

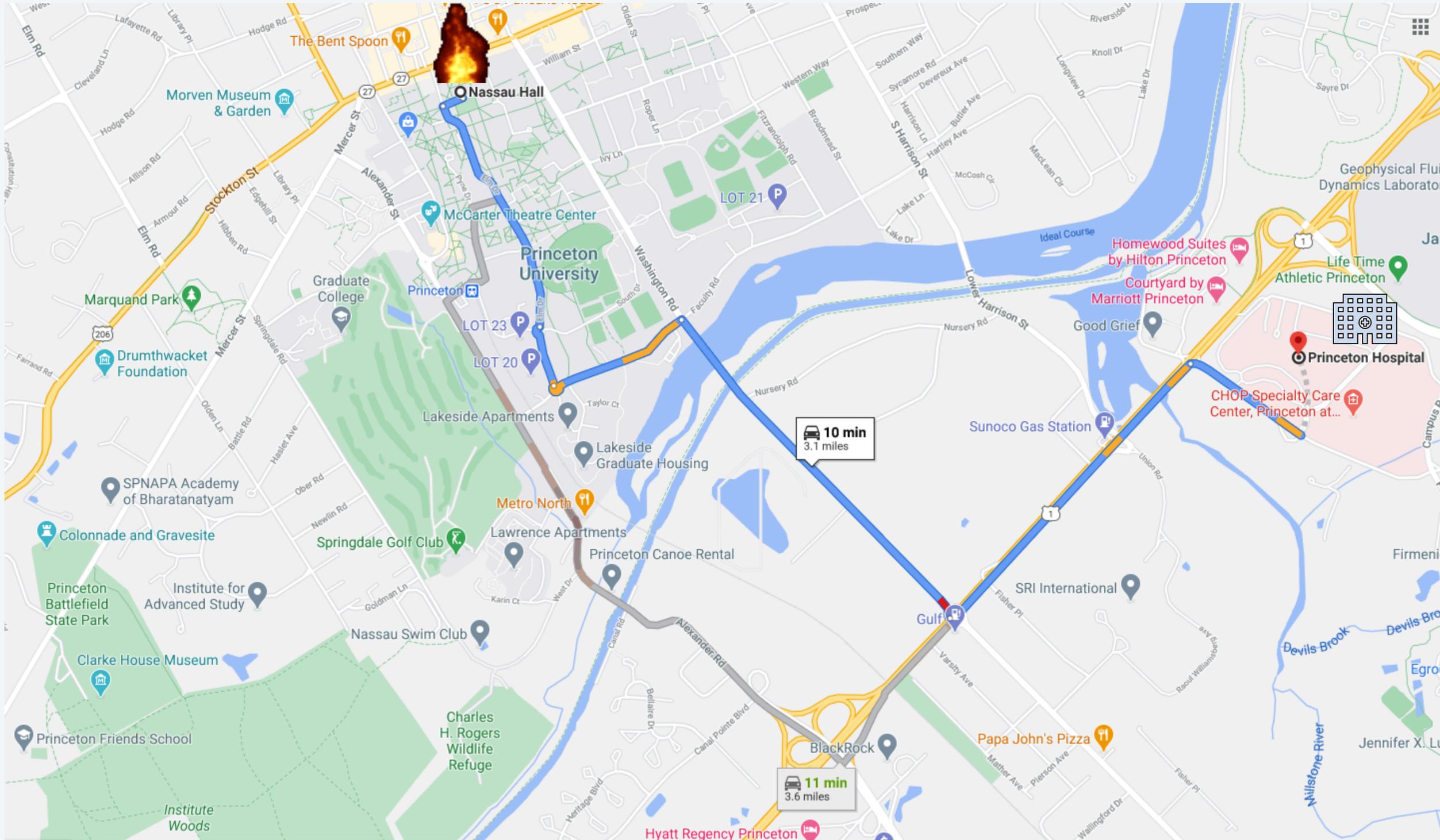
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## 4.4 SHORTEST PATHS

---

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

# Google maps

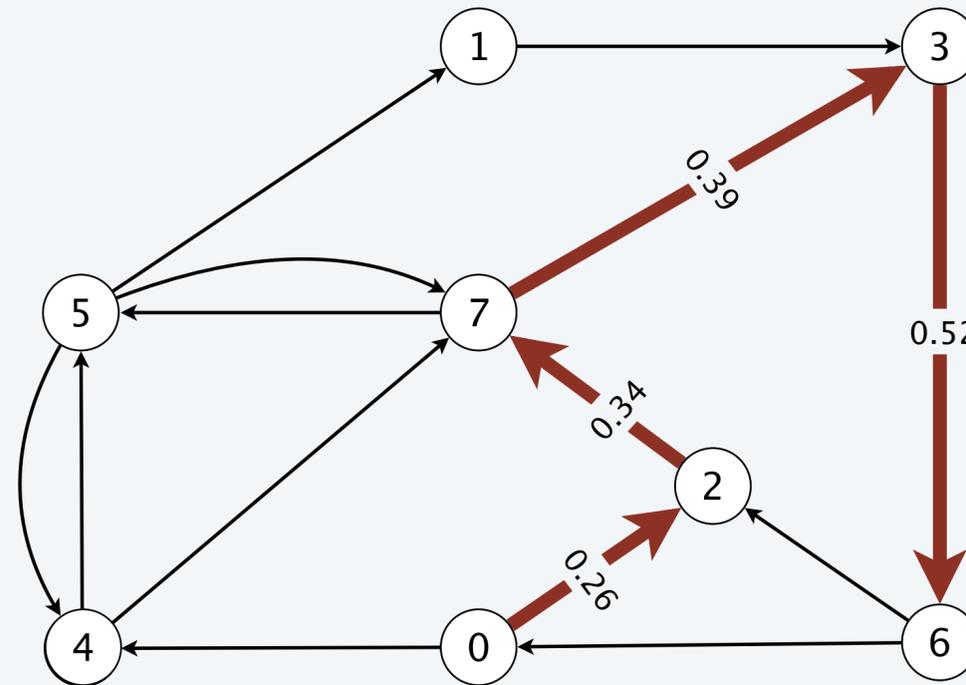


# Shortest path in an edge-weighted digraph

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  $\text{T}_E\text{X}$ .
- 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 (see next lecture)*
- Allow. → *implies that shortest path from  $s$  to  $v$  exists (and that  $E \geq V - 1$ )*

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



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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- ▶ *Dijkstra’s algorithm*

