# Algorithms



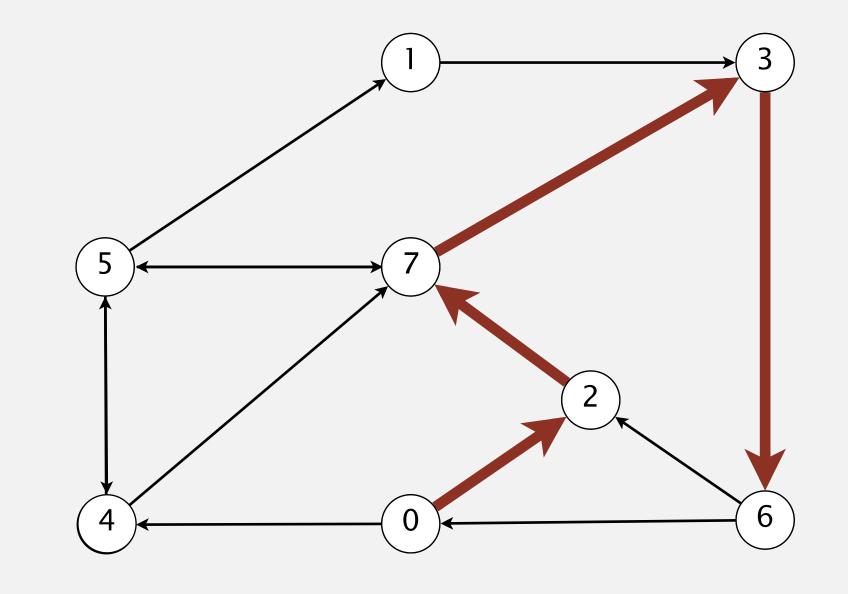
## Shortest paths in an edge-weighted digraph

Given an edge-weighted digraph, find the shortest path from s to t.

#### edge-weighted digraph

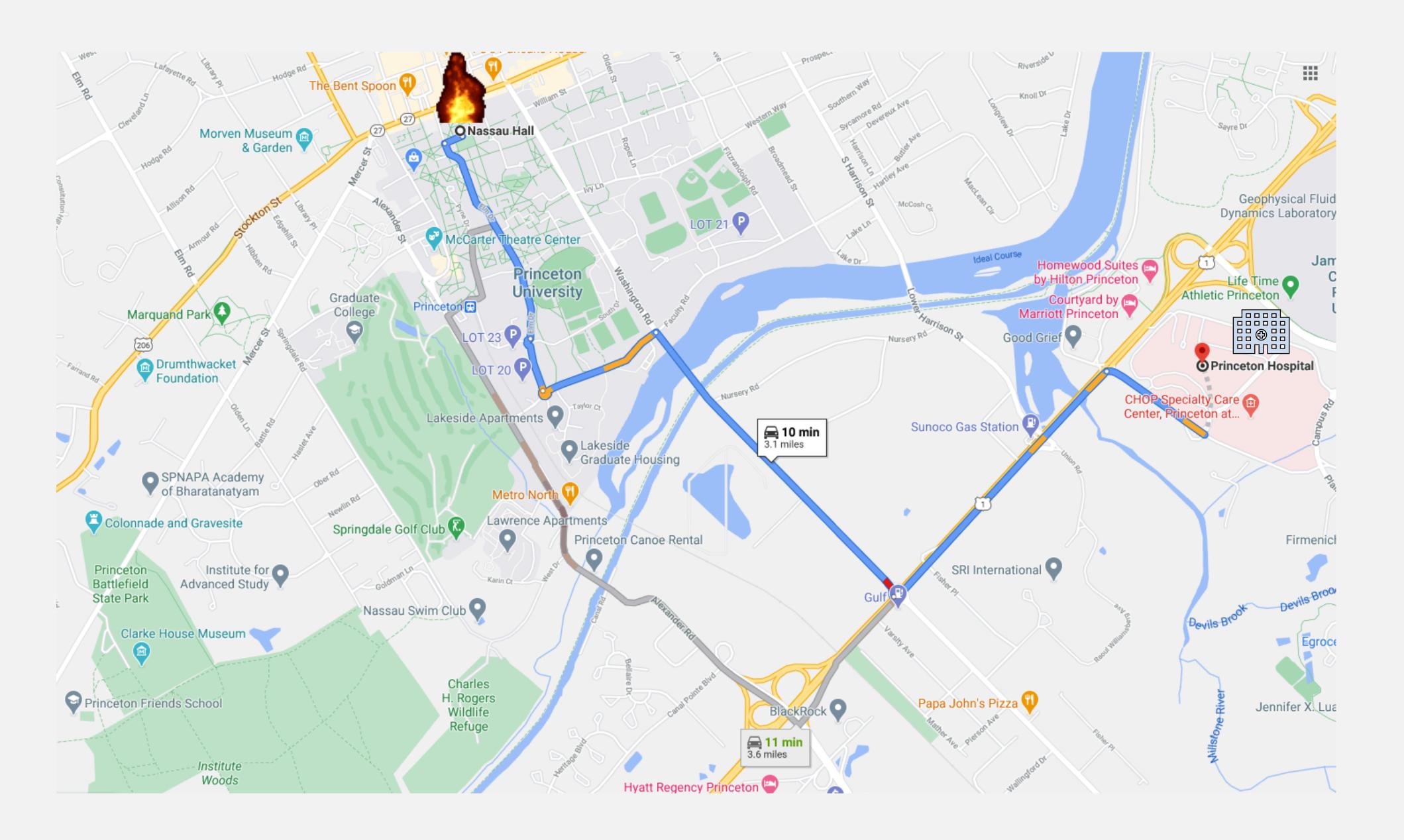
4->5	0.35
5->4	0.35
4->7	0.37
5->7	0.28
7->5	0.28
5->1	0.32
0->4	0.38
0->2	0.26
7->3	0.39
1->3	0.29
2->7	0.34
6->2	0.40
3->6	0.52
6->0	0.58

 $6 -> 4 \quad 0.93$ 



shortest path from 0 to 6 length of path = 1.51  $0 \rightarrow 2 \rightarrow 7 \rightarrow 3 \rightarrow 6$  (0.26 + 0.34 + 0.39 + 0.52)

## Google maps



### Shortest path applications

- PERT/CPM.
- Map routing.
- Seam carving. ← see Assignment 6
- Texture mapping.
- Robot navigation.
- Typesetting in  $T_EX$ .
- 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

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?

