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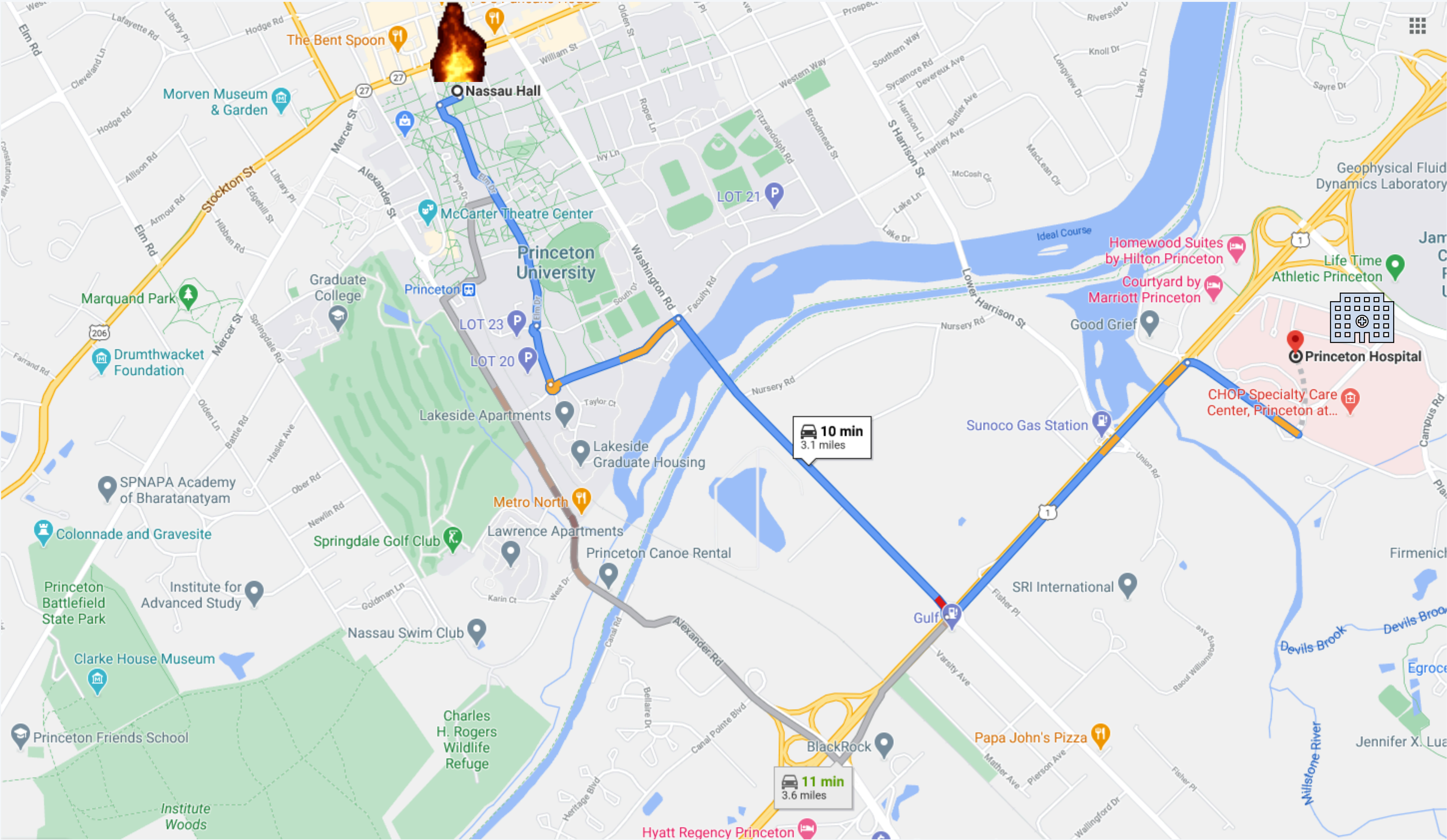
## 4.4 SHORTEST PATHS

---

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



# Google maps



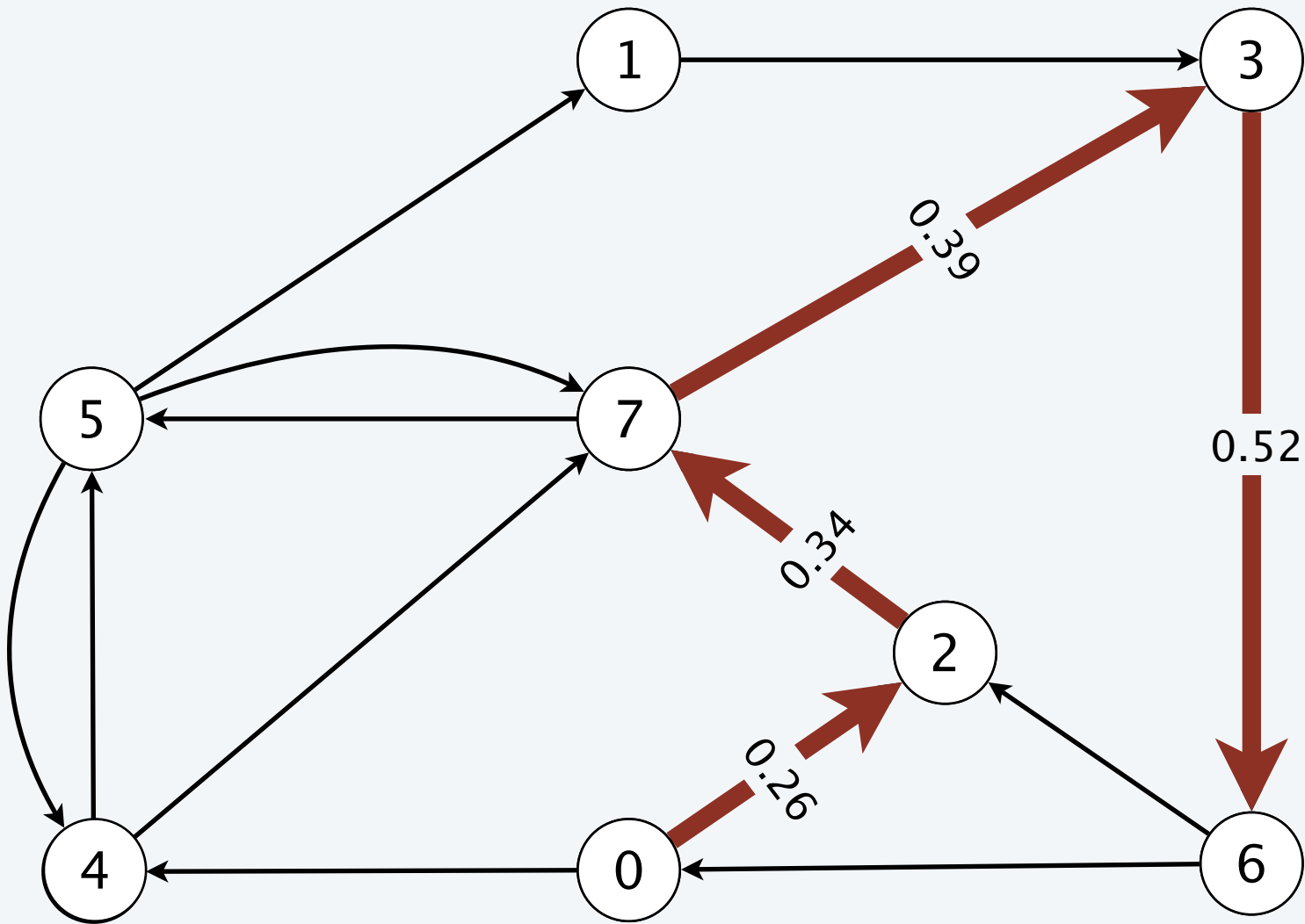


# Shortest path in an edge-weighted digraph

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.



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 in DAGs  
(see next lecture)*
- Allow.

Simplifying assumption. Each vertex is reachable from  $s$ .  *implies that shortest path from  $s$  to  $v$  exists  
(and that  $E \geq V - 1$ )*



Which shortest path variant for 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

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# Data structures for single-source shortest paths

