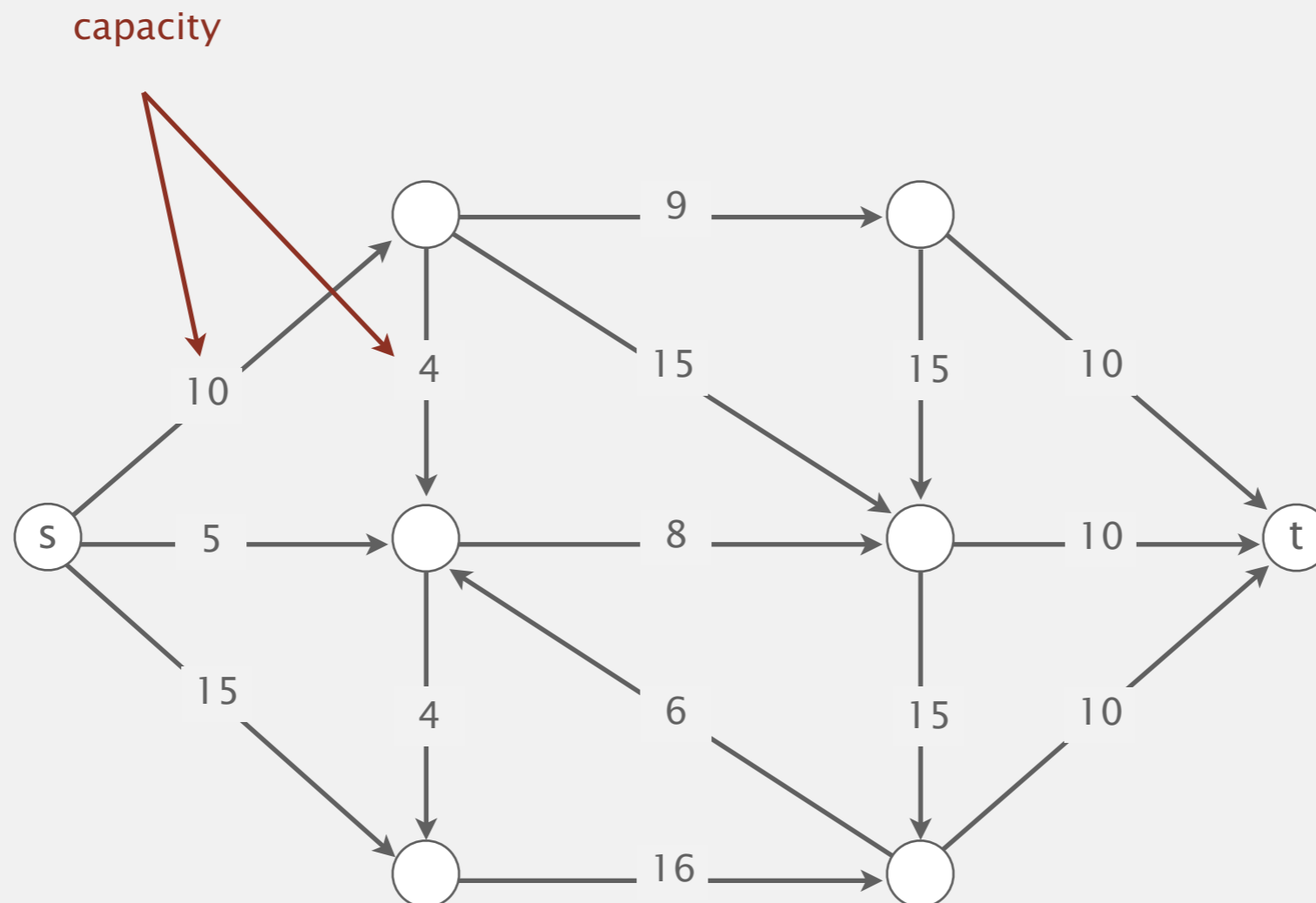
- ▶ *introduction*
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# Mincut problem

---

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

each edge has a  
positive capacity

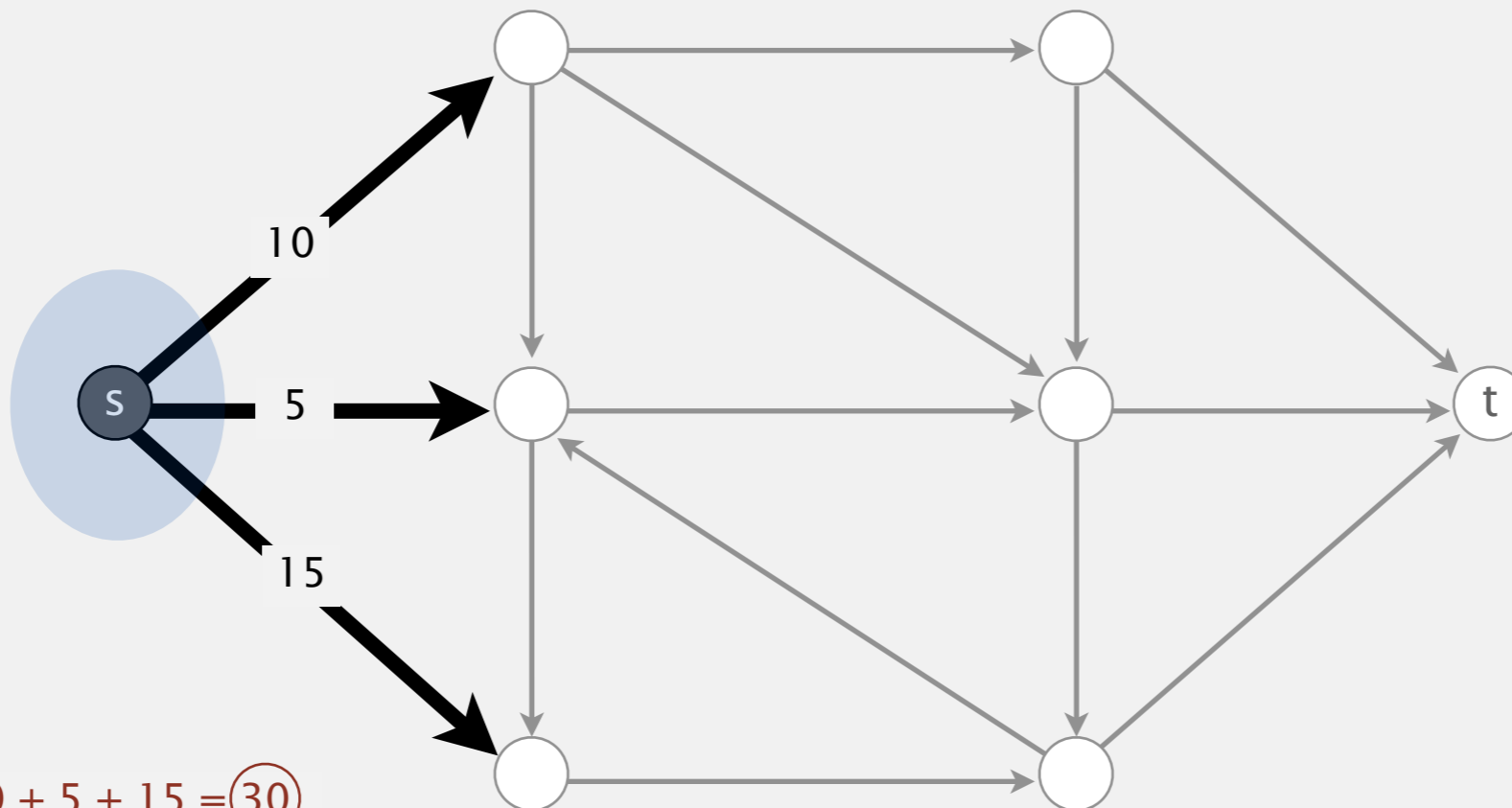


# Mincut problem

---

**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  $B$ .



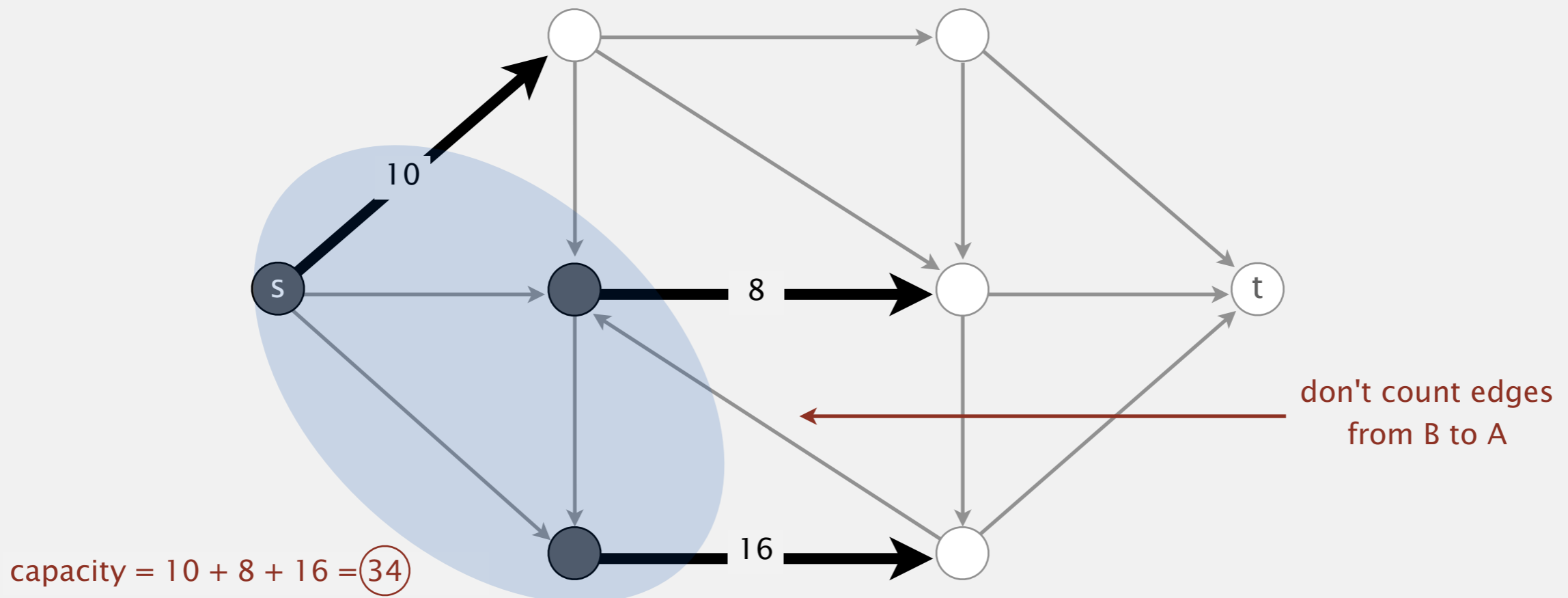
capacity =  $10 + 5 + 15 = 30$

# Mincut problem

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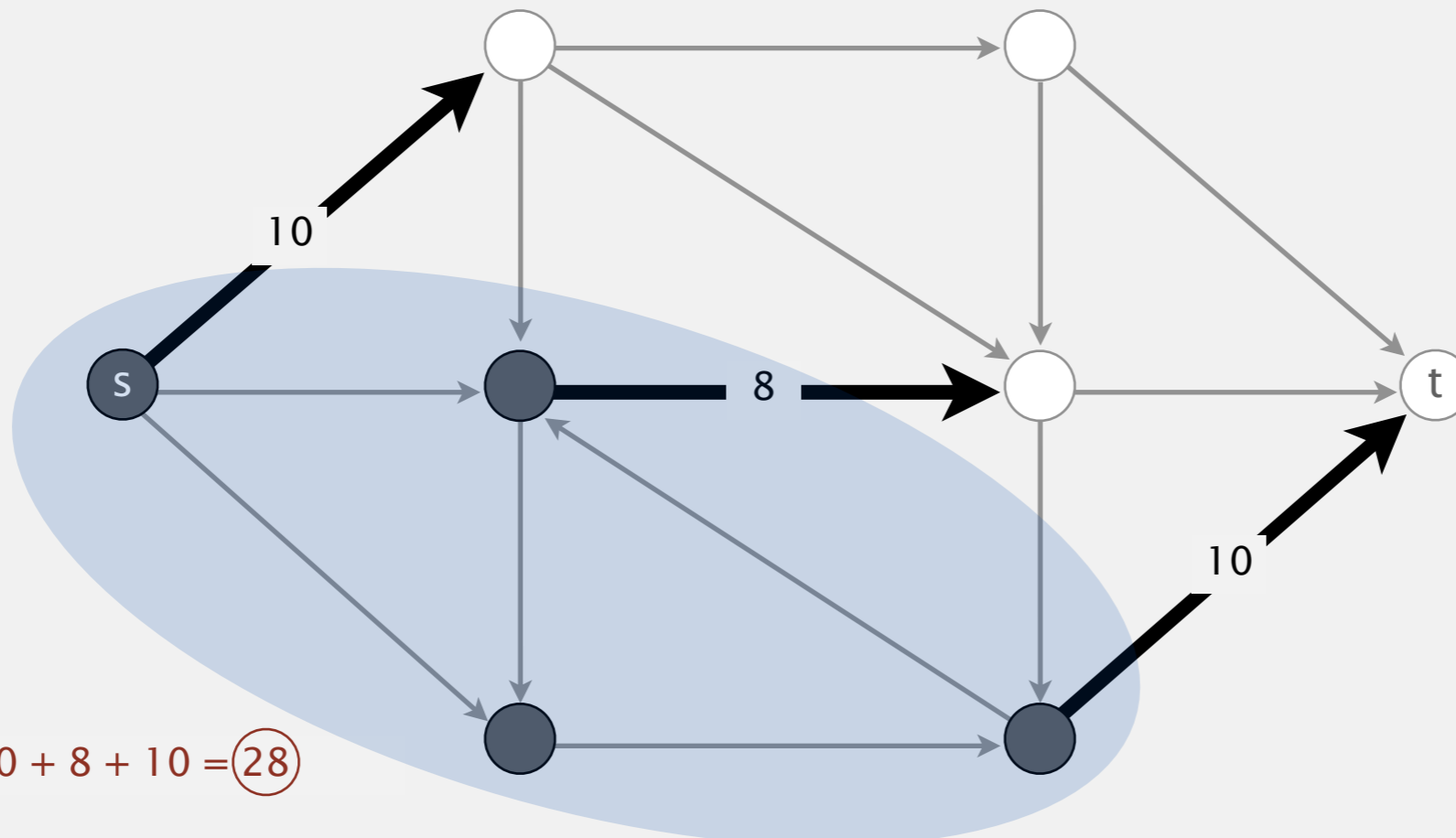
# Mincut problem

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

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



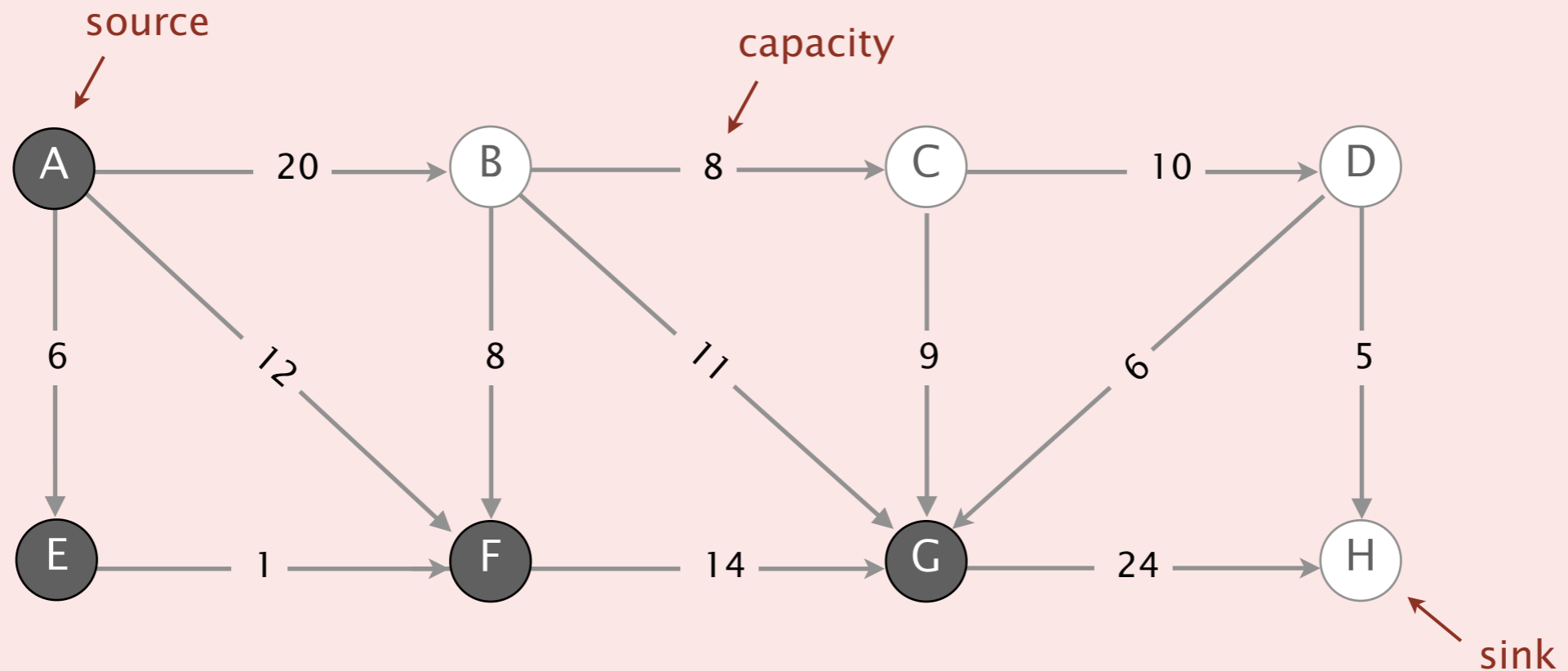
capacity =  $10 + 8 + 10 = 28$

# Maxflow: quiz 1

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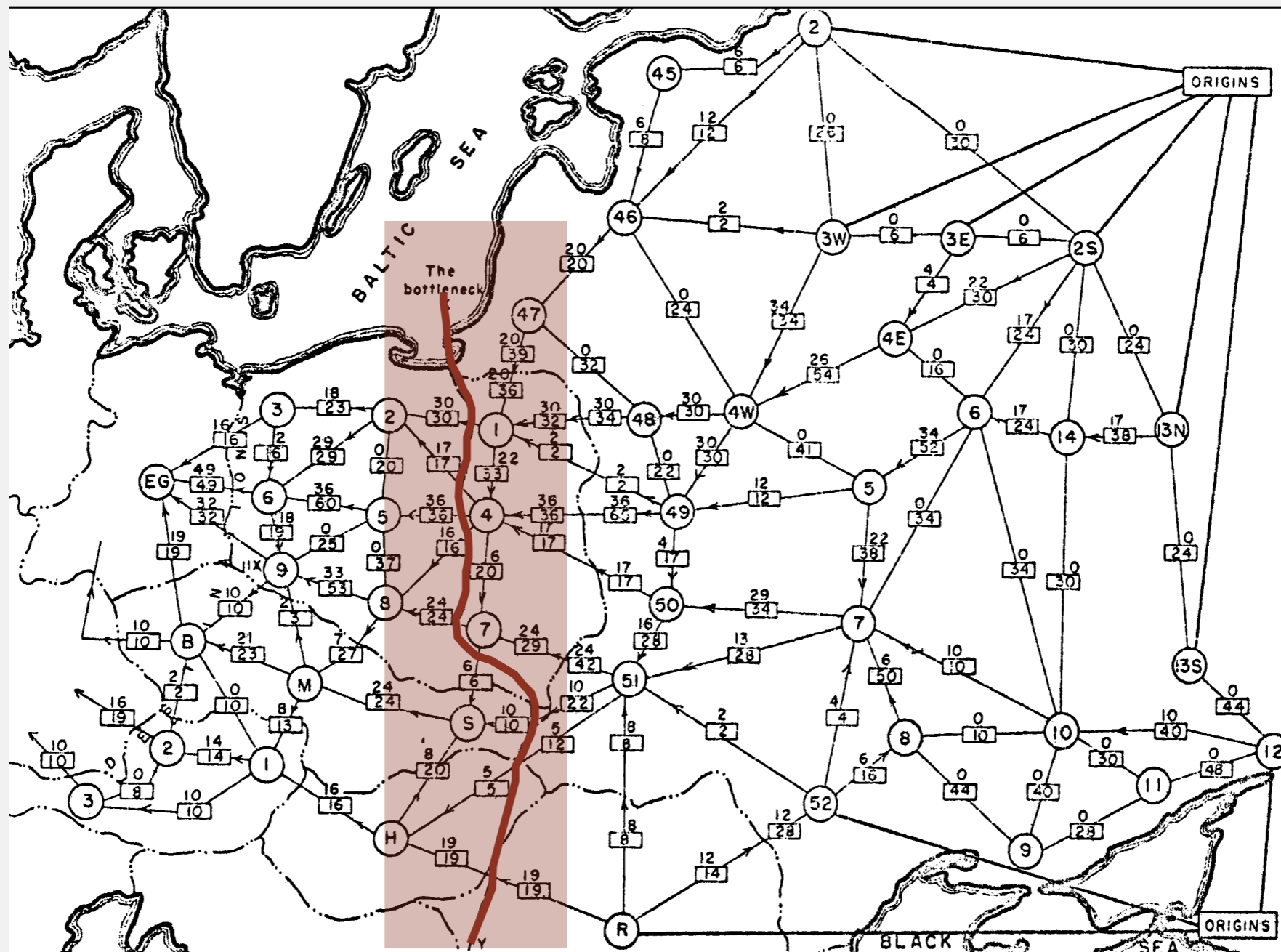
What is the capacity of the st-cut  $\{A, E, F, G\}$ ?