- Single source: from one vertex *s* to every other vertex.
- Single sink: from every vertex to one vertex t.
- Source–sink: from one vertex s to another t.
- All pairs: between all pairs of vertices.

#### Restrictions on edge weights?

- Euclidean weights.
- Arbitrary weights.

#### Cycles?

- No directed cycles.
- No "negative cycles."

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

Simplifying assumption. Each vertex is reachable from s.

## Shortest paths: quiz 1



#### Which variant in car GPS?

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





#### Data structures for single-source shortest paths

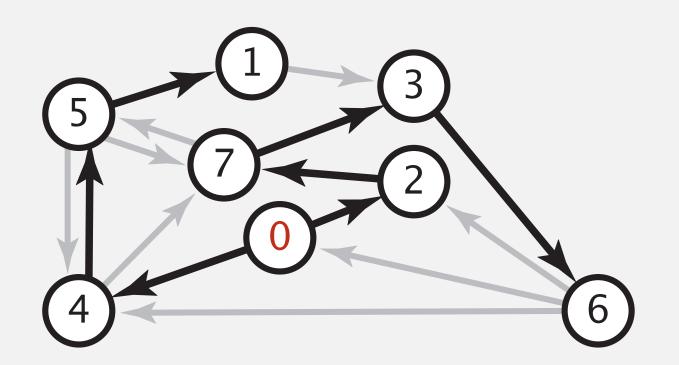
Goal. Find a shortest path from s to every other vertex.

Observation 1. There exists a shortest path from s to v that is simple.  $\longrightarrow$  no repeated vertices  $\Rightarrow \leq V-1$  edges

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

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

- distTo[v] is length of a shortest path from s to v.
- edgeTo[v] is last edge on a shortest path from s to v.



	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

shortest-paths tree from 0

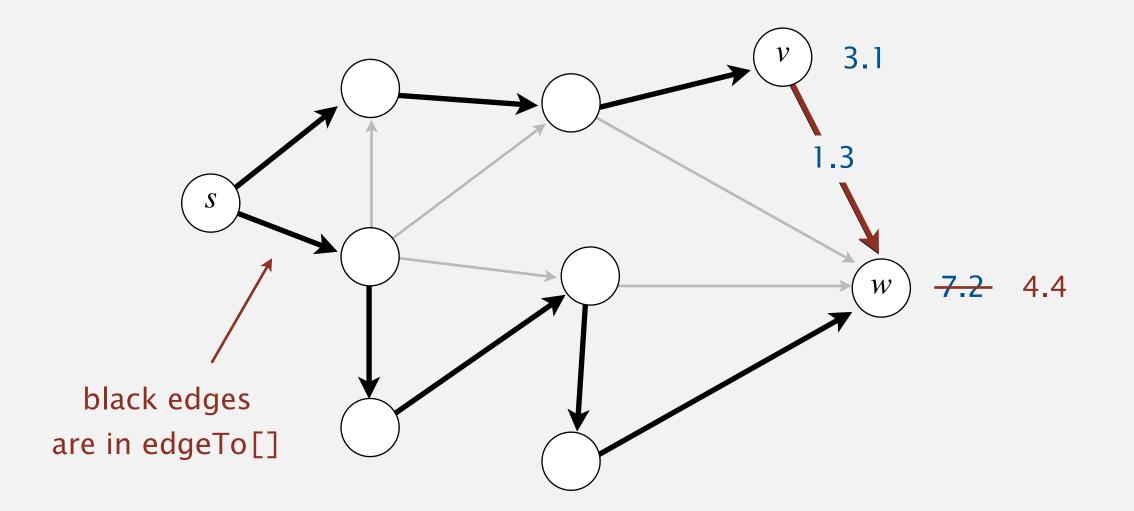
parent-link representation

## Edge relaxation

#### Relax edge $e = v \rightarrow w$ .

- distTo[v] is length of shortest known path from s to v.
- distTo[w] is length of shortest known path from s to w.
- 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 distTo[w] and edgeTo[w].

#### relax edge e = v→w

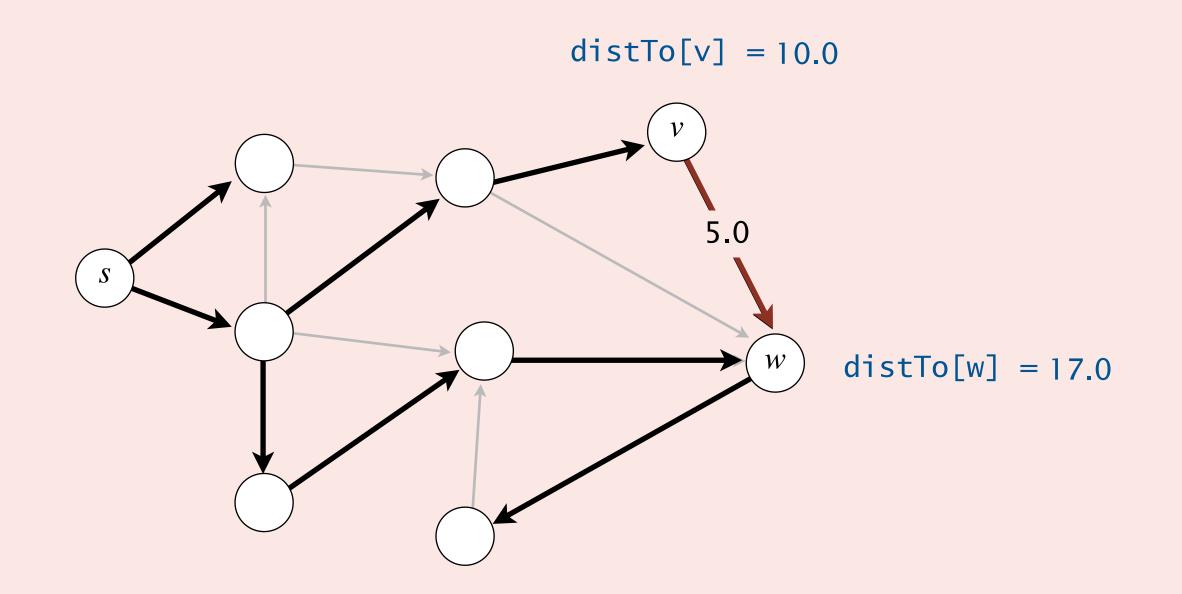


## Shortest paths: quiz 2



### What are the values of distTo[v] and distTo[w] after relaxing $e = v \rightarrow w$ ?

- 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:  $distTo[v] = \infty$ .

