



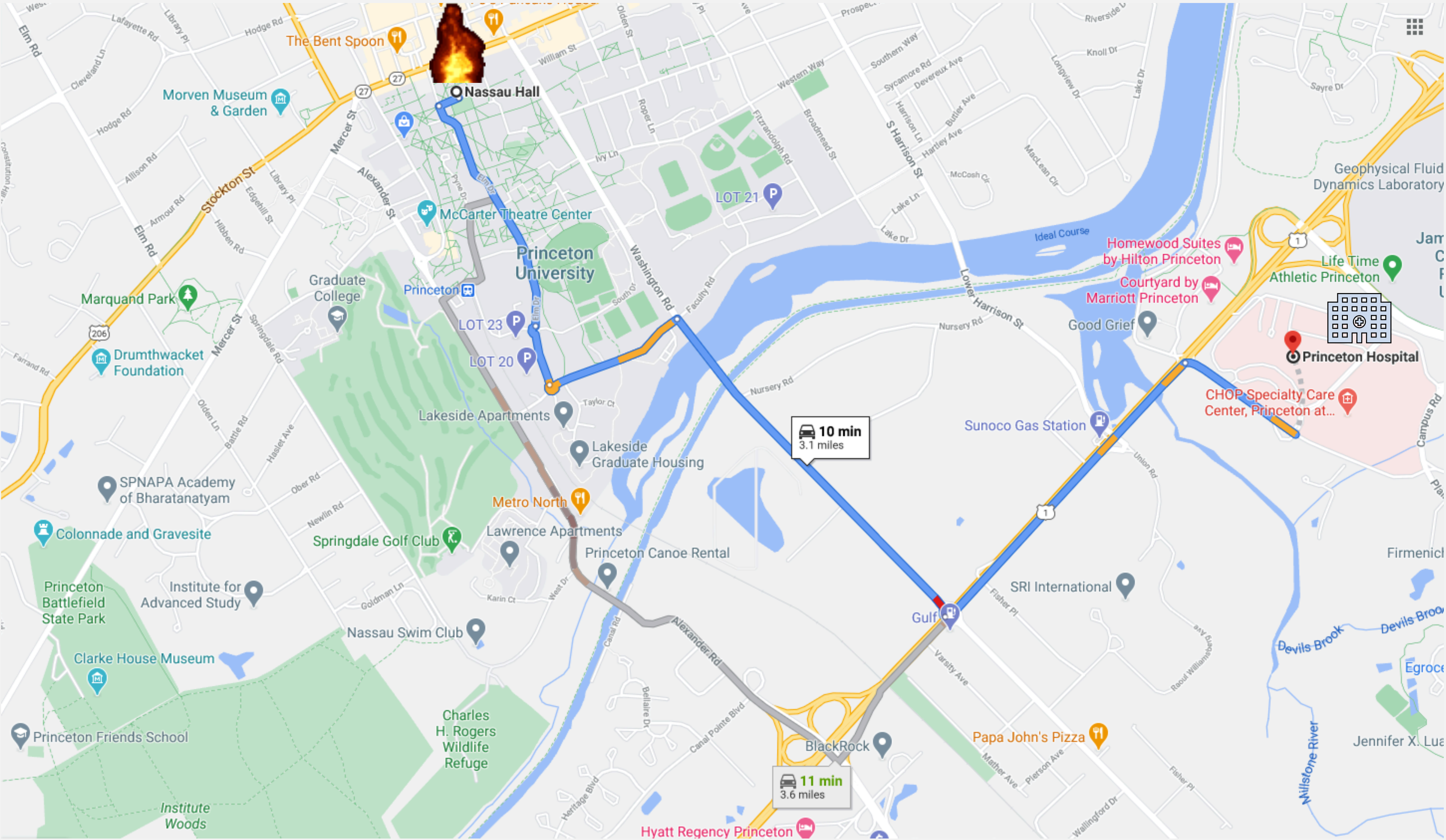
<https://algs4.cs.princeton.edu>

## 4.4 SHORTEST PATHS

---

- *properties*
- *APIs*
- *Bellman–Ford algorithm*
- *Dijkstra’s algorithm*

# Google maps

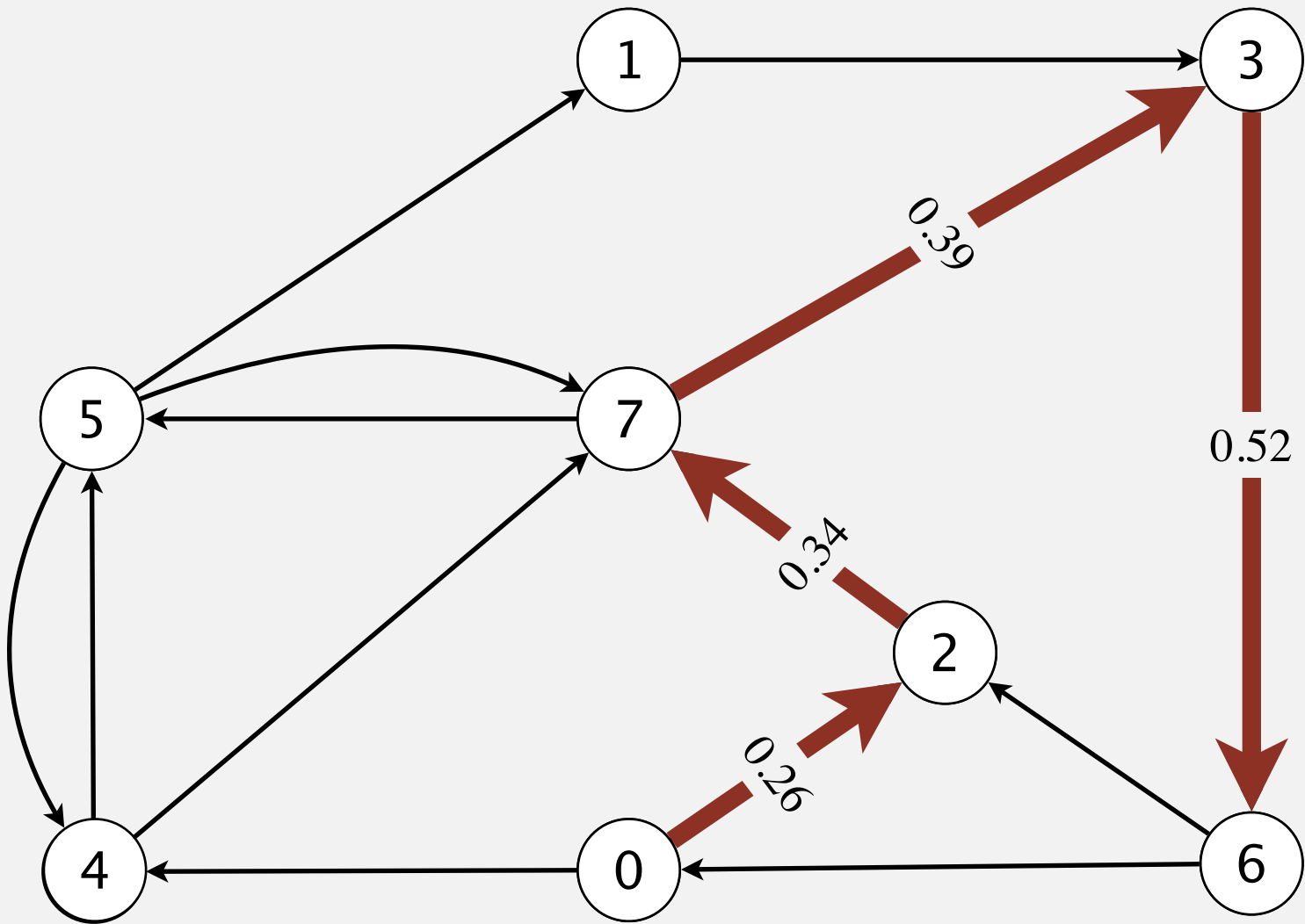


# Shortest path in an edge-weighted digraph

Given an edge-weighted digraph, find a shortest path from one vertex to another vertex.

edge-weighted digraph

4->5	0.35
5->4	0.35
4->7	0.37
5->7	0.28
7->5	0.28
5->1	0.32
0->4	0.38
0->2	0.26
7->3	0.39
1->3	0.29
2->7	0.34
6->2	0.40
3->6	0.52
6->0	0.58
6->4	0.93



shortest path from 0 to 6

0 → 2 → 7 → 3 → 6

length of path = 1.51

(0.26 + 0.34 + 0.39 + 0.52)

# Shortest path applications

---

- PERT/CPM.
- Map routing.
- Seam carving. ← see Assignment 6
- Texture mapping.
- Robot navigation.
- Typesetting in  $\text{T}_{\text{E}}\text{X}$ .
- Currency exchange.
- Urban traffic planning.
- Optimal pipelining of VLSI chip.
- Telemarketer operator scheduling.
- Routing of telecommunications messages.
- Network routing protocols (OSPF, BGP, RIP).
- Optimal truck routing through given traffic congestion pattern.



[https://en.wikipedia.org/wiki/Seam\\_carving](https://en.wikipedia.org/wiki/Seam_carving)

Reference: Network Flows: Theory, Algorithms, and Applications, R. K. Ahuja, T. L. Magnanti, and J. B. Orlin, Prentice Hall, 1993.

# Shortest path variants

---

## Which vertices?

- Source-destination: from one vertex to another vertex.
- Single source: from one vertex to every vertex.
- Single destination: from every vertex to one vertex.
- All pairs: between all pairs of vertices.

## Restrictions on edge weights?

- Non-negative weights.

← we assume this in today's lecture  
(except as noted)

- Euclidean weights.
- Arbitrary weights.

## Directed cycles?

- Prohibit.

← can derive faster algorithms  
(see dynamic programming lecture)

- Allow.

implies that shortest path from  $s$  to  $v$  exists  
(and that  $E \geq V - 1$ )

**Simplifying assumption.** Each vertex is reachable from  $s$ .



**Which variant in car GPS? Hint: drivers make wrong turns occasionally.**

- A.** Source-destination: from one vertex to another vertex.
- B.** Single source: from one vertex to every vertex.
- C.** Single destination: from every vertex to one vertex.
- D.** All pairs: between all pairs of vertices.