- A. 34 (8 + 11 + 9 + 6)
- B. 44 (20 + 24)
- C. 78 (20 + 8 + 11 + 9 + 6 + 24)
- D. *I don't know.*



# Mincut application (RAND 1950s)

"Free world" goal. Cut supplies (if cold war turns into real war).



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



# Potential minicut application (2010s)

---

Government-in-power's goal. Cut off communication to set of people.



**Though maximum flow algorithms have a long history, revolutionary progress is still being made.**

BY ANDREW V. GOLDBERG AND ROBERT E. TARJAN

# Efficient Maximum Flow Algorithms

gorithms in more detail. We restrict ourselves to basic maximum flow algorithms and do not cover interesting special cases (such as undirected graphs, planar graphs, and bipartite matchings) or generalizations (such as minimum-cost and multi-commodity flow problems).

Before formally defining the maximum flow and the minimum cut problems, we give a simple example of each problem: For the maximum flow example, suppose we have a graph that represents an oil pipeline network from an oil well to an oil depot. Each arc has a capacity, or maximum number of liters per second that can flow through the corresponding pipe. The goal is to find the maximum number of liters per second (maximum flow) that can be shipped from well to depot. For the minimum cut problem, we want to find the set of pipes of the smallest total capacity such that removing the pipes disconnects the oil well from the oil depot (minimum cut).

The maximum flow, minimum cut

Efficient Maximum Flow Algorithms by Andrew Goldberg and Bob Tarjan

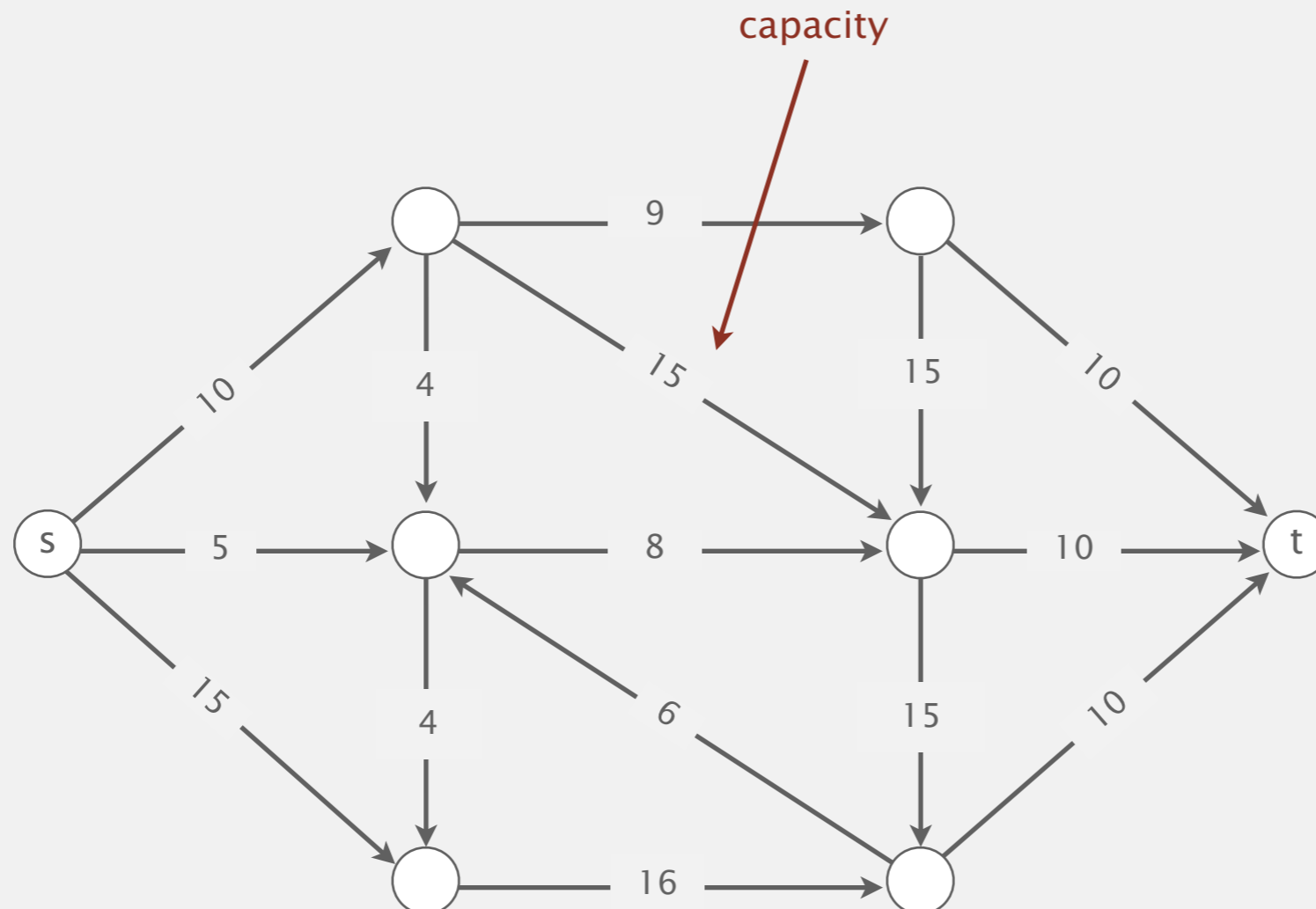
<http://vimeo.com/100774435>

# Maxflow problem

---

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

each edge has a  
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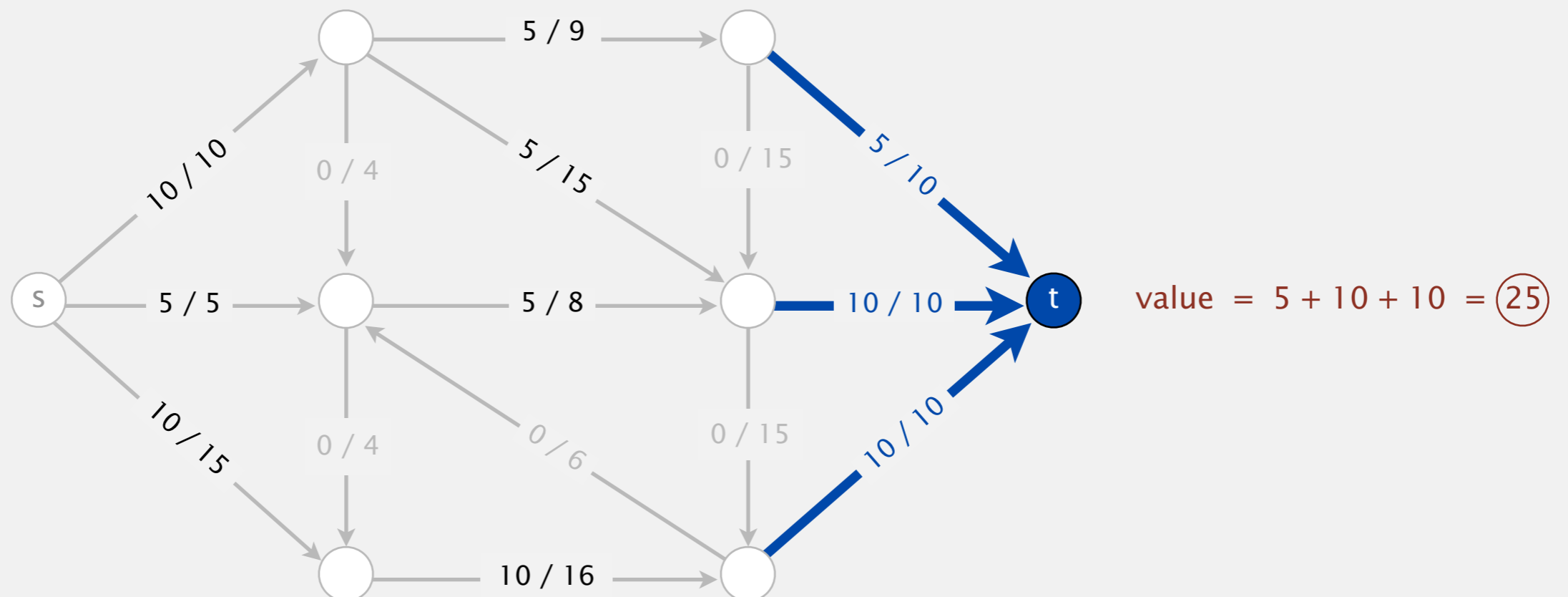
# Maxflow problem

**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  $t$ ).

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

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



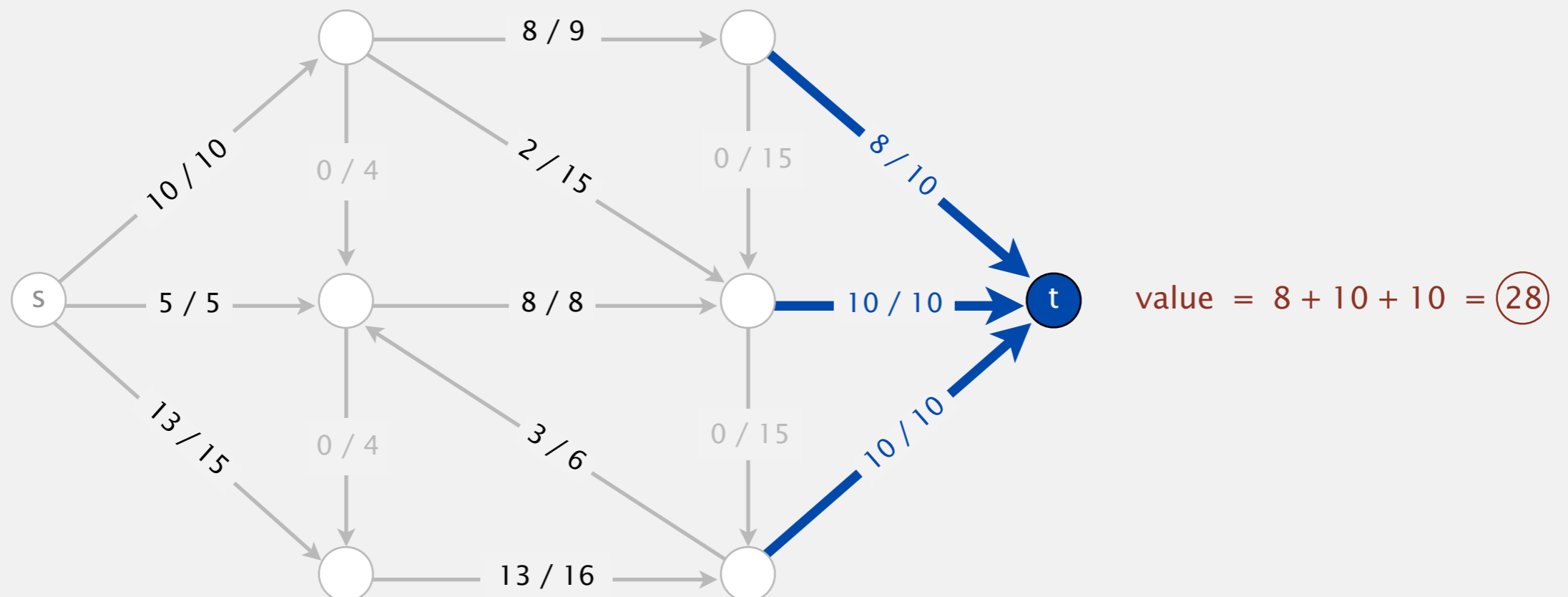
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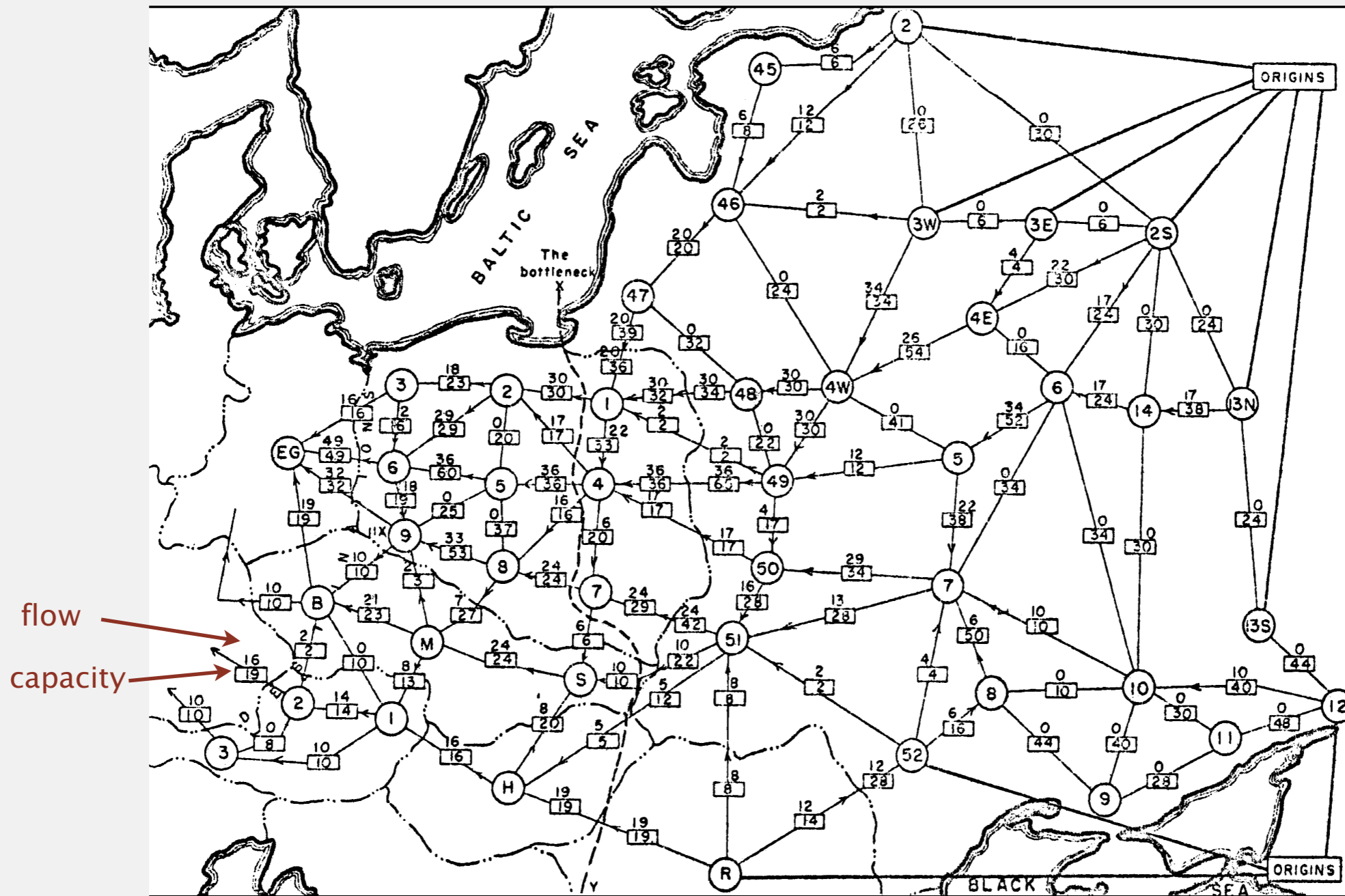
**Def.** The *value* of a flow is the inflow at  $t$ .

**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)

---

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



facebook graph



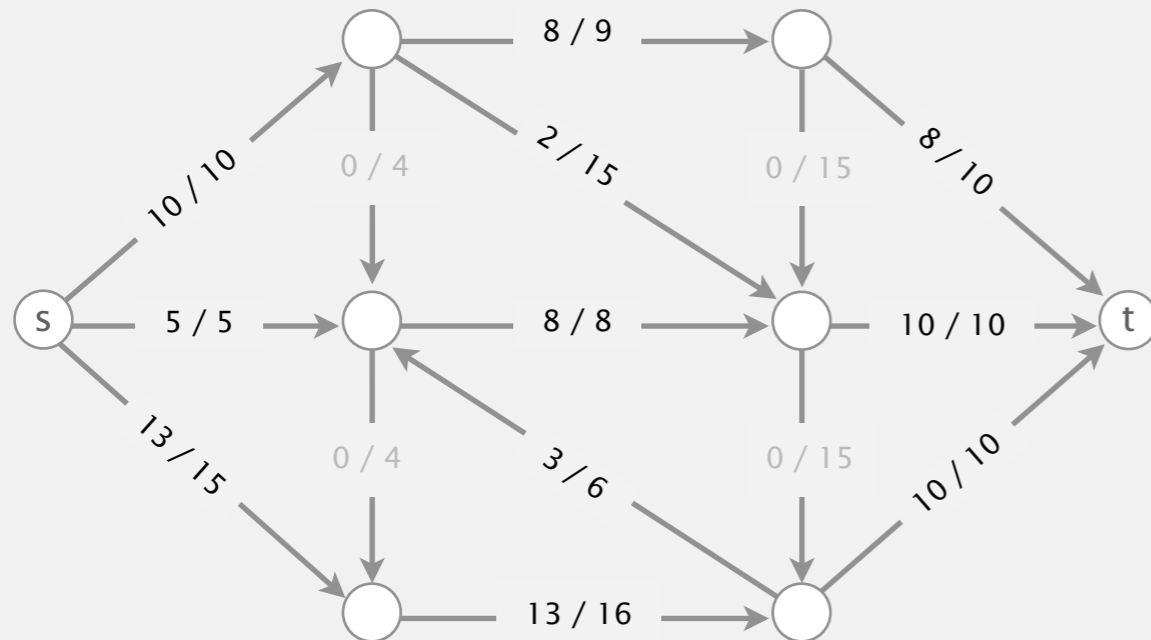
# Summary

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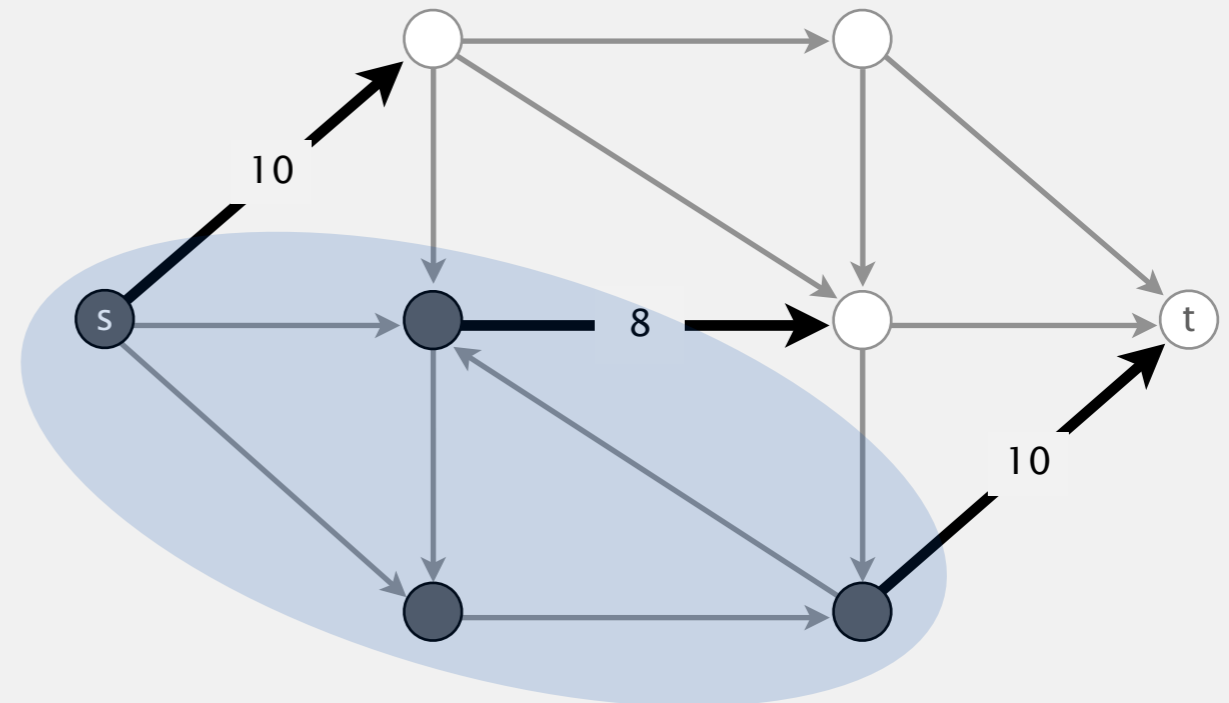
**Input.** A weighted digraph, source vertex  $s$ , and target vertex  $t$ .