For each vertex v: edgeTo[v] = null.

distTo[s] = 0.

Repeat until done:

- Relax any edge.

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

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

## Framework for shortest-paths algorithm

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

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

For each vertex v: edgeTo[v] = null.

distTo[s] = 0.

Repeat until done:

- Relax any edge.

#### Efficient implementations.

- Which edge to relax next?
- How many edge relaxations needed?
- Ex 1. Bellman-Ford algorithm.
- Ex 2. Dijkstra's algorithm.
- Ex 3. Topological sort algorithm.



## Weighted directed edge API

#### 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.

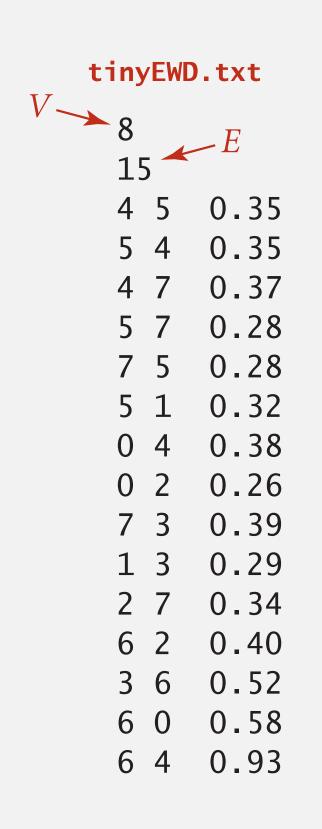
```
public class DirectedEdge
   private final int v, w;
   private final double weight;
   public DirectedEdge(int v, int w, double weight)
     this.v = v;
     this.w = w;
     this.weight = weight;
   public int from()
                                                                      from() and to() replace
   { return v; }
                                                                      either() and other()
   public int to()
   { return w; }
   public double weight()
   { return weight; }
```

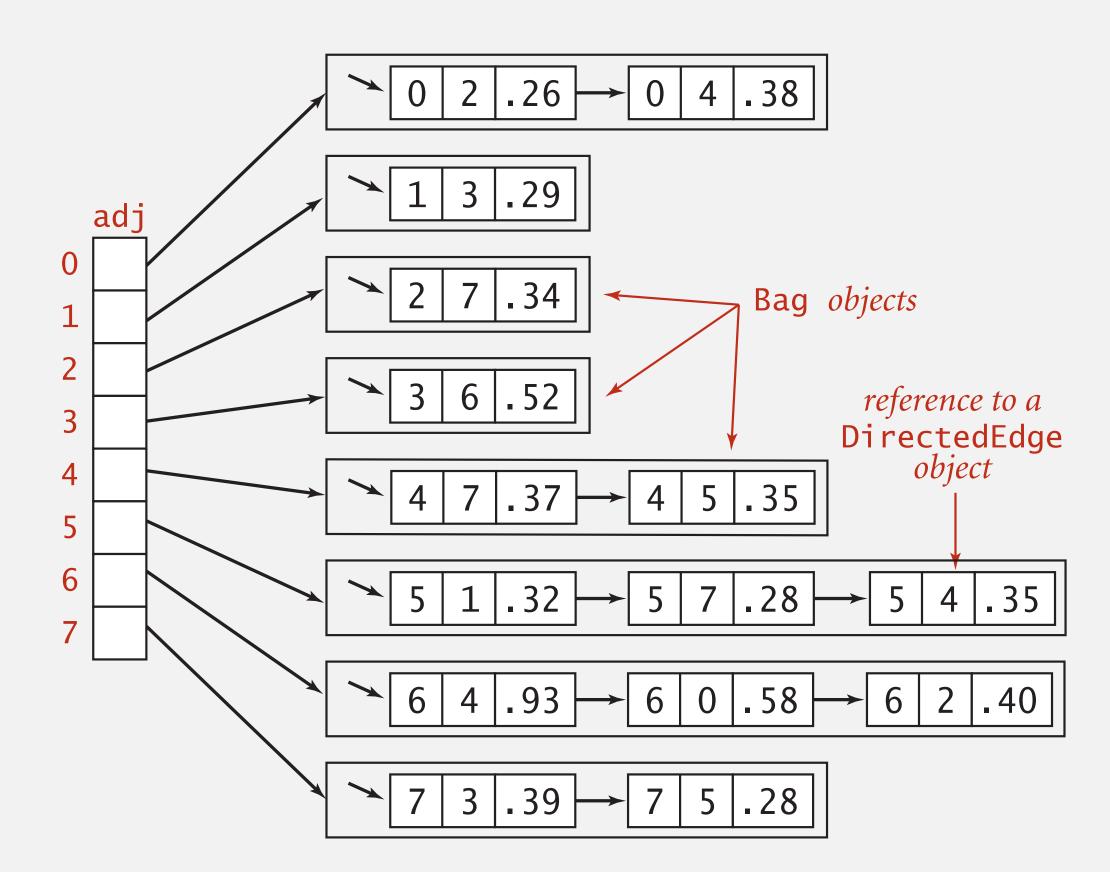
# Edge-weighted digraph API

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></directededge>	adj(int v)	edges incident from v
int	V()	number of vertices
	• •	•