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

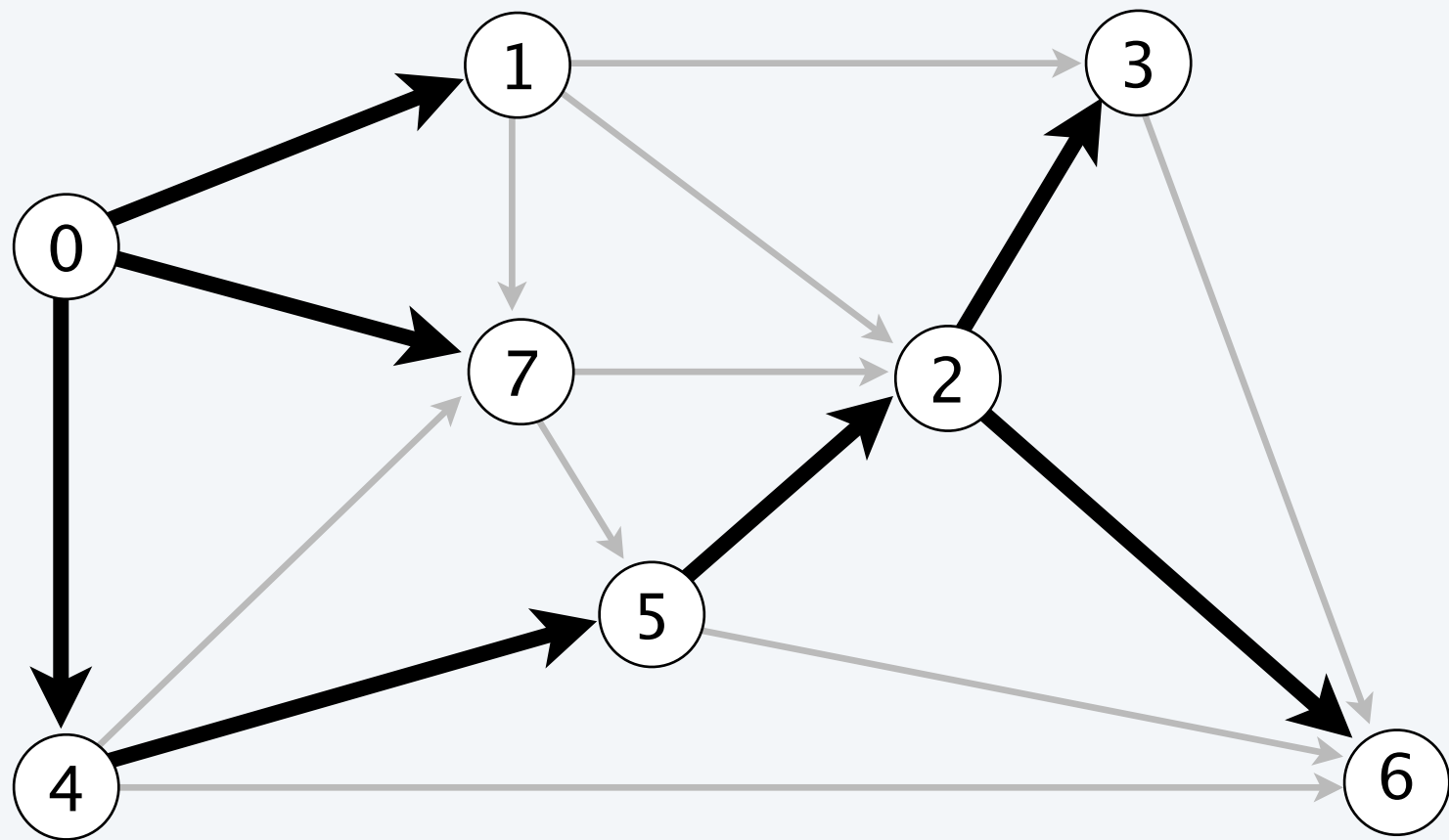
*no repeated vertices*  
 $\Rightarrow \leq V - 1$  edges

**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 shortest paths 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

$v$	$\text{distTo}[]$	$\text{edgeTo}[]$
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

parent-link representation

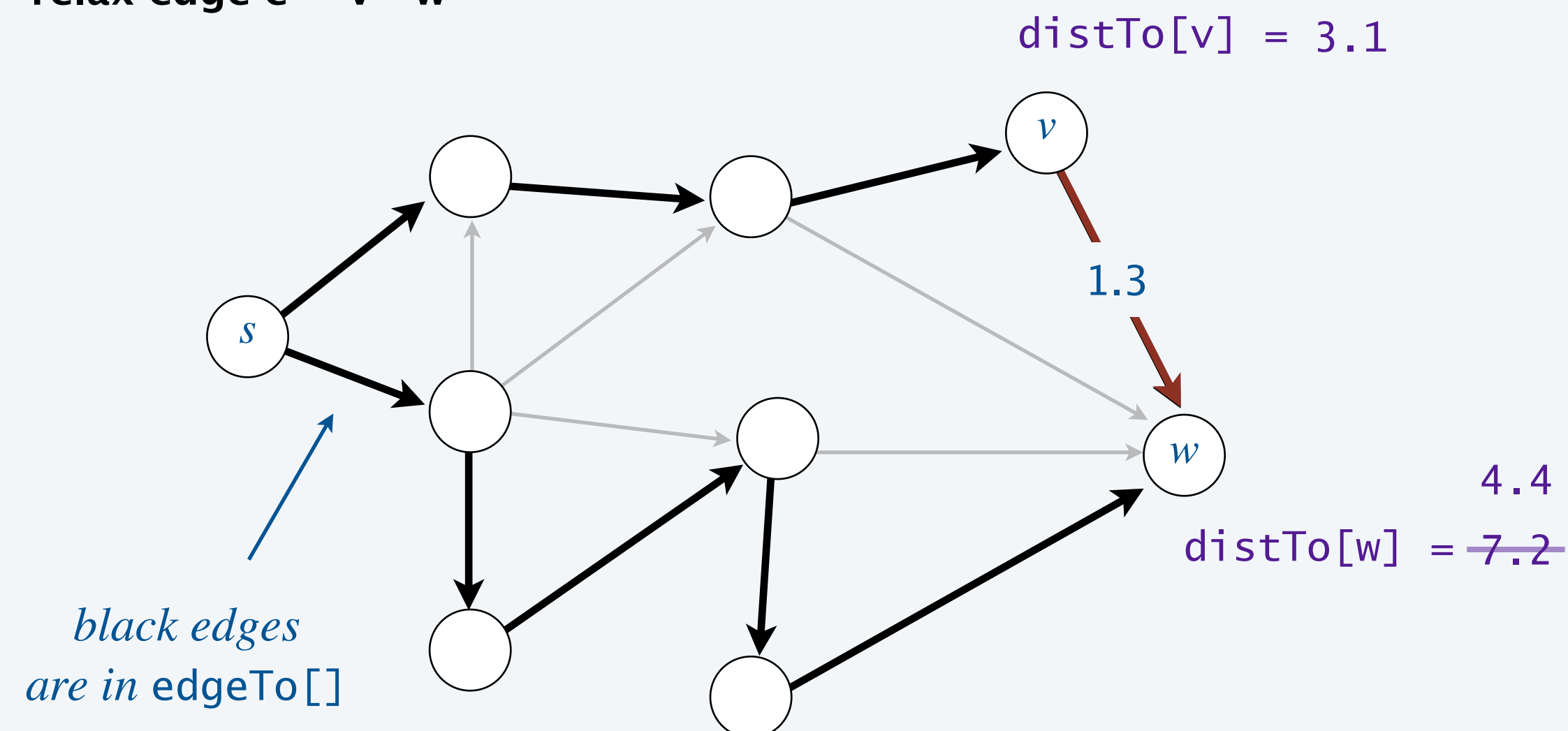


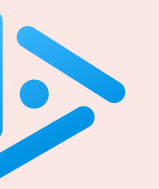
# Edge relaxation

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  $\text{edgeTo}[w]$ .

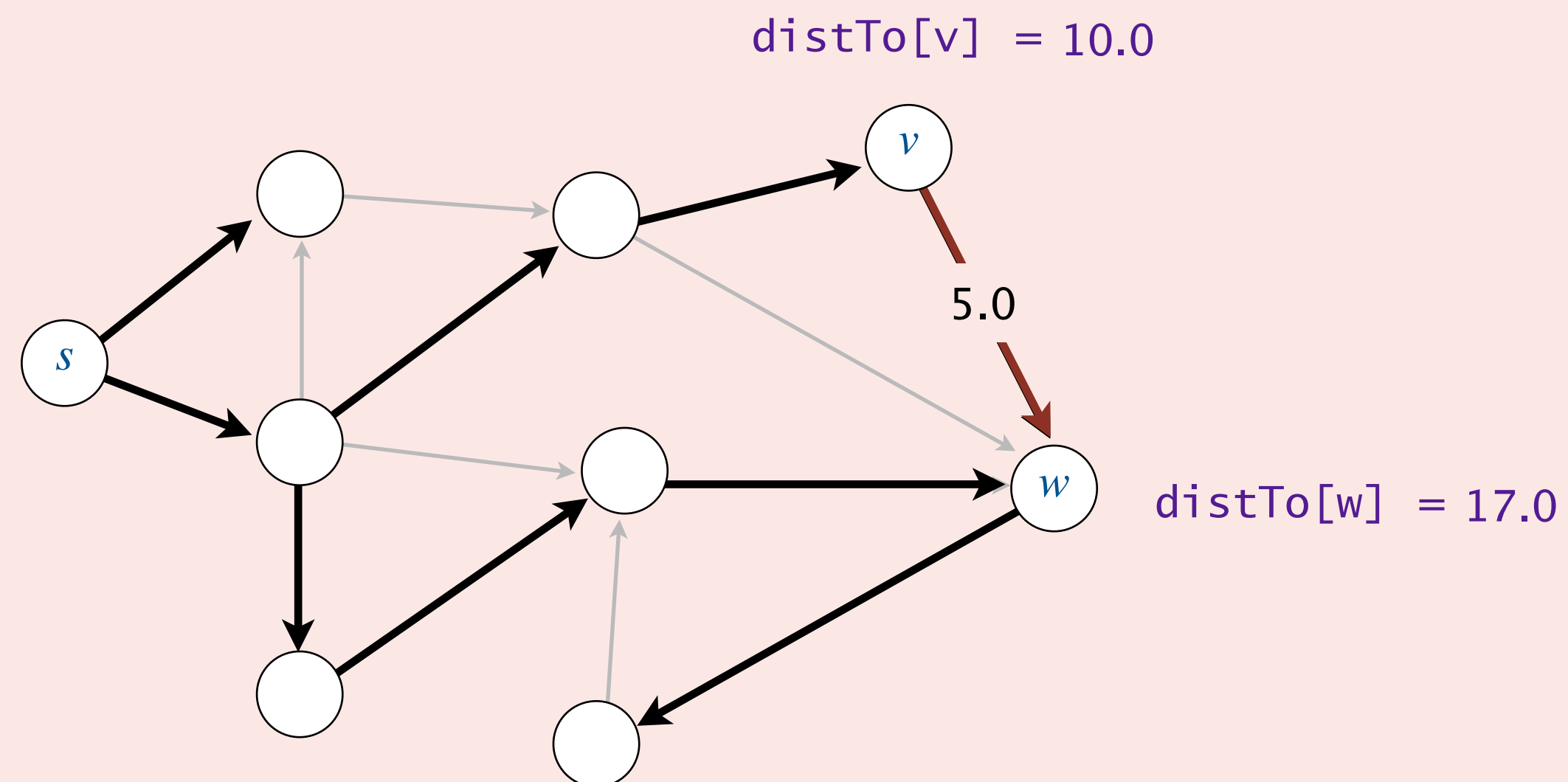
relax edge  $e = v \rightarrow w$





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.  $\longleftarrow$  *next lecture*