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

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



value of flow = 28



capacity of cut = 28

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



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## 6.4 MAXIMUM FLOW

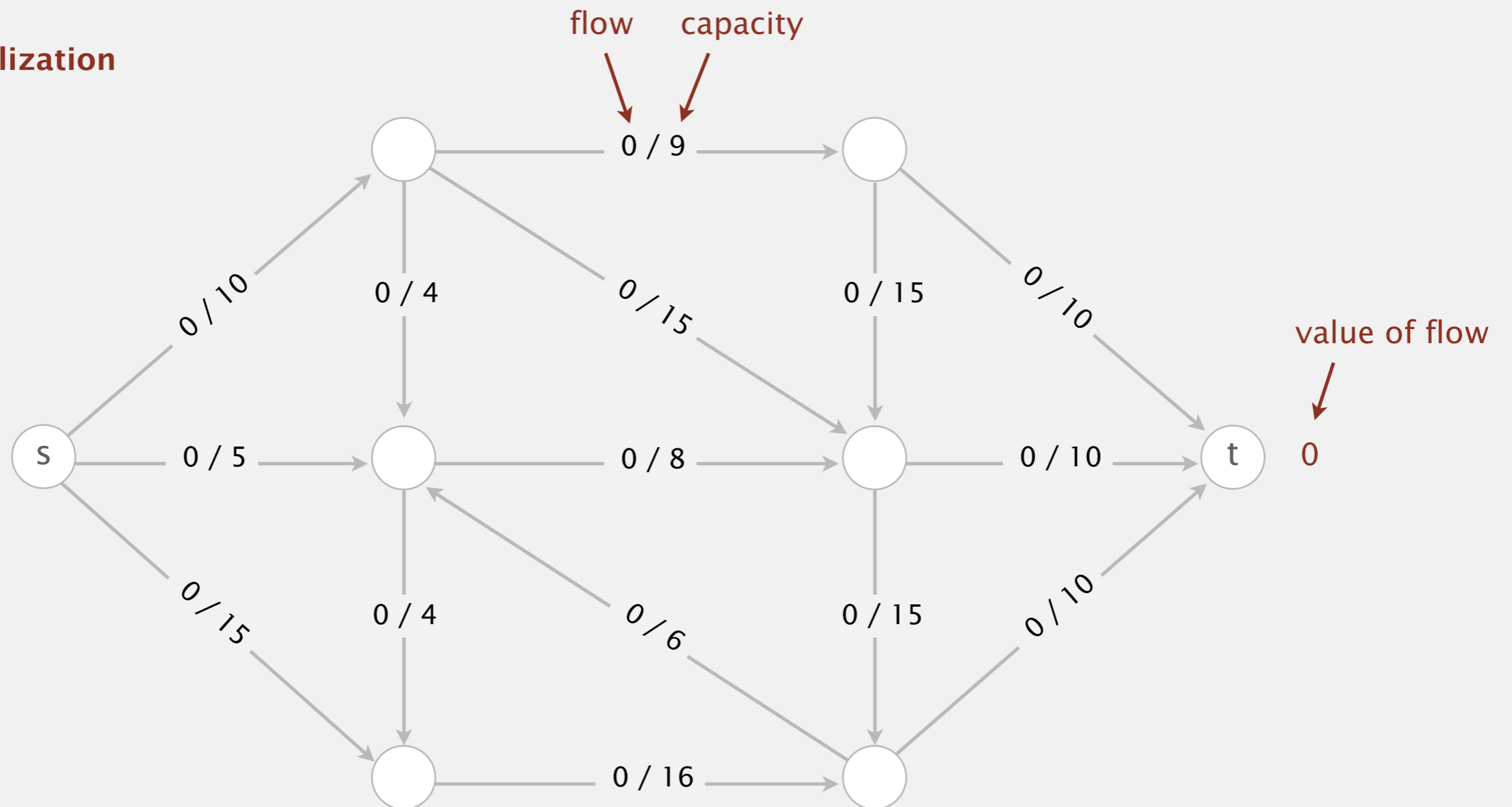
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- ▶ *introduction*
- ▶ ***Ford-Fulkerson algorithm***
- ▶ *maxflow-minicut theorem*
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- ▶ ***applications***

# Ford-Fulkerson algorithm

Initialization. Start with 0 flow.

initialization

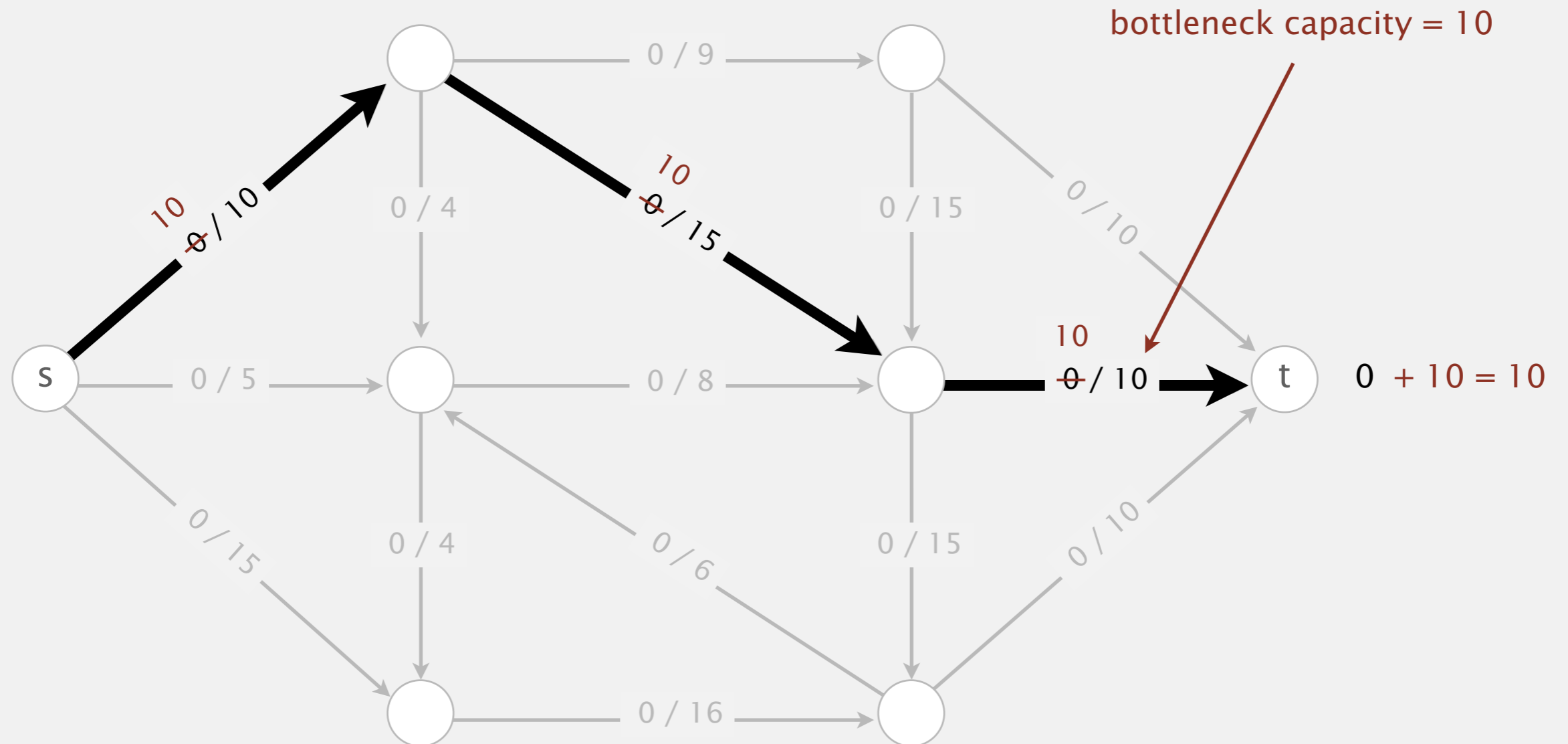


# Idea: increase flow along augmenting paths

**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

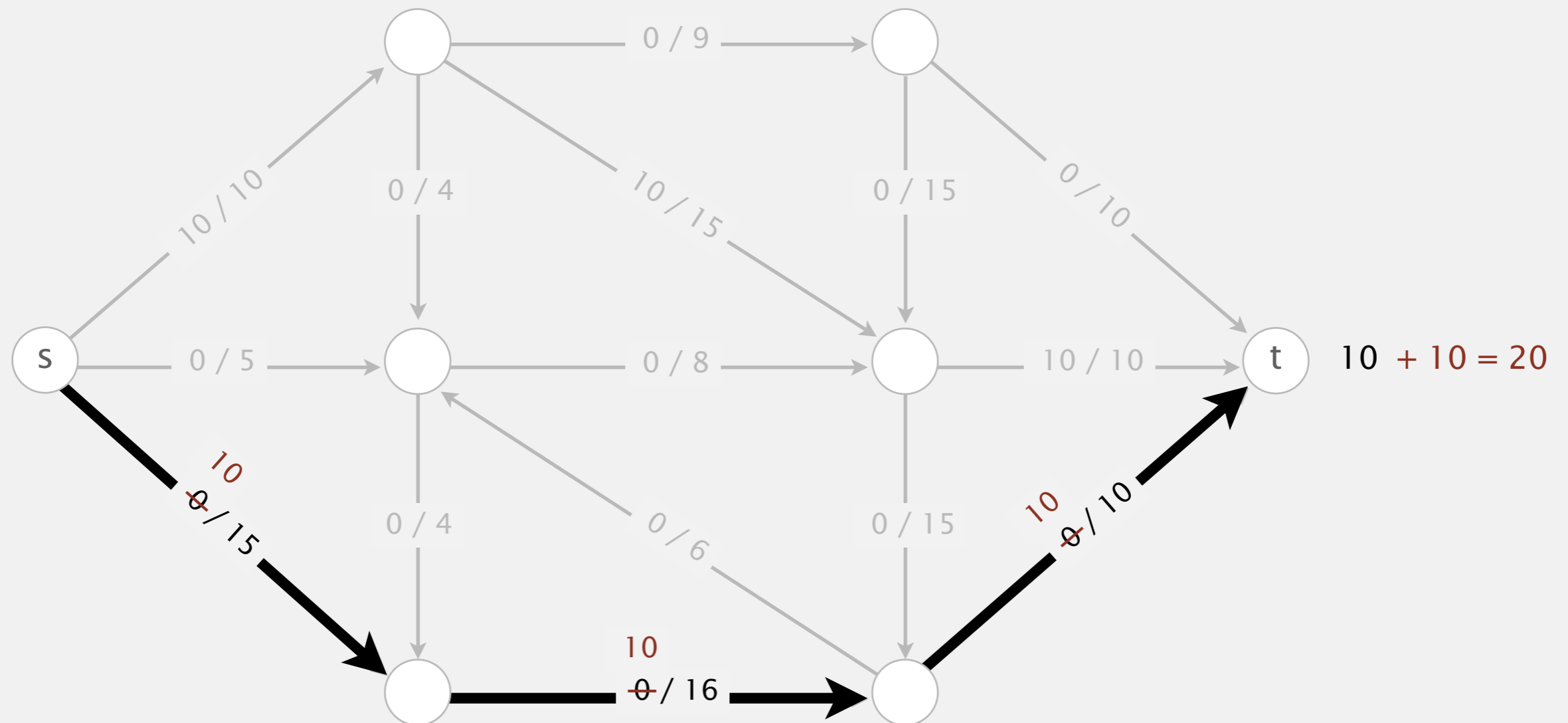


# Idea: increase flow along augmenting paths

**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**

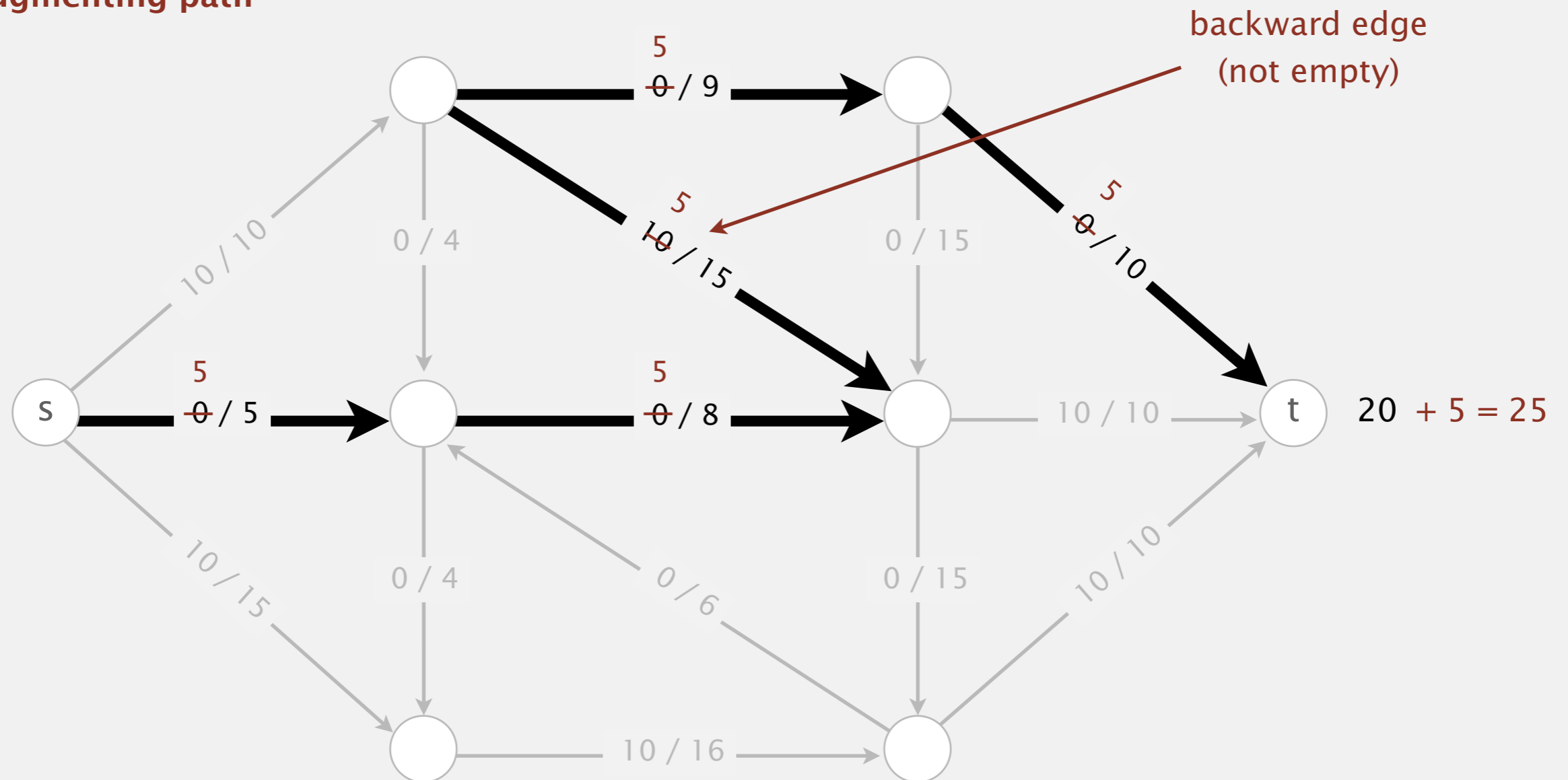


# Idea: increase flow along augmenting paths

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

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

**3<sup>rd</sup> augmenting path**

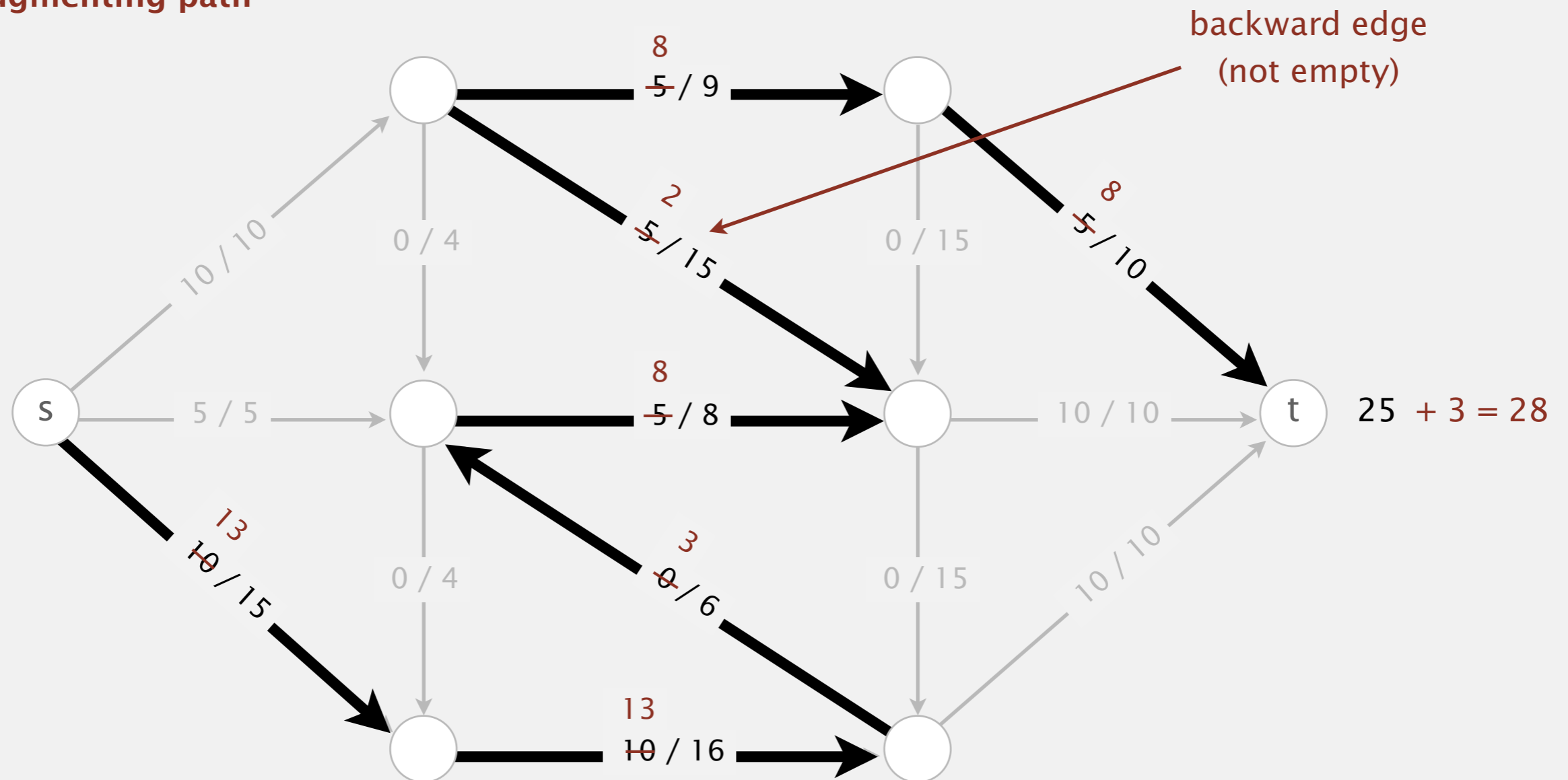


# Idea: increase flow along augmenting paths

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

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

4<sup>th</sup> augmenting path

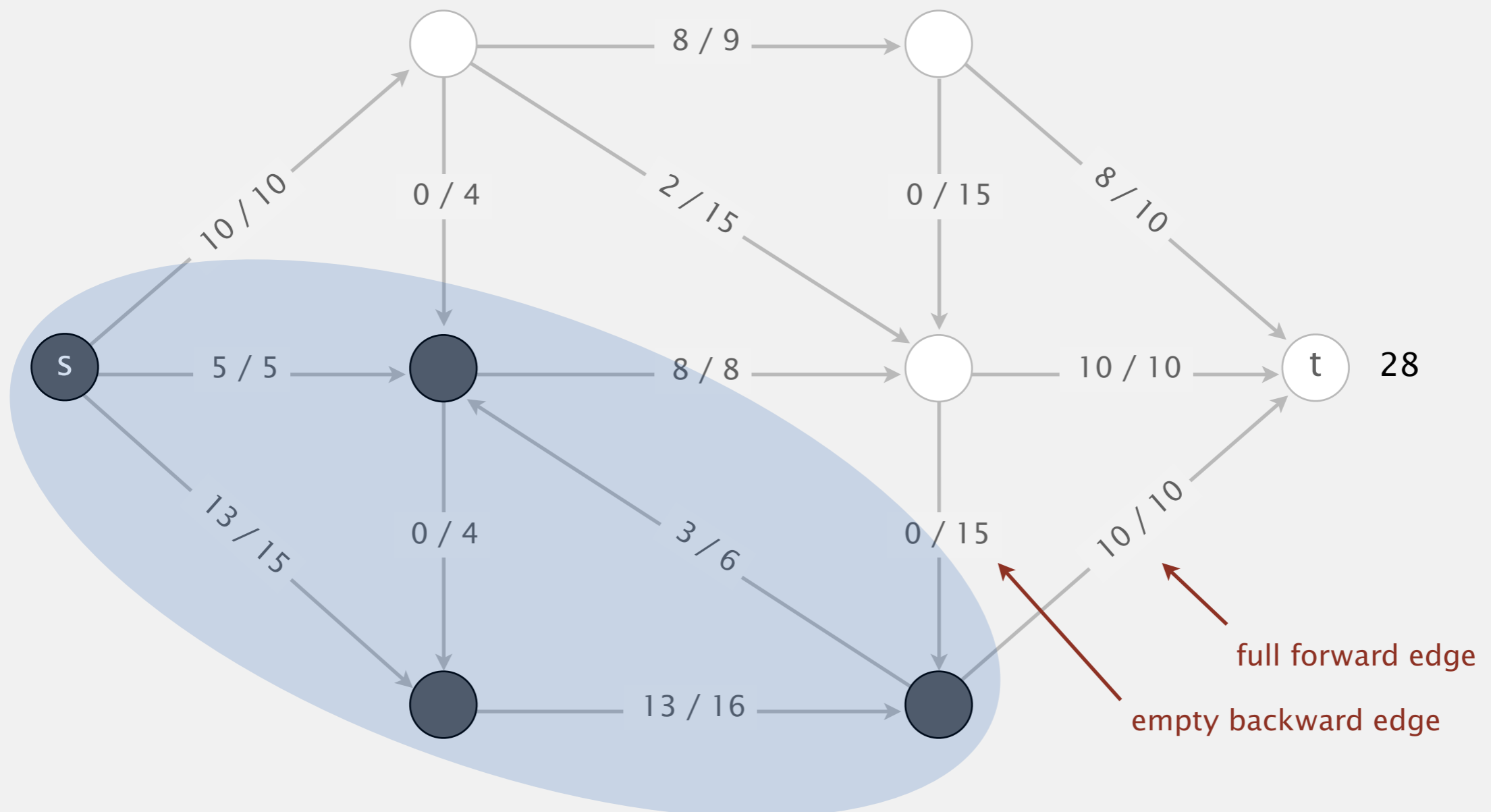


# Idea: increase flow along augmenting paths

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

- Full forward edge.
- Empty backward edge.