# Data structures for single-source shortest paths

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

*no repeated vertices*

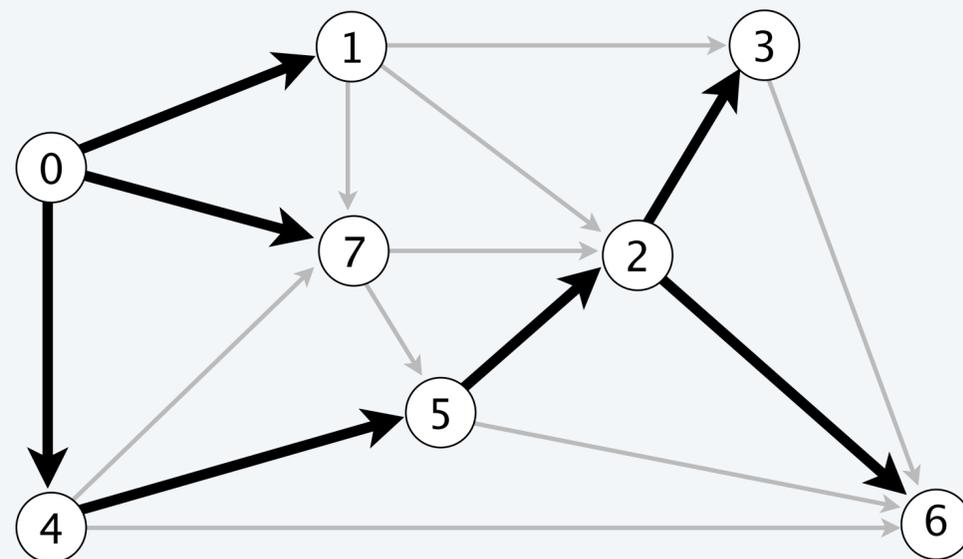
$\Rightarrow \leq V - 1$  edges

**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

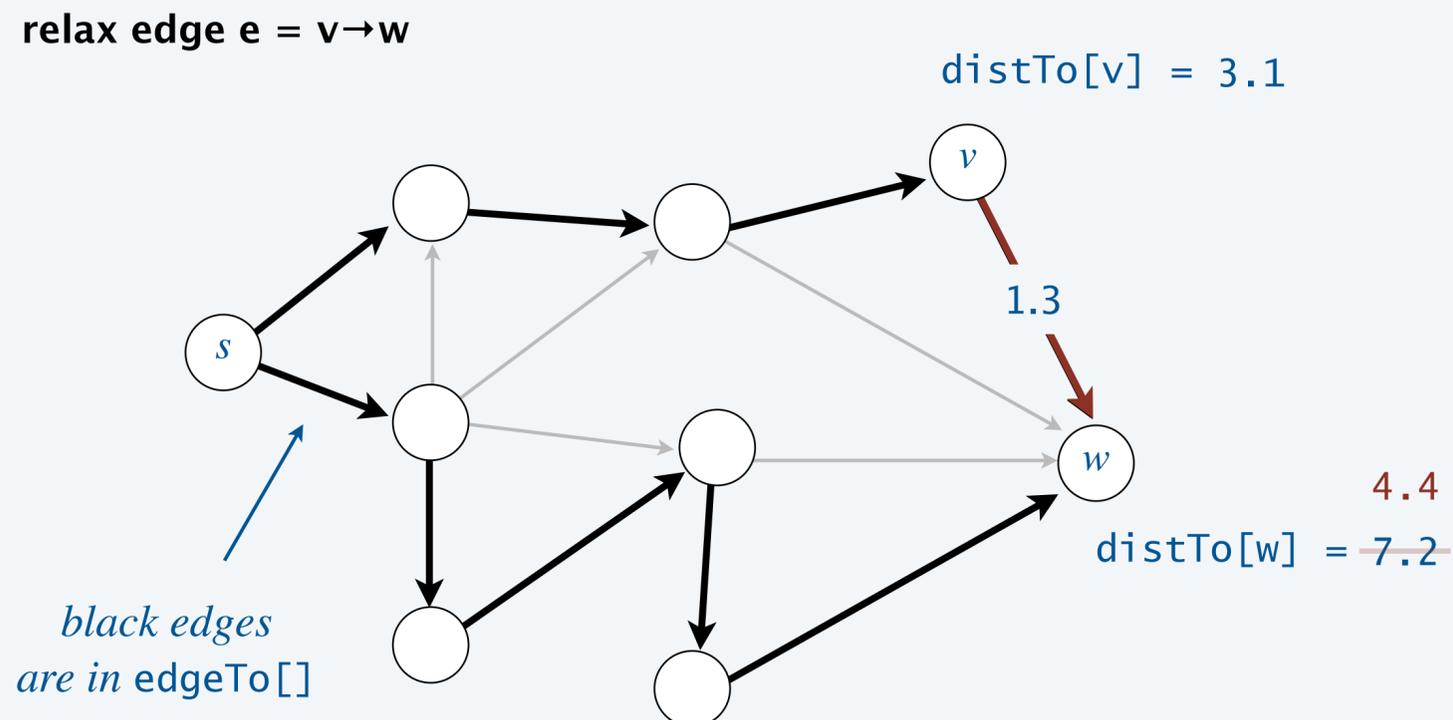
$v$	$\text{distTo}[]$	$\text{edgeTo}[]$
0	0.0	-
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0	2→6
7	8.0	0→7

parent-link representation

# Edge relaxation

Relax edge  $e = v \rightarrow w$ .

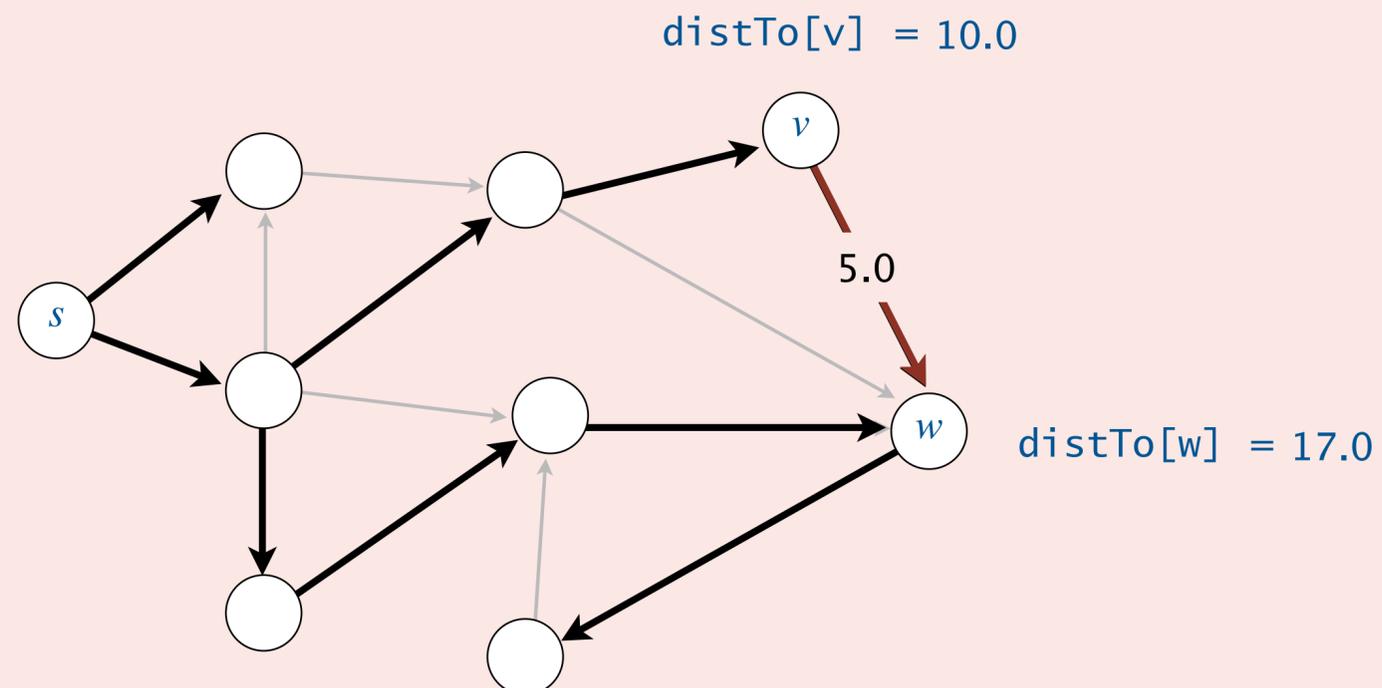
- $\text{distTo}[v]$  is length of shortest **known** path from  $s$  to  $v$ .
- $\text{distTo}[w]$  is length of shortest **known** path from  $s$  to  $w$ .
- $\text{edgeTo}[w]$  is last edge on shortest **known** path from  $s$  to  $w$ .
- If  $e = v \rightarrow w$  yields shorter path from  $s$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