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## 4.4 SHORTEST PATHS

---

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

# Weighted directed edge API

## API.

```
public class DirectedEdge
```

```
    DirectedEdge(int v, int w, double weight)    create weighted edge  $v \rightarrow w$ 
```

```
    int    from()                               vertex  $v$ 
```

```
    int    to()                                 vertex  $w$ 
```

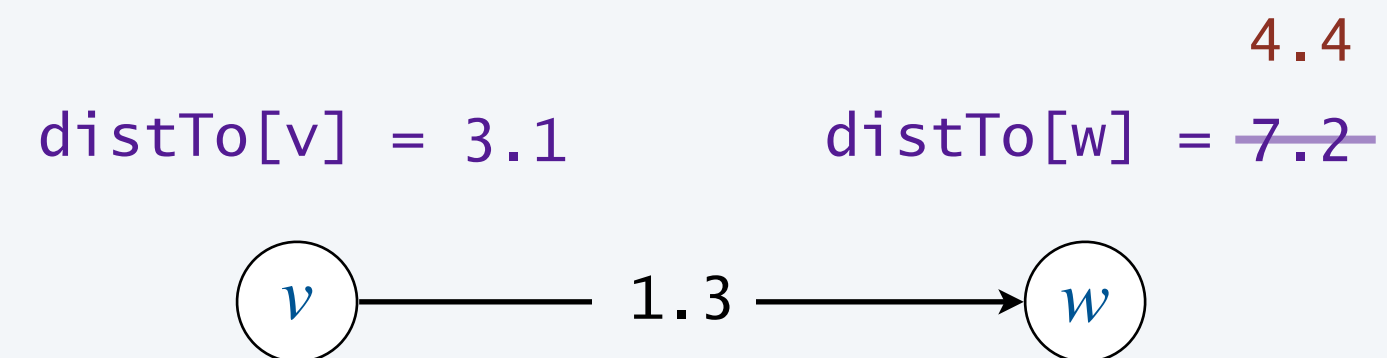
```
    double weight()                             weight of this edge
```

```
        ⋮
```

```
        ⋮
```

Ex. Relax 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

---

```
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;  
    }
```

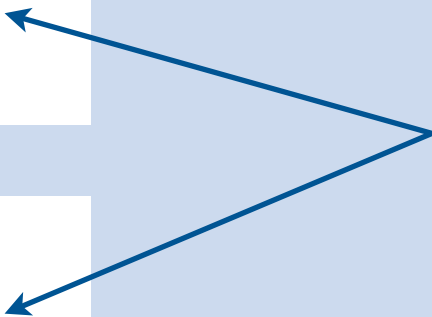
```
    public int from() {  
        return v;  
    }
```

```
    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)	<i>edge-weighted digraph with V vertices (and no edges)</i>
void	addEdge(DirectedEdge e)	<i>add weighted directed edge e</i>
Iterable<DirectedEdge>	adj(int v)	<i>edges incident from v</i>
int	V()	<i>number of vertices</i>
	⋮	⋮

# 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  $e = v \rightarrow w$  only  
to  $v$ '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)	<i>shortest paths from s in digraph G</i>
--	----------------------------------	---

double	distTo(int v)	<i>length of shortest path from s to v</i>
--------	---------------	--

Iterable <DirectedEdge>	pathTo(int v)	<i>shortest path from s to v</i>
-------------------------	---------------	----------------------------------

boolean	hasPathTo(int v)	<i>is there a path from s to v ?</i>
---------	------------------	--------------------------------------



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## 4.4 SHORTEST PATHS

---

- *properties*
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- *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.

---

```
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  $i = E$*

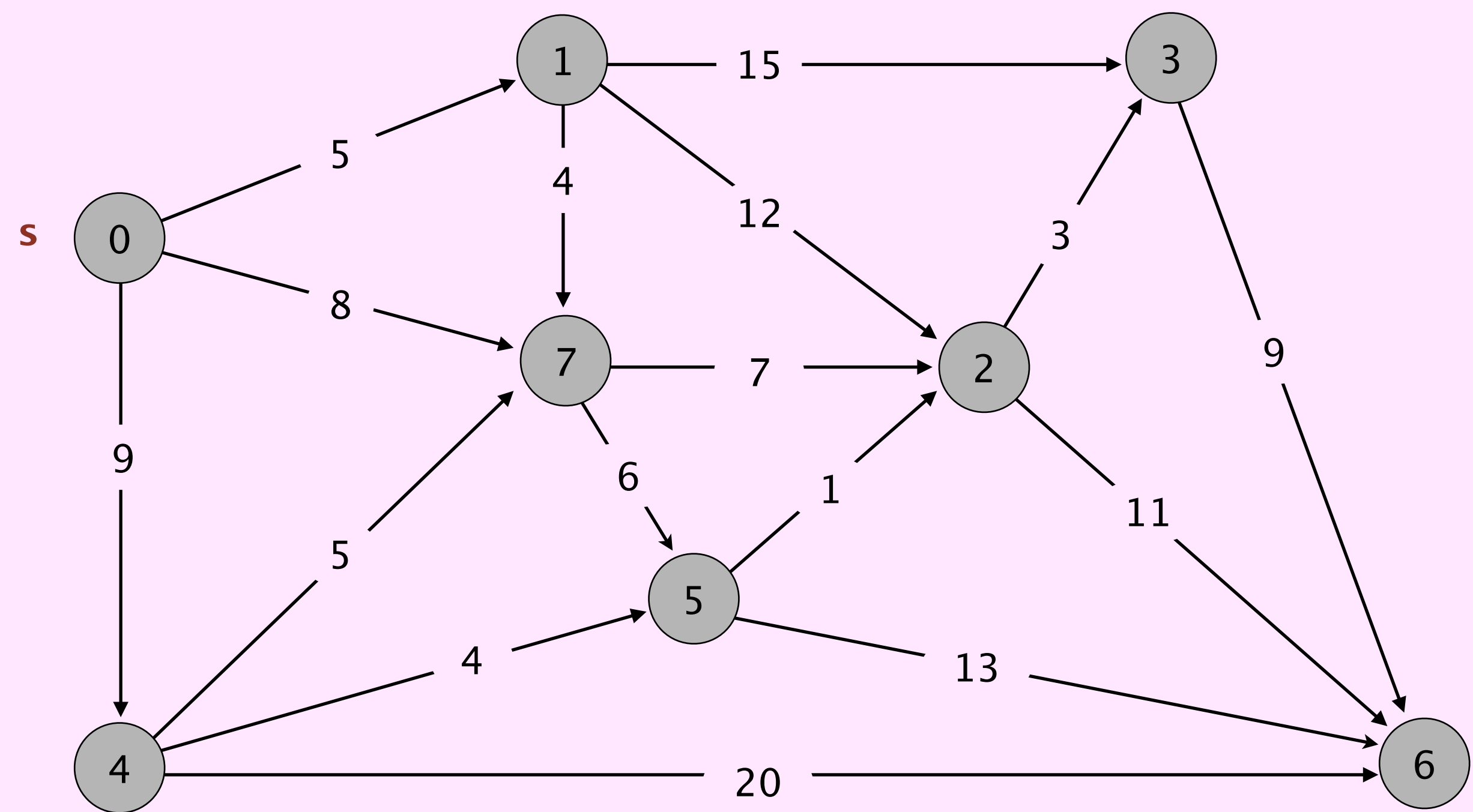
**Running time.** Algorithm takes  $\Theta(EV)$  time and uses  $\Theta(V)$  extra space.



# Bellman–Ford algorithm demo



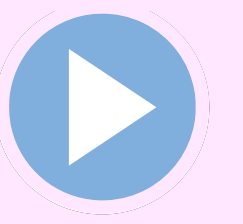
Repeat  $V - 1$  times: relax all  $E$  edges.