no more augmenting paths



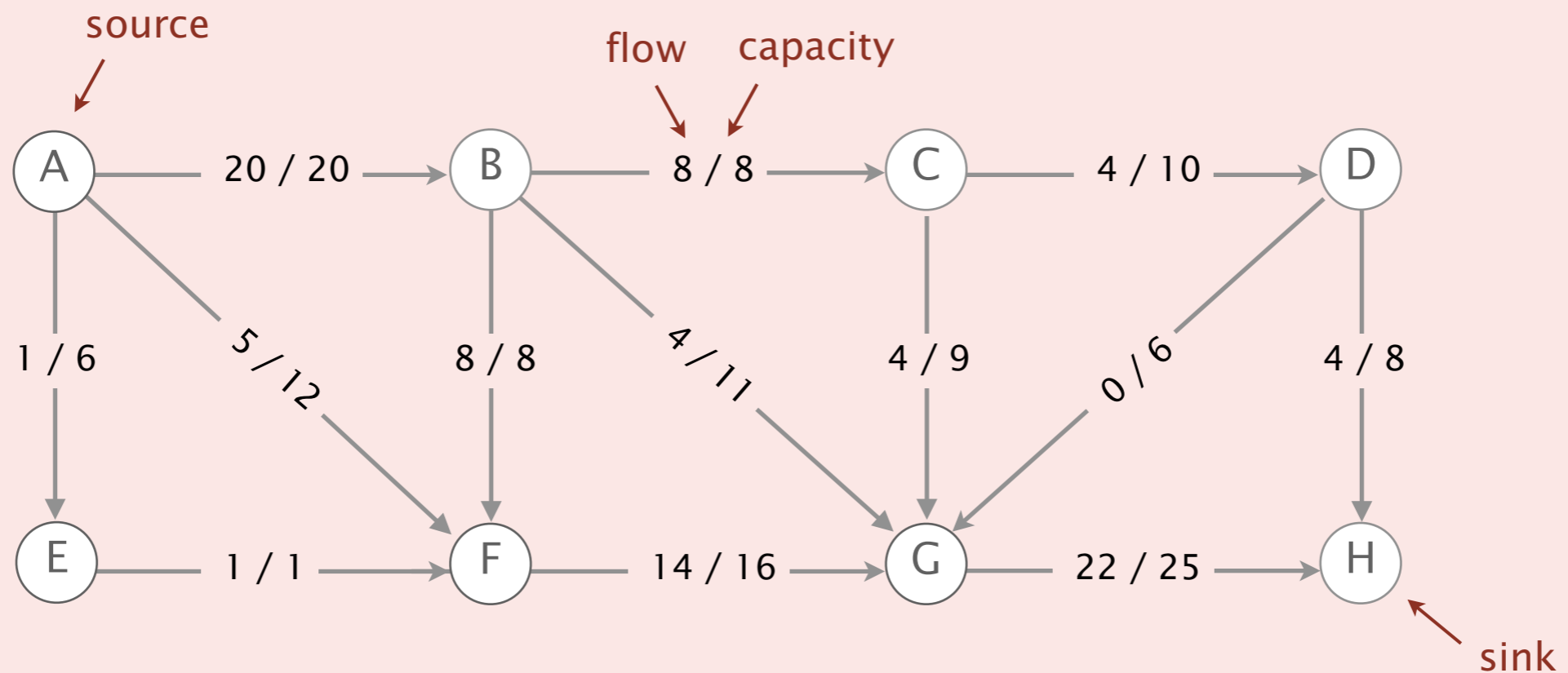


# Maxflow: quiz 2

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Which is an 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.  $A \rightarrow F \rightarrow B \rightarrow G \rightarrow D \rightarrow H$
- D. *I don't know.*



# 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?
- 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?



# Algorithms

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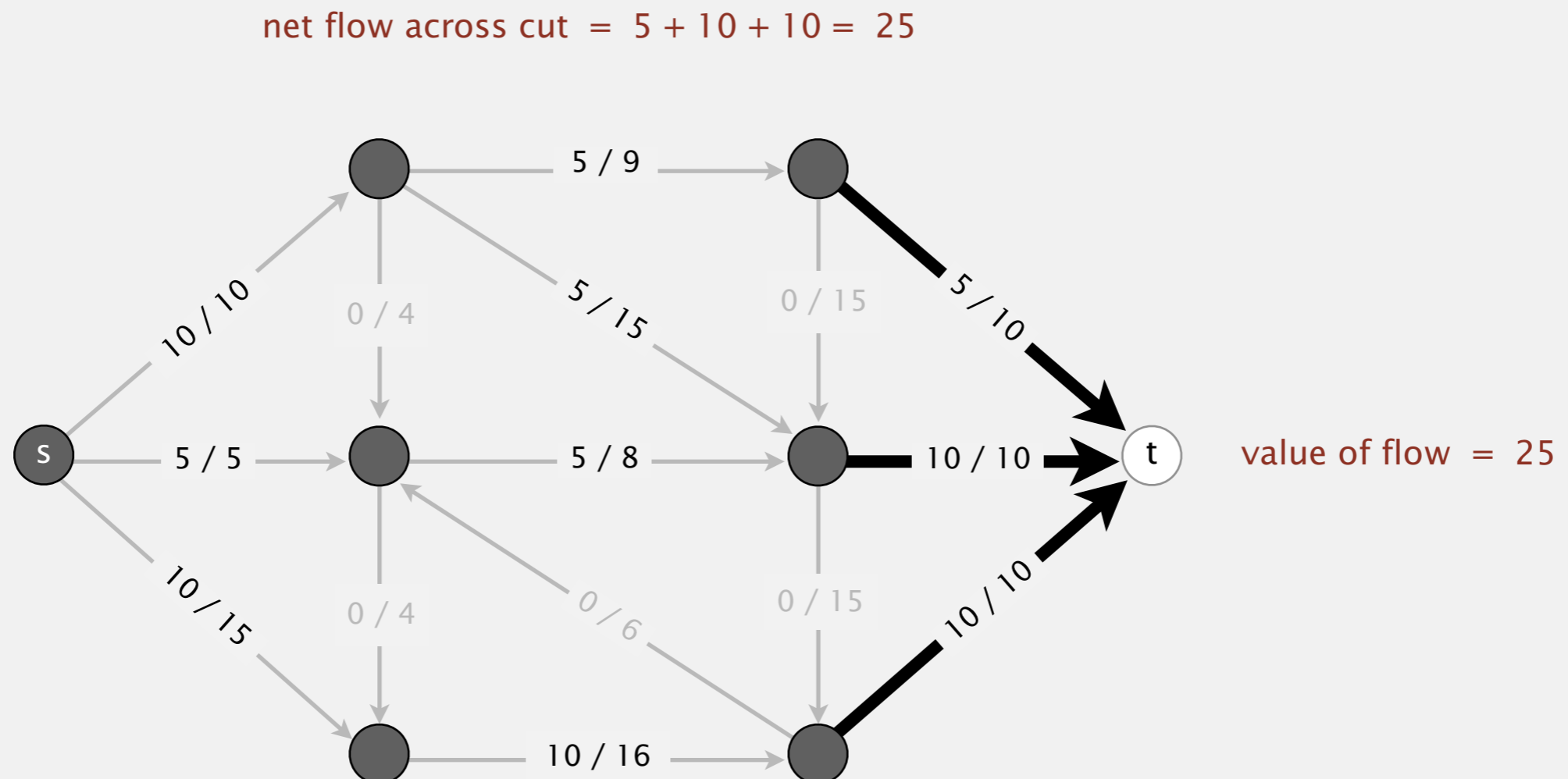
## 6.4 MAXIMUM FLOW

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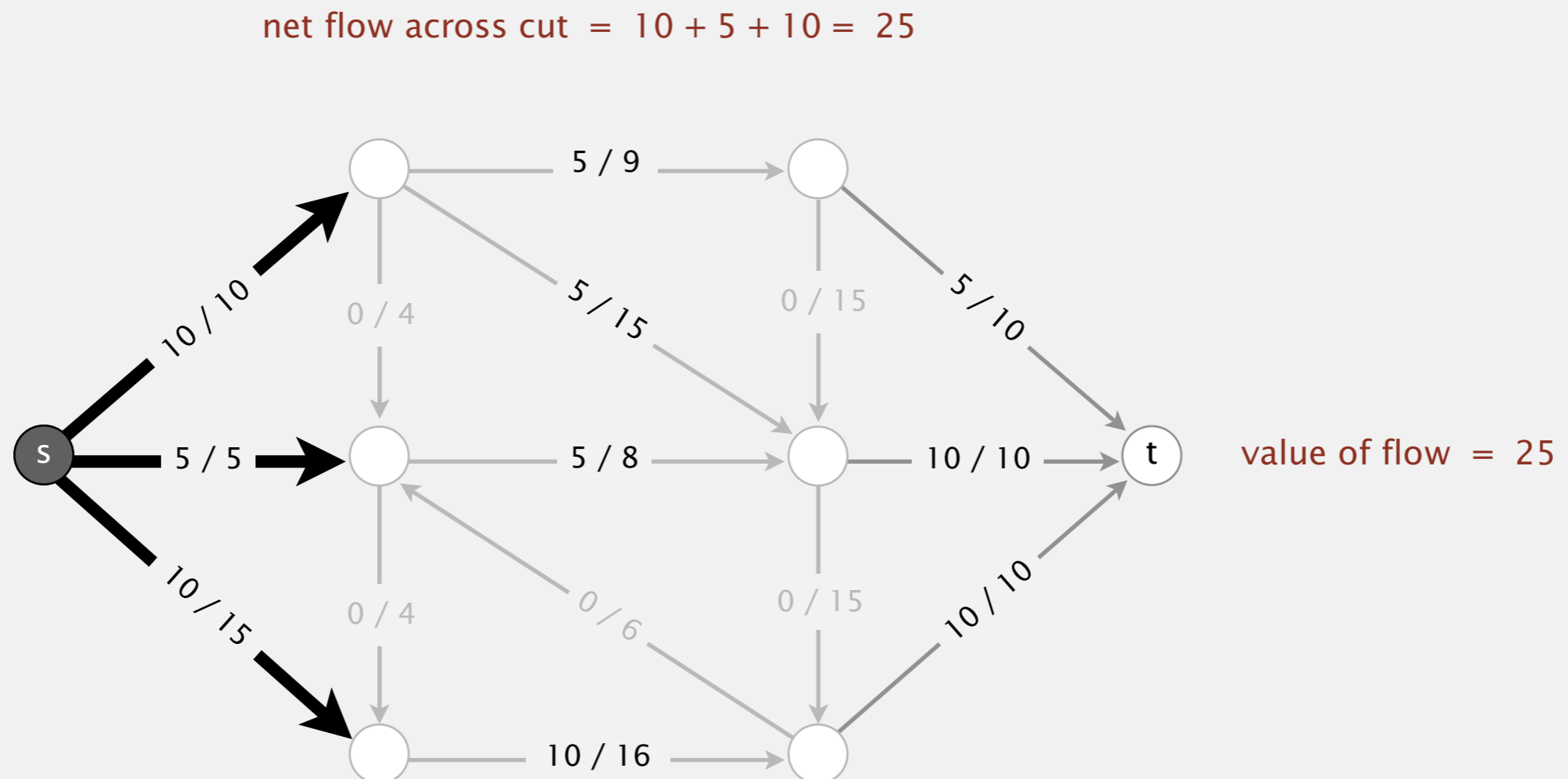
# Relationship between flows and cuts

**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

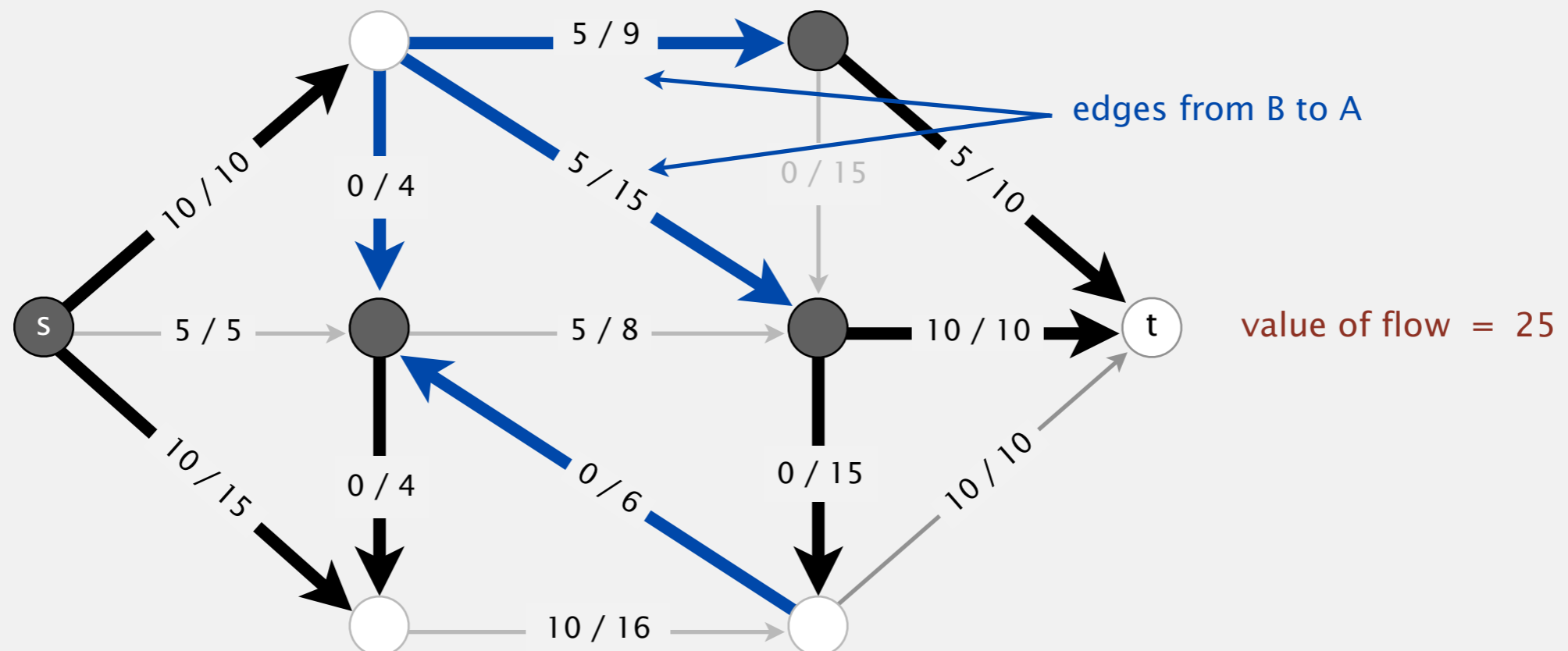
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# Relationship between flows and cuts

**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  $A$ .

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



# Relationship between flows and cuts

---

**Flow-value lemma.** 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  $B$ .

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

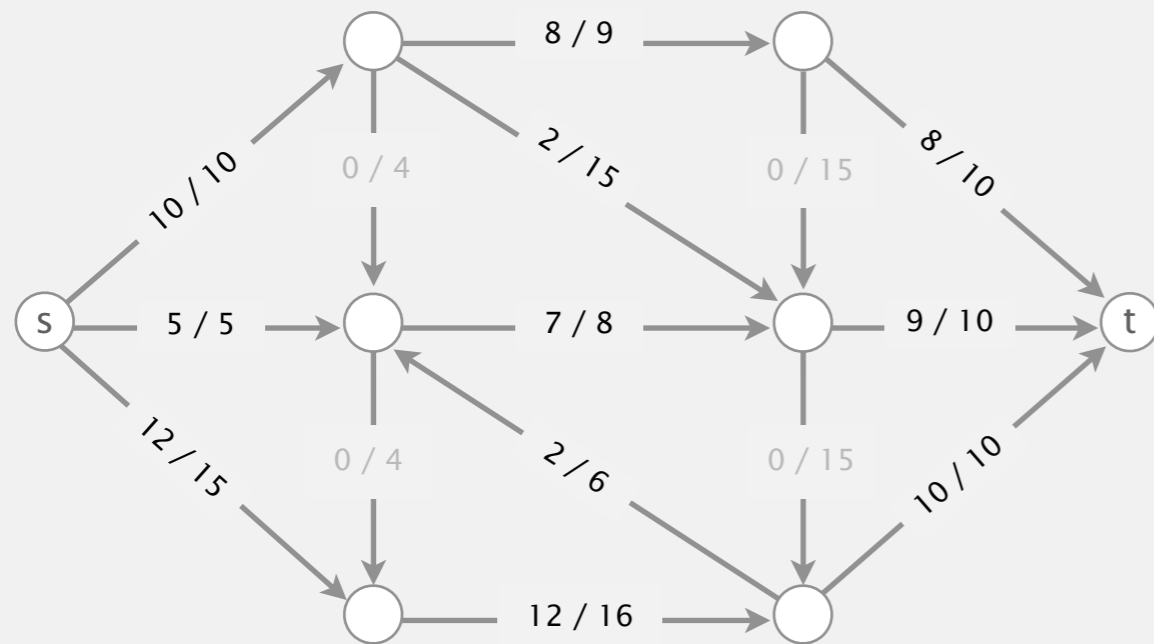
# Relationship between flows and cuts

**Weak duality.** Let  $f$  be any flow and let  $(A, B)$  be any cut. Then, the value of the flow  $\leq$  the capacity of the cut.