## Edge-weighted digraph: adjacency-lists representation





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

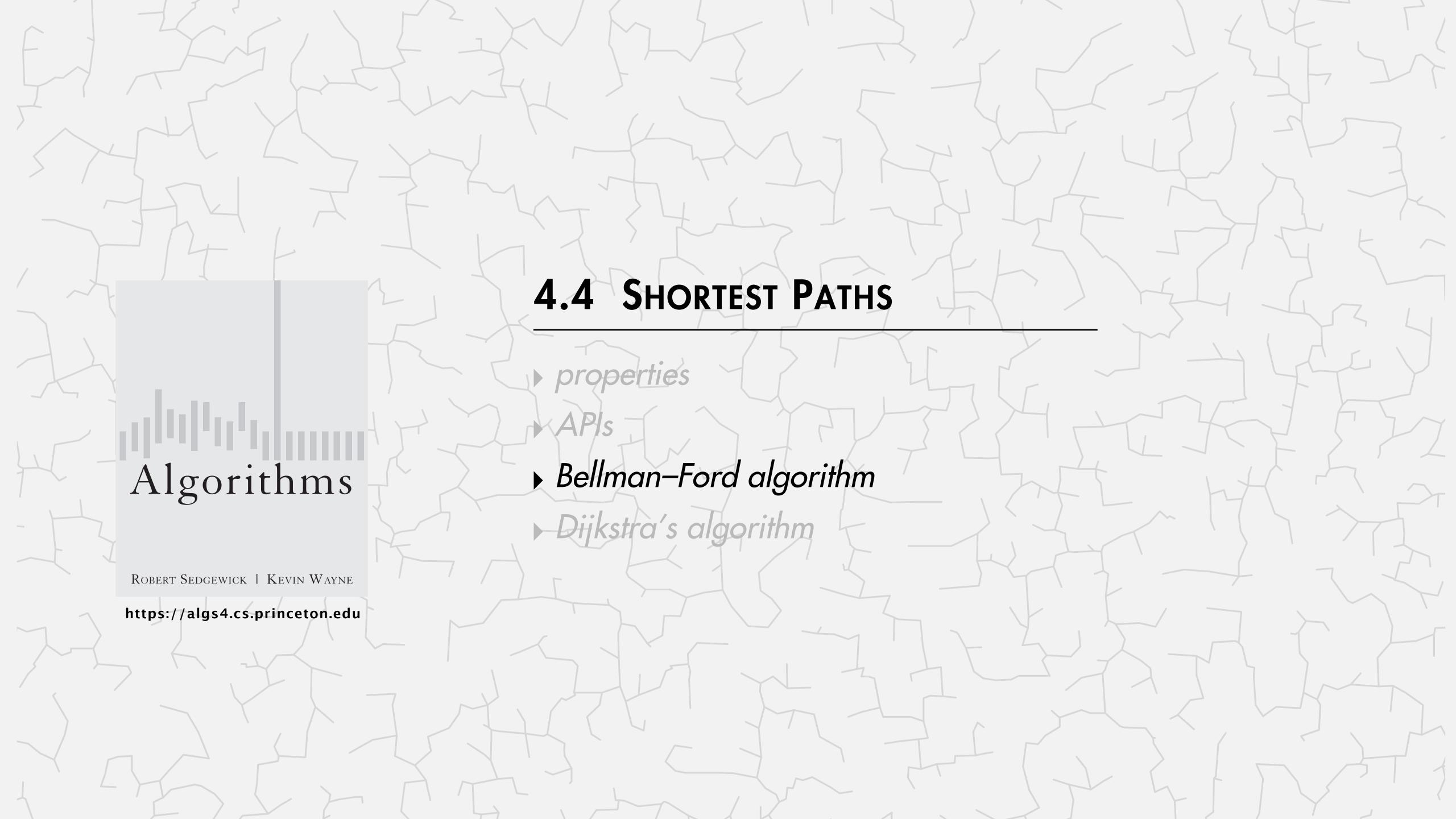
Implementation. Almost identical to EdgeWeightedGraph.

```
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(), w = e.to();
                                                             add edge e = v \rightarrow w to
                                                             only v's adjacency list
     adj[v].add(e);
   public Iterable<DirectedEdge> adj(int v)
   { return adj[v]; }
```

# 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	<pre>distTo(int v)</pre>	length of shortest path from s to v
Iterable <directededge></directededge>	pathTo(int v)	shortest path from s to v
boolean	hasPathTo(int v)	is there a path from s to v?



## Bellman-Ford algorithm

#### Bellman-Ford algorithm

```
For each vertex v: distTo[v] = \infty.
```

For each vertex v: edgeTo[v] = null.

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))