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

---

## API.

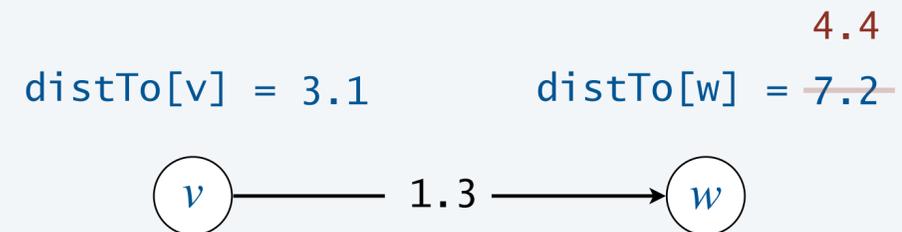
```
public class DirectedEdge
```

---

DirectedEdge(int v, int w, double weight)	<i>create weighted edge <math>v \rightarrow w</math></i>
int from()	<i>vertex <math>v</math></i>
int to()	<i>vertex <math>w</math></i>
double weight()	<i>weight of this edge</i>
⋮	⋮

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;  
    }
```

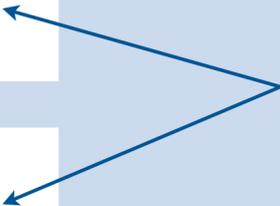
```
    public int from() {  
        return v;  
    }
```