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## 4.4 SHORTEST PATHS

---

- *properties*
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- *Bellman–Ford algorithm*
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# Data structures for single-source shortest paths

**Goal.** Find a shortest path from  $s$  to every vertex.

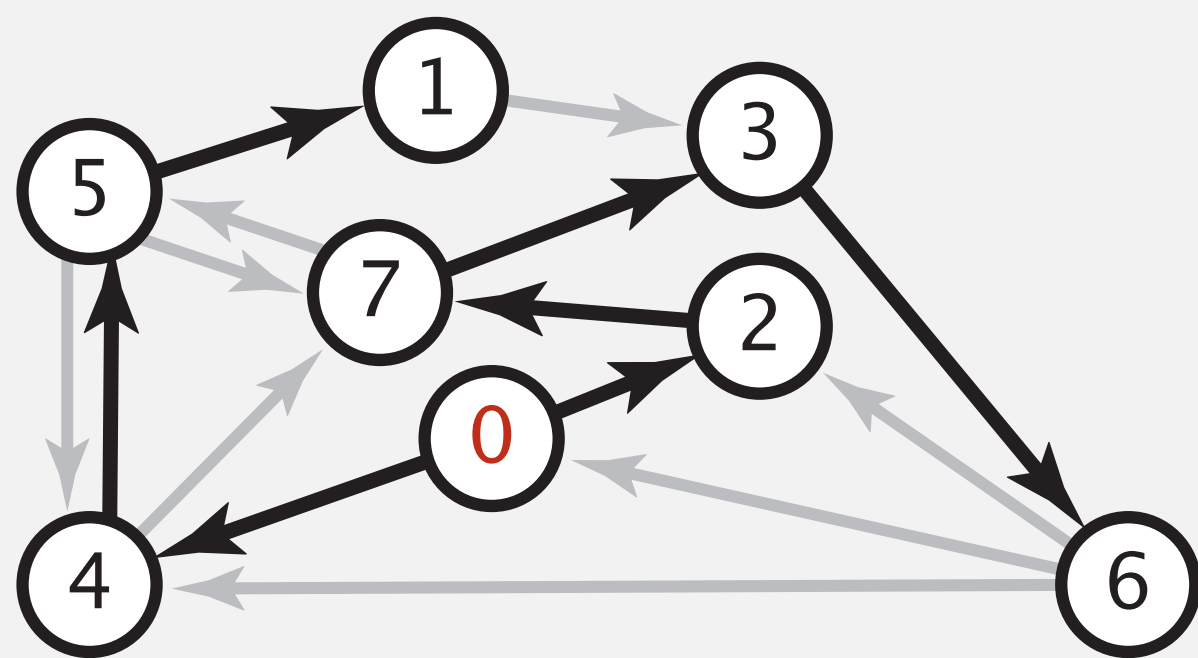
no repeated vertices  
 $\Rightarrow \leq V - 1$  edges

**Observation 1.** There exists a shortest path from  $s$  to  $v$  that is simple.

**Observation 2.** A **shortest-paths tree** (SPT) solution exists. Why?

**Consequence.** Can represent a SPT with two vertex-indexed arrays:

- $\text{distTo}[v]$  is length of a shortest path from  $s$  to  $v$ .
- $\text{edgeTo}[v]$  is last edge on a shortest path from  $s$  to  $v$ .



shortest-paths tree from 0

	distTo[]	edgeTo[]
0	0	null
1	1.05	5->1 0.32
2	0.26	0->2 0.26
3	0.97	7->3 0.37
4	0.38	0->4 0.38
5	0.73	4->5 0.35
6	1.49	3->6 0.52
7	0.60	2->7 0.34

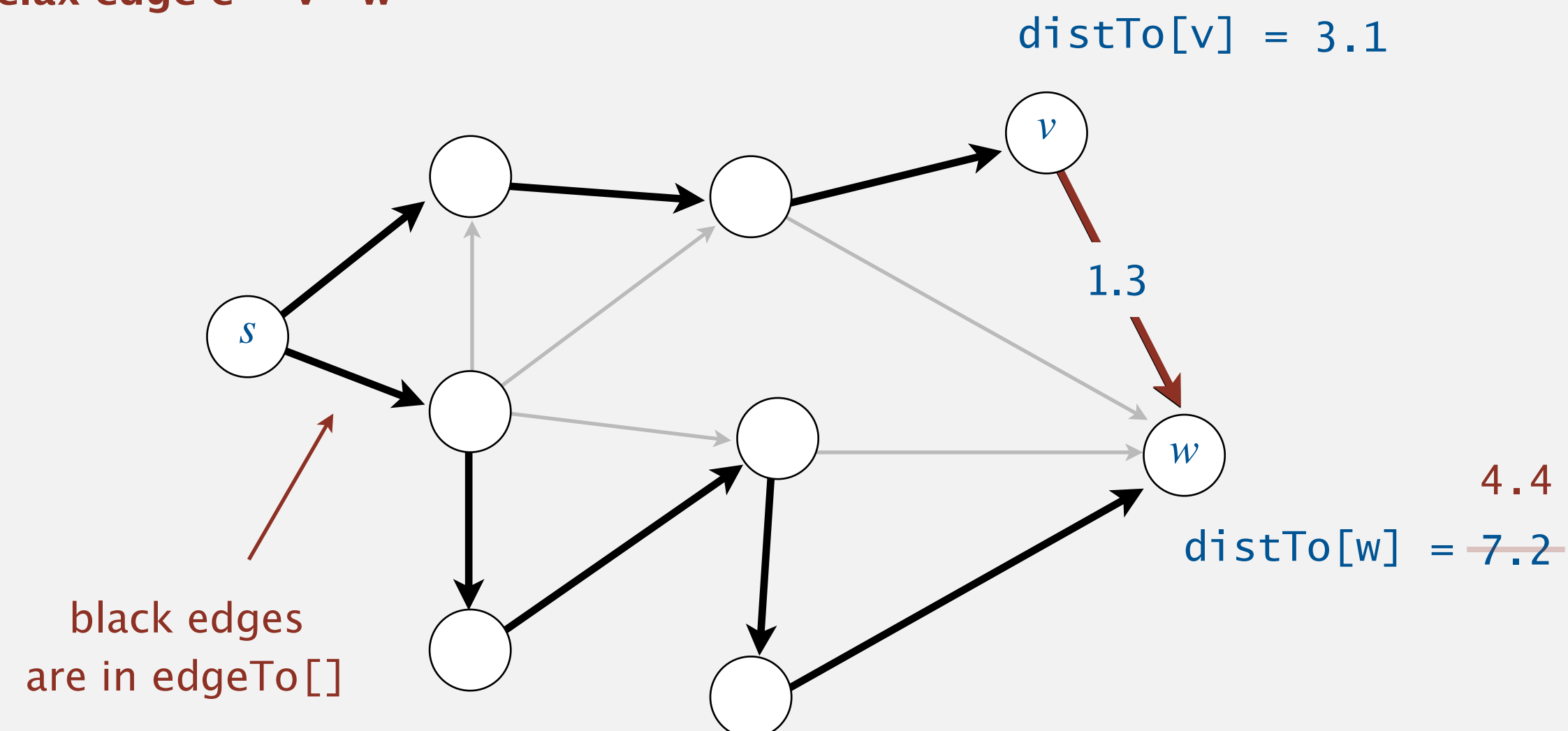
parent-link representation

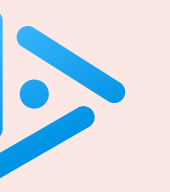
# Edge relaxation

Relax edge  $e = v \rightarrow w$ .

- $\text{distTo}[v]$  is length of shortest **known** path from  $s$  to  $v$ .
- $\text{distTo}[w]$  is length of shortest **known** path from  $s$  to  $w$ .
- $\text{edgeTo}[w]$  is last edge on shortest **known** path from  $s$  to  $w$ .
- If  $e = v \rightarrow w$  yields shorter path from  $s$  to  $w$ , via  $v$ , update  $\text{distTo}[w]$  and  $\text{edgeTo}[w]$ .

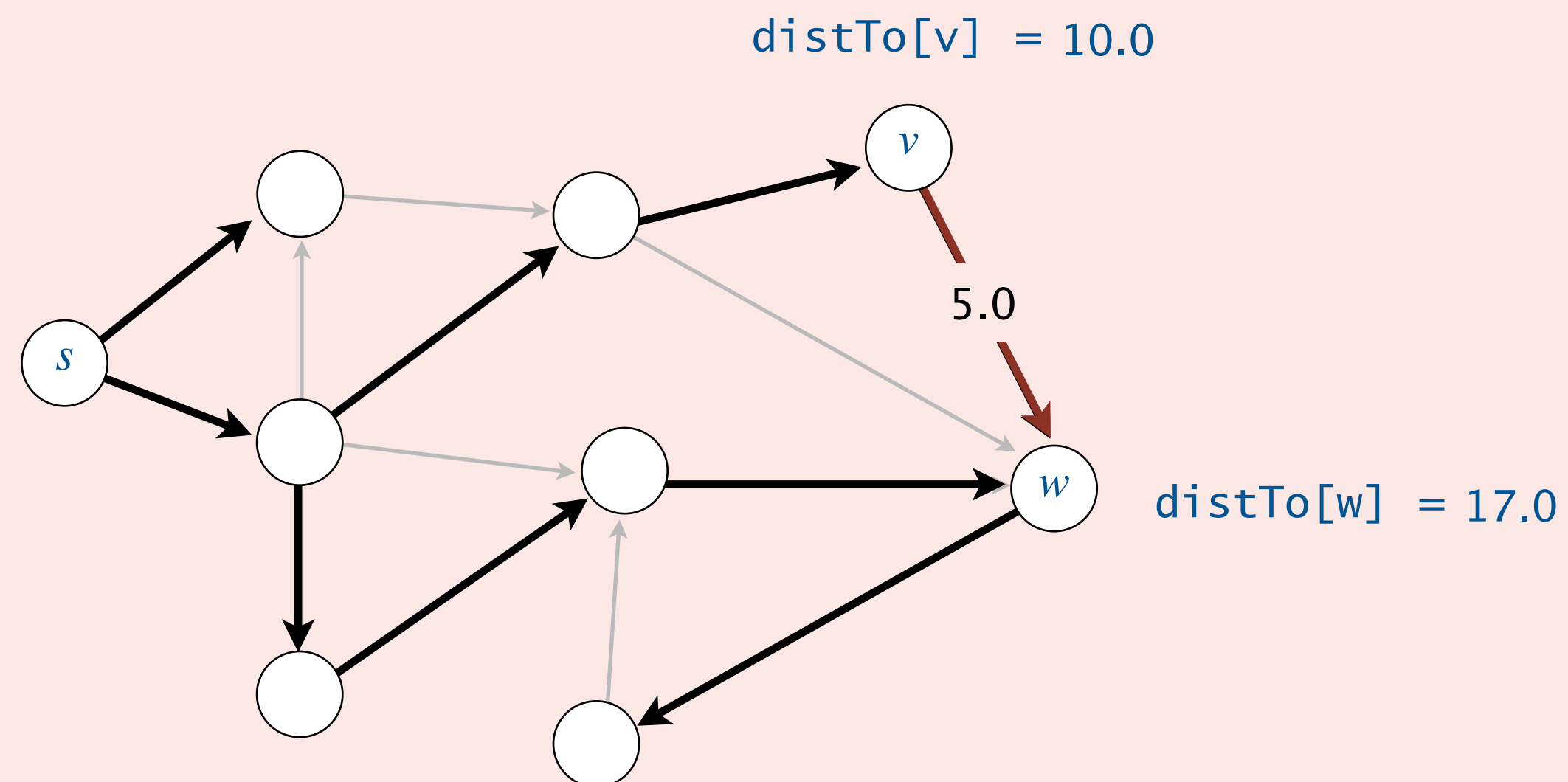
relax edge  $e = v \rightarrow w$





What are the values of  $\text{distTo}[v]$  and  $\text{distTo}[w]$  after relaxing edge  $e = v \rightarrow w$  ?

- A. 10.0 and 15.0
- B. 10.0 and 17.0
- C. 12.0 and 15.0
- D. 12.0 and 17.0



# Framework for shortest-paths algorithm

---

## Generic algorithm (to compute a SPT from $s$ )

---

For each vertex  $v$ :  $\text{distTo}[v] = \infty$ .

For each vertex  $v$ :  $\text{edgeTo}[v] = \text{null}$ .

$\text{distTo}[s] = 0$ .

Repeat until  $\text{distTo}[v]$  values converge:

- Relax any edge.
- 

**Key properties.** Throughout the generic algorithm,

- $\text{distTo}[v]$  is either infinity or the length of a (simple) path from  $s$  to  $v$ .
- $\text{distTo}[v]$  does not increase.