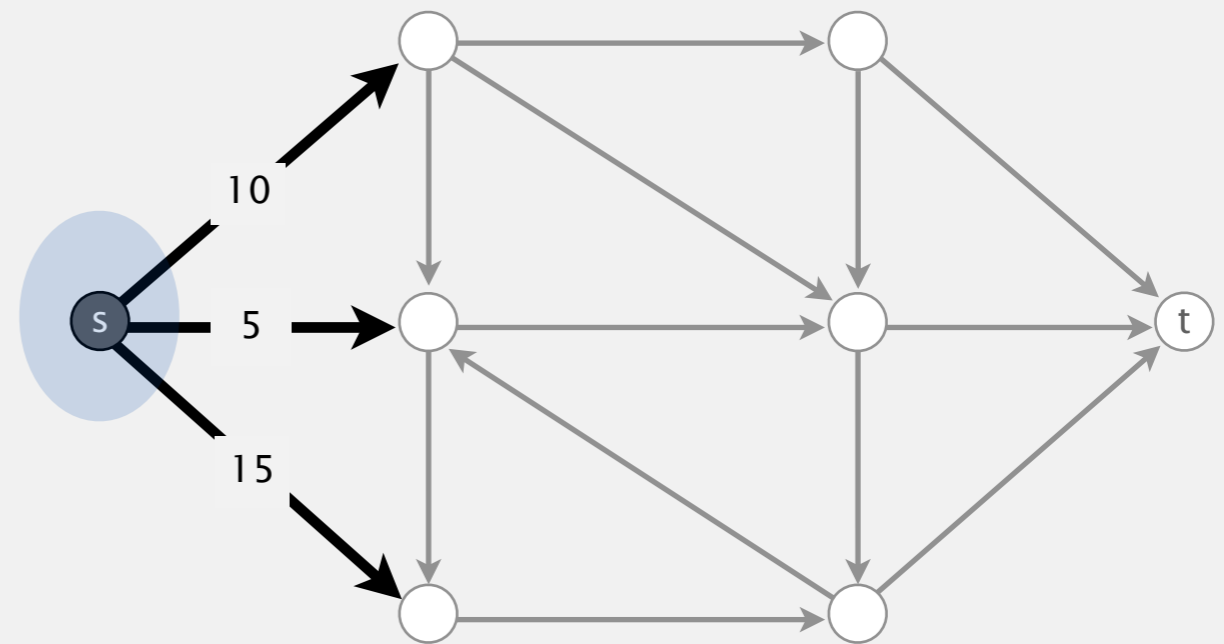
**Pf.** Value of flow  $f =$  net flow across cut  $(A, B) \leq$  capacity of cut  $(A, B)$ .

↑  
flow-value lemma

↑  
flow bounded by capacity



value of flow = 27



capacity of cut = 30



# Maxflow-mincut theorem

---

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

**Maxflow-mincut theorem.** Value of the maxflow = capacity of mincut.

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

- i. There exists a cut whose capacity equals the value of the flow  $f$ .
- ii.  $f$  is a maxflow.
- iii. There is no augmenting path with respect to  $f$ .

[ i  $\Rightarrow$  ii ]

- Suppose that  $(A, B)$  is a cut with capacity equal to the value of  $f$ .
- Then, the value of any flow  $f' \leq$  capacity of  $(A, B) =$  value of  $f$ .
- Thus,  $f$  is a maxflow.

↑  
weak duality

↑  
by assumption

# Maxflow-mincut theorem

---

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

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

- i. There exists a cut whose capacity equals the value of the flow  $f$ .
- ii.  $f$  is a maxflow.
- iii. There is no augmenting path with respect to  $f$ .

[ ii  $\Rightarrow$  iii ] We prove contrapositive:  $\sim$ iii  $\Rightarrow$   $\sim$ ii.

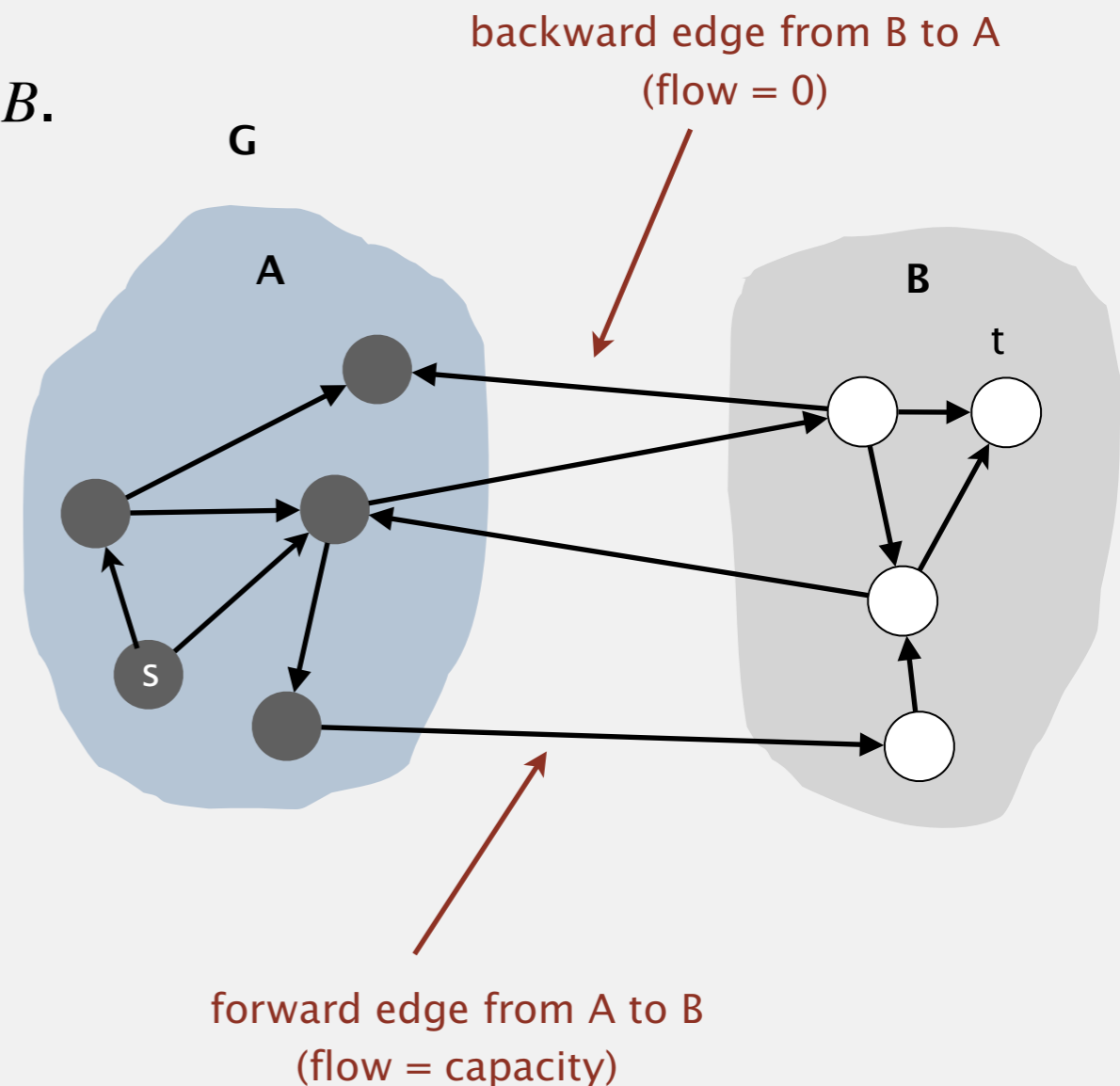
- Suppose that there is an augmenting path with respect to  $f$ .
- Can improve flow  $f$  by sending flow along this path.
- Thus,  $f$  is not a maxflow.

# Maxflow-mincut theorem

[ iii  $\Rightarrow$  i ]

- Let  $f$  be a flow with no augmenting paths.
- Let  $A$  be set of vertices connected to  $s$  by an undirected path with no full forward or empty backward edges
- By definition of cut  $A$ ,  $s$  is in  $A$ .
- By definition of cut  $A$  and flow  $f$ ,  $t$  is in  $B$ .
- Capacity of cut = net flow across cut  
= value of flow  $f$ .

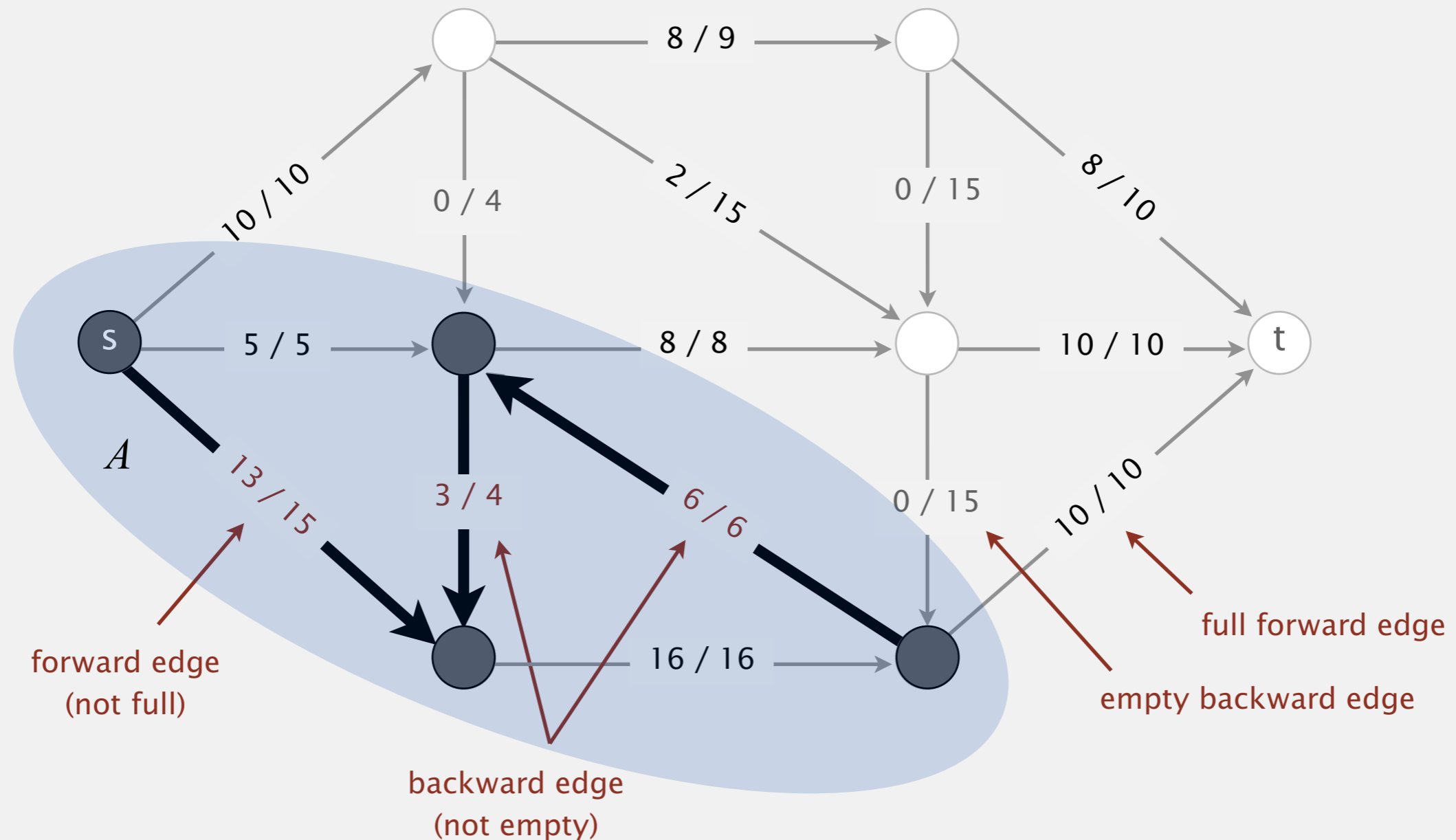
flow-value  
lemma



# Computing a mincut from a maxflow

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

- By augmenting path theorem, no augmenting paths with respect to  $f$ .
- Compute  $A$  = set of vertices connected to  $s$  by an undirected path with no full forward or empty backward edges.





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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

# Ford-Fulkerson algorithm with integer capacities

---

**Important special case.** Edge capacities are integers between 1 and  $U$ .

flow on each edge is an integer

**Invariant.** The flow is integral throughout Ford-Fulkerson.

**Pf.** [by induction]

- Bottleneck capacity is an integer.
- Flow on an edge increases/decreases by bottleneck capacity.

**Proposition.** Number of augmentations  $\leq$  the value of the maxflow.

**Pf.** Each augmentation increases the value by at least 1.

critical for some applications (stay tuned)

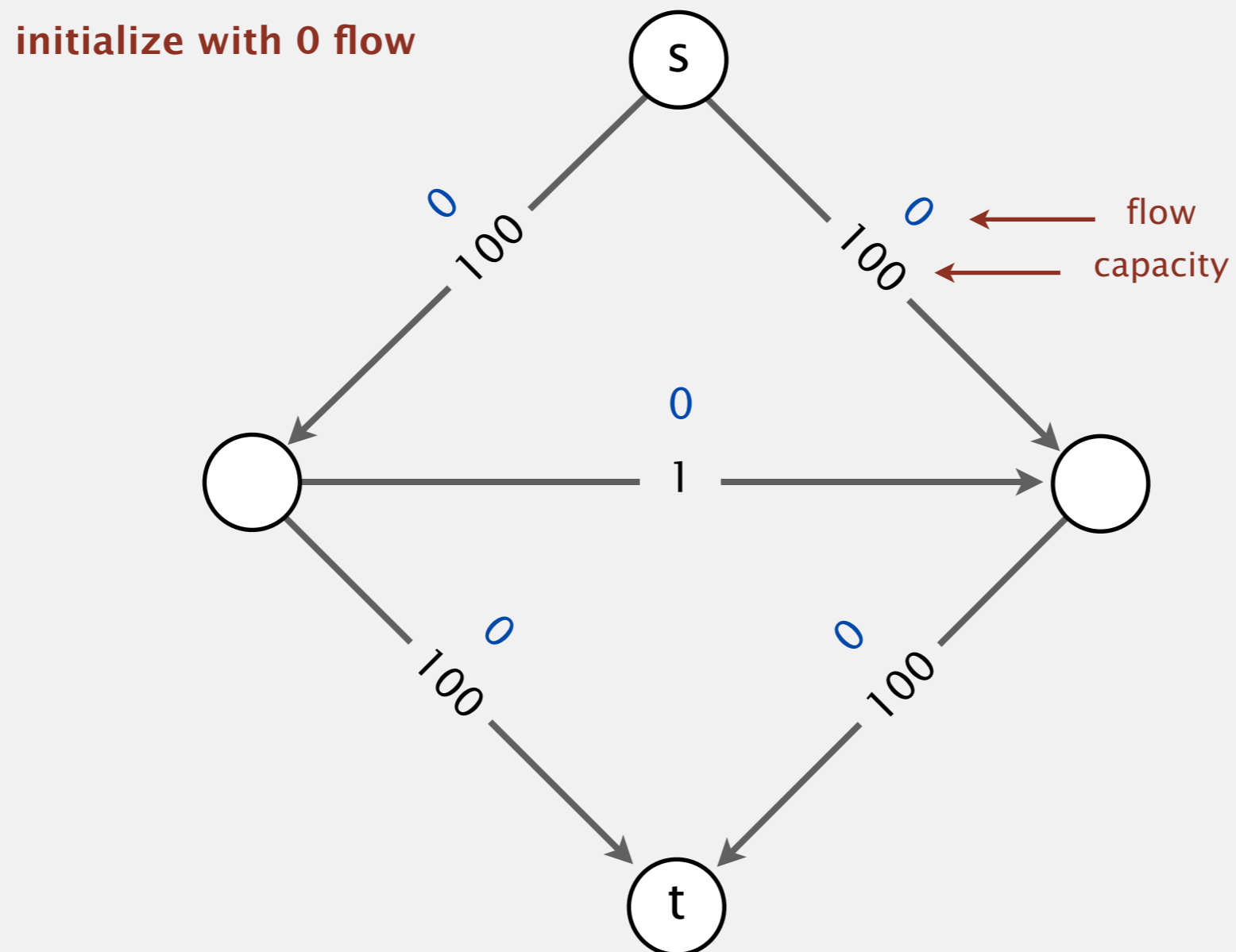
**Integrality theorem.** There exists an integral maxflow.