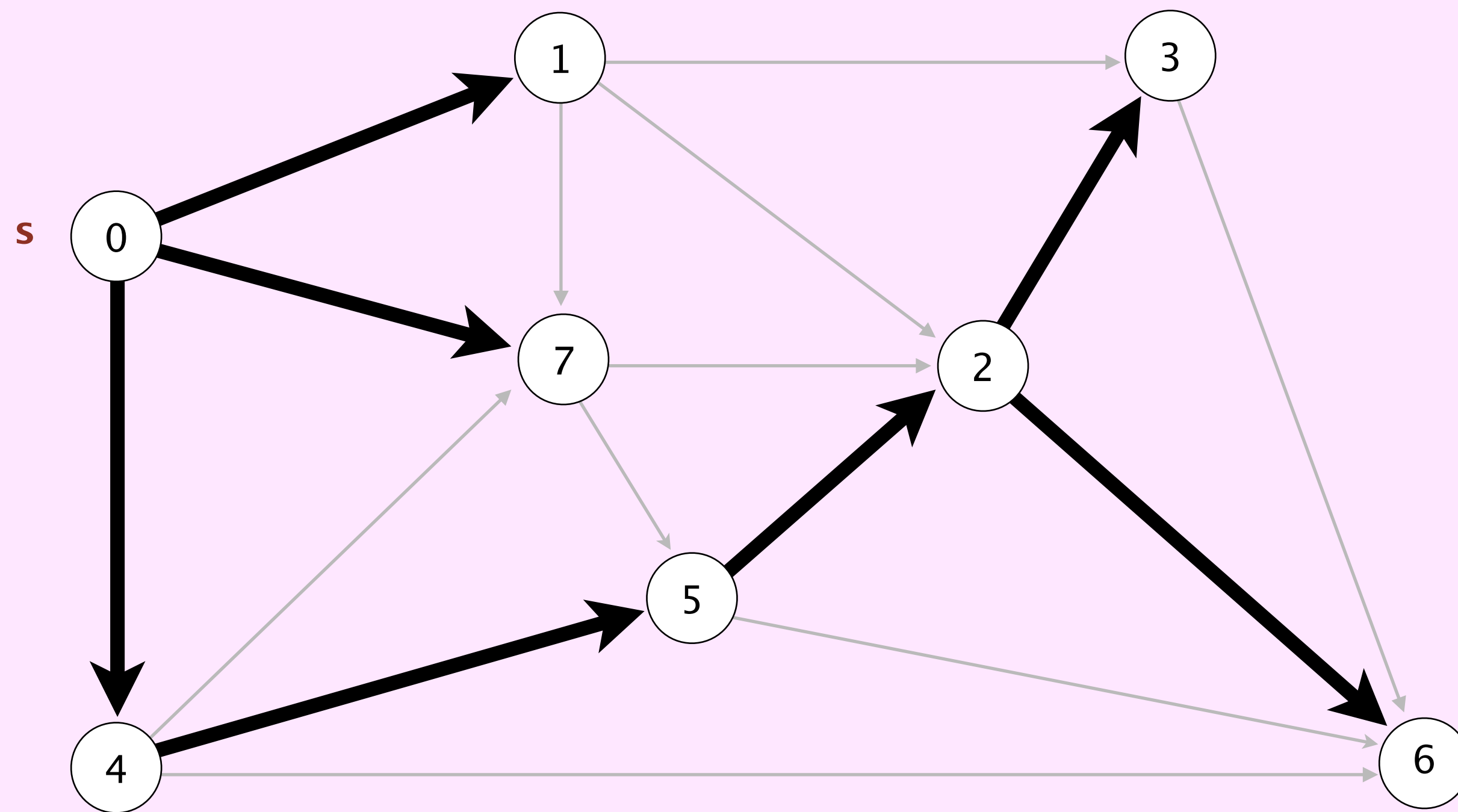
an edge-weighted digraph

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.

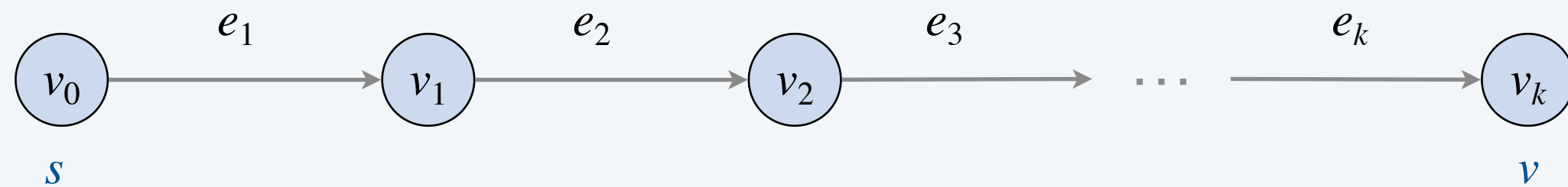


v	distTo[]	edgeTo[]
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0	2→6
7	8.0	0→7

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  $v$  consisting of  $k$  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

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

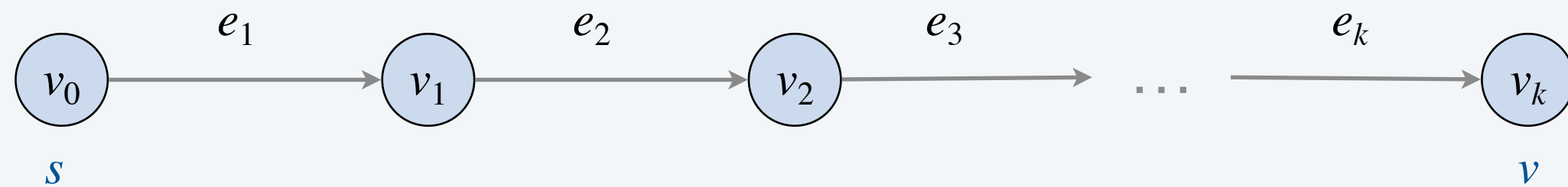
- This inequality continues to hold because  $\text{distTo}[v_{i+1}]$  cannot increase. ■



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



**Corollary.** For each vertex  $v$ , Bellman–Ford computes length of shortest path from  $s$  to  $v$ .

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

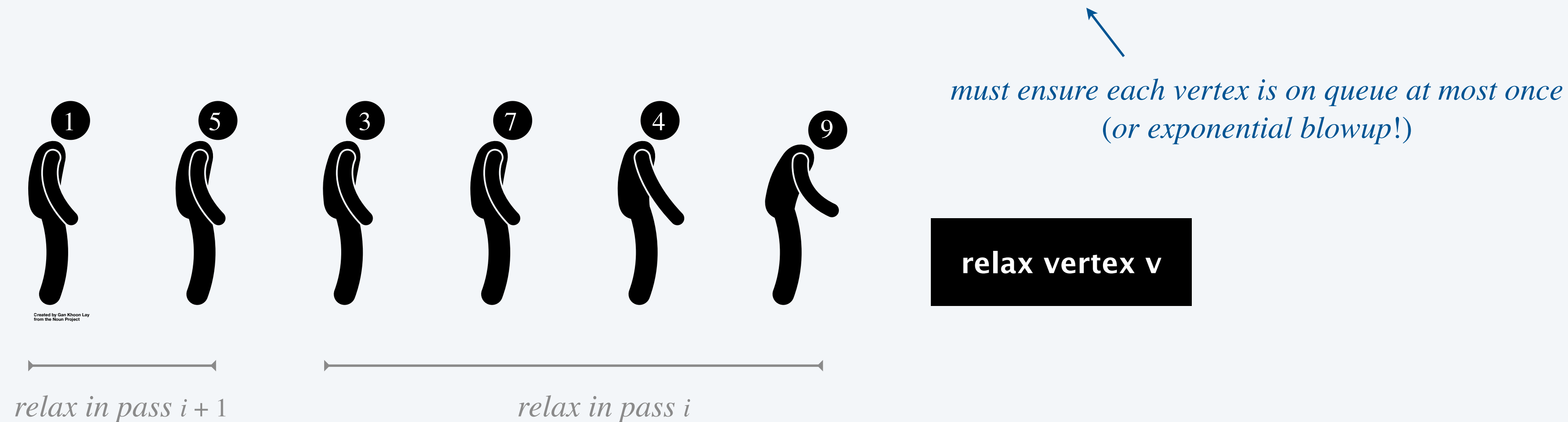
- There exists a simple shortest path  $P^*$  from  $s$  to  $v$ ; it contains  $k \leq V - 1$  edges.
- The Proposition implies that, after at most  $V - 1$  passes,  $\text{distTo}[v] \leq \text{length}(P^*)$ .
- Since  $\text{distTo}[v]$  is the length of some path from  $s$  to  $v$ ,  $\text{distTo}[v]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

# 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  $i + 1$ .

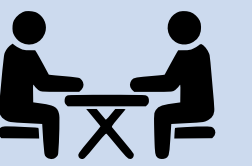
## Queue-based implementation of Bellman–Ford.

- 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.