# Framework for shortest-paths algorithm

---

## Generic algorithm (to compute a SPT from $s$ )

---

For each vertex  $v$ :  $\text{distTo}[v] = \infty$ .

For each vertex  $v$ :  $\text{edgeTo}[v] = \text{null}$ .

$\text{distTo}[s] = 0$ .

Repeat until  $\text{distTo}[v]$  values converge:

- Relax any edge.
- 

## Efficient implementations.

- Which edge to relax next?
- How many edge relaxations needed to guarantee convergence?

Ex 1. Bellman-Ford algorithm.

Ex 2. Dijkstra's algorithm.

Ex 3. Topological sort algorithm.



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## 4.4 SHORTEST PATHS

---

- *properties*
- *APIs*
- *Bellman–Ford algorithm*
- *Dijkstra’s algorithm*

# Weighted directed edge API

```
public class DirectedEdge
```

```
    DirectedEdge(int v, int w, double weight)
```

*weighted edge  $v \rightarrow w$*

```
    int from()
```

*vertex  $v$*

```
    int to()
```

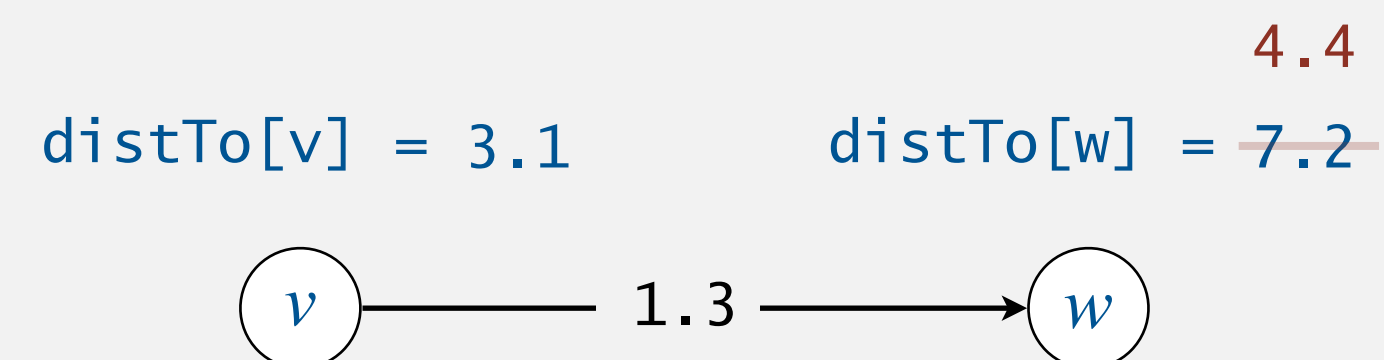
*vertex  $w$*

```
    double weight()
```

*weight of this edge*

Relaxing an edge  $e = v \rightarrow w$ .

```
private void relax(DirectedEdge e)
{
    int v = e.from(), w = e.to();
    if (distTo[w] > distTo[v] + e.weight())
    {
        distTo[w] = distTo[v] + e.weight();
        edgeTo[w] = e;
    }
}
```



# Weighted directed edge: implementation in Java

---

**API.** Similar to [Edge](#) for undirected graphs, but a bit simpler.

```
public class DirectedEdge
{
    private final int v, w;
    private final double weight;

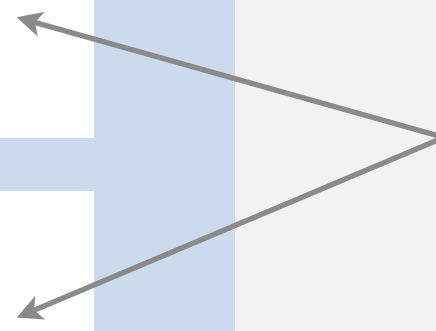
    public DirectedEdge(int v, int w, double weight)
    {
        this.v = v;
        this.w = w;
        this.weight = weight;
    }

    public int from()
    { return v; }

    public int to()
    { return w; }

    public double weight()
    { return weight; }

}
```



from() and to() replace  
either() and other()

# Edge-weighted digraph API

---

**API.** Same as `EdgeWeightedGraph` except with `DirectedEdge` objects.

```
public class EdgeWeightedDigraph
```

```
    EdgeWeightedDigraph(int V)    edge-weighted digraph with V vertices
```

```
    void addEdge(DirectedEdge e)    add weighted directed edge e
```

```
    Iterable<DirectedEdge> adj(int v)    edges incident from v
```

```
    int V()    number of vertices
```

```
    ⋮
```

```
    ⋮
```

# Edge-weighted digraph: adjacency-lists implementation in Java

---

**Implementation.** Almost identical to EdgeWeightedGraph.

```
public class EdgeWeightedDigraph
{
    private final int V;
    private final Bag<DirectedEdge>[] adj;
```

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

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

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

```
}
```

← add edge  $e = v \rightarrow w$  to  
only  $v$ 's adjacency list

# Single-source shortest paths API

---

**Goal.** Find the shortest path from  $s$  to every other vertex.

```
public class SP
```

```
    SP(EdgeWeightedDigraph G, int s)    shortest paths from s in digraph G
```

```
    double distTo(int v)                length of shortest path from s to v
```

```
    Iterable <DirectedEdge> pathTo(int v) shortest path from s to v
```

```
    boolean hasPathTo(int v)            is there a path from s to v?
```



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## 4.4 SHORTEST PATHS

---

- *properties*
- *APIs*
- *Bellman–Ford algorithm*
- *Dijkstra’s algorithm*

# Bellman–Ford algorithm

## Bellman–Ford algorithm

For each vertex  $v$ :  $\text{distTo}[v] = \infty$ .

For each vertex  $v$ :  $\text{edgeTo}[v] = \text{null}$ .

$\text{distTo}[s] = 0$ .

Repeat  $V-1$  times:

- Relax each edge.

```
private void relax(DirectedEdge e)
{
    int v = e.from(), w = e.to();
    if (distTo[w] > distTo[v] + e.weight())
    {
        distTo[w] = distTo[v] + e.weight();
        edgeTo[w] = e;
    }
}
```

```
for (int i = 1; i < G.V(); i++)
    for (int v = 0; v < G.V(); v++)
        for (DirectedEdge e : G.adj(v))
            relax(e);
```

← pass  $i$  (relax each edge once)

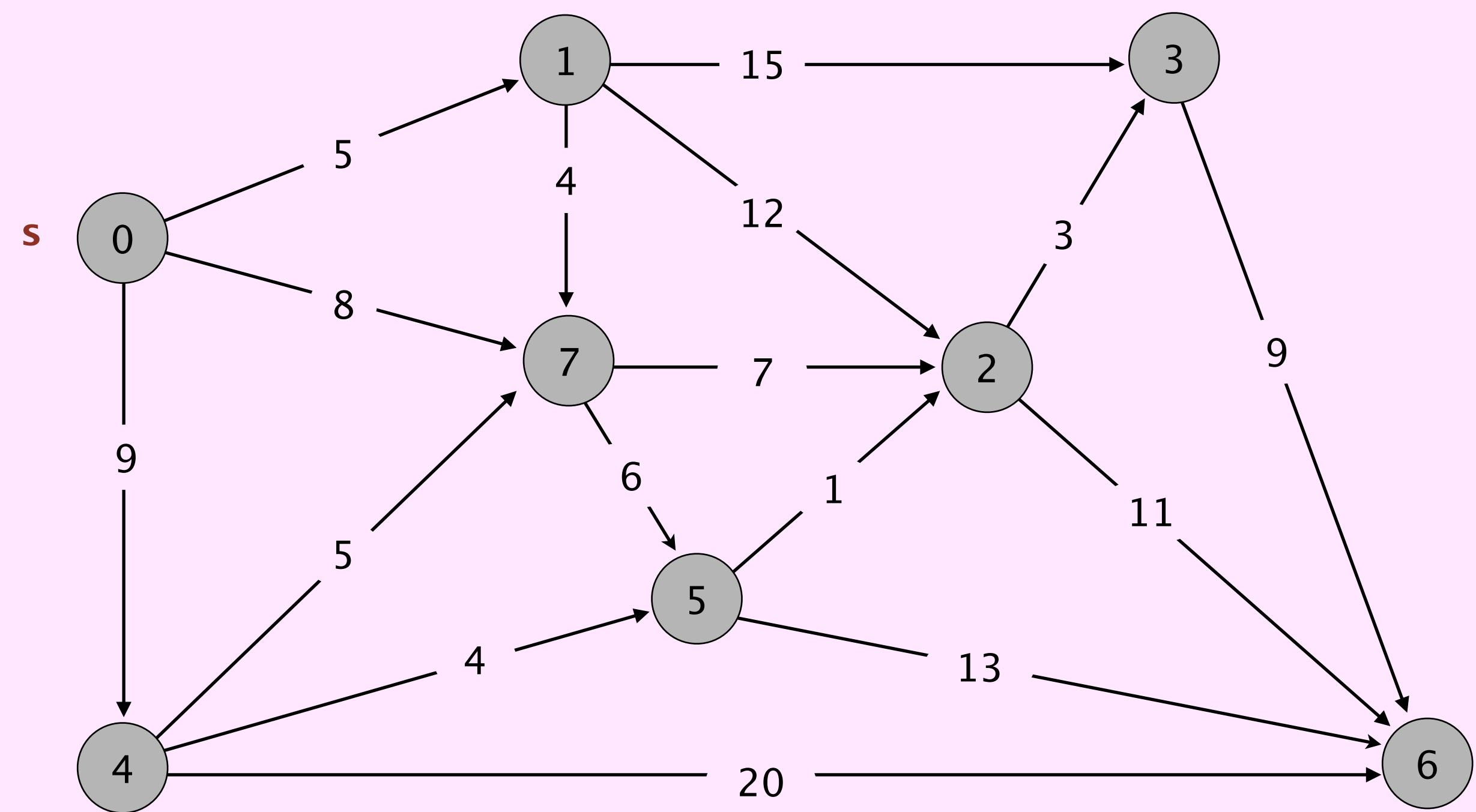
number of calls to `relax()` in pass  $i$  =  
 $\text{outdegree}(0) + \text{outdegree}(1) + \text{outdegree}(2) + \dots = E$

**Running time.** Algorithm takes  $\Theta(E V)$  time and uses  $\Theta(V)$  extra space.

# Bellman–Ford algorithm demo



Repeat  $V - 1$  times: relax all  $E$  edges.