**Pf.** Ford-Fulkerson terminates and maxflow that it finds is integer-valued.

# Bad case for Ford-Fulkerson

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



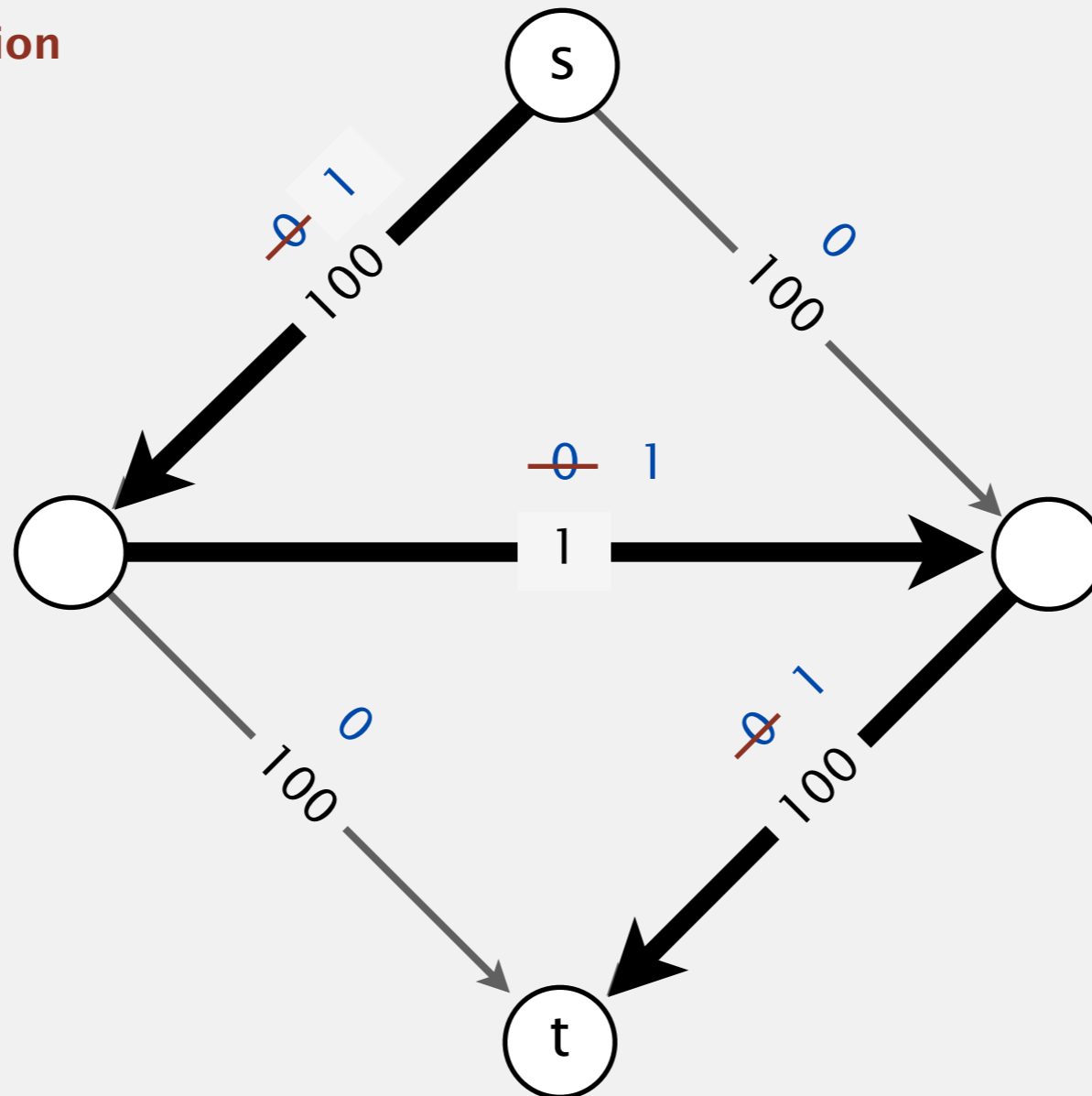


# Bad case for Ford-Fulkerson

---

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

1<sup>st</sup> iteration

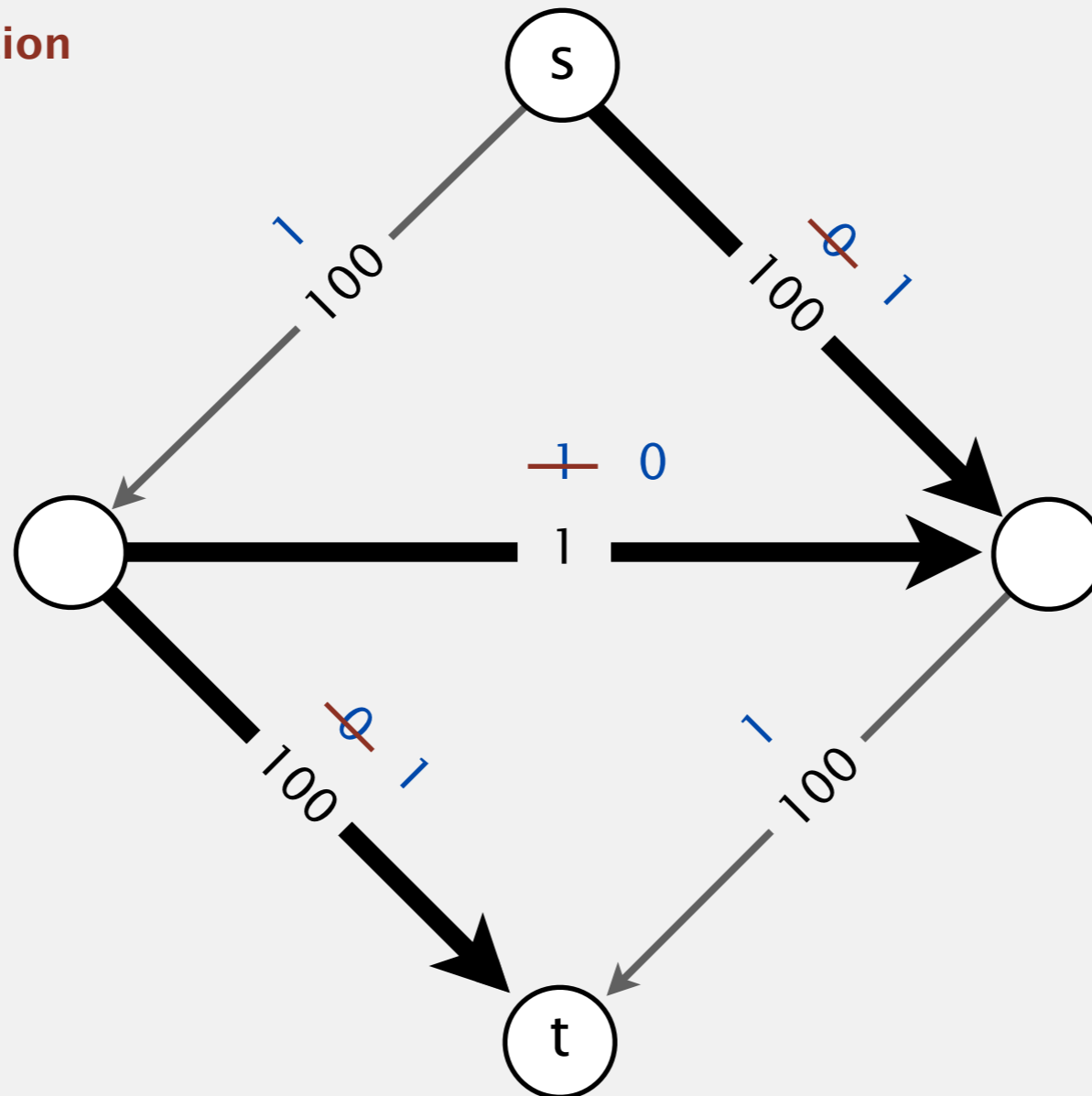


# Bad case for Ford-Fulkerson

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

2<sup>nd</sup> iteration

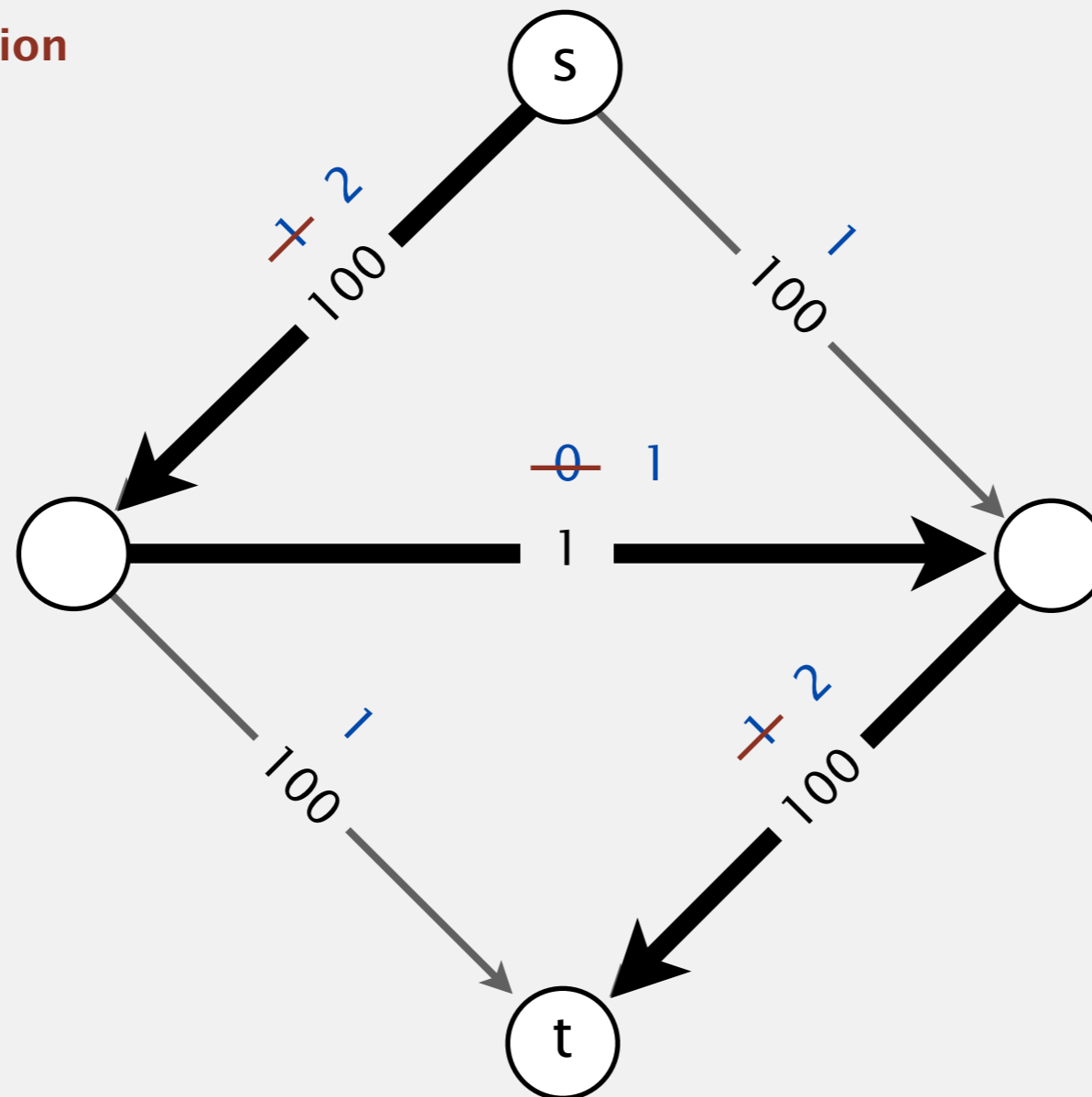


# Bad case for Ford-Fulkerson

---

**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

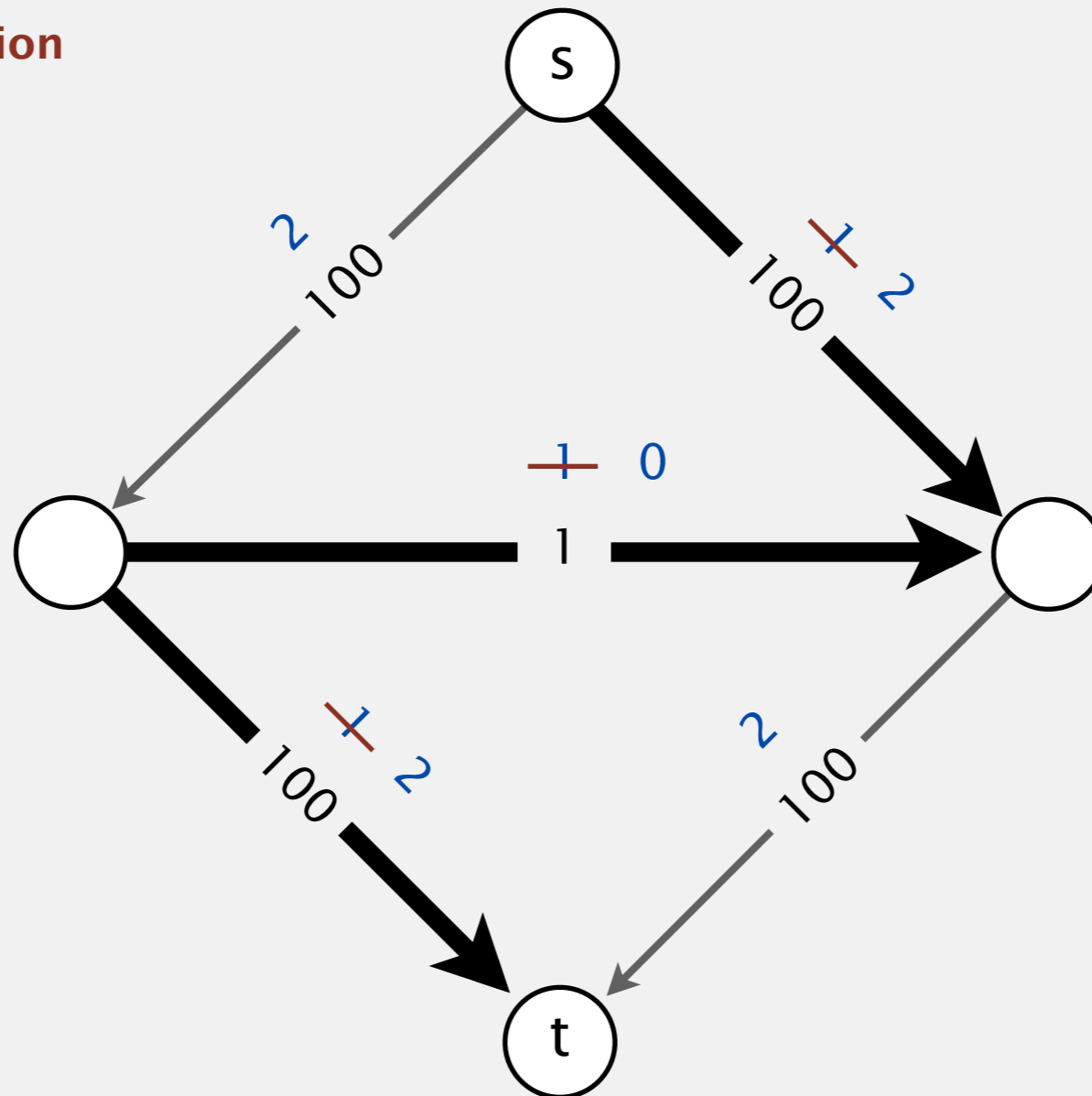


# Bad case for Ford-Fulkerson

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

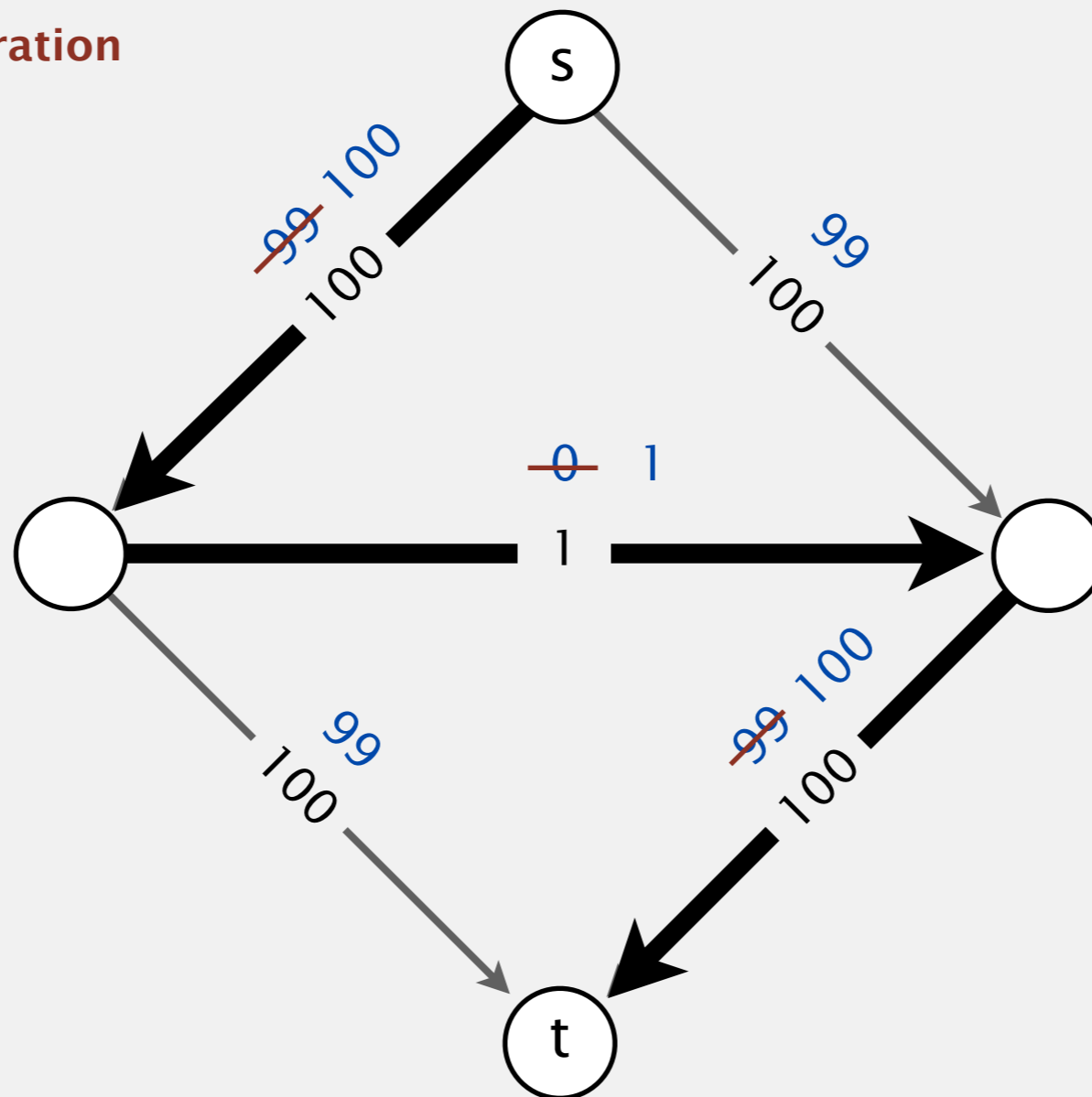


# Bad case for Ford-Fulkerson

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

199<sup>th</sup> iteration

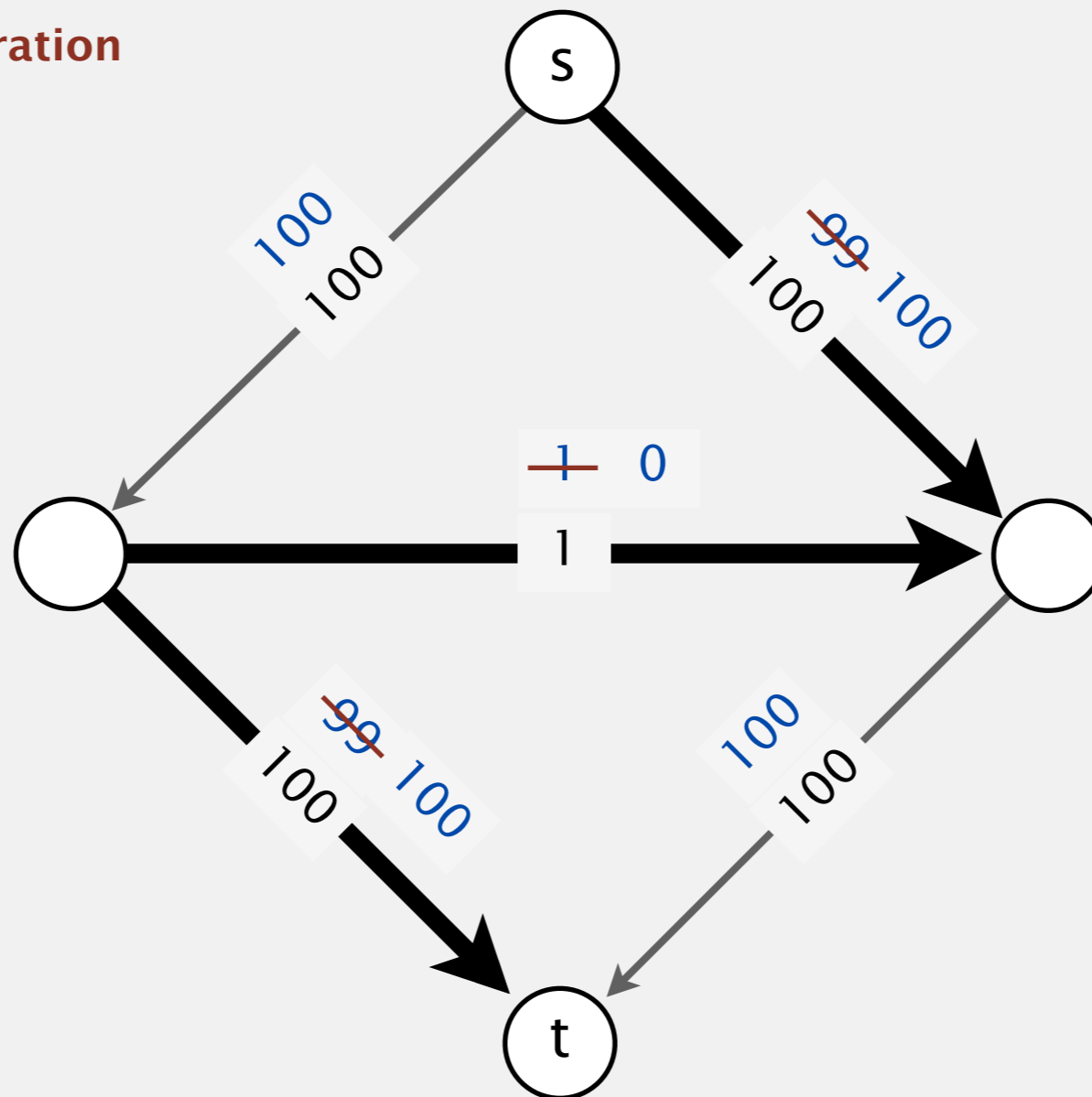


# Bad case for Ford-Fulkerson

---

**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



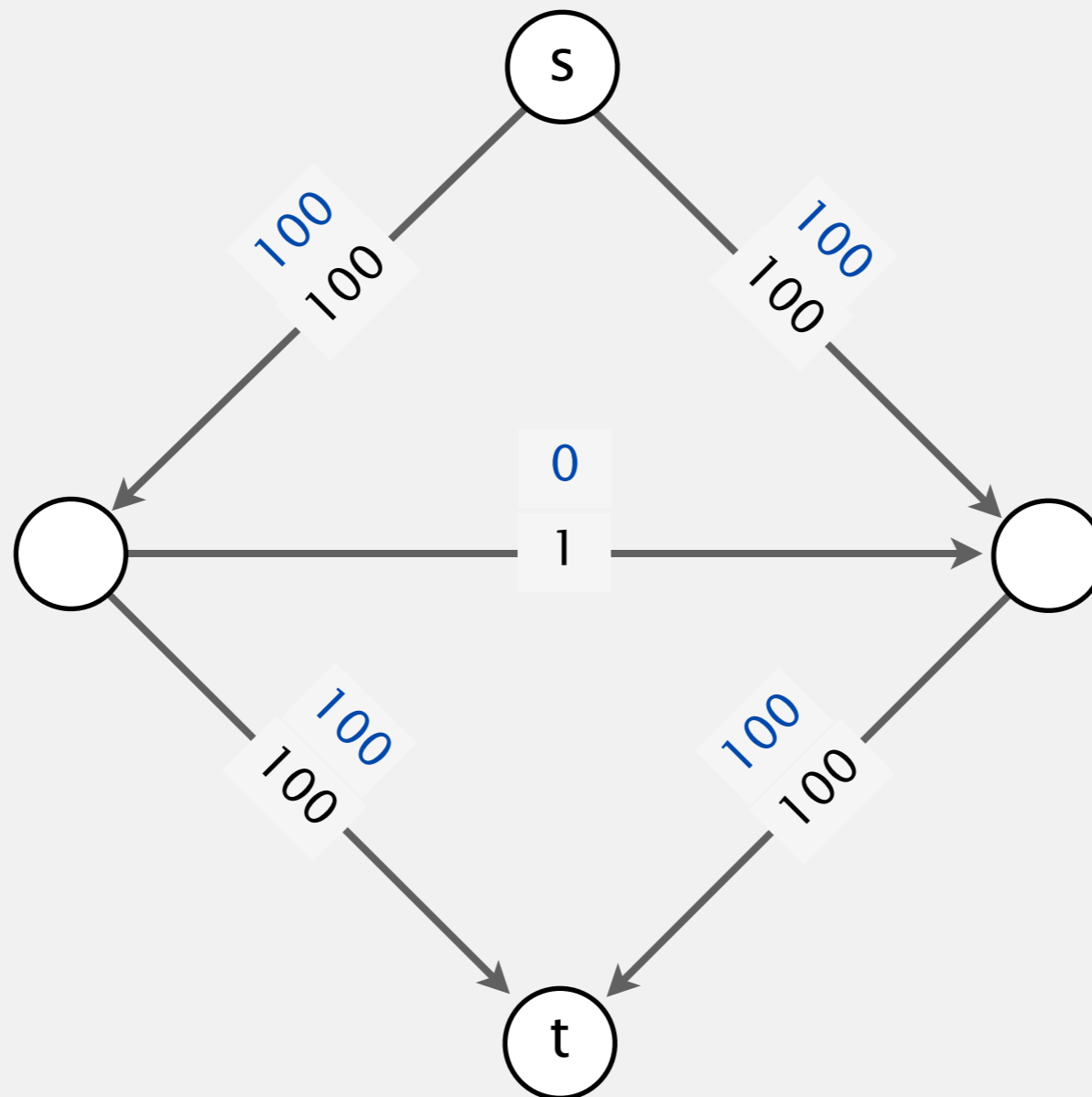
# Bad case for Ford-Fulkerson

---

**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
DFS path	$\leq E U$	stack (DFS)
shortest path	$\leq \frac{1}{2} E V$	queue (BFS)
fattest path	$\leq E \ln(E U)$	priority queue

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



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

## 6.4 MAXIMUM FLOW

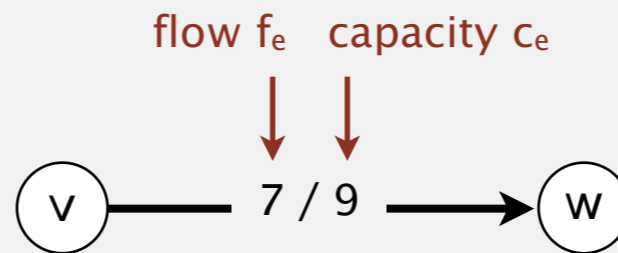
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# Flow network representation

---

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



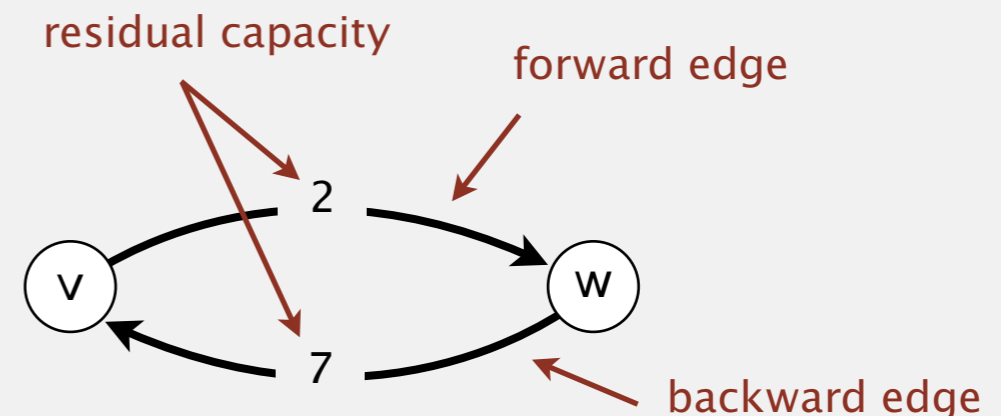
**Flow network data type.** Must be able to process edge  $e = v \rightarrow w$  in either direction: include  $e$  in adjacency lists of both  $v$  and  $w$ .