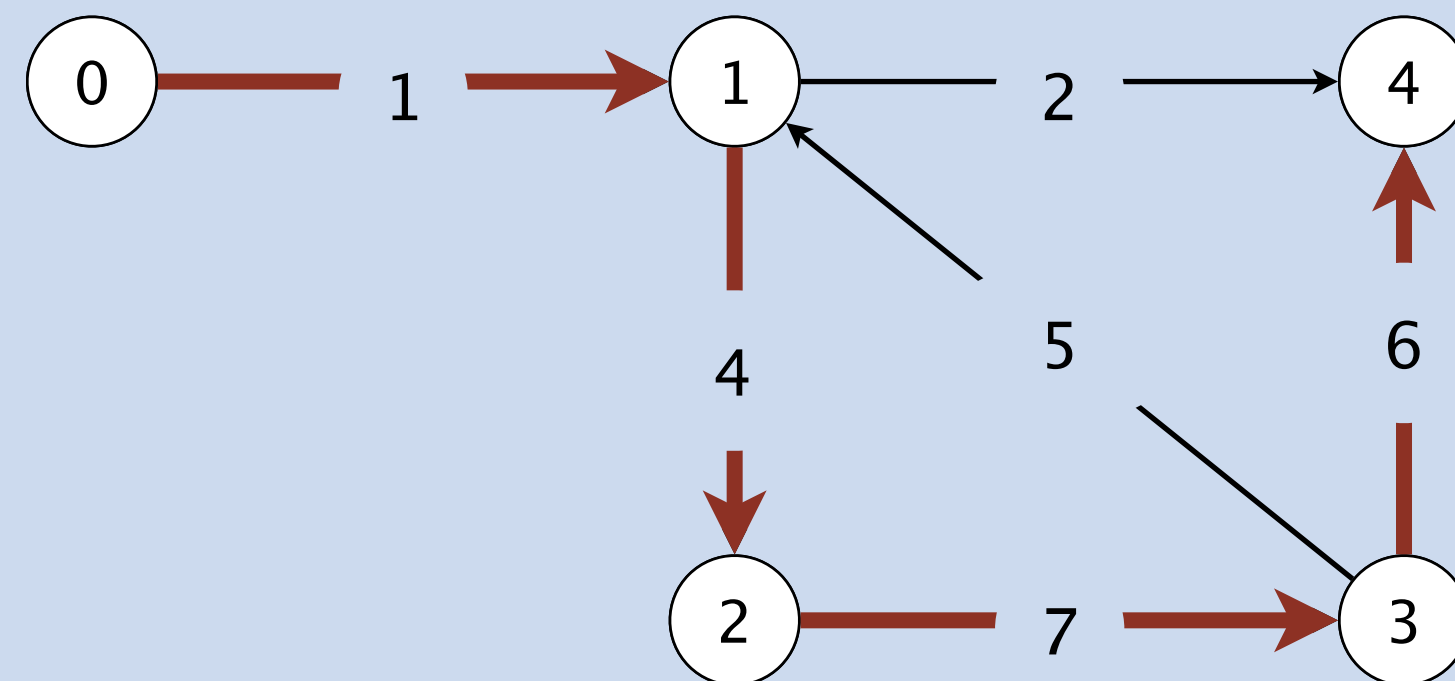
## Impact.

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



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

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



**longest simple path from 0 to 4**

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

**length of path = 18**

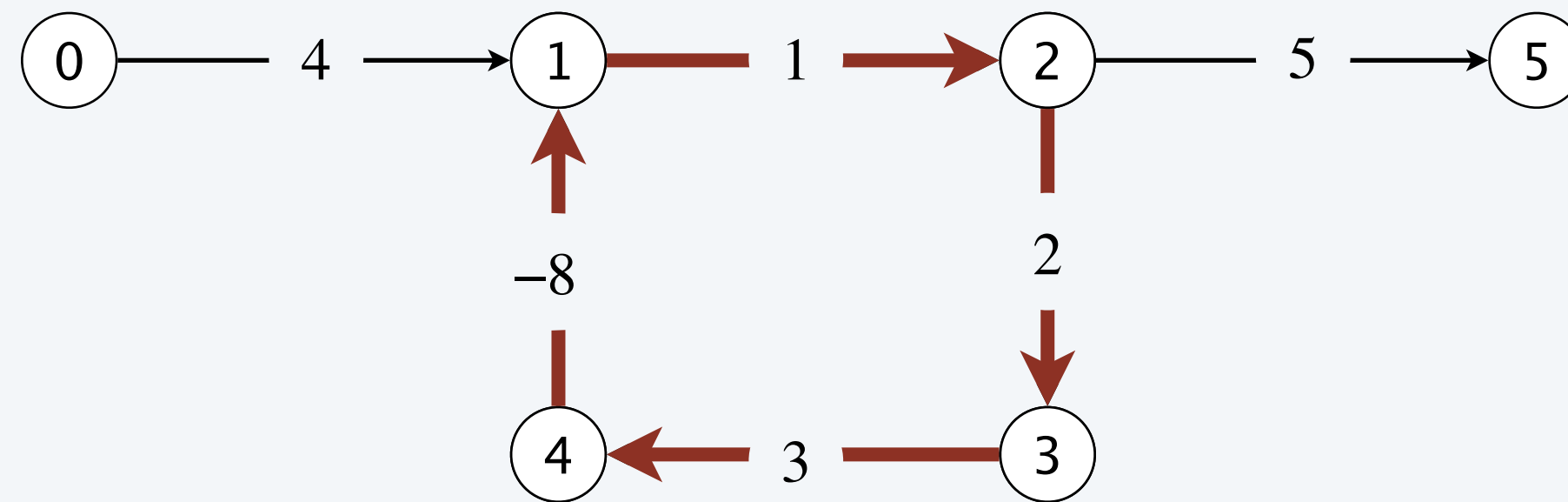
$(1 + 4 + 7 + 6)$

# Bellman–Ford algorithm: negative weights

---

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

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



**negative cycle**  
(length =  $1 + 2 + 3 + -8 = -2 < 0$ )

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

$0 \rightarrow 1 \rightarrow \underline{2 \rightarrow 3 \rightarrow 4 \rightarrow 1} \rightarrow \dots \rightarrow \underline{2 \rightarrow 3 \rightarrow 4 \rightarrow 1} \rightarrow 2 \rightarrow 5$





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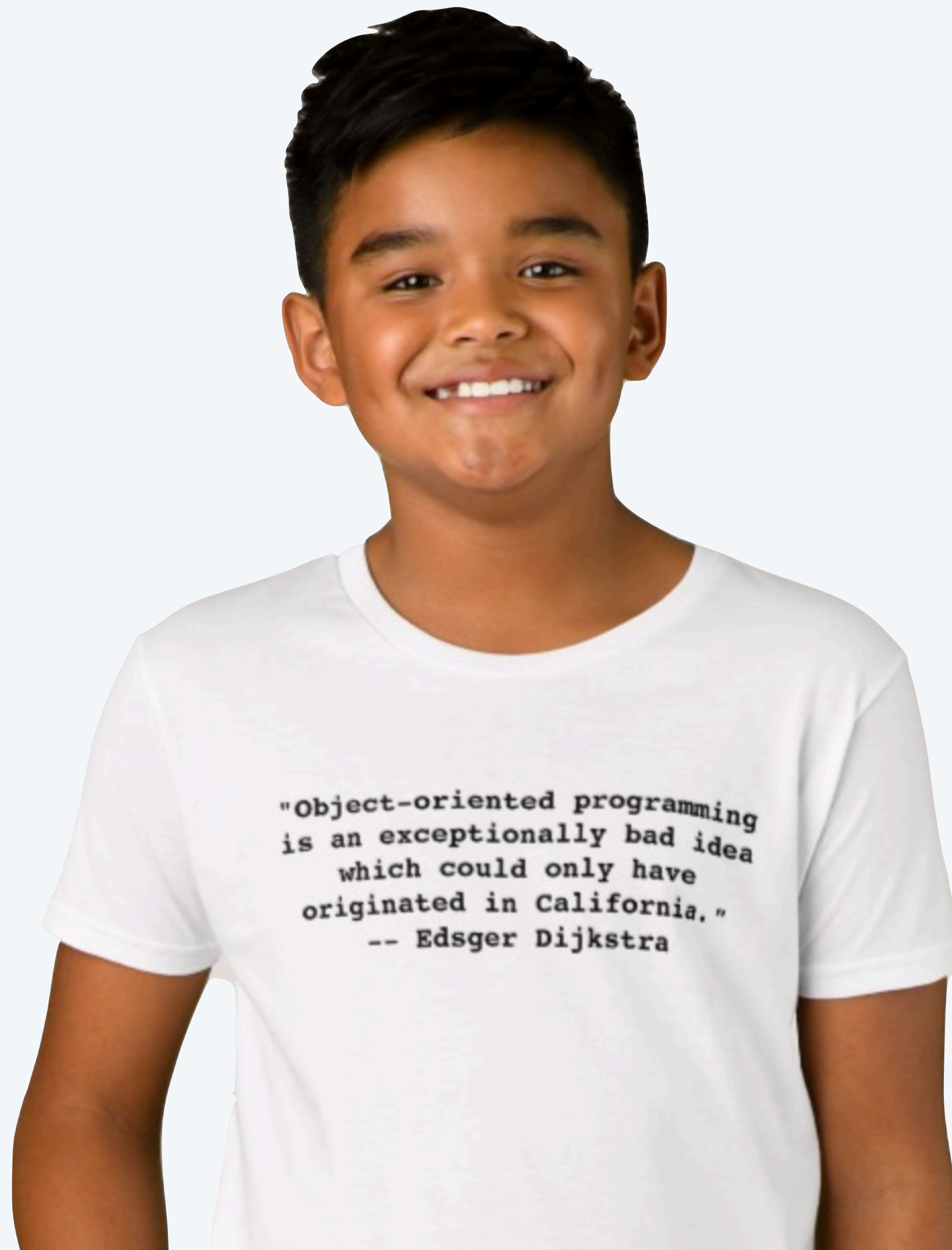
## 4.4 SHORTEST PATHS

---

- *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  $v$ :  $\text{edgeTo}[v] = \text{null}$ .

$T = \emptyset$ .

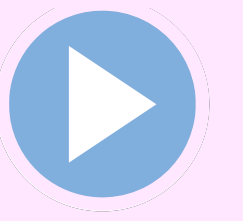
$\text{distTo}[s] = 0$ .