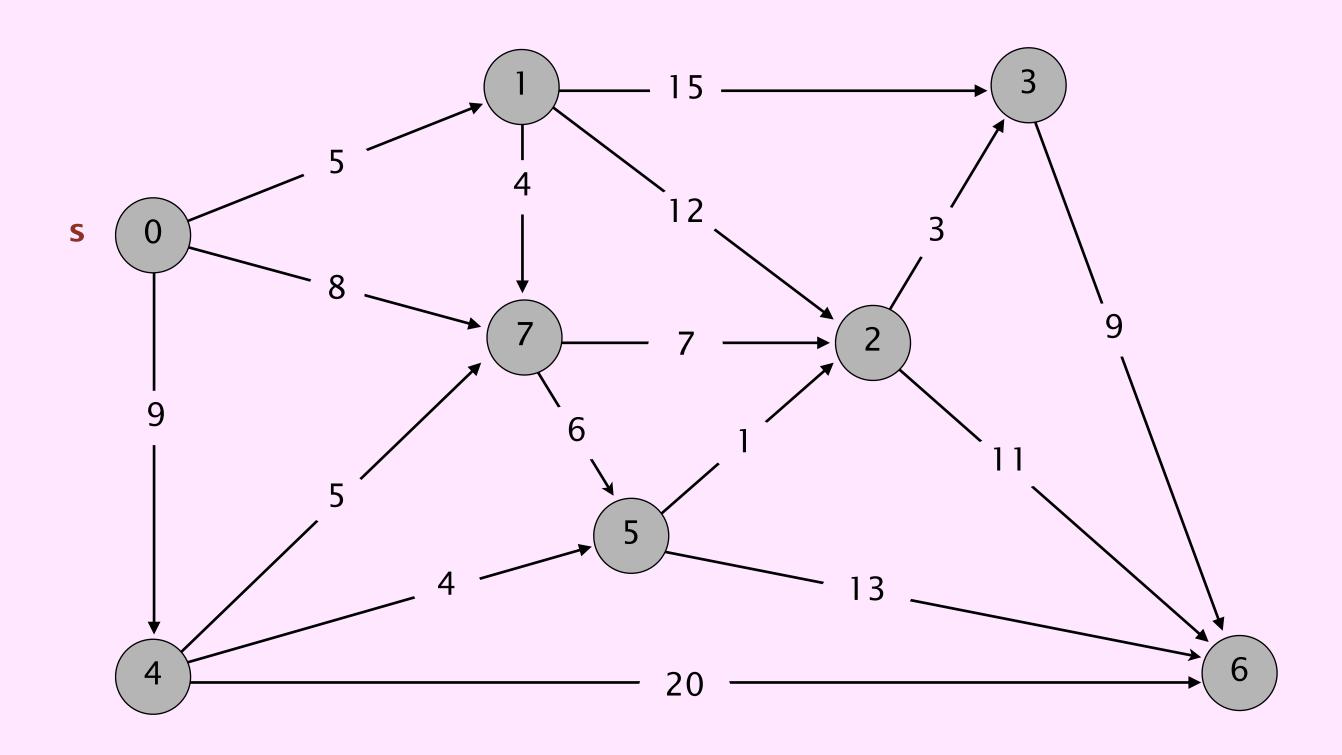
relax(e);
```

Running time. Algorithm takes  $\Theta(E\ V)$  time in both best- and worst-case.

# Bellman-Ford algorithm demo



Repeat V-1 times: relax all E edges.



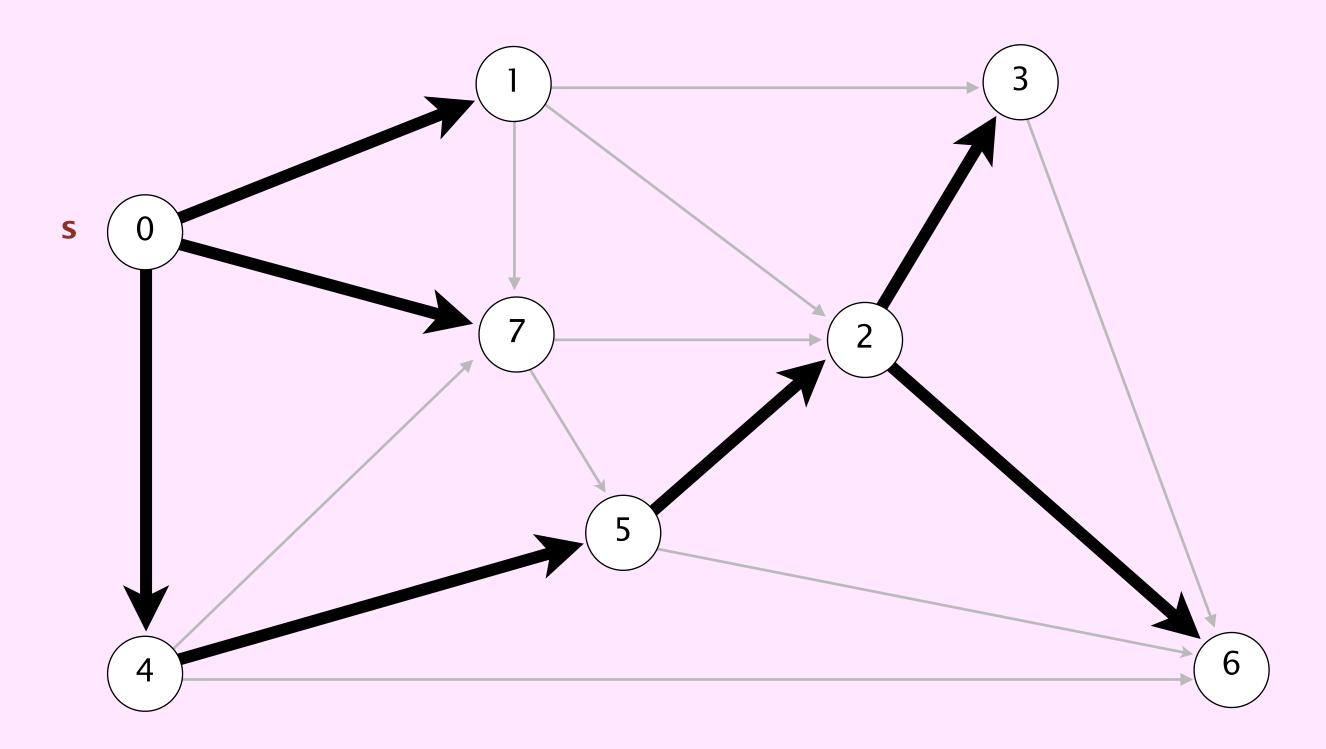
an edge-weighted digraph

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



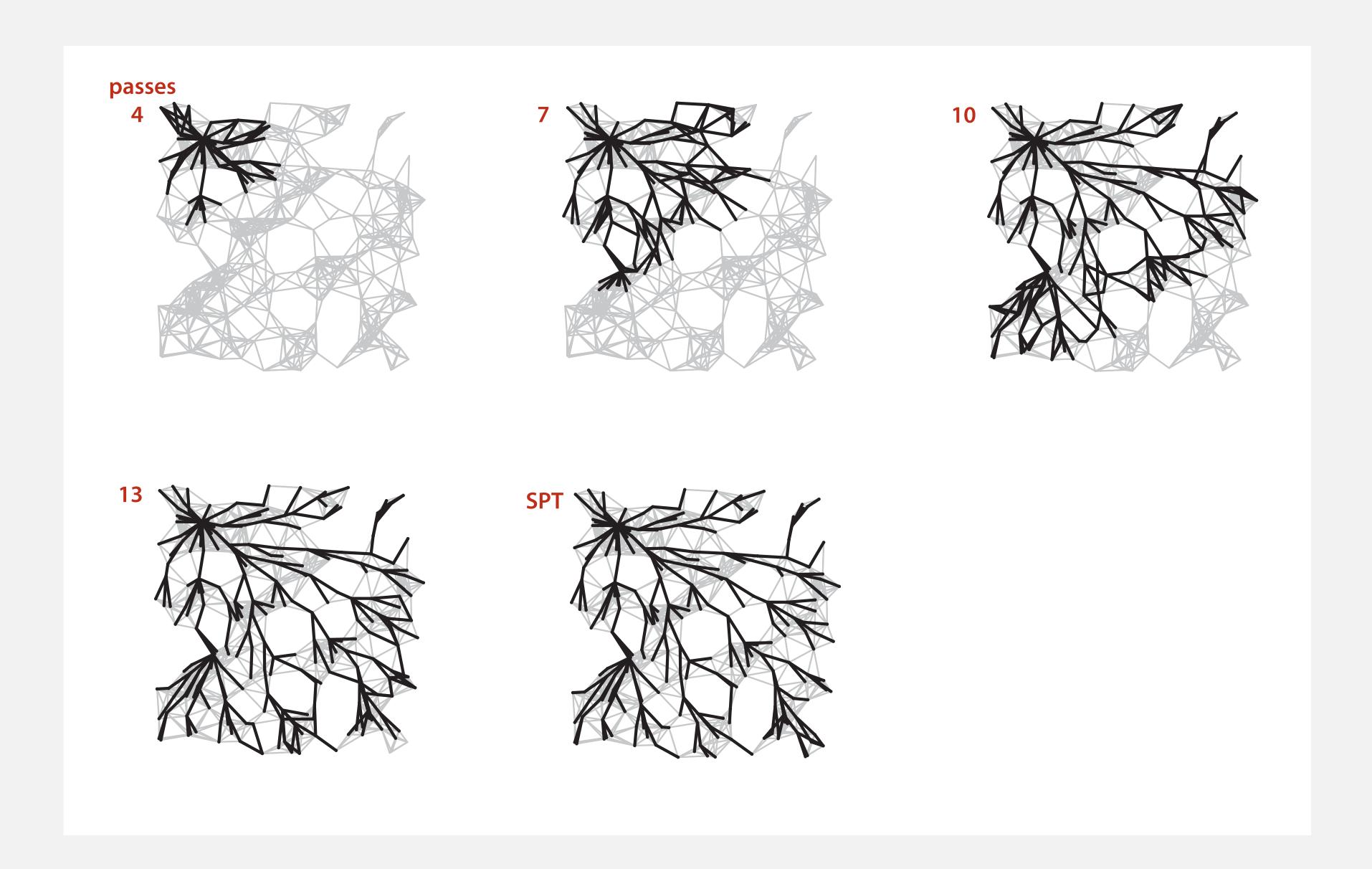
Repeat V-1 times: relax all E edges.



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

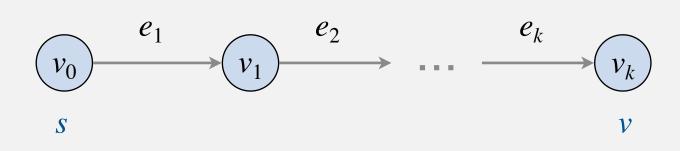
# Bellman-Ford algorithm: visualization



#### Bellman-Ford algorithm: correctness proof

Proposition. Let  $s = v_0 \rightarrow v_1 \rightarrow ... \rightarrow v_k = v$  be any path from s to v.

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



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

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