**Residual (spare) capacity.**

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

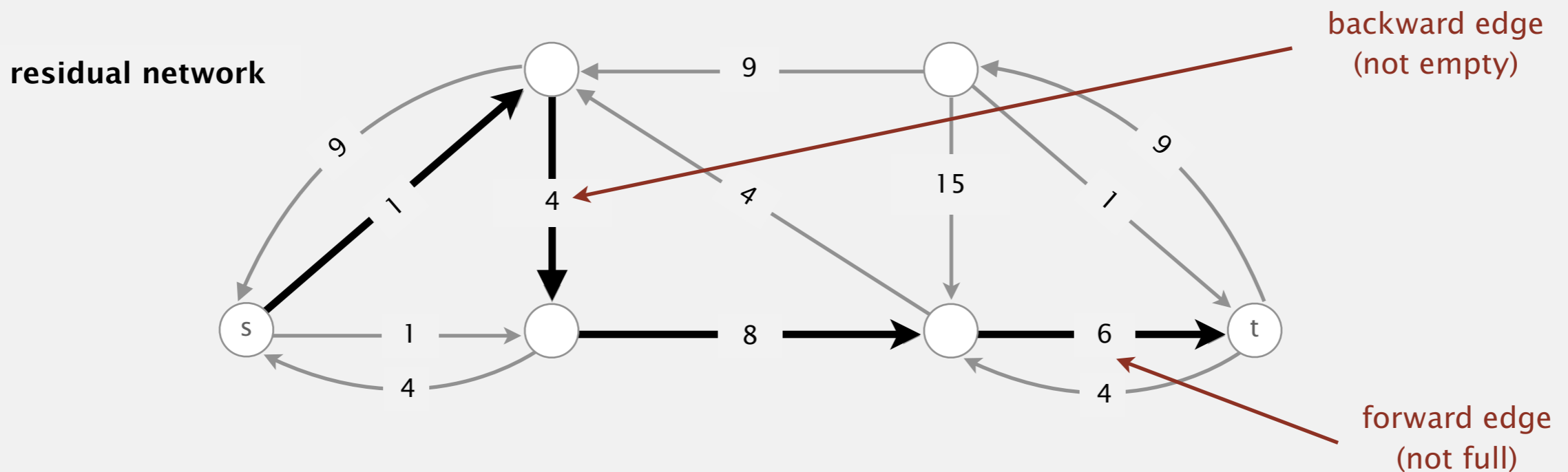
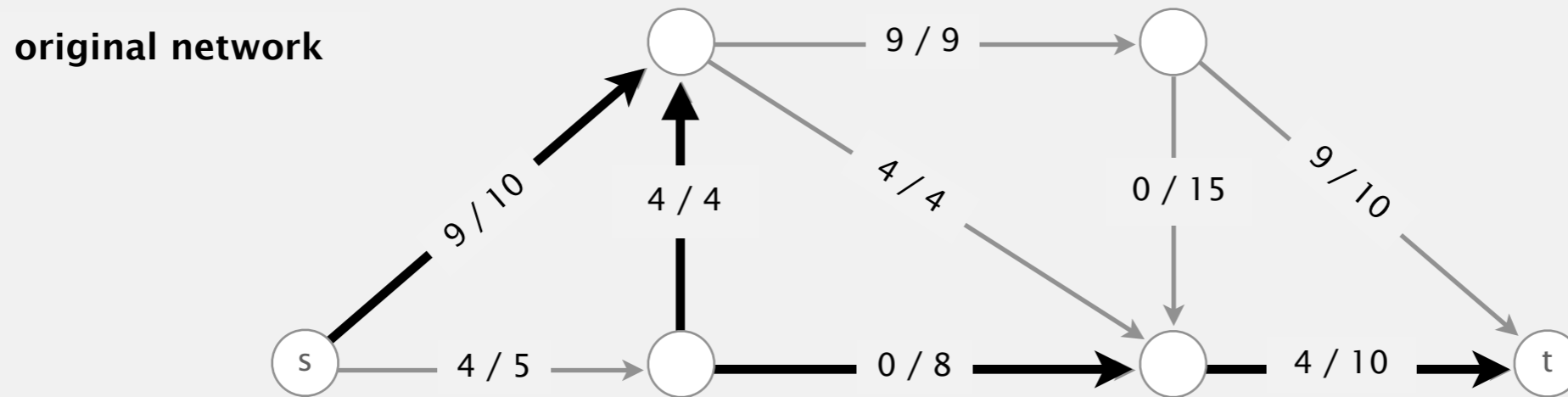
**Augment flow.**

- 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

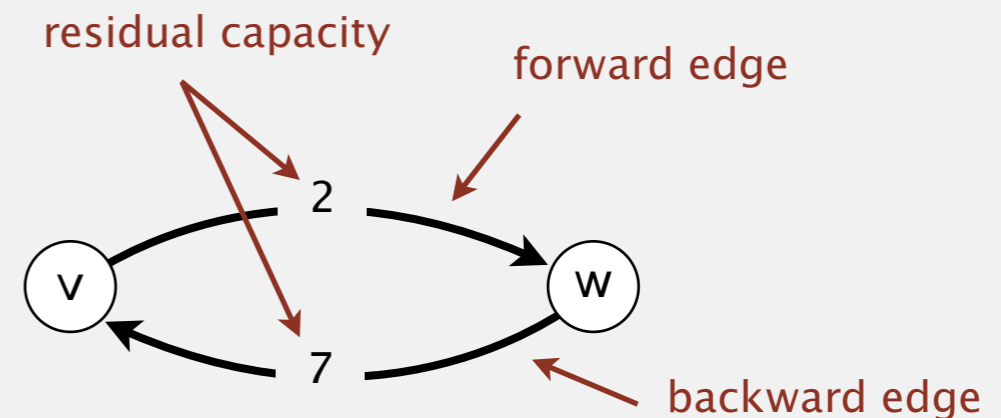
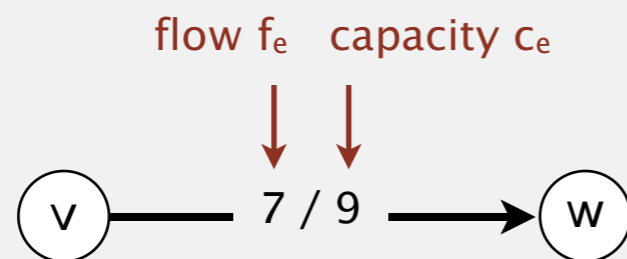


**Key point.** Augmenting paths in original network are in one-to-one correspondence with directed paths in residual network.

# Flow edge API

```
public class FlowEdge
```

```
    FlowEdge(int v, int w, double capacity)    create a flow edge v→w  
  
    int from()                                vertex this edge points from  
  
    int to()                                  vertex this edge points to  
  
    int other(int v)                          other endpoint  
  
    double capacity()                         capacity of this edge  
  
    double flow()                             flow in this edge  
  
    double residualCapacityTo(int v)          residual capacity toward v  
  
    void addResidualFlowTo(int v, double delta) add delta flow toward v
```



# 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

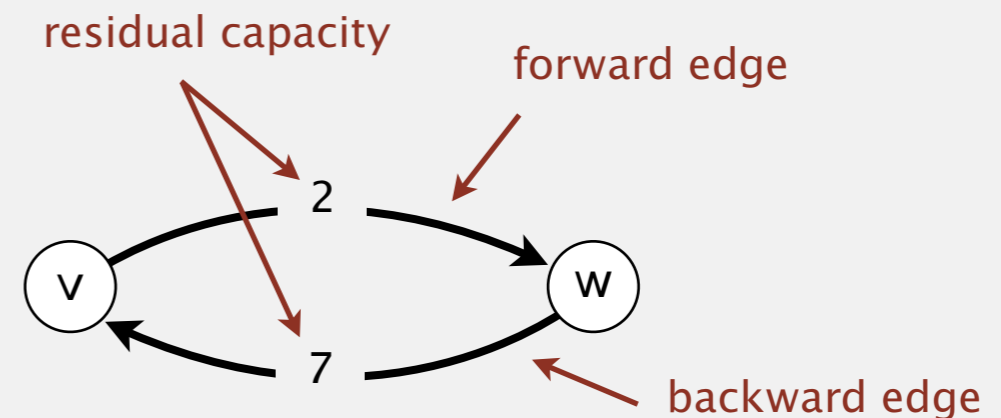
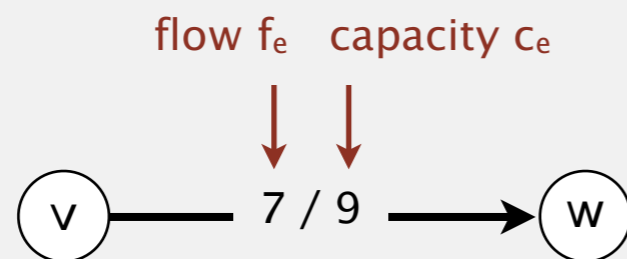
# Flow edge: Java implementation (continued)

```
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/from 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>();
    }
```

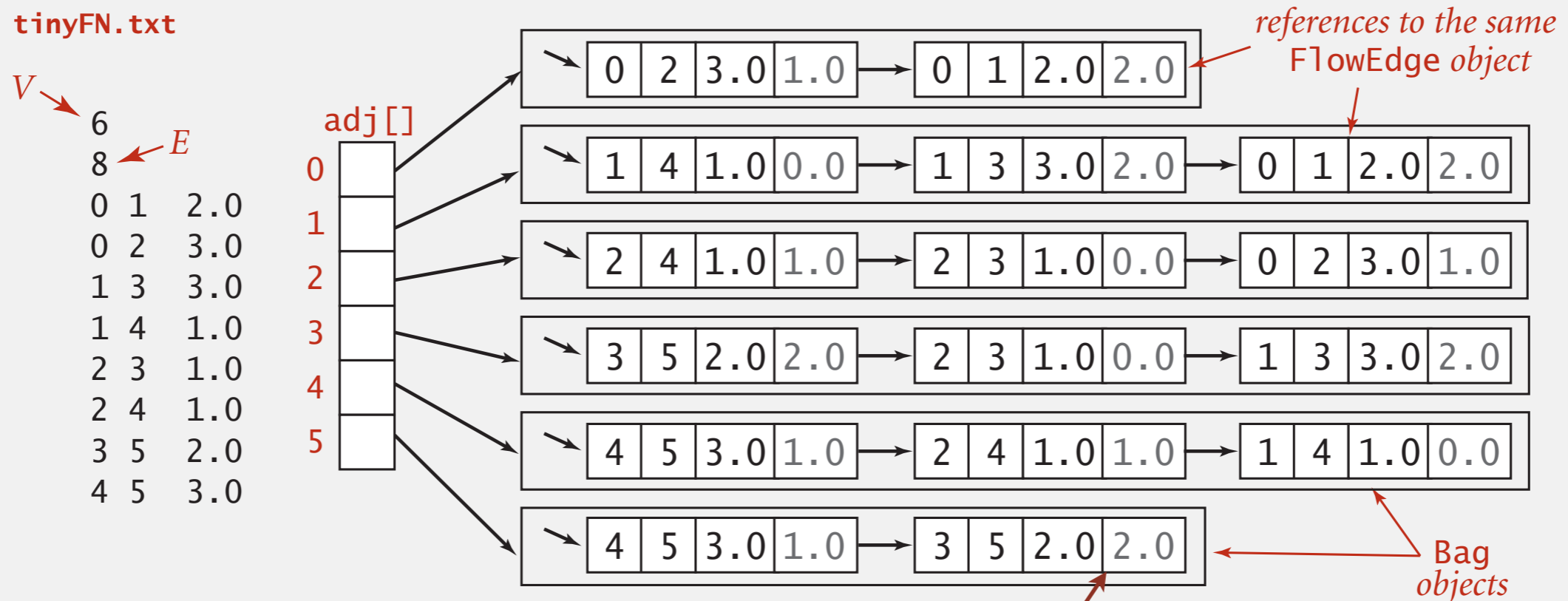
```
    public void addEdge(FlowEdge e)
    {
        int v = e.from();
        int w = e.to();
        adj[v].add(e);
        adj[w].add(e);
    }
```

← 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;
        }
    }
```

compute edgeTo[] and marked[]

compute bottleneck capacity

augment flow