```
    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
```

---

	<code>EdgeWeightedDigraph(int V)</code>	<i>edge-weighted digraph with <math>V</math> vertices (and no edges)</i>
<code>void</code>	<code>addEdge(DirectedEdge e)</code>	<i>add weighted directed edge <math>e</math></i>
<code>Iterable&lt;DirectedEdge&gt;</code>	<code>adj(int v)</code>	<i>edges incident from <math>v</math></i>
<code>int</code>	<code>V()</code>	<i>number of vertices</i>
	<code>⋮</code>	<code>⋮</code>

# 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)    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)*

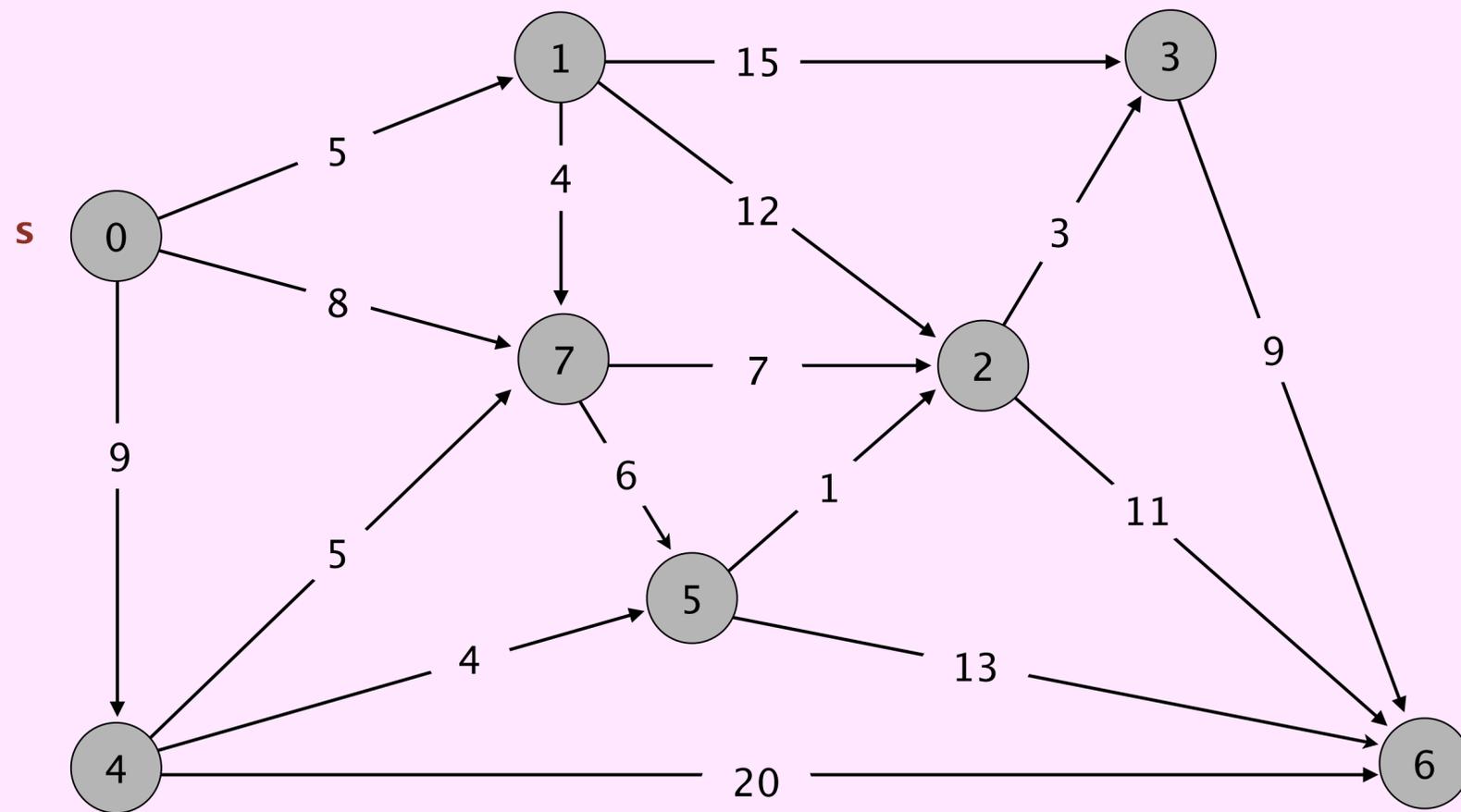
*number of calls to relax() in pass i =  
outdegree(0) + outdegree(1) + outdegree(2) + ... = E*

**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.



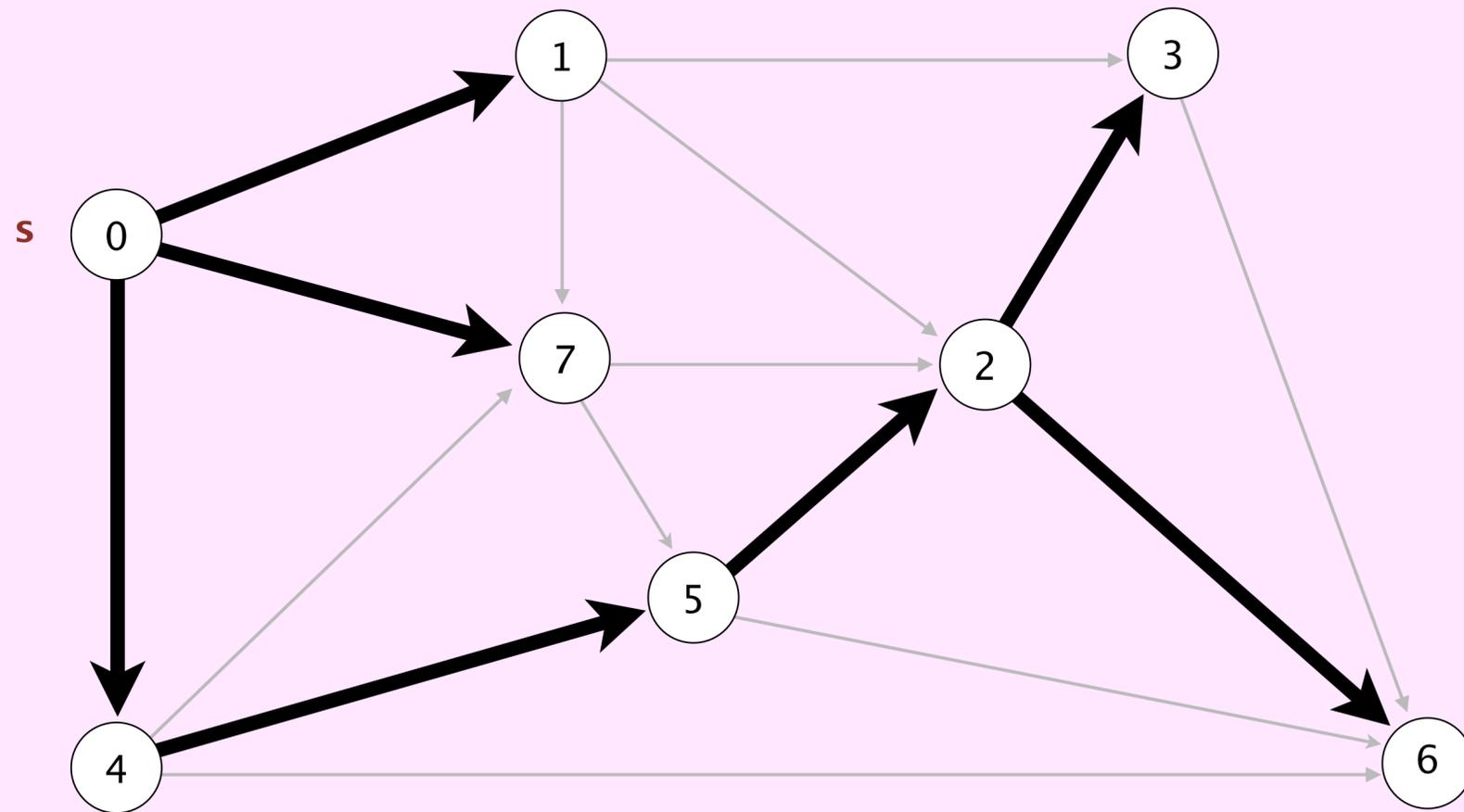
an edge-weighted digraph

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

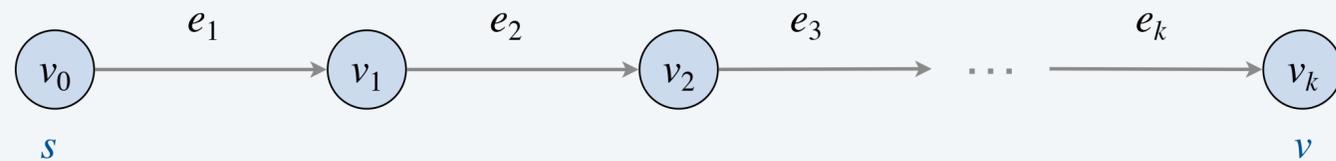
shortest-paths tree from vertex  $s$

# 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  $v$  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

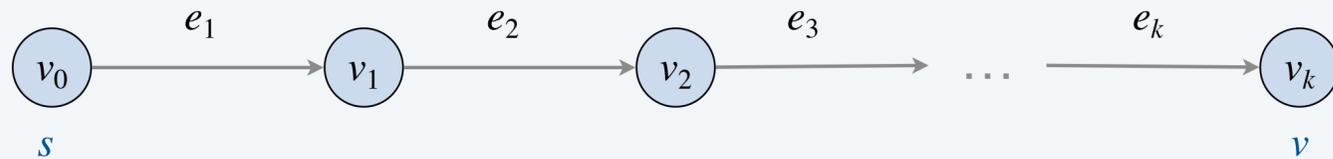