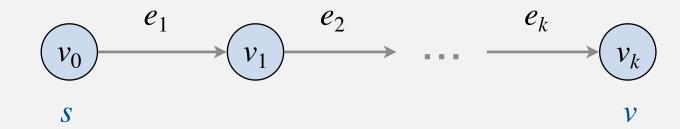
$$\mathsf{distTo}[v_{i+1}] \leq \mathsf{distTo}[v_i] + weight(e_{i+1}) \longleftarrow \mathsf{edge} \ \mathsf{relaxation}$$
 
$$\leq weight(e_1) + weight(e_2) + \dots + weight(e_i) + weight(e_{i+1}). \longleftarrow \mathsf{inductive} \ \mathsf{hypothesis}$$

• This inequality continues to hold because distTo[ $v_{i+1}$ ] does not increase.  $\blacksquare$ 

#### Bellman-Ford algorithm: correctness proof

Proposition. Let  $s = v_0 \rightarrow v_1 \rightarrow ... \rightarrow v_k = v$  be any path from s to v.

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



Corollary. Bellman-Ford computes shortest path distances.

Pf.

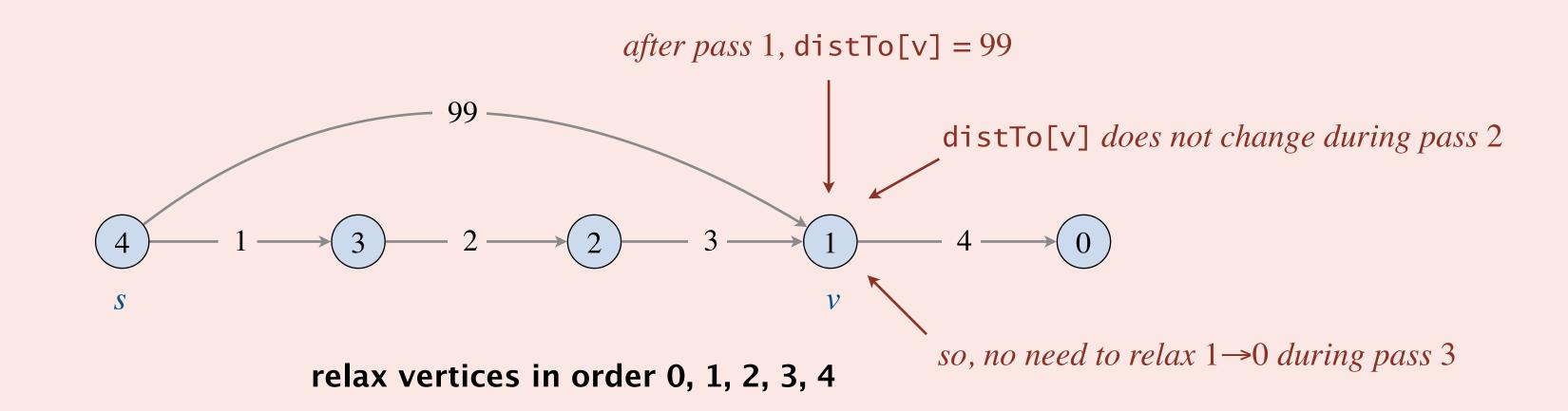
- There exists a shortest path  $P^*$  from s to v with at most V-1 edges.
- From Proposition, distTo[v]  $\leq length(P^*)$ .  $\leftarrow$  Bellman-Ford runs for V-1 passes