```
    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]; }
```

is v reachable from s in residual network?



# Algorithms

ROBERT SEDGEWICK | KEVIN WAYNE

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

## 6.4 MAXIMUM FLOW

---

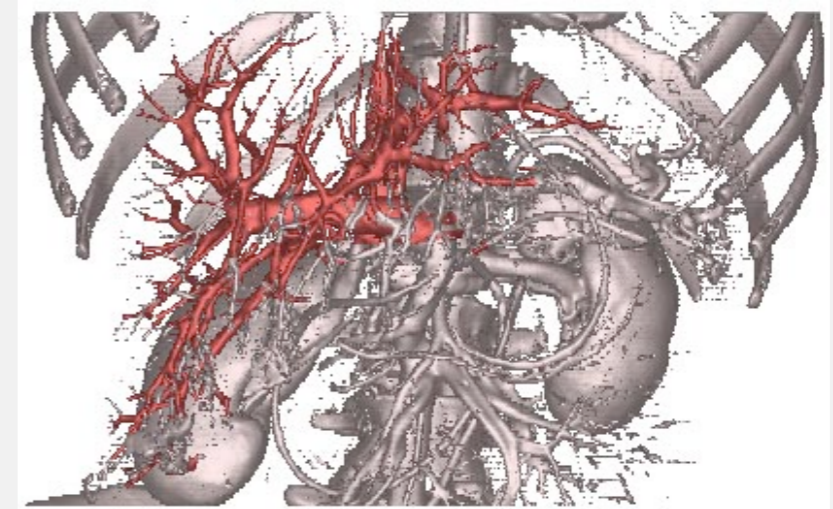
- ▶ *introduction*
- ▶ *Ford-Fulkerson algorithm*
- ▶ *maxflow-minicut theorem*
- ▶ *analysis of running time*
- ▶ *Java implementation*
- ▶ ***applications***

# Maxflow and mincut applications

---

Maxflow/mincut is a widely applicable problem-solving model.

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



**liver and hepatic vascularization segmentation**

# 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



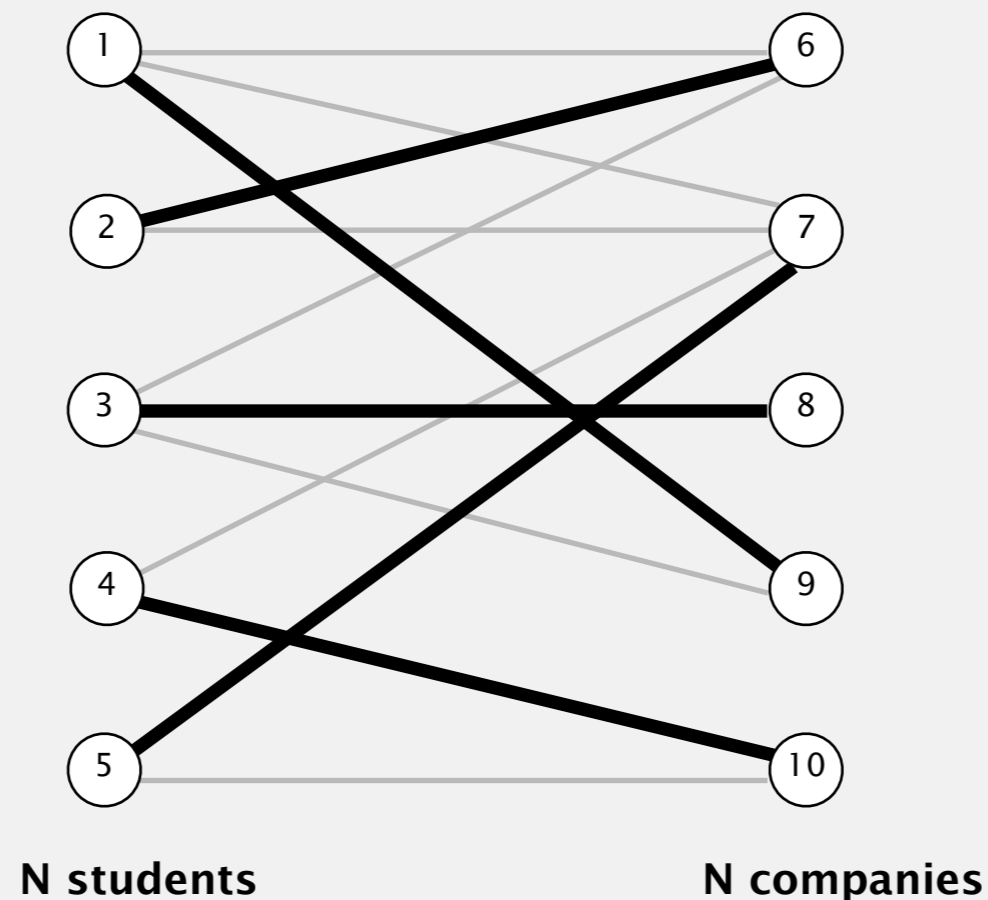
# Bipartite matching problem

Given a bipartite graph, find a perfect matching.

## perfect matching (solution)

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

## bipartite graph

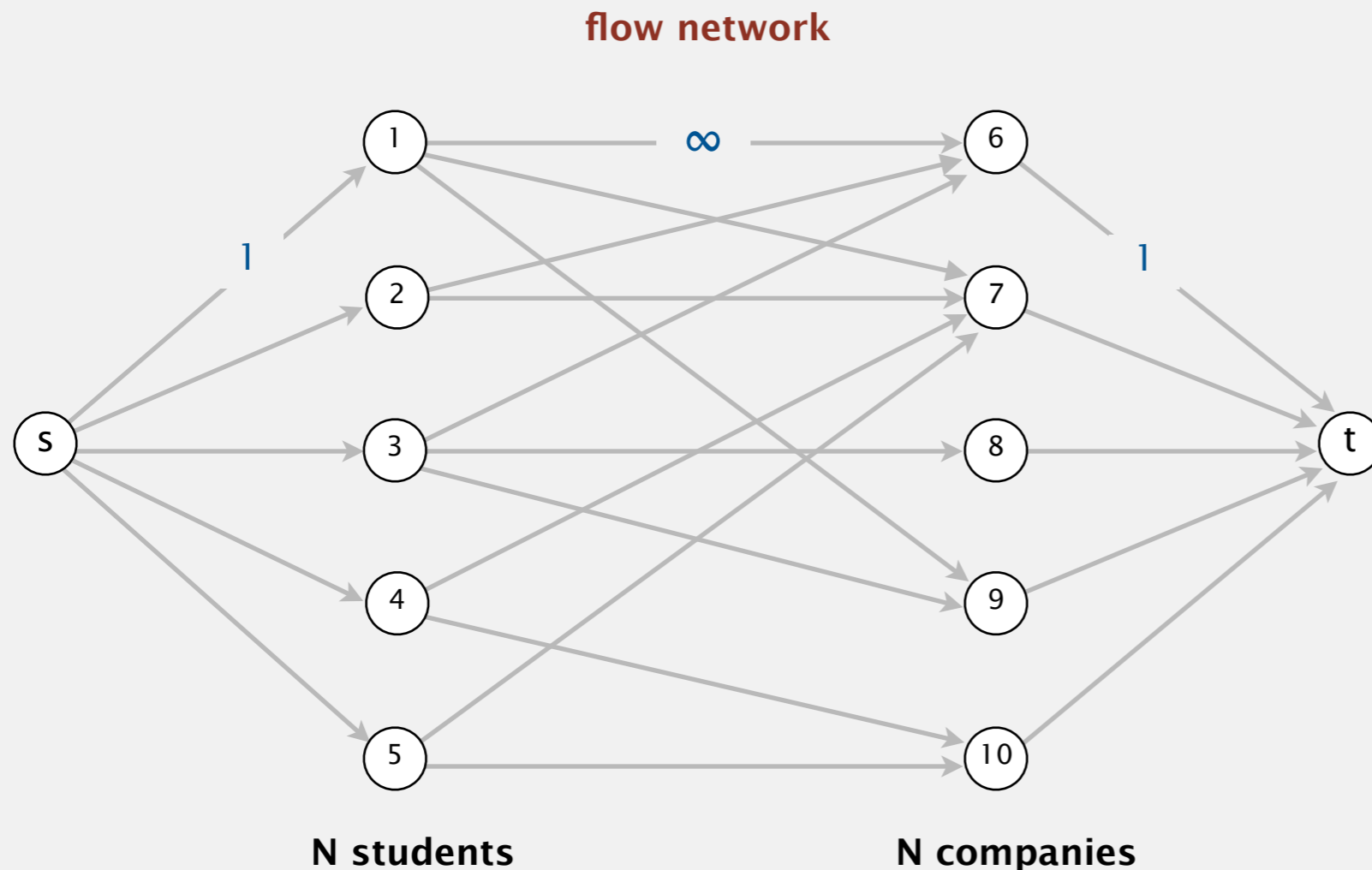


## bipartite matching problem

- |   |          |    |          |
|---|----------|----|----------|
| 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).

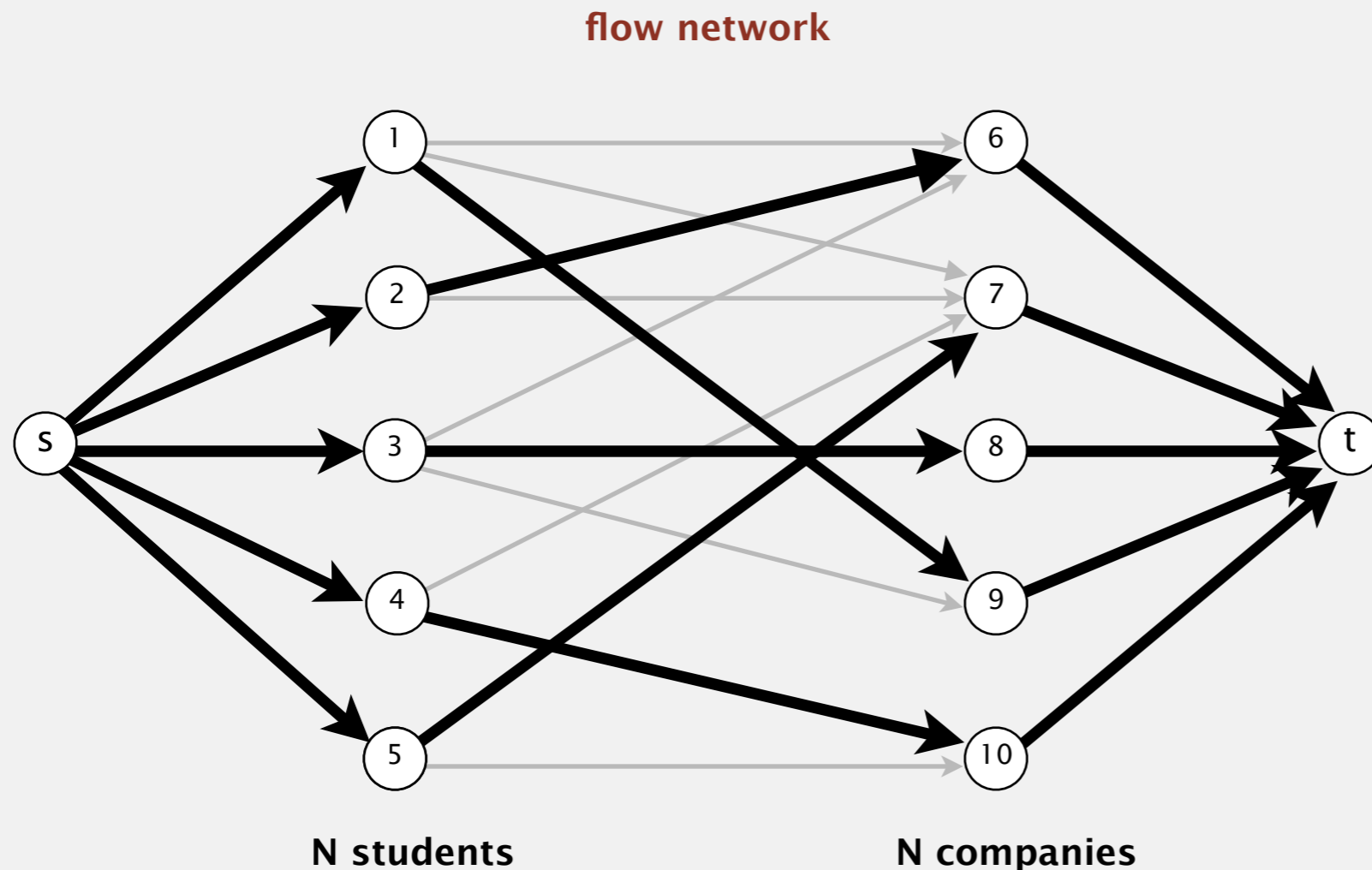


**bipartite matching problem**

1	Alice	6	Adobe
	Adobe		Alice
	Amazon		Bob
	Google		Carol
2	Bob	7	Amazon
	Adobe		Alice
	Amazon		Bob
3	Carol		Dave
	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

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



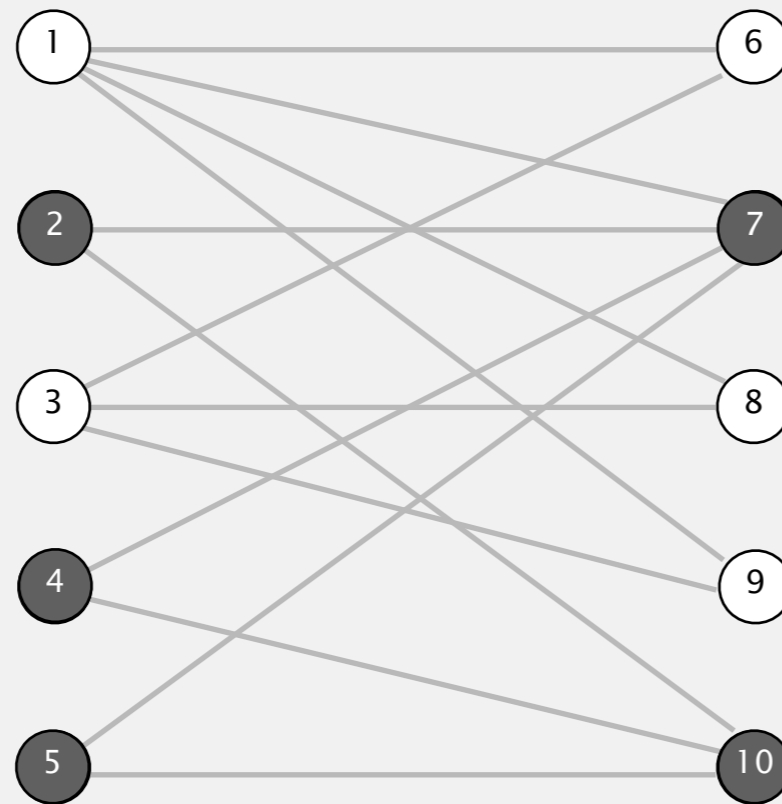
## bipartite matching problem

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

# What the mincut tells us

---

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



**no perfect matching exists**

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

$T = \{ 7, 10 \}$

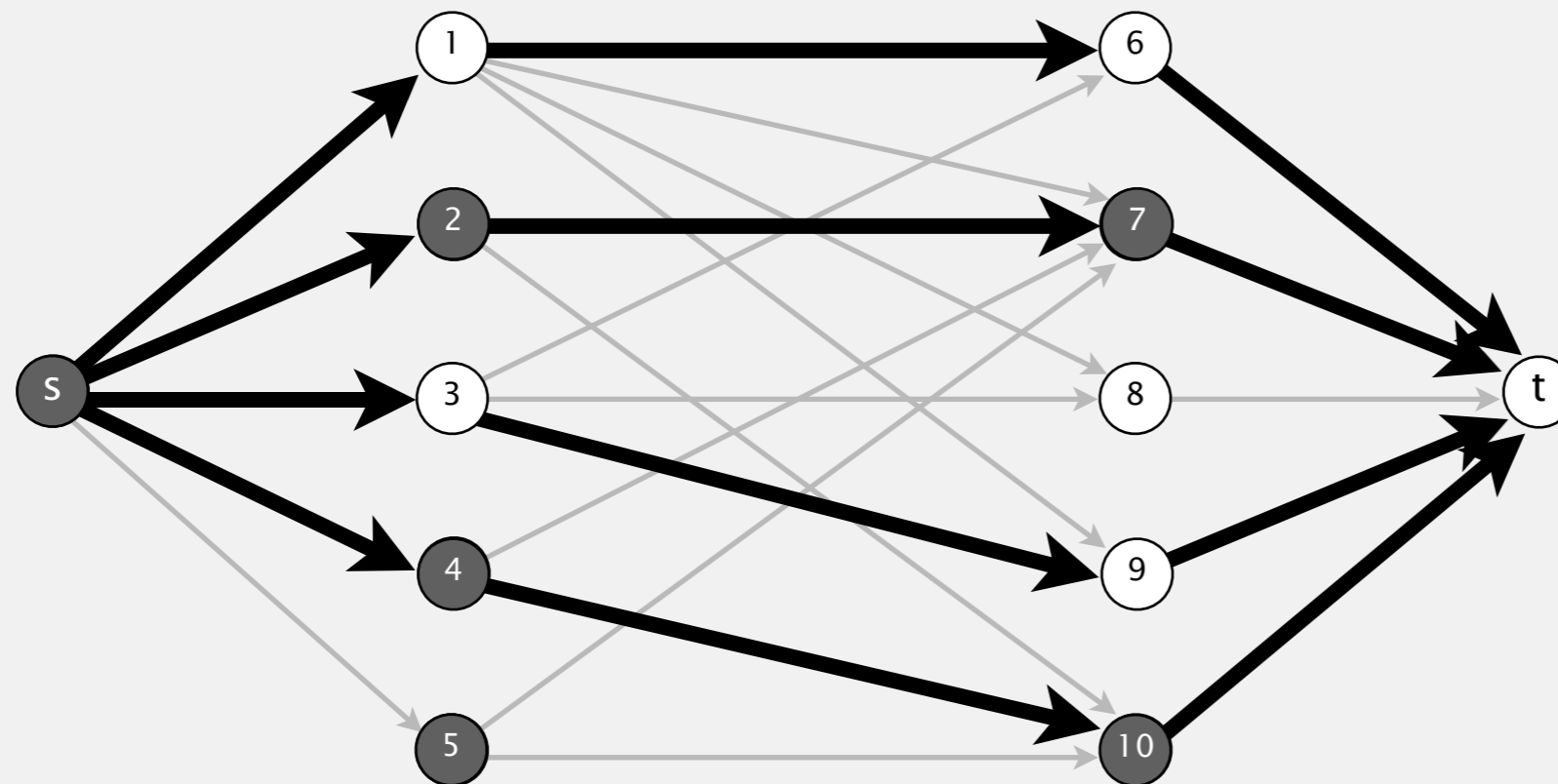
student in  $S$   
can be matched  
only to  
companies in  $T$

$|S| > |T|$

# 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.

# Baseball elimination problem

---

Q. Which teams have a chance of finishing the season with the most wins?

i	team	wins	losses	to play	ATL	PHI	NYM	WAS	
0		Atlanta	83	71	8	–	1	6	1
1		Philly	80	79	3	1	–	0	2
2		New York	78	78	6	6	0	–	0
3		Washington	77	82	3	1	2	0	–


Washington is mathematically eliminated.

- Washington finishes with  $\leq 80$  wins.
- Atlanta already has 83 wins.

# Baseball elimination problem

---

Q. Which teams have a chance of finishing the season with the most wins?

i	team	wins	losses	to play	ATL	PHI	NYM	WAS	
0		Atlanta	83	71	8	–	1	6	1
1		Philly	80	79	3	1	–	0	2
2		New York	78	78	6	6	0	–	0
3		Washington	77	82	3	1	2	0	–






Philadelphia is mathematically eliminated.

- Philadelphia finishes with  $\leq 83$  wins.
- Either New York or Atlanta will finish with  $\geq 84$  wins.

**Observation.** Answer depends not only on how many games already won and left to play, but on **whom** they're against.

# Baseball elimination problem

Q. Which teams have a chance of finishing the season with the most wins?

i	team	wins	losses	to play	NYY	BAL	BOS	TOR	DET
0	 New York	75	59	28	–	3	8	7	3
1	 Baltimore	71	63	28	3	–	2	7	4
2	 Boston	69	66	27	8	2	–	0	0
3	 Toronto	63	72	27	7	7	0	–	0
4	 Detroit	49	86	27	3	4	0	0	–

AL East (August 30, 1996)

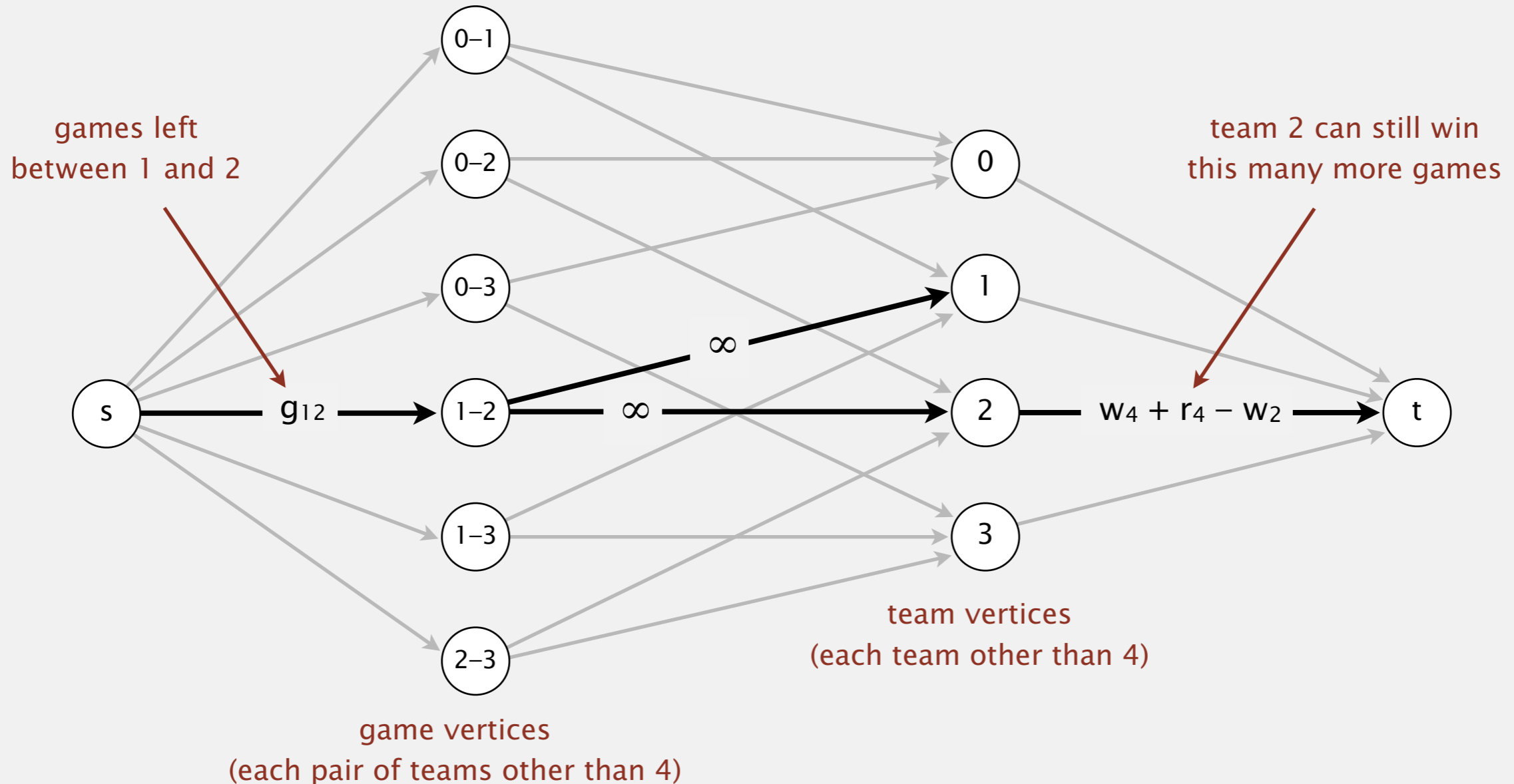
**Detroit is mathematically eliminated.**

- Detroit finishes with  $\leq 76$  wins.
- Wins for  $R = \{ \text{NYY, BAL, BOS, TOR} \} = 278$ .
- Remaining games among  $\{ \text{NYY, BAL, BOS, TOR} \} = 3 + 8 + 7 + 2 + 7 = 27$ .
- Average team in  $R$  wins  $305/4 = 76.25$  games.



# Baseball elimination problem: maxflow formulation

**Intuition.** Remaining games flow from  $s$  to  $t$ .



**Fact.** Team 4 not eliminated iff all edges pointing from  $s$  are full in maxflow.

# Maximum flow algorithms: theory

---

(Yet another) holy grail for theoretical computer scientists.

year	method	worst case	discovered by
1951	<b>simplex</b>	$E^3 U$	Dantzig
1955	<b>augmenting path</b>	$E^2 U$	Ford-Fulkerson
1970	<b>shortest augmenting path</b>	$E^3$	Dinitz, Edmonds-Karp
1970	<b>fattest augmenting path</b>	$E^2 \log E \log(E U)$	Dinitz, Edmonds-Karp
1977	<b>blocking flow</b>	$E^{5/2}$	Cherkasky
1978	<b>blocking flow</b>	$E^{7/3}$	Galil
1983	<b>dynamic trees</b>	$E^2 \log E$	Sleator-Tarjan
1985	<b>capacity scaling</b>	$E^2 \log U$	Gabow
1997	<b>length function</b>	$E^{3/2} \log E \log U$	Goldberg-Rao
2012	<b>compact network</b>	$E^2 / \log E$	Orlin
?	?	$E$	?

maxflow algorithms for sparse networks with  $E$  edges, integer capacities between 1 and  $U$

# Maximum flow algorithms: practice

---

**Warning.** Worst-case order-of-growth is generally not useful for predicting or comparing maxflow algorithm performance in practice.

**Best in practice.** Push-relabel method with gap relabeling:  $E^{3/2}$ .

## On Implementing Push-Relabel Method for the Maximum Flow Problem

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<sup>2</sup> Computer Science Department, Stanford University  
Stanford, CA 94305, USA  
*goldberg@cs.stanford.edu*

**Abstract.** We study efficient implementations of the push-relabel method for the maximum flow problem. The resulting codes are faster than the previous codes, and much faster on some problem families. The speedup is due to the combination of heuristics used in our implementations. We also exhibit a family of problems for which the running time of all known methods seem to have a roughly quadratic growth rate.



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OF OPERATIONAL  
RESEARCH

Theory and Methodology

## Computational investigations of maximum flow algorithms

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Received 30 August 1995; accepted 27 June 1996

# 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!