Repeat until all vertices are marked:

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

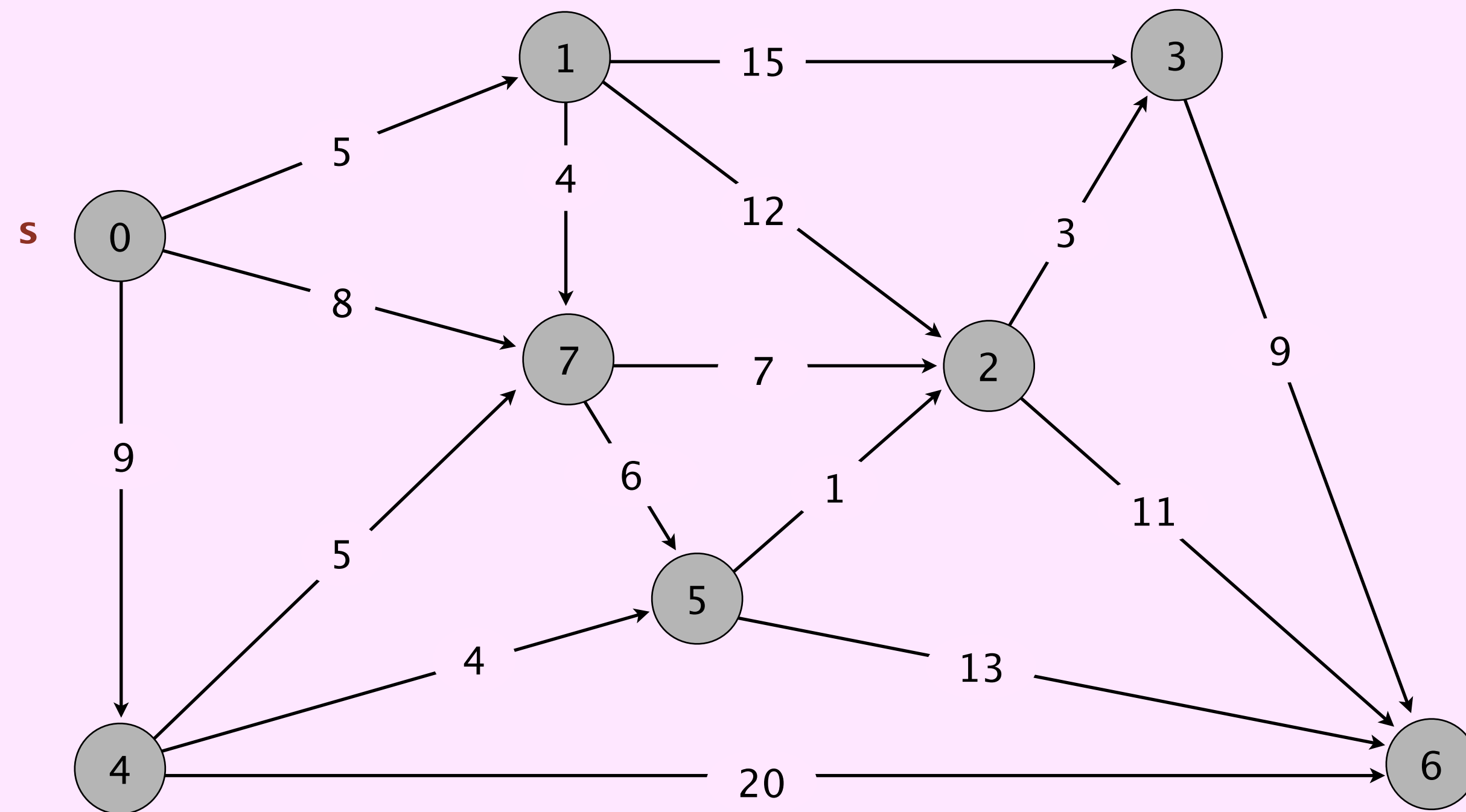
Key difference with Bellman–Ford. Each edge gets relaxed exactly once!

# Dijkstra's algorithm demo



Repeat until all vertices are marked:

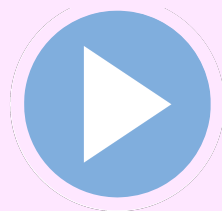
- Select unmarked vertex  $v$  with the smallest `distTo[]` value.
- Mark  $v$  and relax all edges incident from  $v$ .



an edge-weighted digraph

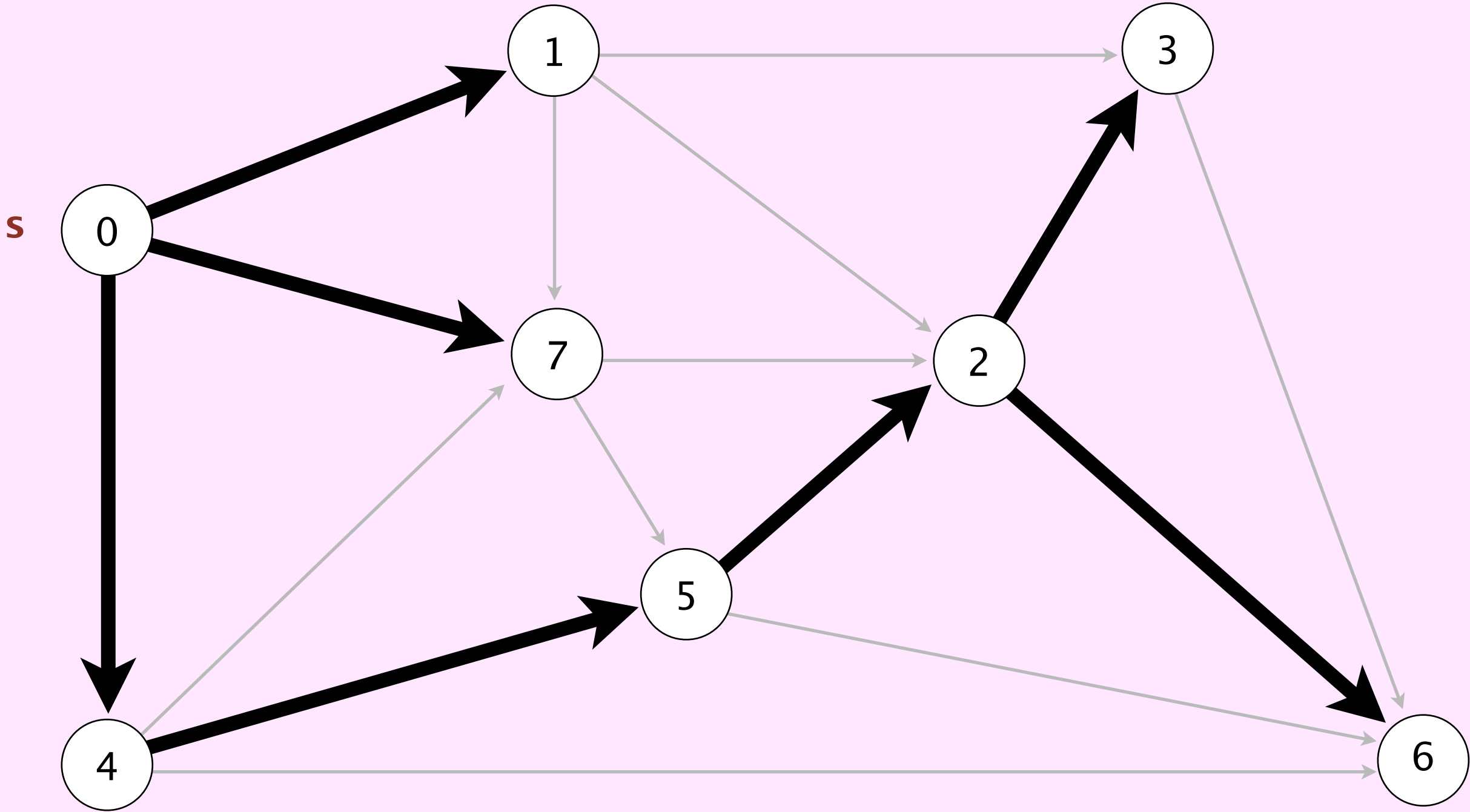
0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0





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

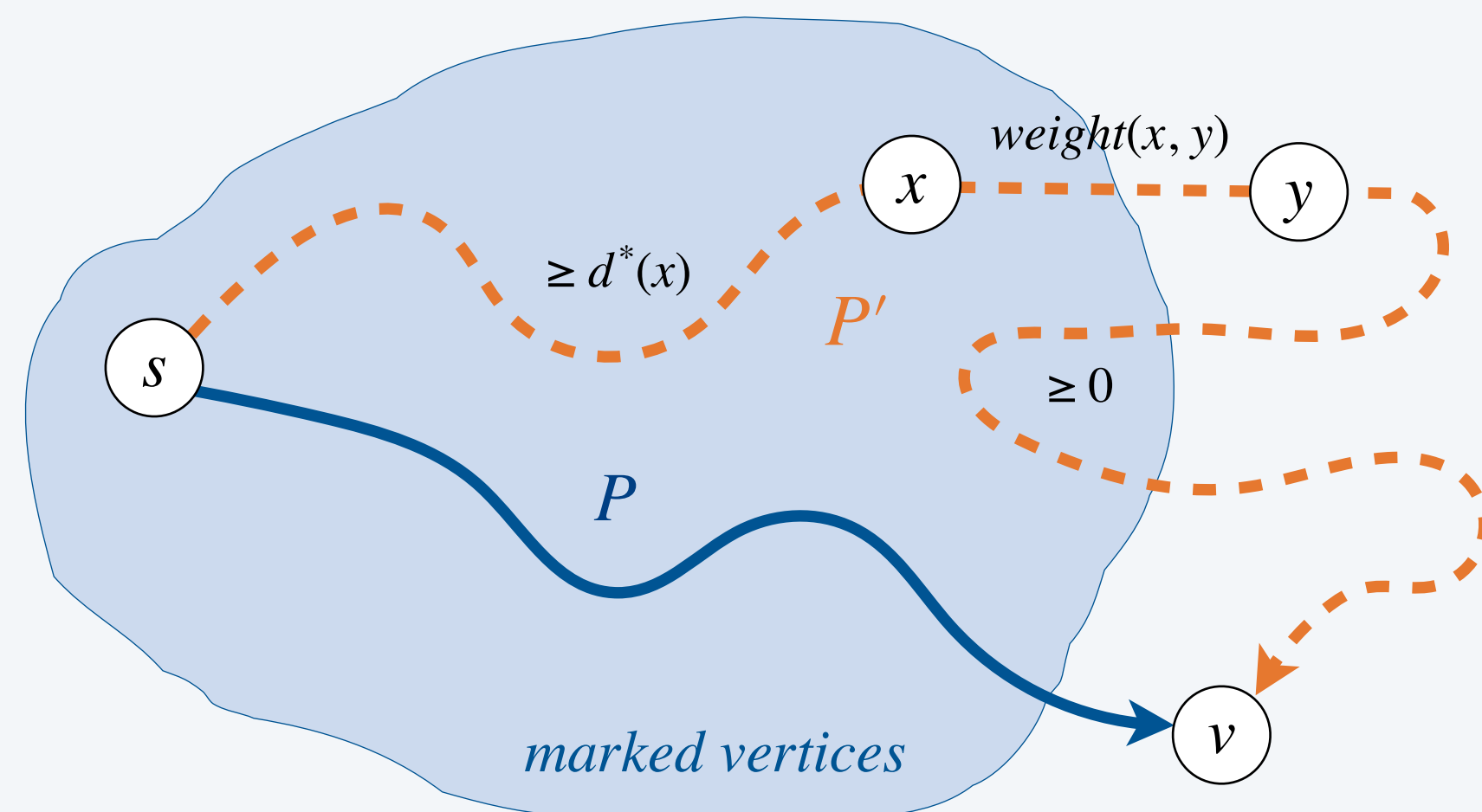
# Dijkstra's algorithm: correctness proof