$$\begin{aligned} \text{distTo}[v_{i+1}] &\leq \text{distTo}[v_i] + \text{weight}(e_{i+1}) \quad \longleftarrow \text{edge relaxation} \\ &\leq \text{weight}(e_1) + \text{weight}(e_2) + \dots + \text{weight}(e_i) + \text{weight}(e_{i+1}). \quad \longleftarrow \text{inductive hypothesis} \end{aligned}$$

- 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  $v$  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  $v$  ]

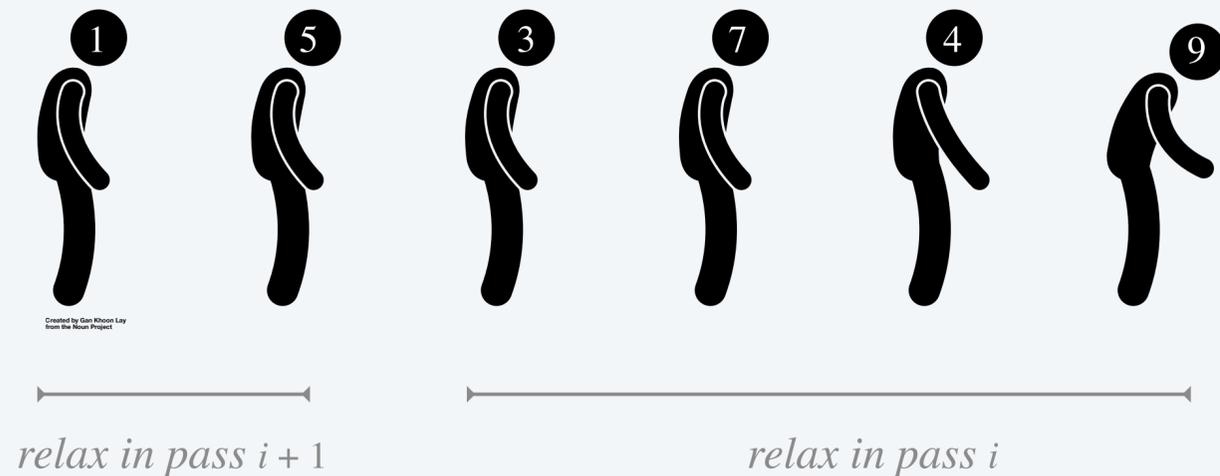
- There exists a simple shortest path  $P^*$  from  $s$  to  $v$ ; it contains  $k \leq V - 1$  edges.
- The Proposition implies that, after at most  $V - 1$  passes,  $\text{distTo}[v] \leq \text{length}(P^*)$ .
- Since  $\text{distTo}[v]$  is the length of some path from  $s$  to  $v$ ,  $\text{distTo}[v]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

# 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.

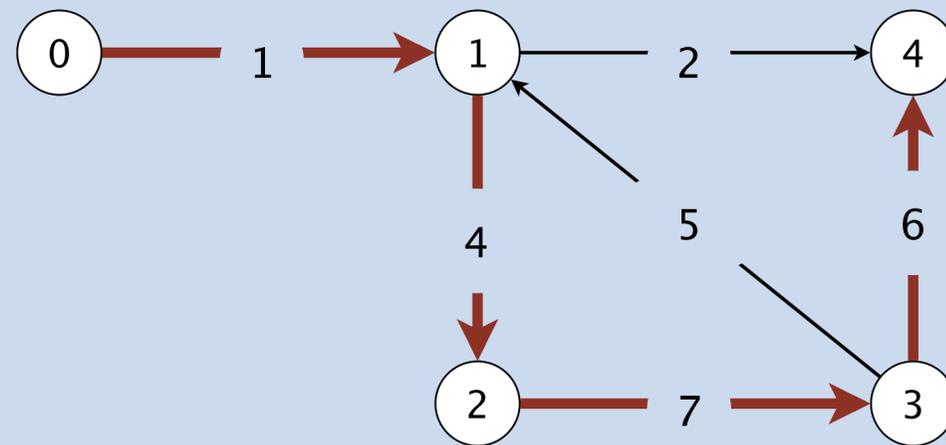
- In the worst case, the running time is still  $\Theta(E V)$ .
- But much faster in practice on typical inputs.

# 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 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**

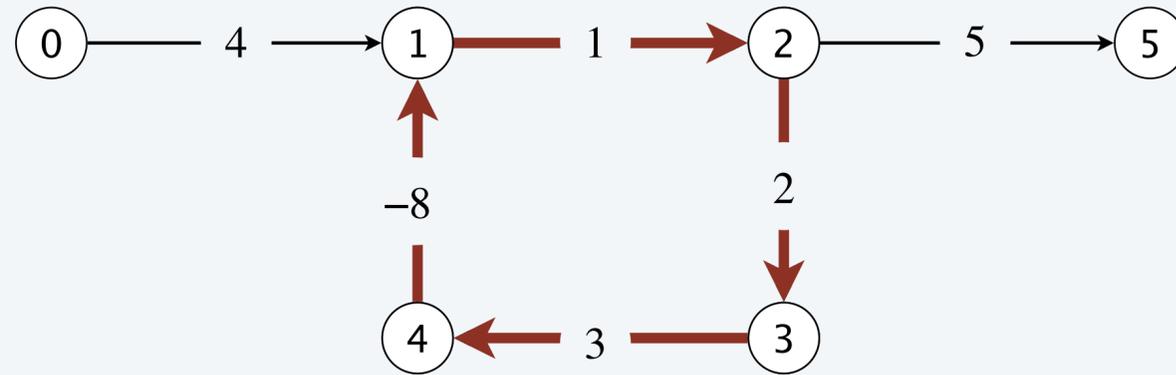
$(1 + 4 + 7 + 6)$

# Bellman–Ford algorithm: negative weights

---

**Remark.** The Bellman–Ford algorithm works even if some 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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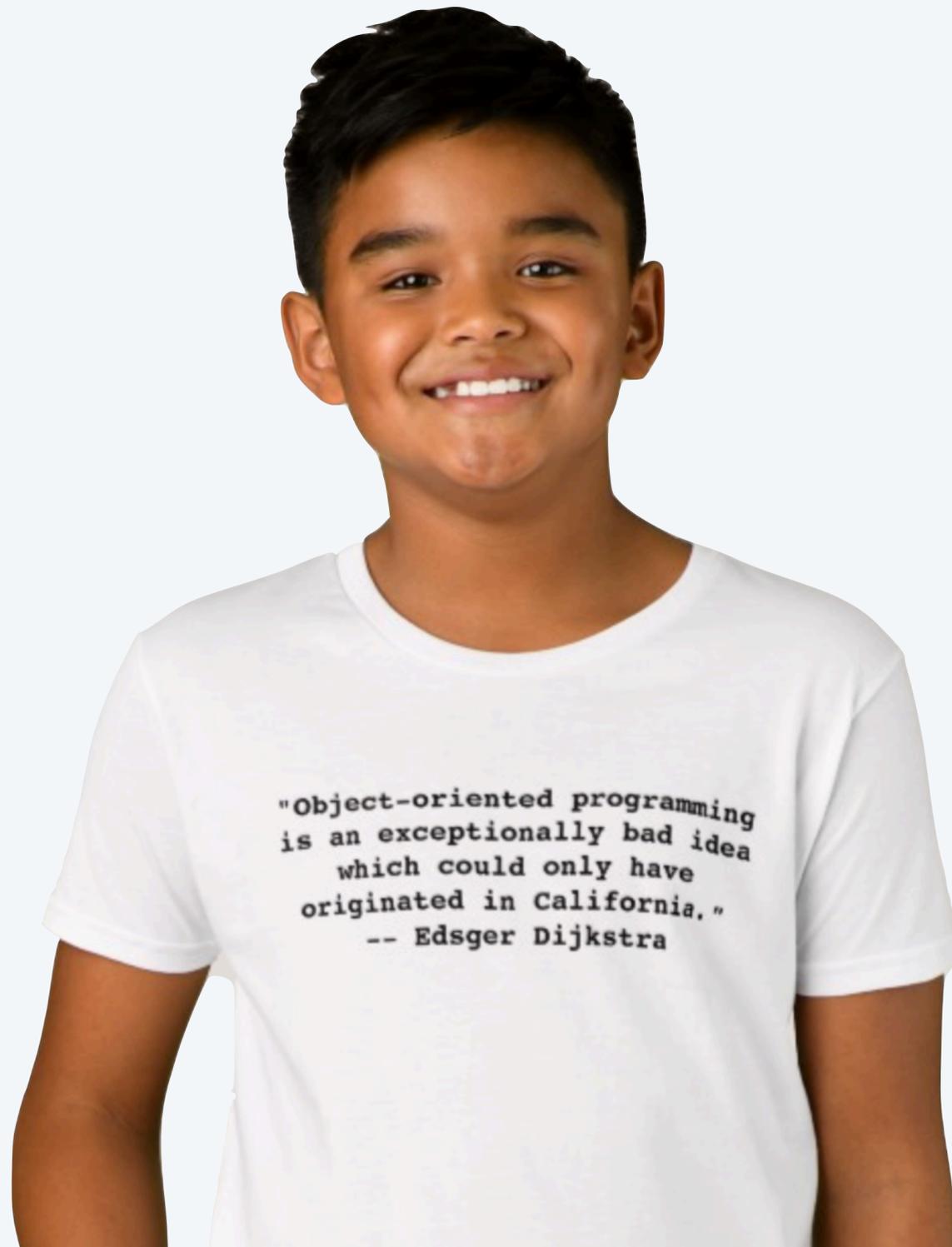
## 4.4 SHORTEST PATHS

---