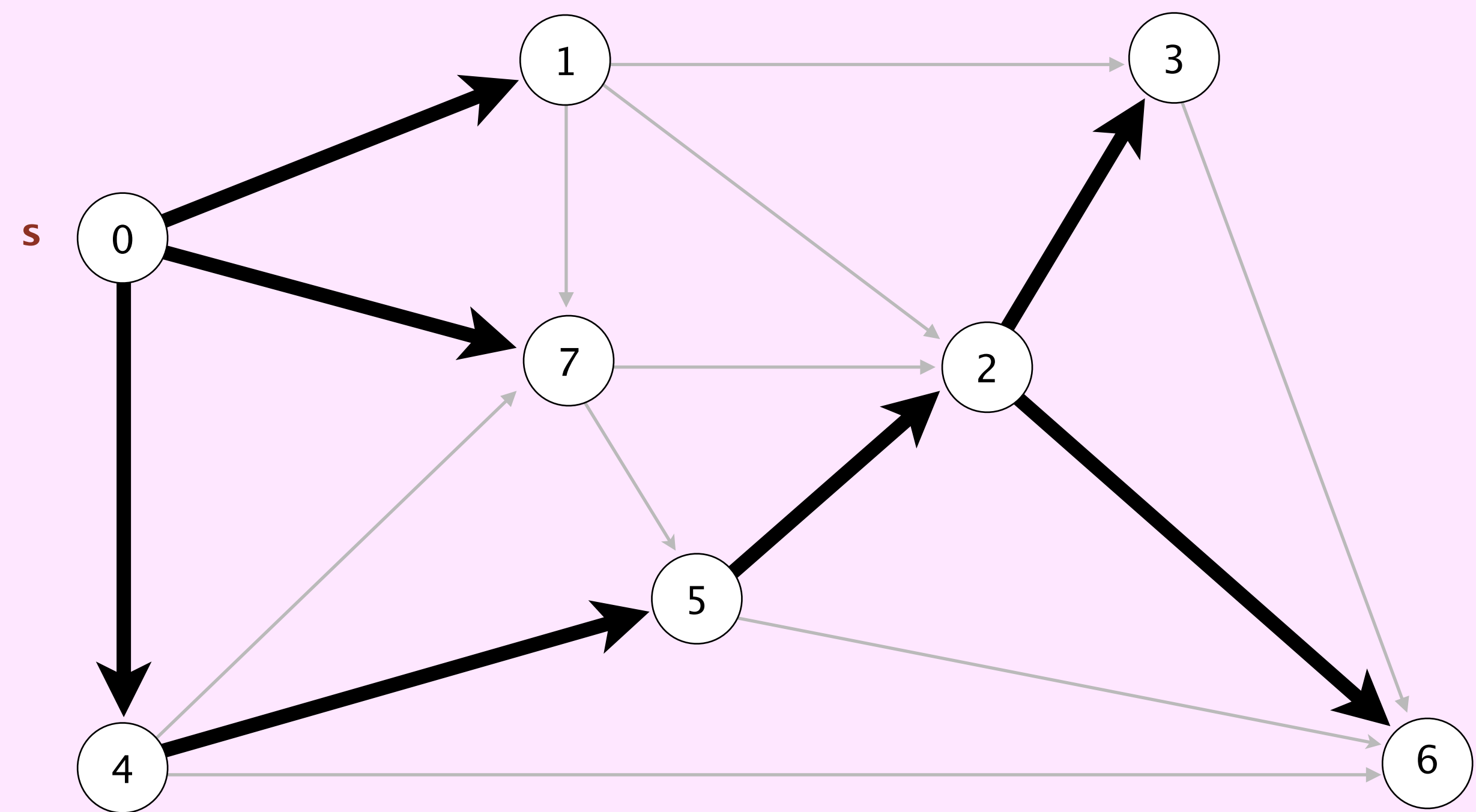
an edge-weighted digraph

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# Bellman–Ford algorithm demo



Repeat  $V - 1$  times: relax all  $E$  edges.



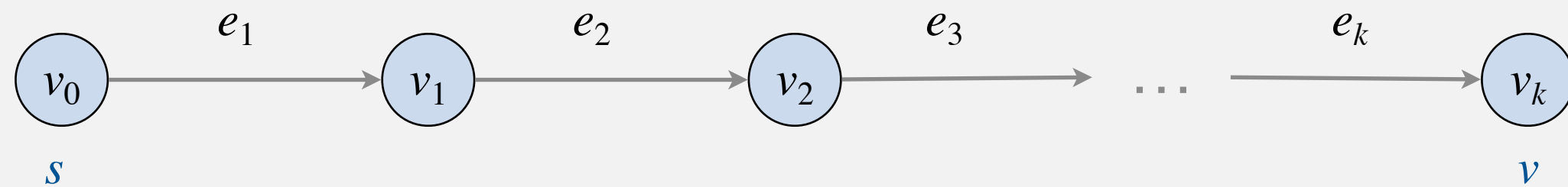
v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0	2→6
7	8.0	0→7

shortest-paths tree from vertex s

# Bellman–Ford algorithm: correctness proof

**Proposition.** Let  $s = v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_k = v$  be any path from  $s$  to  $v$  containing  $k$  edges.

Then, after pass  $k$ ,  $\text{distTo}[v_k] \leq \text{weight}(e_1) + \text{weight}(e_2) + \dots + \text{weight}(e_k)$ .



**Pf.** [ by induction on number of passes  $i$  ]

- Base case: initially,  $0 = \text{distTo}[v_0] \leq 0$ .
- Inductive hypothesis: after pass  $i$ ,  $\text{distTo}[v_i] \leq \text{weight}(e_1) + \text{weight}(e_2) + \dots + \text{weight}(e_i)$ .
- This inequality continues to hold because  $\text{distTo}[v_i]$  cannot increase.
- Immediately after relaxing edge  $e_{i+1}$  in pass  $i+1$ , we have

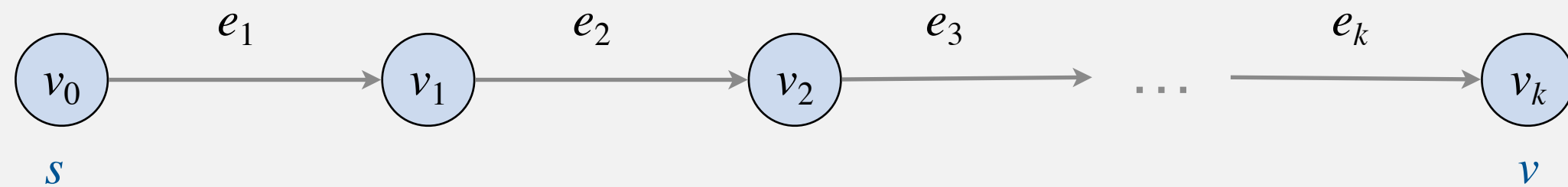
$$\begin{aligned} \text{distTo}[v_{i+1}] &\leq \text{distTo}[v_i] + \text{weight}(e_{i+1}) && \longleftarrow \text{edge relaxation} \\ &\leq \text{weight}(e_1) + \text{weight}(e_2) + \dots + \text{weight}(e_i) + \text{weight}(e_{i+1}). && \longleftarrow \text{inductive hypothesis} \end{aligned}$$

- This inequality continues to hold because  $\text{distTo}[v_{i+1}]$  cannot increase. ■

# Bellman–Ford algorithm: correctness proof

---

**Proposition.** Let  $s = v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_k = v$  be any path from  $s$  to  $v$  containing  $k$  edges. Then, after pass  $k$ ,  $\text{distTo}[v_k] \leq \text{weight}(e_1) + \text{weight}(e_2) + \dots + \text{weight}(e_k)$ .



**Corollary.** Bellman–Ford computes shortest path distances.

**Pf.** [ apply Proposition to a shortest path from  $s$  to  $v$  ]

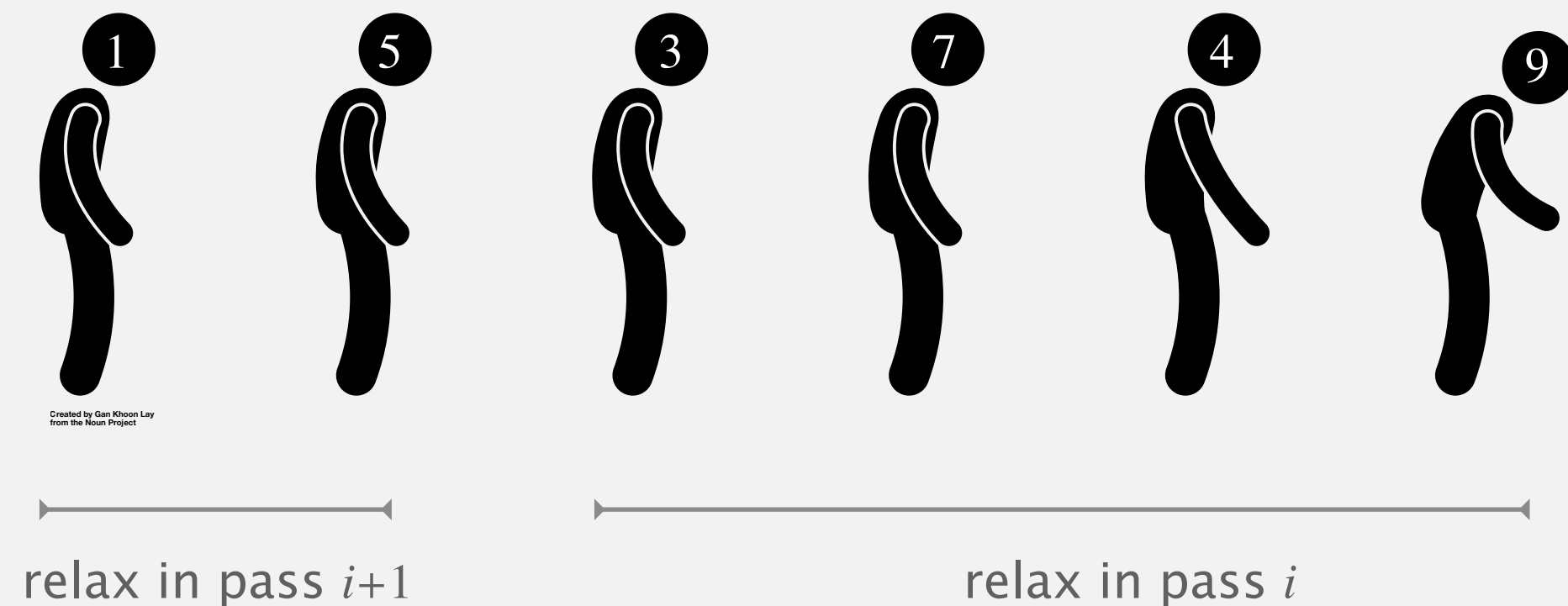
- There exists a simple shortest path  $P^*$  from  $s$  to  $v$ ; it contains  $k \leq V - 1$  edges.
- The Proposition implies that, after at most  $V - 1$  passes,  $\text{distTo}[v] \leq \text{length}(P^*)$ .
- Since  $\text{distTo}[v]$  is the length of some path from  $s$  to  $v$ ,  $\text{distTo}[v] = \text{length}(P^*)$ . ■

# Bellman–Ford algorithm: practical improvement

**Observation.** If `distTo[v]` does not change during pass  $i$ ,  
not necessary to relax any edges incident from  $v$  in pass  $i + 1$ .

## Queue-based implementation of Bellman–Ford.

- Perform **vertex** relaxations.  $\longleftarrow$  relax vertex  $v$  = relax all edges incident from  $v$
- Maintain **queue** of vertices whose `distTo[]` values changed since it was last relaxed.