**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  $y$ :



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

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

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

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

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

# 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 vertex  $w$  to PQ.
- Found better path from  $s$  to  $w$ : decrease priority associated with vertex  $w$  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*

*index*      *priority*

Q. How to efficiently 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:*

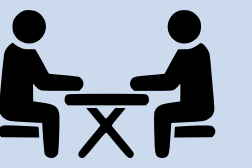
$n = V,$   
 $index = vertex,$   
 $key = distance\ from\ s$

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

---

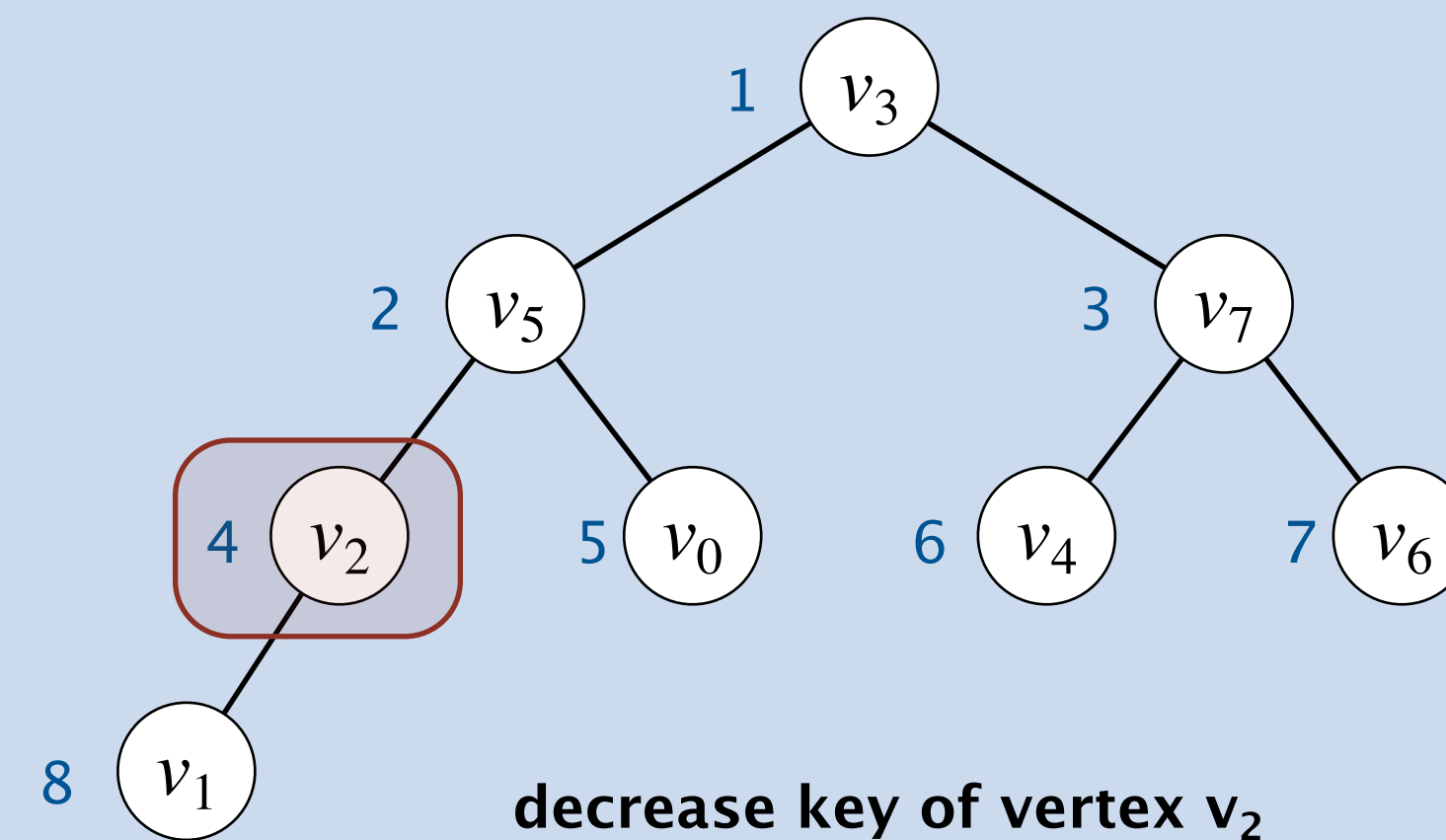
IndexMinPQ(int n)	<i>create PQ with indices 0, 1, ..., n - 1</i>
void insert(int i, Key key)	<i>associate key with index i</i>
int delMin()	<i>remove min key and return associated index</i>
void decreaseKey(int i, Key key)	<i>decrease the key associated with index i</i>
boolean isEmpty()	<i>is the priority queue empty ?</i>
⋮	⋮

# Decreasing the key in a binary heap

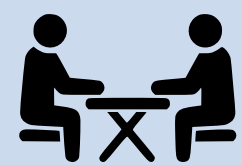


**Goal.** Implement DECREASE-KEY operation in a min-oriented binary heap.

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



# Decreasing the key in a binary heap



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

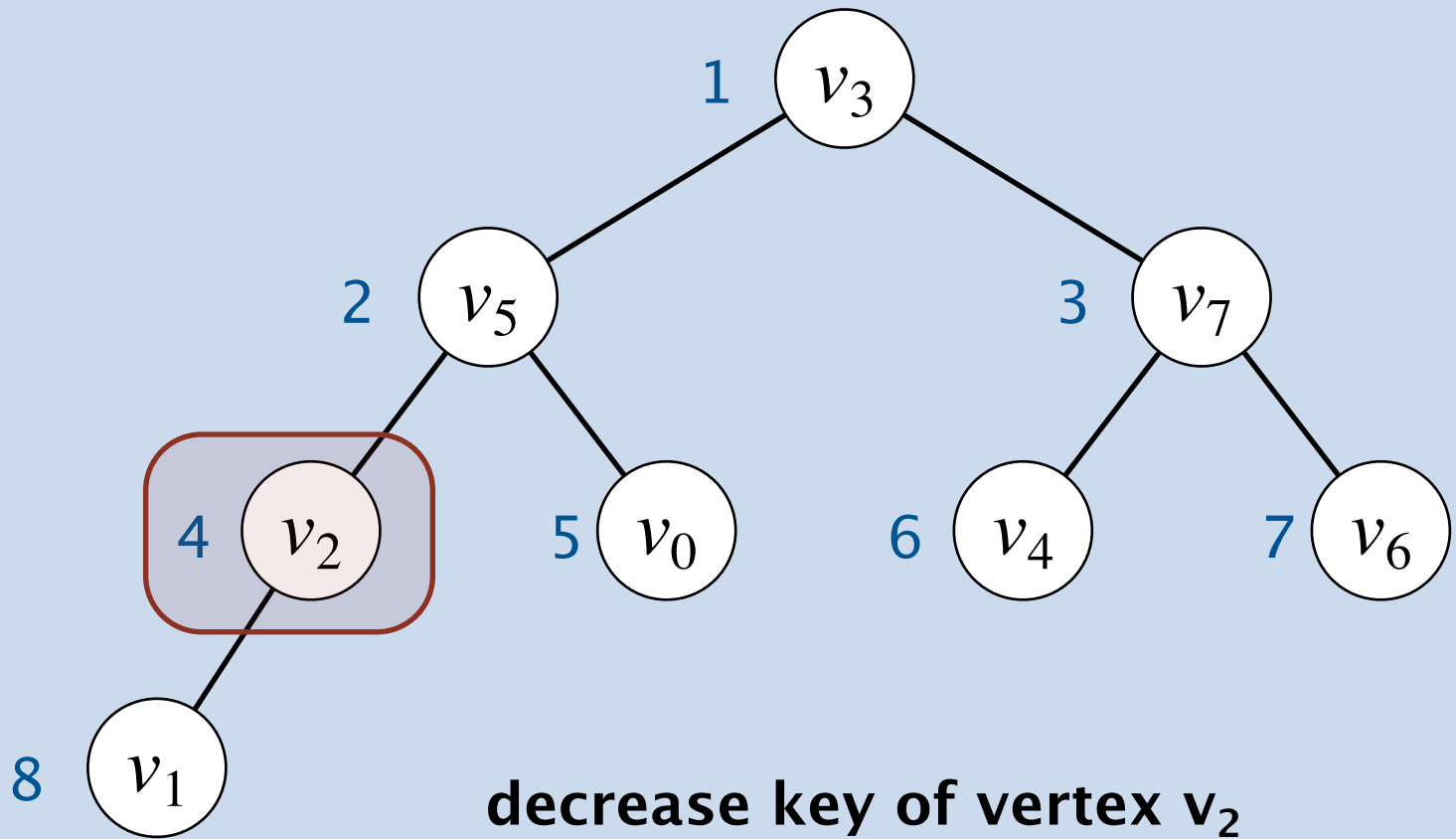
**Solution.**

- Find vertex in binary 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
<code>pq[]</code>	—	$v_3$	$v_5$	$v_7$	$v_2$	$v_0$	$v_4$	$v_6$	$v_1$
<code>qp[]</code>	5	8	4	1	6	2	4	3	—
<code>keys[]</code>	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 probably not worth implementing.