- ▶ *properties*
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- ▶ *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  $v$ .
- 

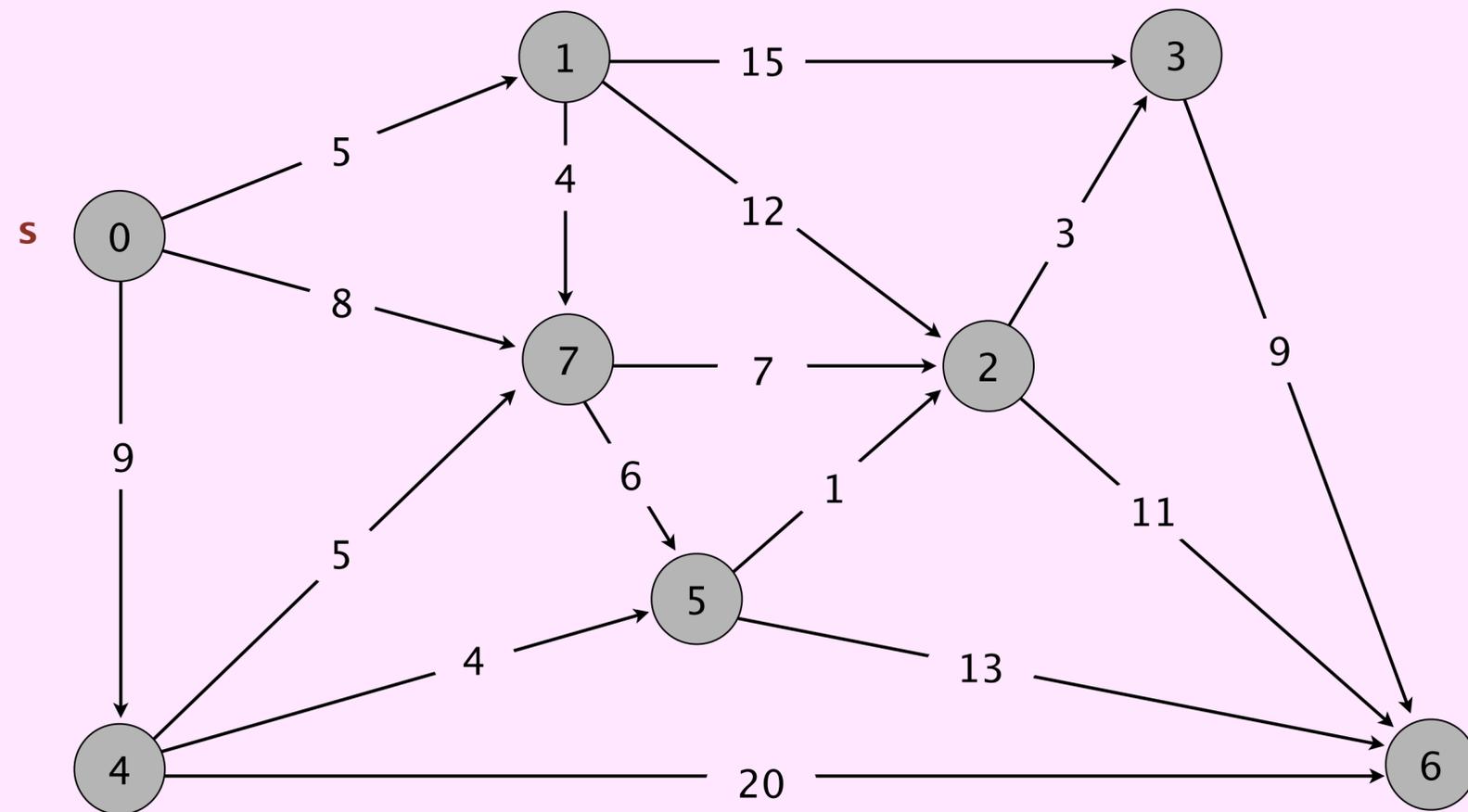
Key difference with Bellman–Ford. Each edge gets relaxed exactly once!

# 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

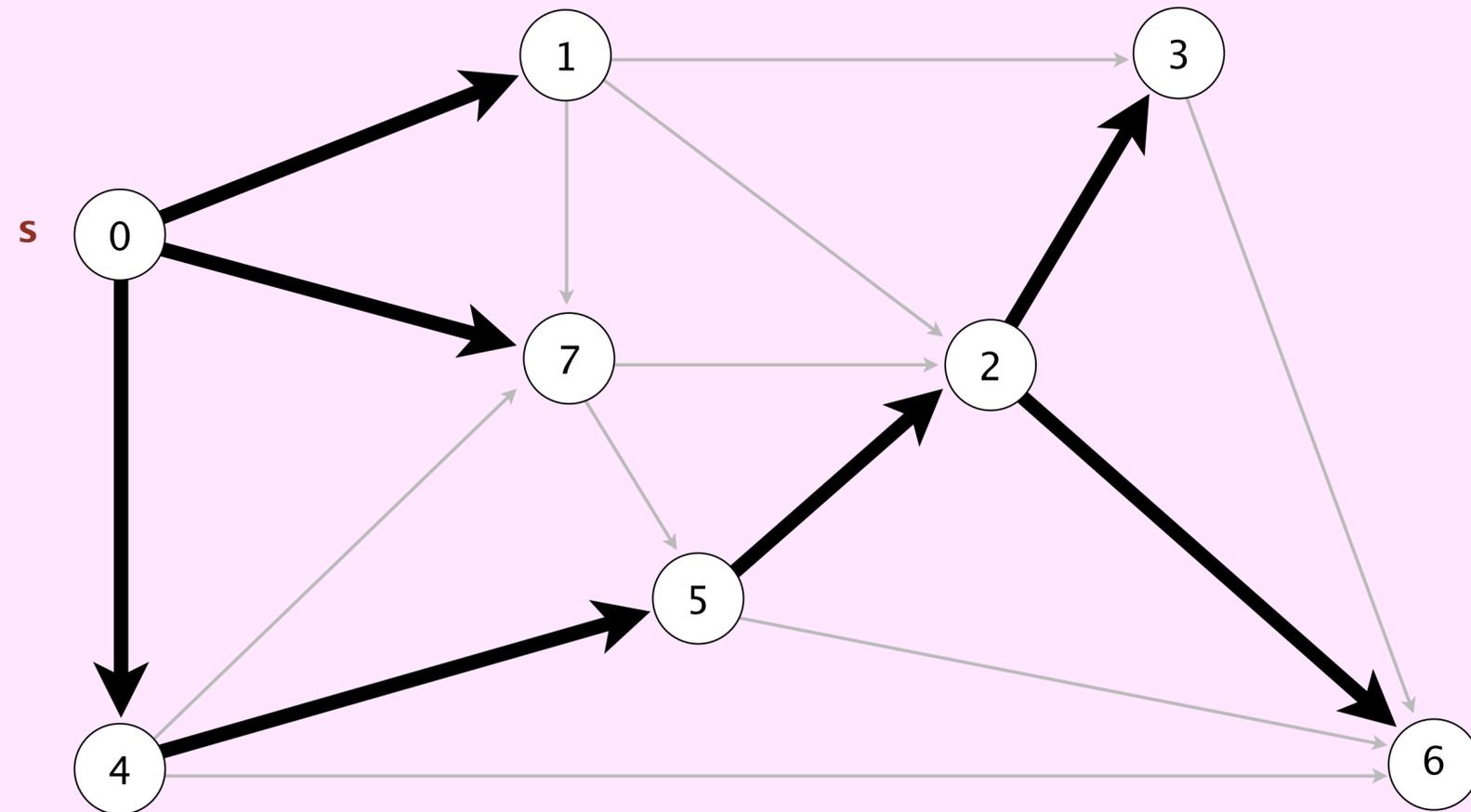
0→1	5.0
0→4	9.0
0→7	8.0
1→2	12.0
1→3	15.0
1→7	4.0
2→3	3.0
2→6	11.0
3→6	9.0
4→5	4.0
4→6	20.0
4→7	5.0
5→2	1.0
5→6	13.0
7→5	6.0
7→2	7.0

# 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  $v$ .



$v$	<code>distTo[]</code>	<code>edgeTo[]</code>
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  $s$

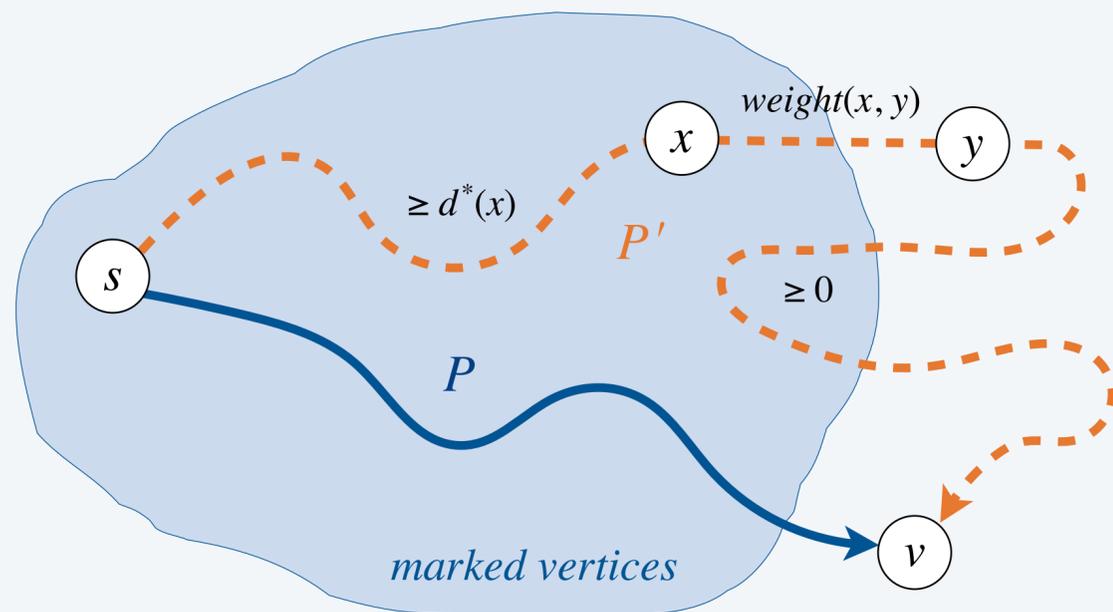
# Dijkstra's algorithm: correctness proof

**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  $y$*

$$\longrightarrow \leq \text{distTo}[y]$$

*vertex  $x$  is marked (so it was relaxed)*

$$\longrightarrow \leq \text{distTo}[x] + \text{weight}(x, y)$$

*induction*

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

*$P'$  is a path from  $s$  to  $x$ , followed by edge  $x \rightarrow y$ , followed by non-negative edges*

$$\longrightarrow \leq \text{length}(P') \quad \blacksquare$$

# 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>>
```