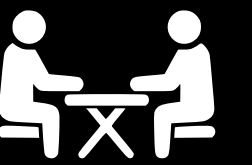
must ensure each vertex is on queue at most once  
(or exponential blowup!)

relax vertex  $v$

## Impact.

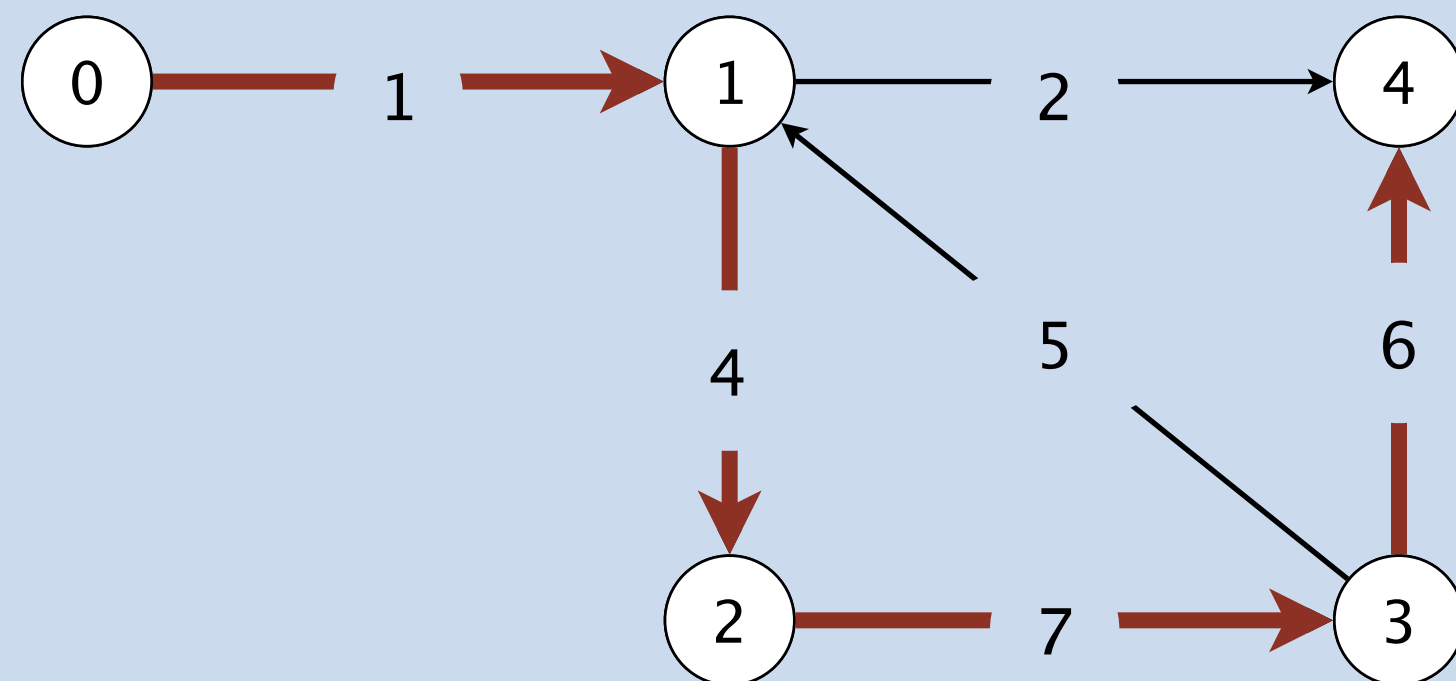
- In the worst case, the running time is still  $\Theta(E V)$ .
- But much faster in practice on typical inputs.

# LONGEST PATH



**Problem.** Given a digraph  $G$  with positive edge weights and vertex  $s$ , find a **longest simple path** from  $s$  to every other vertex.

**Goal.** Design algorithm that takes  $\Theta(E V)$  time in the worst case.



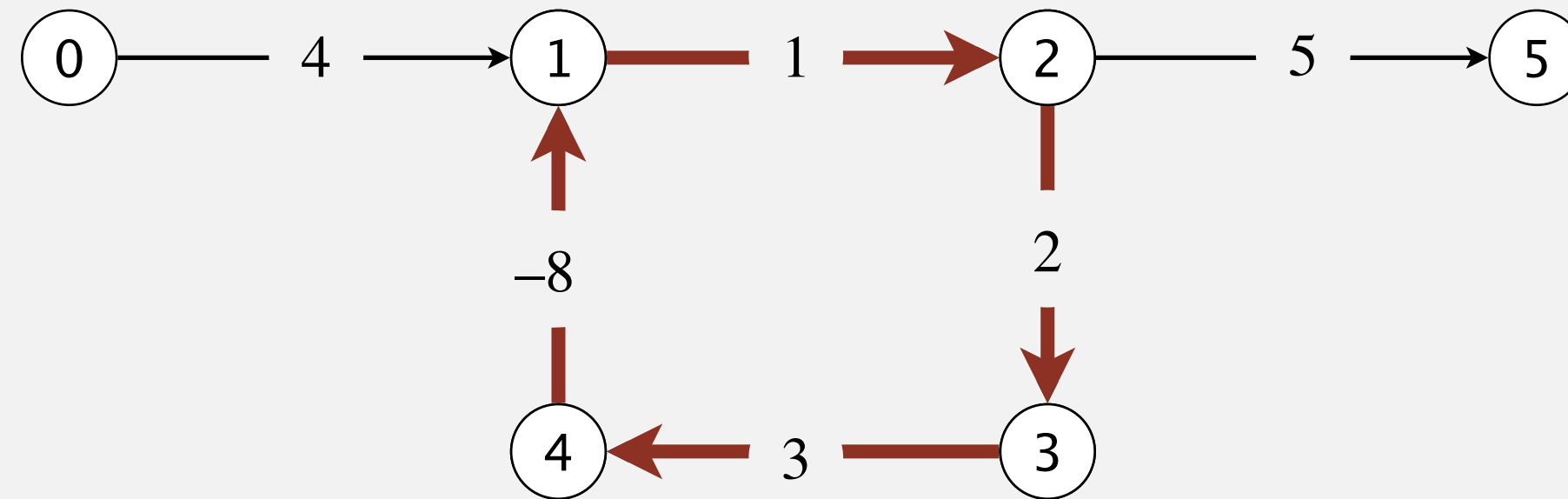
longest simple path from 0 to 4:  $0 \rightarrow 1 \rightarrow 2 \rightarrow 3 \rightarrow 4$

# Bellman–Ford algorithm: negative weights

---

**Remark.** The Bellman–Ford algorithm works even if some weights are negative, provided there are no **negative cycles**.

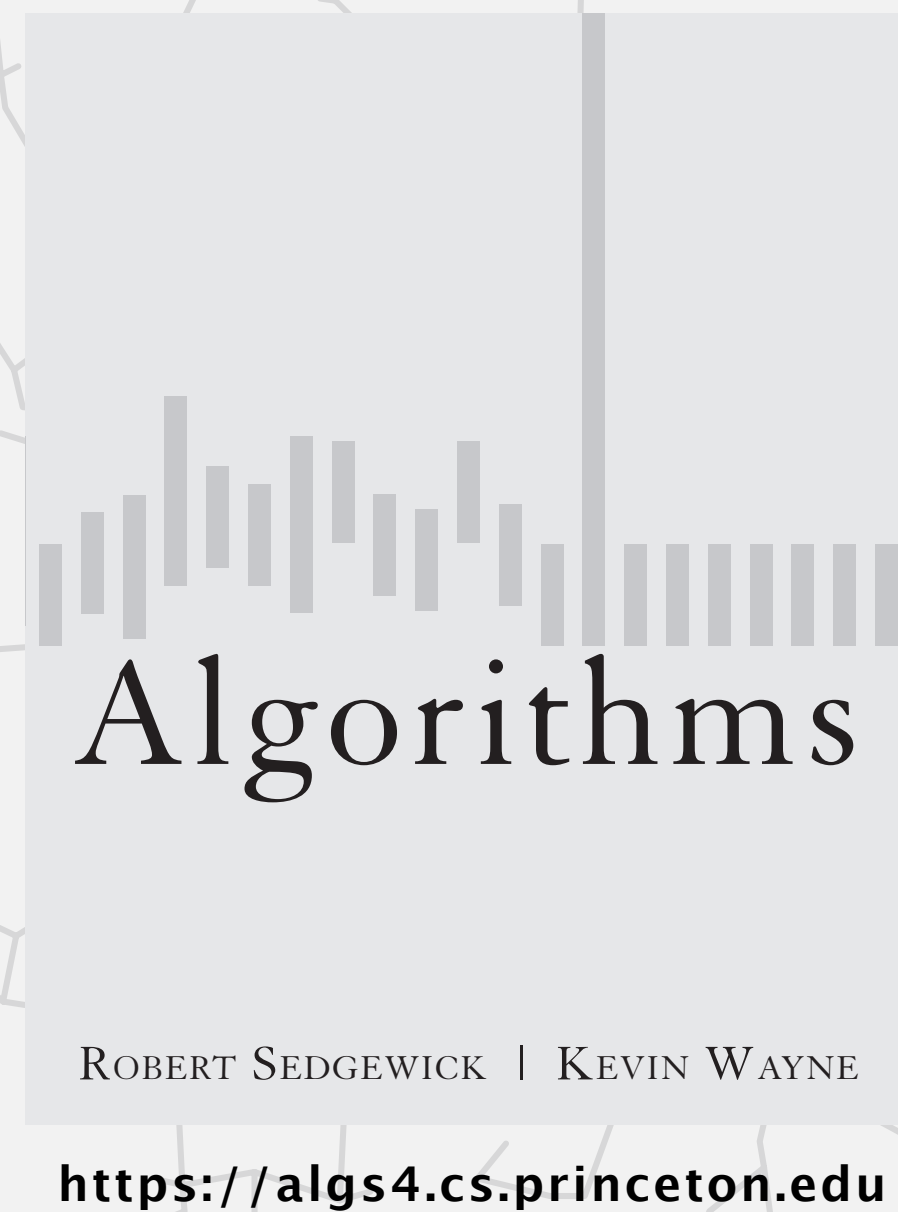
**Negative cycle.** A directed cycle whose length is negative.



$$\text{length of negative cycle} = 1 + 2 + 3 + -8 = -2$$

**Negative cycles and shortest paths.** Length of path can be made arbitrarily negative by using negative cycle.

$$0 \rightarrow 1 \rightarrow 2 \rightarrow 3 \rightarrow 4 \rightarrow 1 \rightarrow \dots \rightarrow 2 \rightarrow 3 \rightarrow 4 \rightarrow 1 \rightarrow 2 \rightarrow 5$$



## 4.4 SHORTEST PATHS

---

- *properties*
- *APIs*
- *Bellman–Ford algorithm*
- *Dijkstra's algorithm*

## Edsger W. Dijkstra: select quote

---



# Dijkstra's algorithm

---

## Dijkstra's algorithm

---

For each vertex  $v$ :  $\text{distTo}[v] = \infty$ .

For each vertex  $v$ :  $\text{edgeTo}[v] = \text{null}$ .

$T = \emptyset$ .

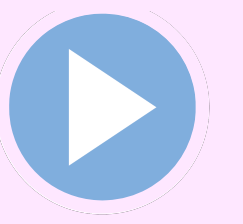
$\text{distTo}[s] = 0$ .