- Since distTo[v] is the length of some path from s to v, distTo[v] =  $length(P^*)$ .

## Shortest paths: quiz 4



# Suppose that distTo[v] does not change during pass i of Bellman-Ford. Which of the following are true?

- A. distTo[v] is the length of a shortest path from s to v.
- **B.** Not necessary to relax any edges incident to v in pass i + 1.
- C. Not necessary to relax any edges incident from v in pass i + 1.
- **D.** All of the above.

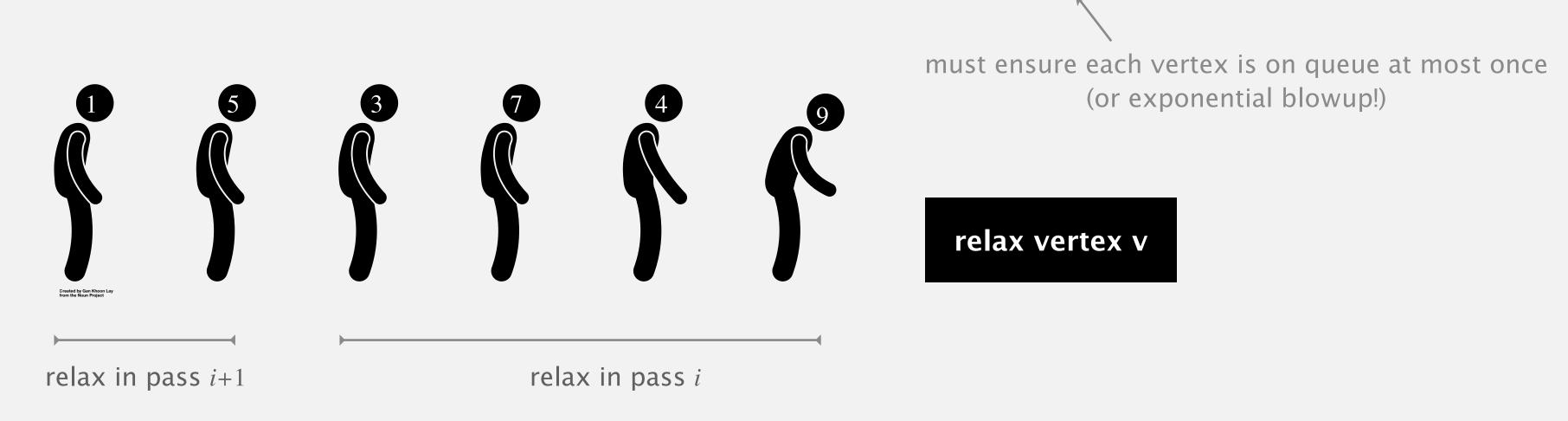


#### Bellman-Ford algorithm: practical improvement

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.  $\leftarrow$  relax all edges incident from v
- Maintain queue of vertices whose distTo[] values changed since it was last relaxed.



#### Impact.

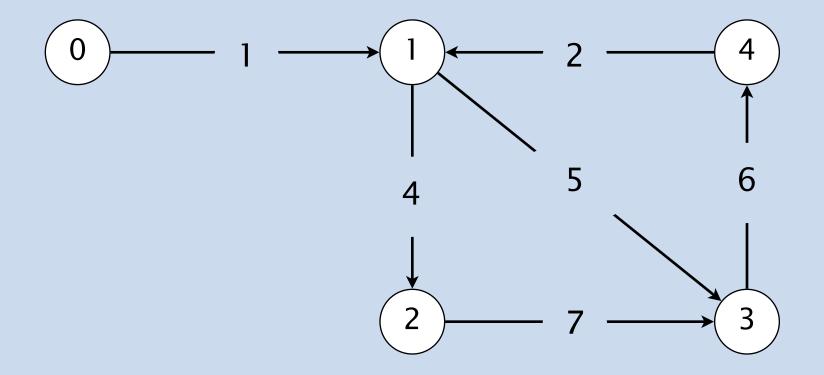
- In the worst case, the running time is still  $\Theta(E\ V)$ .