---

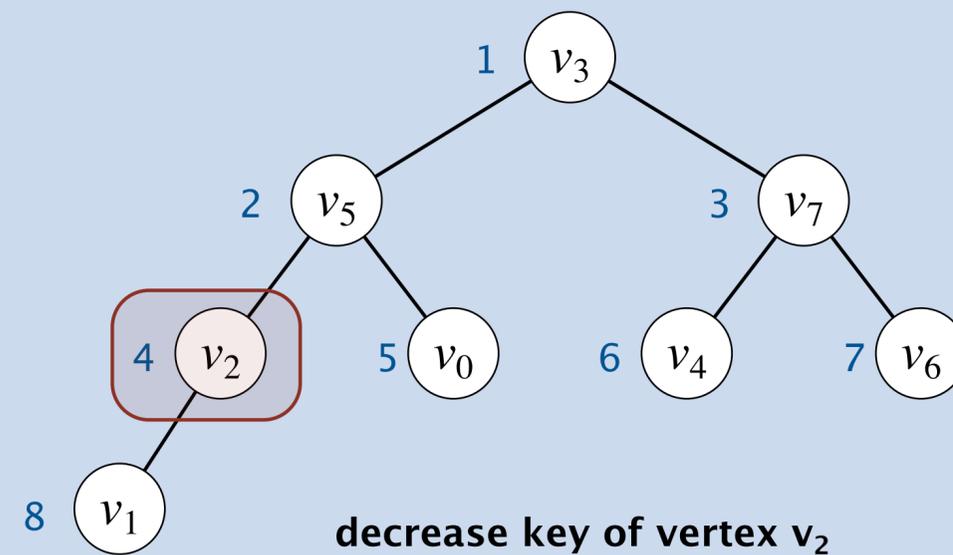
<code>IndexMinPQ(int n)</code>	<i>create PQ with indices 0, 1, ..., n - 1</i>
<code>void insert(int i, Key key)</code>	<i>associate key with index i</i>
<code>int delMin()</code>	<i>remove min key and return associated index</i>
<code>void decreaseKey(int i, Key key)</code>	<i>decrease the key associated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