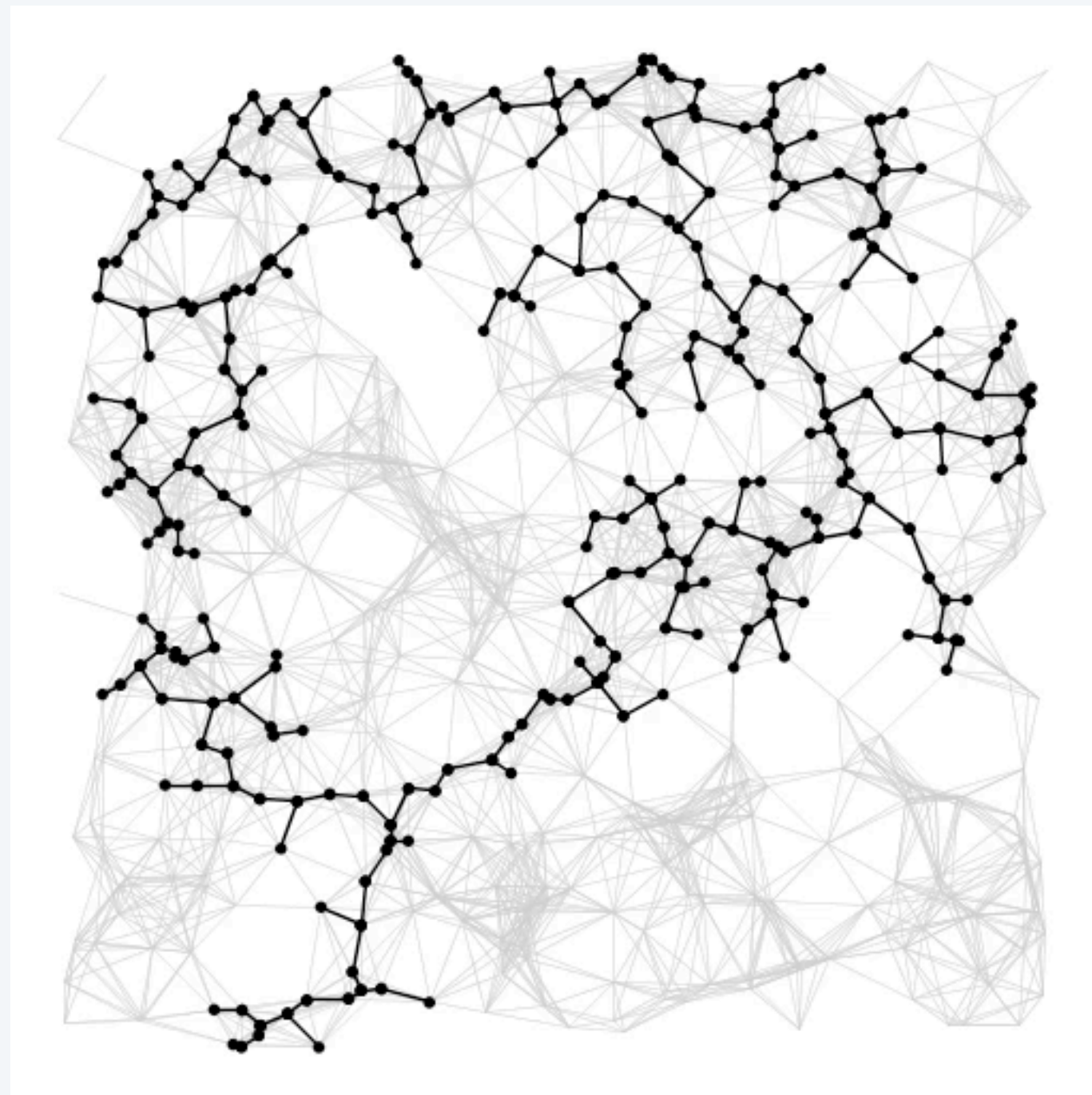


# Priority-first search

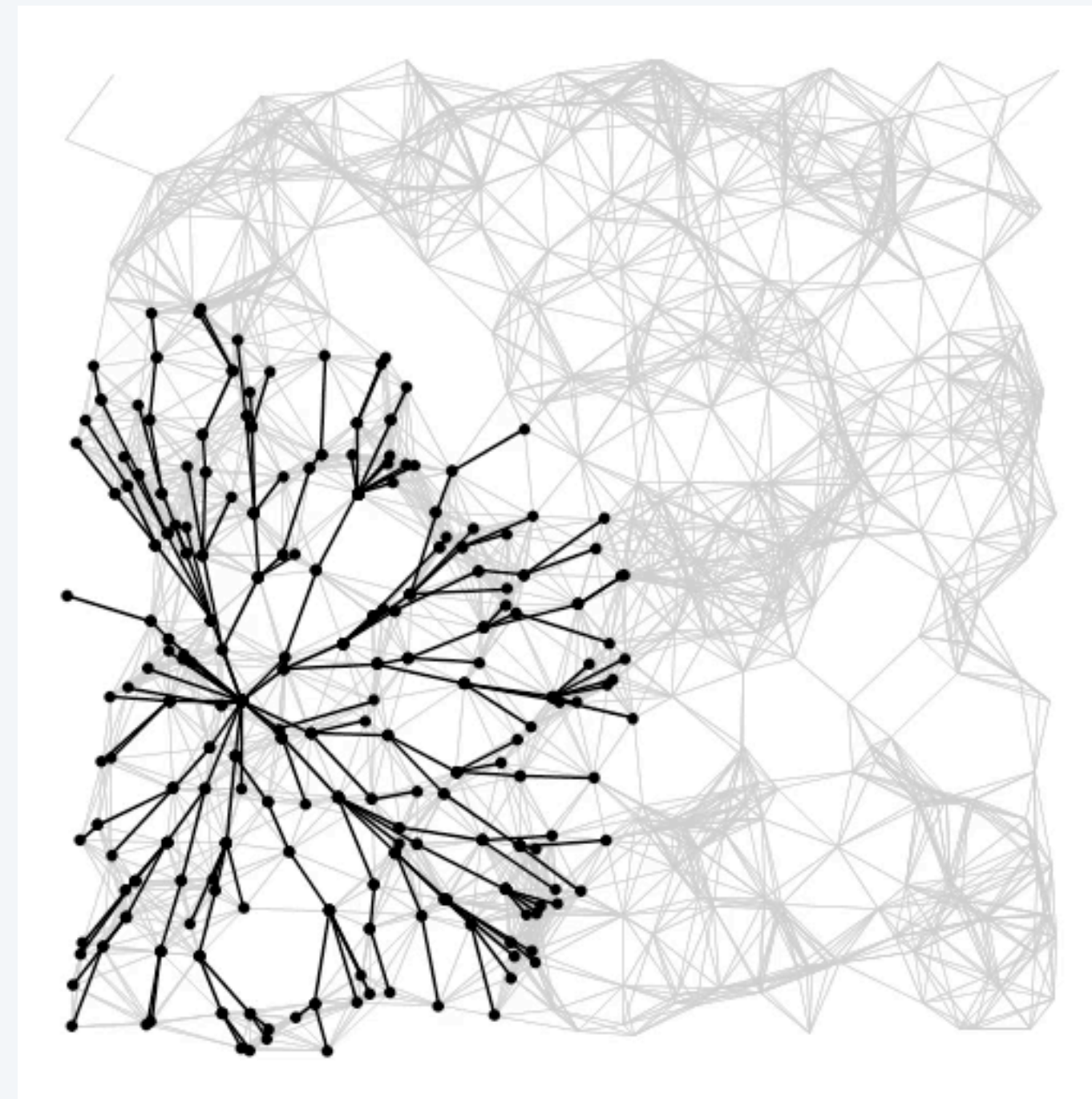
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**Observation.** Prim and Dijkstra are essentially the same algorithm.

- 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	$\Theta(E V)$	✓	✓
Dijkstra	$\Theta(E \log V)$		✓
topological sort	$E$	✓	

<sup>†</sup> no negative cycles

# Which shortest paths algorithm to use?

---

Select algorithm based on properties of edge-weighted digraph.

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

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

<sup>†</sup> no negative cycles

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## A final thought

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“ *Do only what only you can do.* ”

— Edsger W. Dijkstra