Repeat until all vertices are marked:

- Select unmarked vertex  $v$  with the smallest  $\text{distTo}[]$  value.
  - Mark  $v$ .
  - Relax each edge incident from  $v$ .
- 

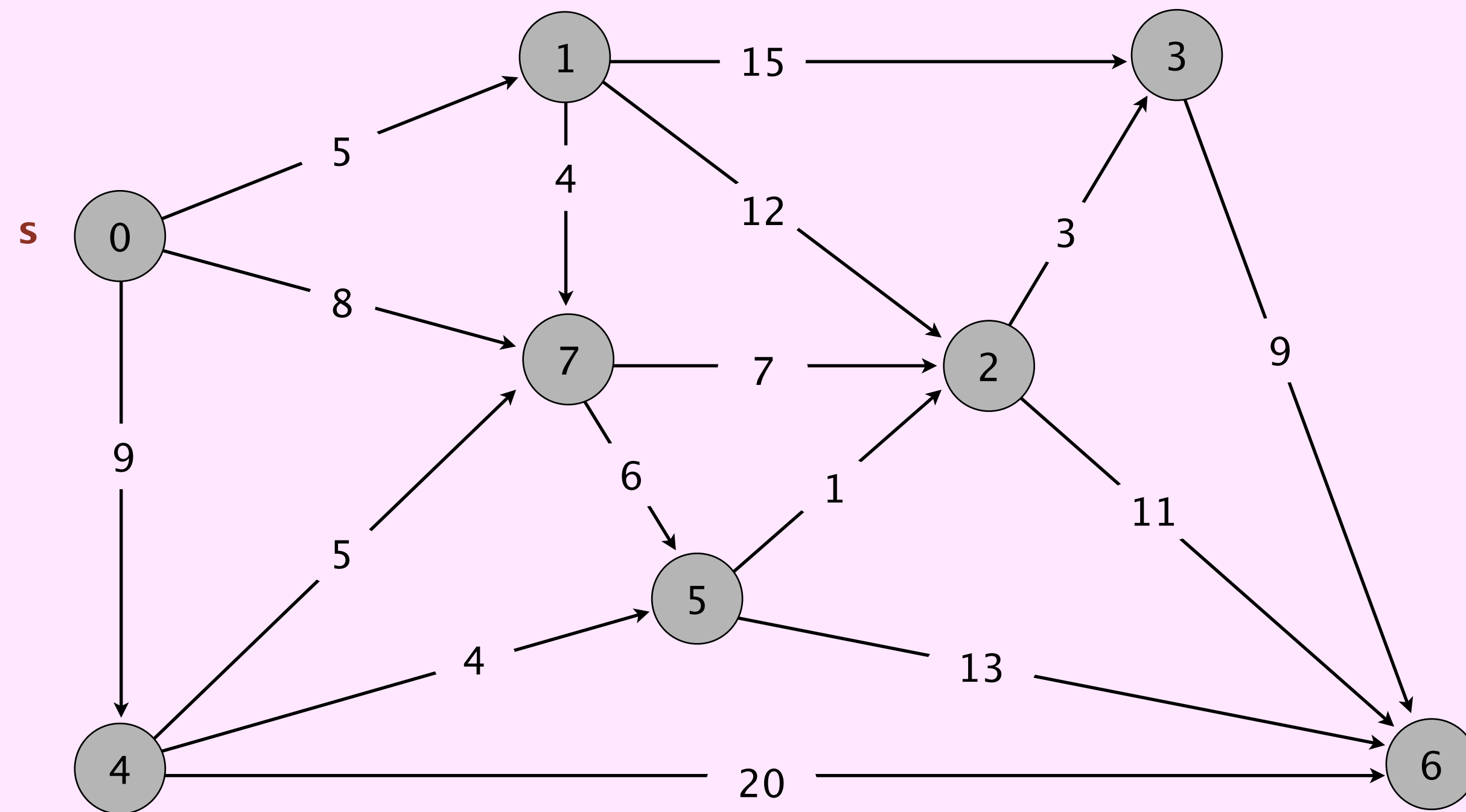
Key difference with Bellman–Ford. Each edge gets relaxed exactly once!

# Dijkstra's algorithm demo



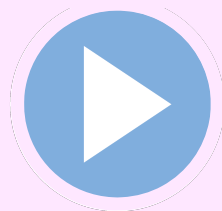
Repeat until all vertices are marked:

- Select unmarked vertex  $v$  with the smallest `distTo[]` value.
- Mark  $v$  and relax all edges incident from  $v$ .



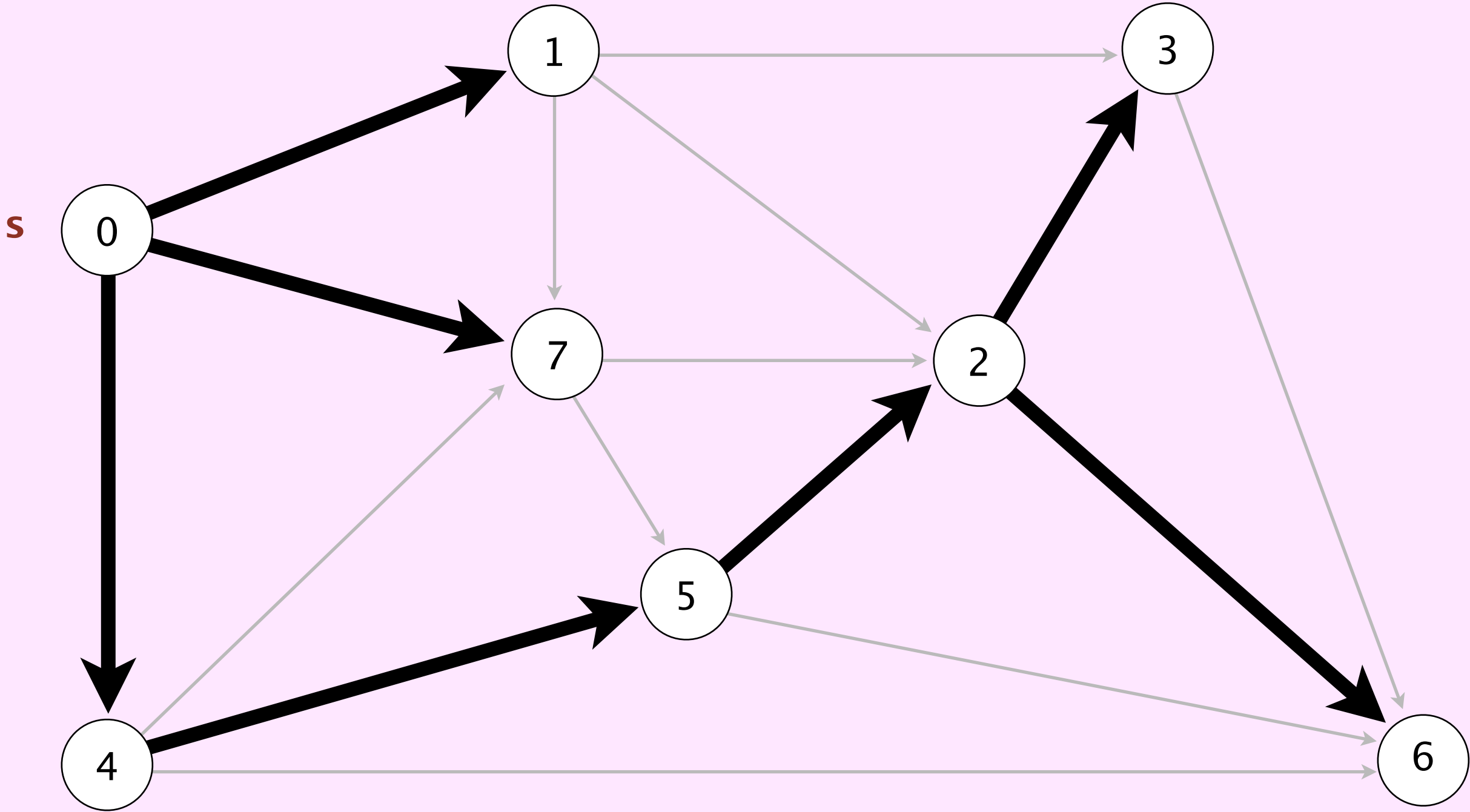
an edge-weighted digraph

0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0



Repeat until all vertices are marked:

- Select unmarked vertex  $v$  with the smallest `distTo[]` value.
- Mark  $v$  and relax all edges incident from  $v$ .



v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0	2→6
7	8.0	0→7

shortest-paths tree from vertex s

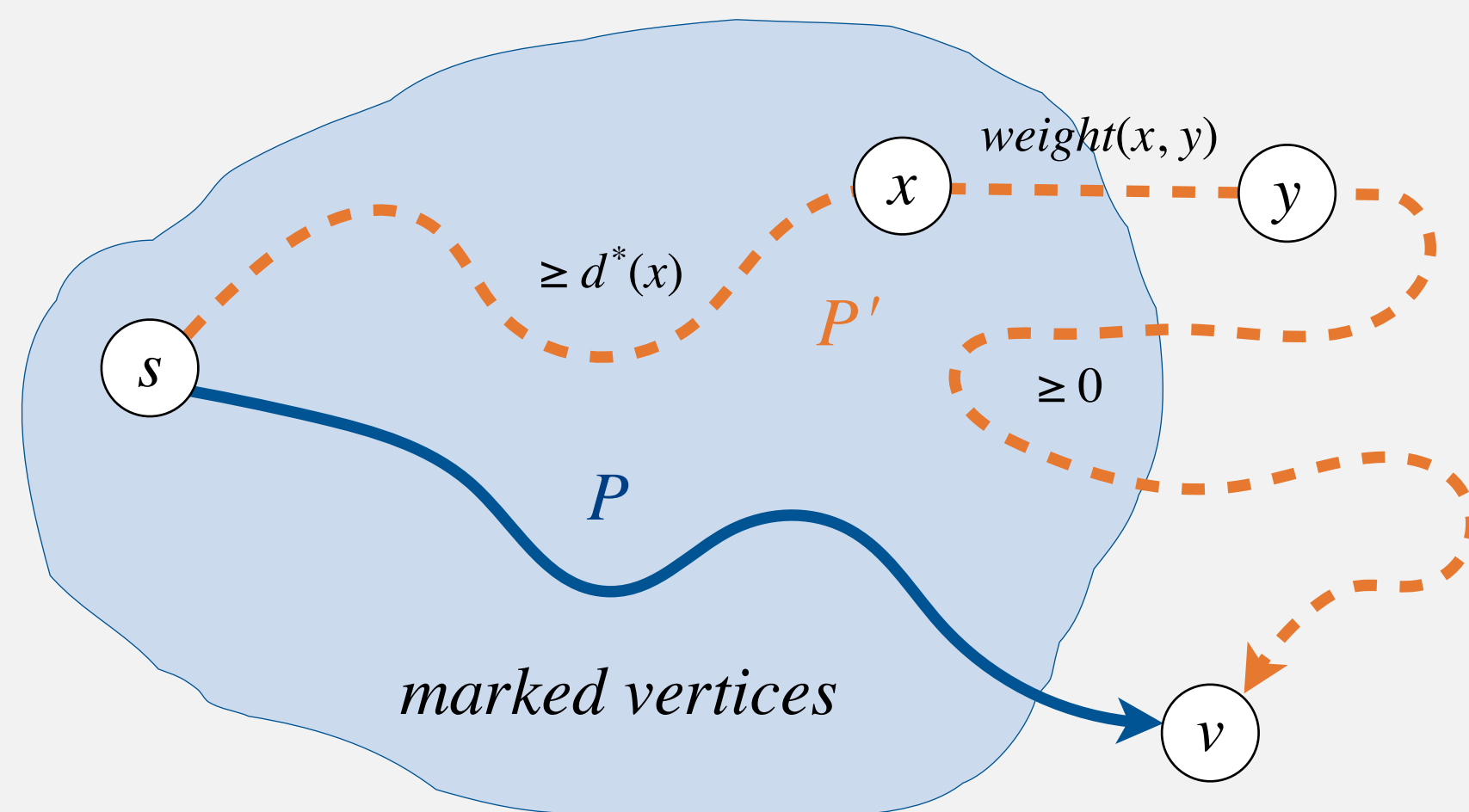
# Dijkstra's algorithm: correctness proof