# 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.

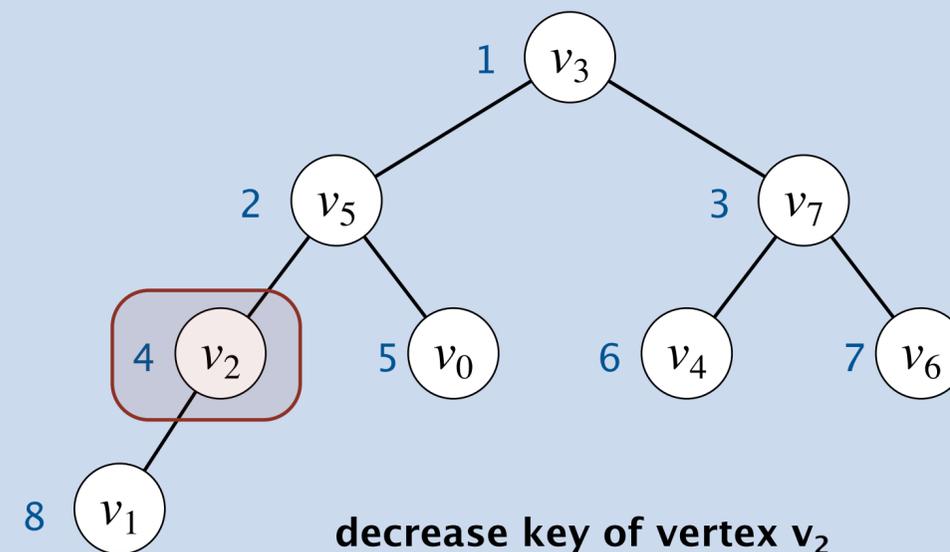
**Solution.**

- 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
<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.

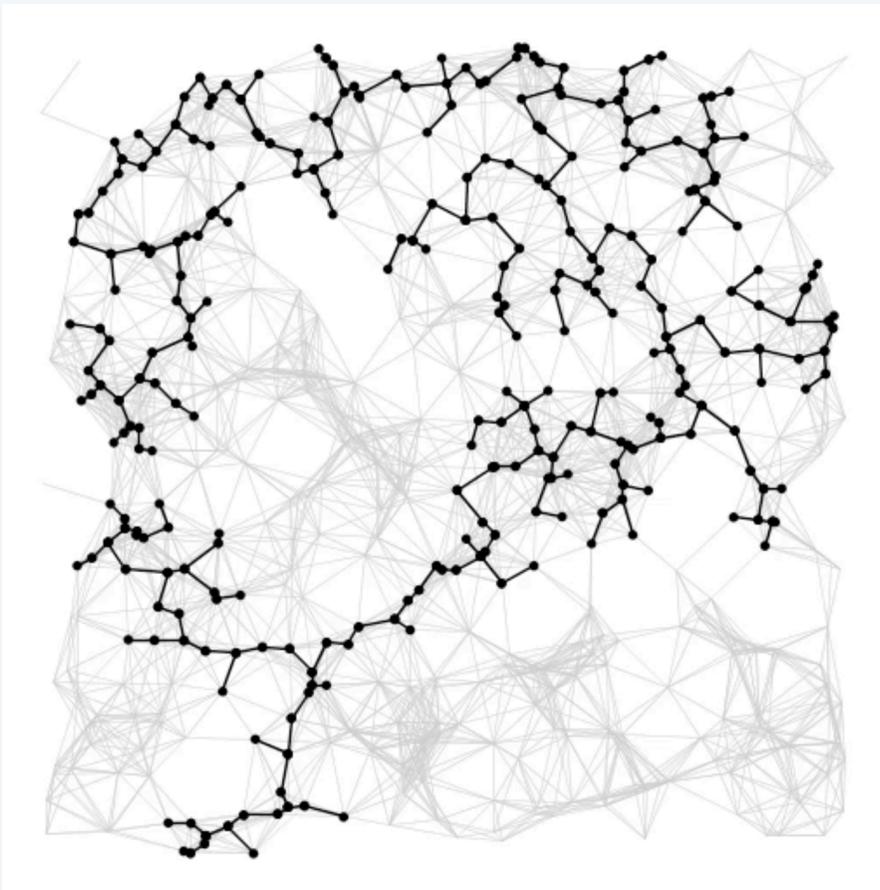
- Array implementation optimal for complete digraphs.
- Binary heap much faster for sparse digraphs.
- 4-way heap worth the trouble in performance-critical situations.
- Fibonacci heap best in theory, but probably not worth implementing.

# Priority-first search

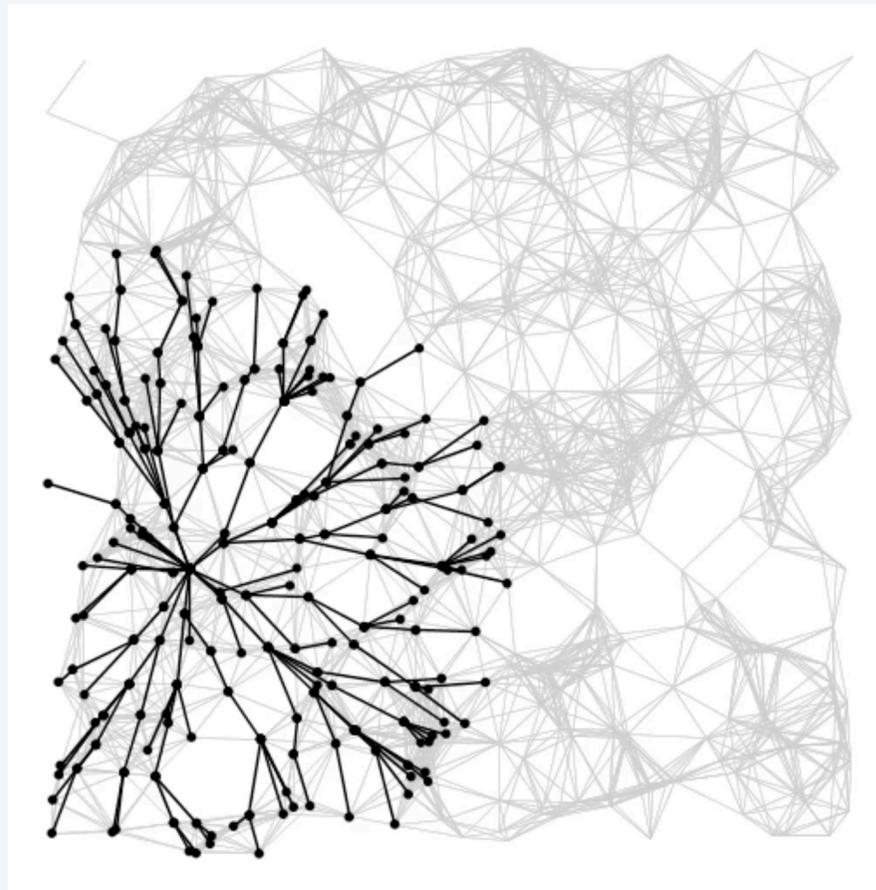
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**Observation.** Prim and Dijkstra are essentially the same algorithm.

- 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 †	directed cycles
<b>Bellman–Ford</b>	$E V$	✓	✓
<b>Dijkstra</b>	$E \log V$		✓
topological sort	$E$	✓	

† 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
<b>Bellman–Ford</b>	$E V$	✓	✓
<b>Dijkstra</b>	$E \log V$		✓
topological sort	$E$	✓	

† no negative cycles

# Credits

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## A final thought

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*“ Do only what only you can do. ”*

— Edsger W. Dijkstra