**Invariant.** For each marked vertex  $v$ :  $\text{distTo}[v] = d^*(v)$ .

length of shortest path from  $s$  to  $v$

**Pf.** [ by induction on number of marked vertices ]

- Let  $v$  be next vertex marked.
- Let  $P$  be the path from  $s$  to  $v$  of length  $\text{distTo}[v]$ .
- Consider any other path  $P'$  from  $s$  to  $v$ .
- Let  $x \rightarrow y$  be first edge in  $P'$  with  $x$  marked and  $y$  unmarked.
- $P'$  is already as long as  $P$  by the time it reaches  $y$ :



by construction

$$\text{length}(P) = \text{distTo}[v]$$

Dijkstra chose  $v$  instead of  $y$   $\rightarrow \leq \text{distTo}[y]$

vertex  $x$  is marked  
(so it was relaxed)  $\rightarrow \leq \text{distTo}[x] + \text{weight}(x, y)$

induction  $\rightarrow = d^*(x) + \text{weight}(x, y)$

$P'$  is a path from  $s$  to  $x$ ,  
followed by edge  $x \rightarrow y$ ,  
followed by non-negative edges  $\rightarrow \leq \text{length}(P') \quad \blacksquare$

# Dijkstra's algorithm: correctness proof

---

**Invariant.** For each marked vertex  $v$ :  $\text{distTo}[v] = d^*(v)$ .

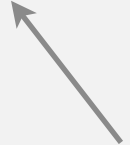
length of shortest path from  $s$  to  $v$



**Corollary 1.** Dijkstra's algorithm computes shortest path distances.

**Corollary 2.** Dijkstra's algorithm relaxes vertices in increasing order of distance from  $s$ .

generalizes both  
level-order traversal in a tree  
and breadth-first search in a graph



# Dijkstra's algorithm: Java implementation

```
public class DijkstraSP
{
    private DirectedEdge[] edgeTo;
    private double[] distTo;
    private IndexMinPQ<Double> pq;

    public DijkstraSP(EdgeWeightedDigraph G, int s)
    {
        edgeTo = new DirectedEdge[G.V()];
        distTo = new double[G.V()];
        pq = new IndexMinPQ<Double>(G.V());

        for (int v = 0; v < G.V(); v++)
            distTo[v] = Double.POSITIVE_INFINITY;
        distTo[s] = 0.0;

        pq.insert(s, 0.0);
        while (!pq.isEmpty())
        {
            int v = pq.delMin();
            for (DirectedEdge e : G.adj(v))
                relax(e);
        }
    }
}
```

PQ that supports  
decreasing the key  
(stay tuned)

PQ contains the  
unmarked vertices  
with finite distTo[] values

relax vertices in increasing order  
of distance from  $s$

# Dijkstra's algorithm: Java implementation

---

When relaxing an edge, also update PQ:

- Found first path from  $s$  to  $w$ : add  $w$  to PQ.
- Found better path from  $s$  to  $w$ : decrease key of  $w$  in PQ.

```
private void relax(DirectedEdge e)
{
    int v = e.from(), w = e.to();
    if (distTo[w] > distTo[v] + e.weight())
    {
        distTo[w] = distTo[v] + e.weight();
        edgeTo[w] = e;

        if (!pq.contains(w)) pq.insert(w, distTo[w]);
        else
            pq.decreaseKey(w, distTo[w]);
    }
}
```

← update PQ

Q. How to implement DECREASE-KEY operation in a priority queue?

## Indexed priority queue (Section 2.4)

---

Associate an index between 0 and  $n - 1$  with each key in a priority queue.

- Insert a key associated with a given index.
- Delete a minimum key and return associated index.
- **Decrease the key** associated with a given index.

for Dijkstra's algorithm:

$n = V,$

index = vertex,

key = distance from  $s$

```
public class IndexMinPQ<Key extends Comparable<Key>>
```

```
    IndexMinPQ(int n)
```

*create PQ with indices 0, 1, ...,  $n - 1$*

```
    void insert(int i, Key key)
```

*associate key with index  $i$*

```
    int delMin()
```

*remove min key and return associated index*

```
    void decreaseKey(int i, Key key)
```

*decrease the key associated with index  $i$*

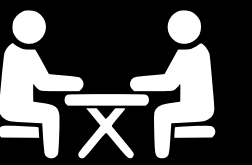
```
    boolean isEmpty()
```

*is the priority queue empty?*

```
    ⋮
```

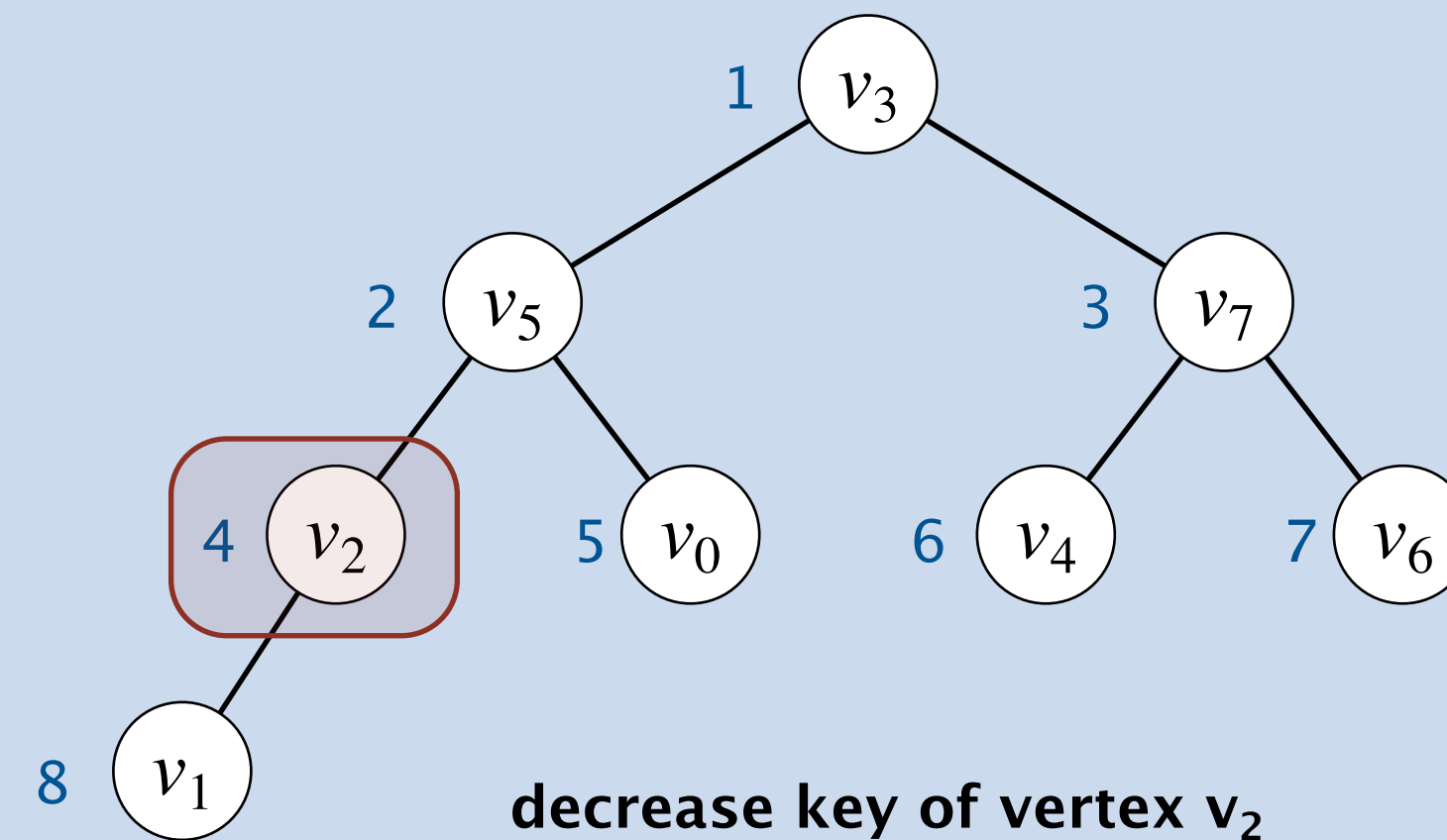
⋮

# DECREASE-KEY IN A BINARY HEAP



**Goal.** Implement DECREASE-KEY operation in a binary heap.

	0	1	2	3	4	5	6	7	8
pq[]	-	$v_3$	$v_5$	$v_7$	$v_2$	$v_0$	$v_4$	$v_6$	$v_1$



# DECREASE-KEY IN A BINARY HEAP



**Goal.** Implement DECREASE-KEY operation in a binary heap.

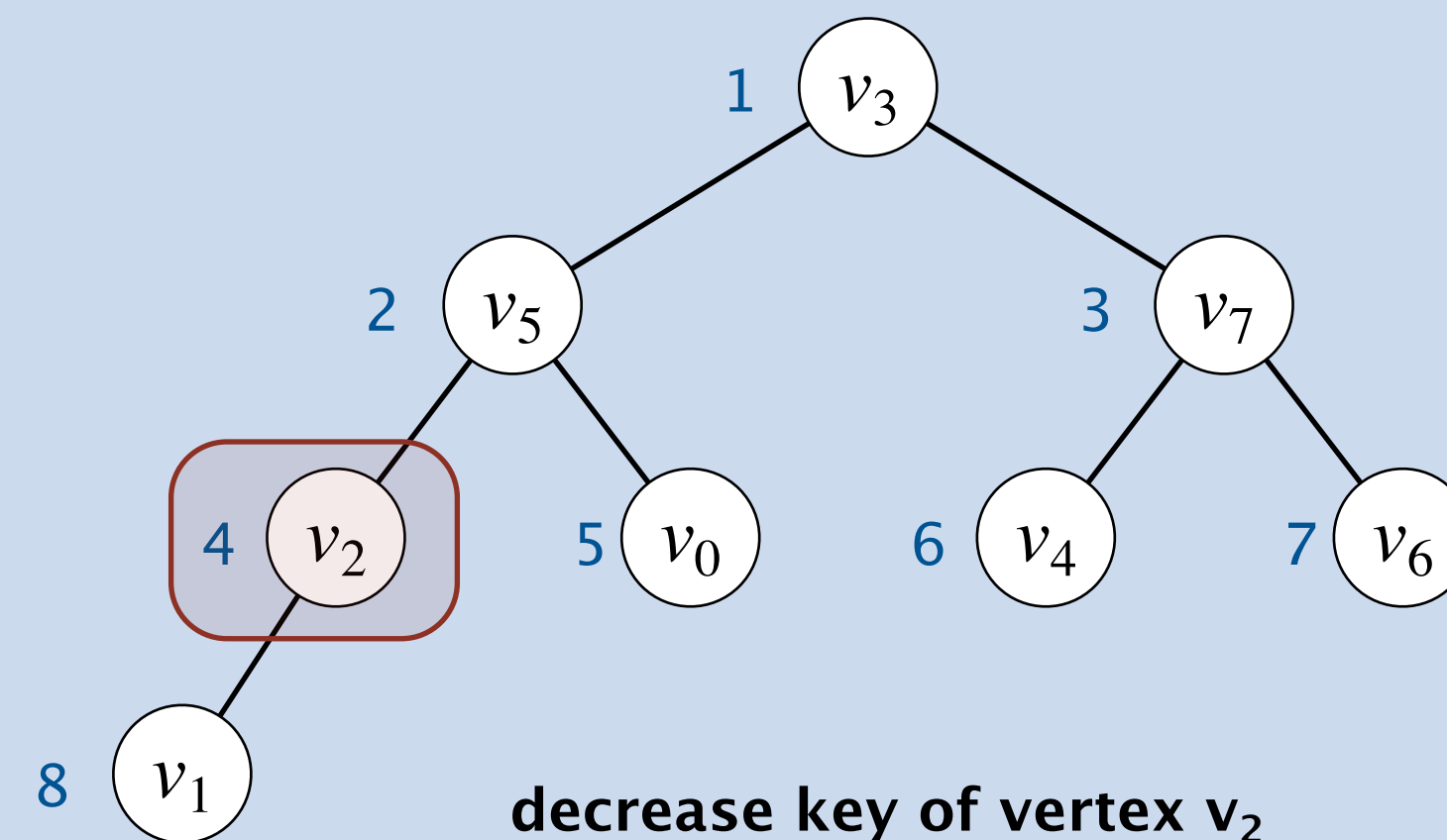
**Solution.**

- Find vertex in heap. How?
- Change priority of vertex and call `swim()` to restore heap invariant.

**Extra data structure.** Maintain an inverse array `qp[]` that maps from the vertex to the binary heap node index.

	0	1	2	3	4	5	6	7	8
pq[]	–	$v_3$	$v_5$	$v_7$	$v_2$	$v_0$	$v_4$	$v_6$	$v_1$
qp[]	5	8	4	1	6	2	4	3	–
keys[]	1.0	2.0	3.0	0.0	6.0	8.0	4.0	2.0	–

vertex 2 has priority 3.0  
and is at heap index 4



## Dijkstra's algorithm: which priority queue?

---

Number of PQ operations:  $V$  INSERT,  $V$  DELETE-MIN,  $\leq E$  DECREASE-KEY.

PQ implementation	INSERT	DELETE-MIN	DECREASE-KEY	total
unordered array	1	$V$	1	$V^2$
binary heap	$\log V$	$\log V$	$\log V$	$E \log V$
d-way heap	$\log_d V$	$d \log_d V$	$\log_d V$	$E \log_{E/V} V$
Fibonacci heap	$1^\dagger$	$\log V^\dagger$	$1^\dagger$	$E + V \log V$

$^\dagger$  amortized

### Bottom line.

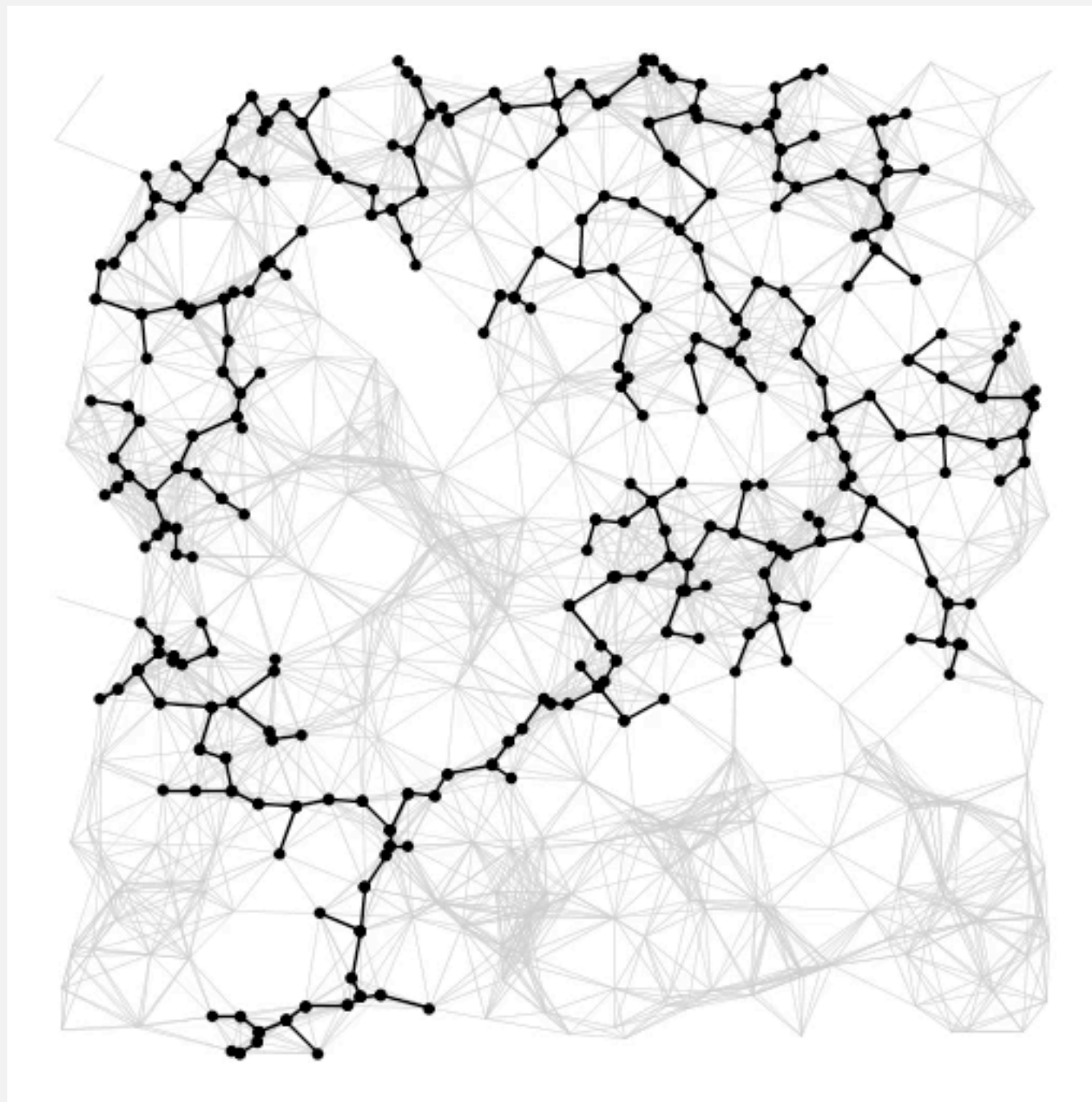
- Array implementation optimal for complete digraphs.
- Binary heap much faster for sparse digraphs.
- 4-way heap worth the trouble in performance-critical situations.
- Fibonacci heap best in theory, but not worth implementing.

# Priority-first search

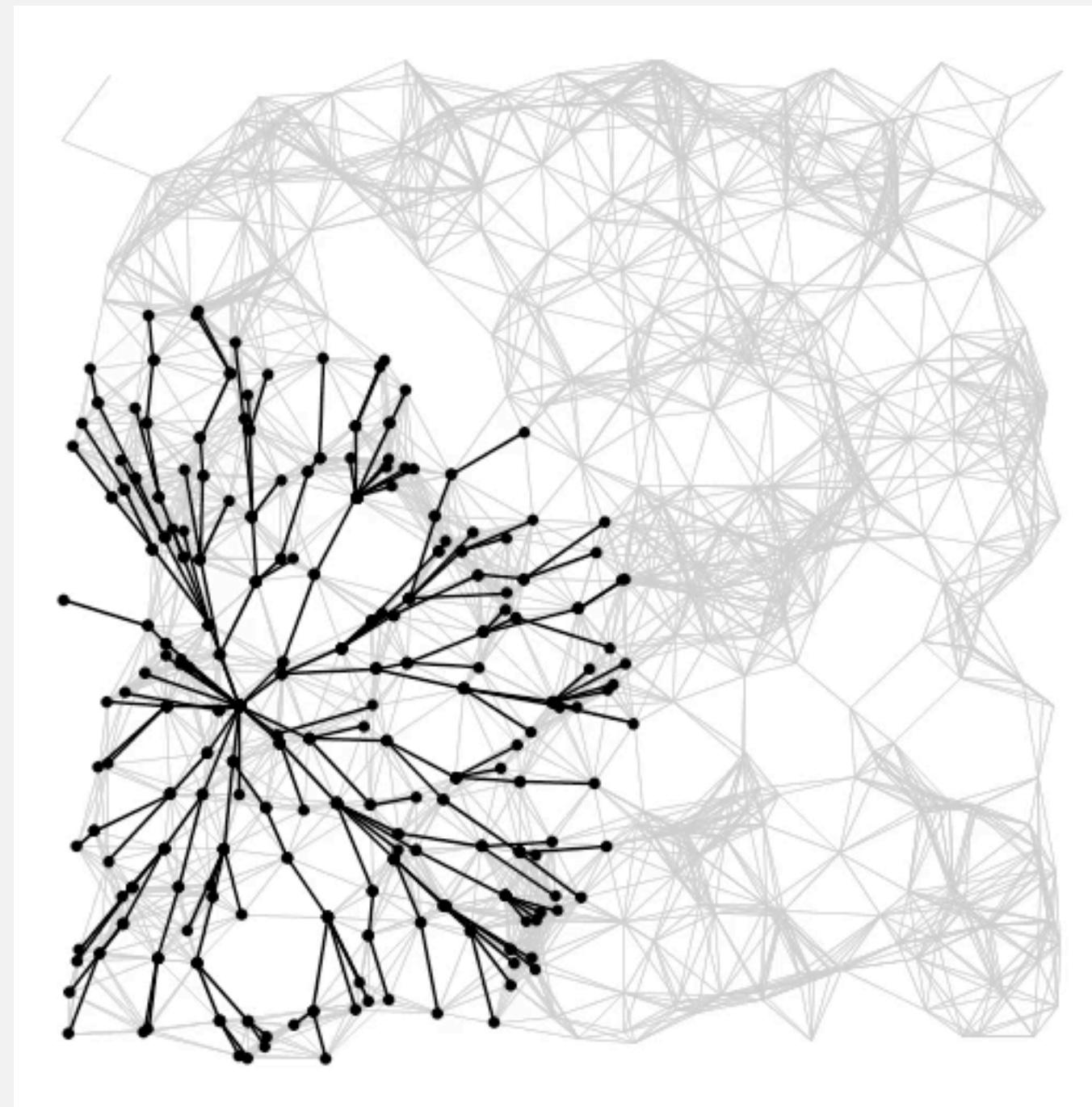
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**Observation.** Prim and Dijkstra are essentially the same algorithm.

- Prim: Choose next vertex that is closest to **any vertex in the tree** (via an undirected edge).
- Dijkstra: Choose next vertex that is closest to the **source vertex** (via a directed path).



Prim's algorithm

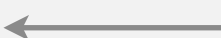


Dijkstra's algorithm

# Algorithms for shortest paths

---

## Variations on a theme: vertex relaxations.

- Bellman–Ford: relax all vertices; repeat  $V - 1$  times.
- Dijkstra: relax vertices in order of distance from  $s$ .
- Topological sort: relax vertices in topological order.  see Section 4.4 and next lecture

algorithm	worst-case running time	negative weights <sup>†</sup>	directed cycles
Bellman–Ford	$E V$	✓	✓
Dijkstra	$E \log V$		✓
topological sort	$E$	✓	

<sup>†</sup> no negative cycles

# Which shortest paths algorithm to use?

---

Select algorithm based on properties of edge-weighted digraph.

- Negative weights (but no “negative cycles”): Bellman–Ford.
- Non-negative weights: Dijkstra.
- DAG: topological sort.

algorithm	worst-case running time	negative weights †	directed cycles
Bellman–Ford	$E V$	✓	✓
Dijkstra	$E \log V$		✓
topological sort	$E$	✓	

† no negative cycles

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