- But much faster in practice on typical inputs.

# 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.

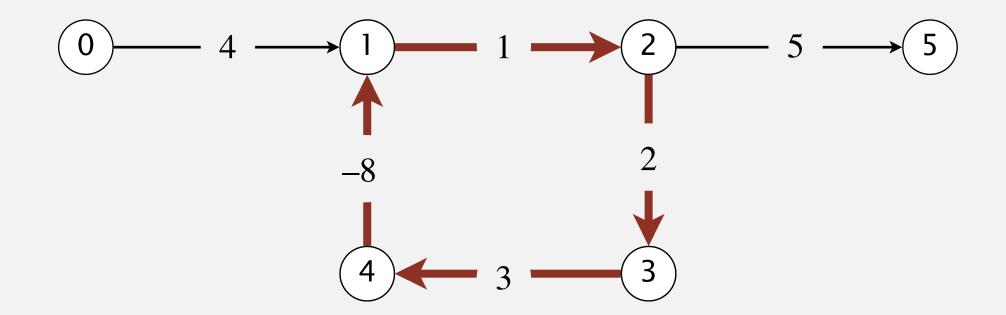


longest simple path from 0 to 4:  $0 \rightarrow 1 \rightarrow 2 \rightarrow 3 \rightarrow 4$ 

## 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.



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.

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



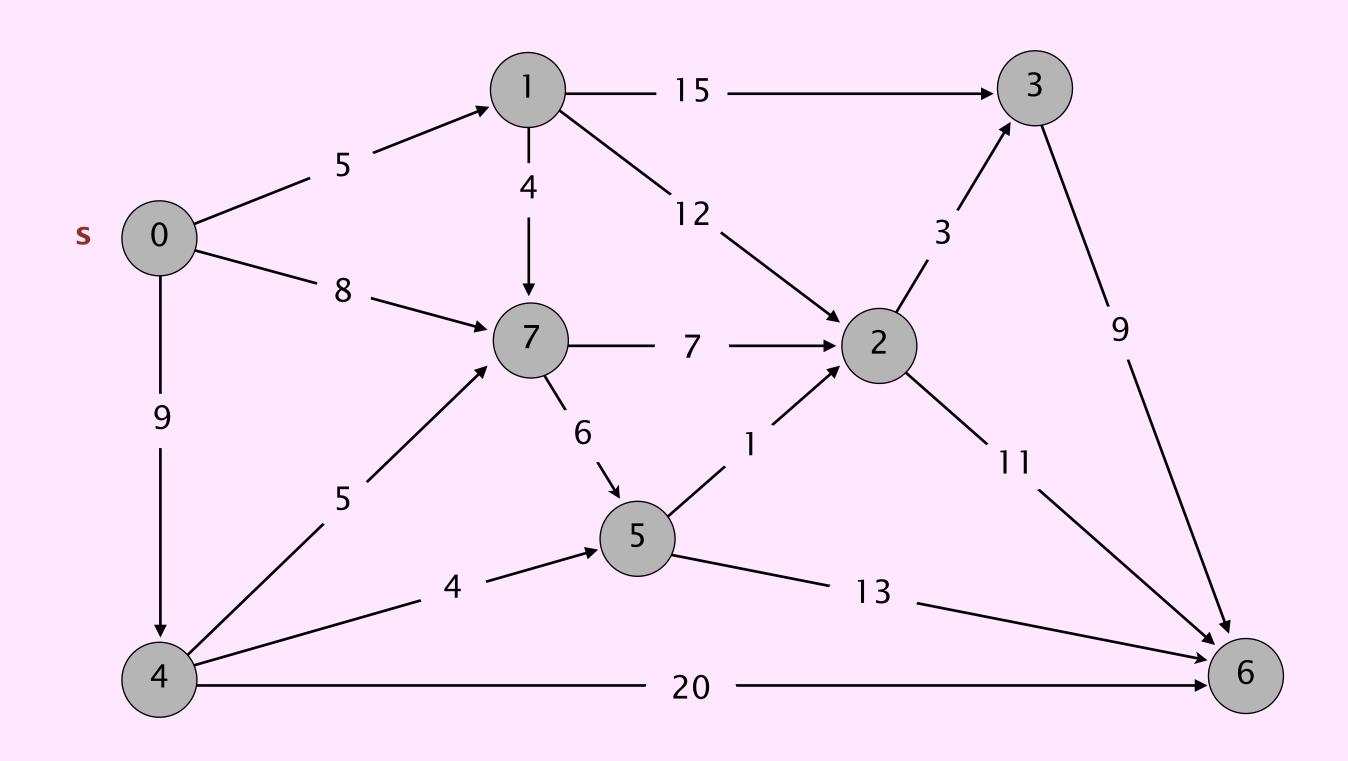
## Edsger W. Dijkstra: select quotes



## Dijkstra's algorithm demo



- Consider vertices in increasing order of distance from s (non-tree vertex with the lowest distTo[] value).
- · Add vertex to tree and relax all edges incident from that vertex.



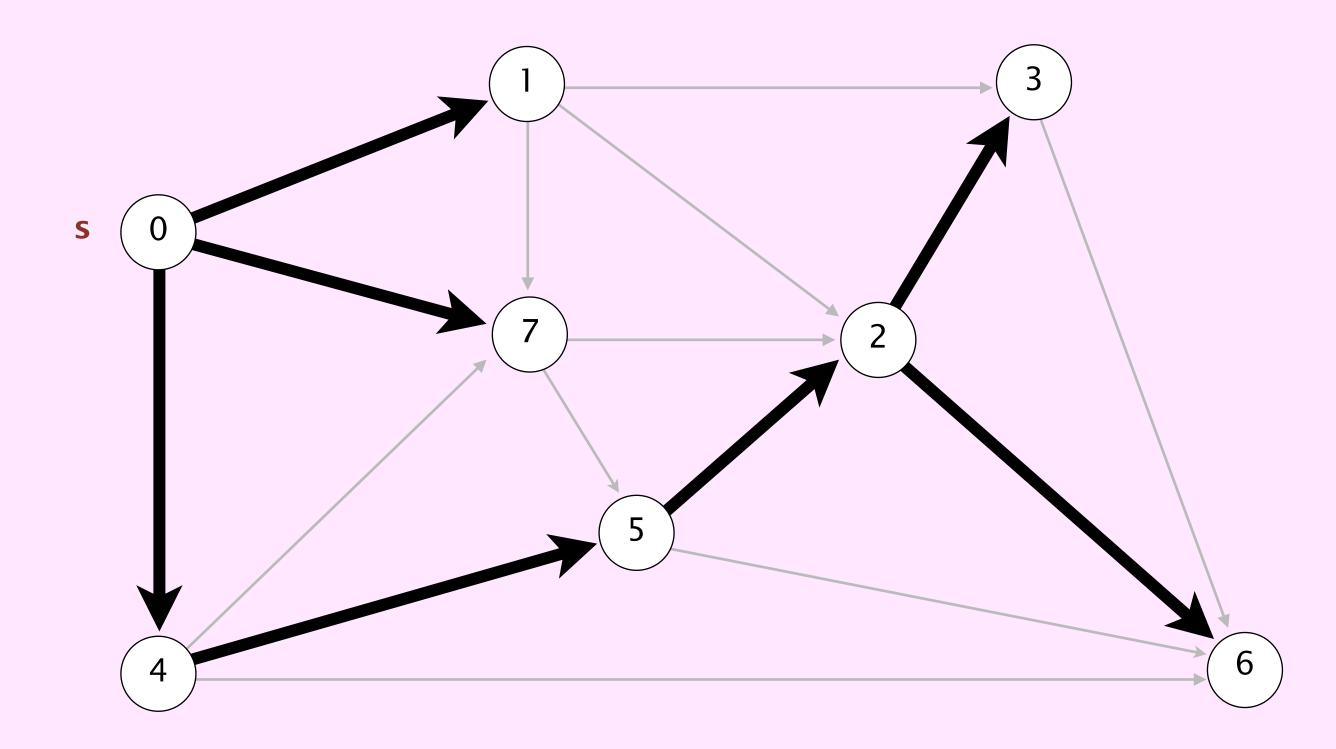
an edge-weighted digraph

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



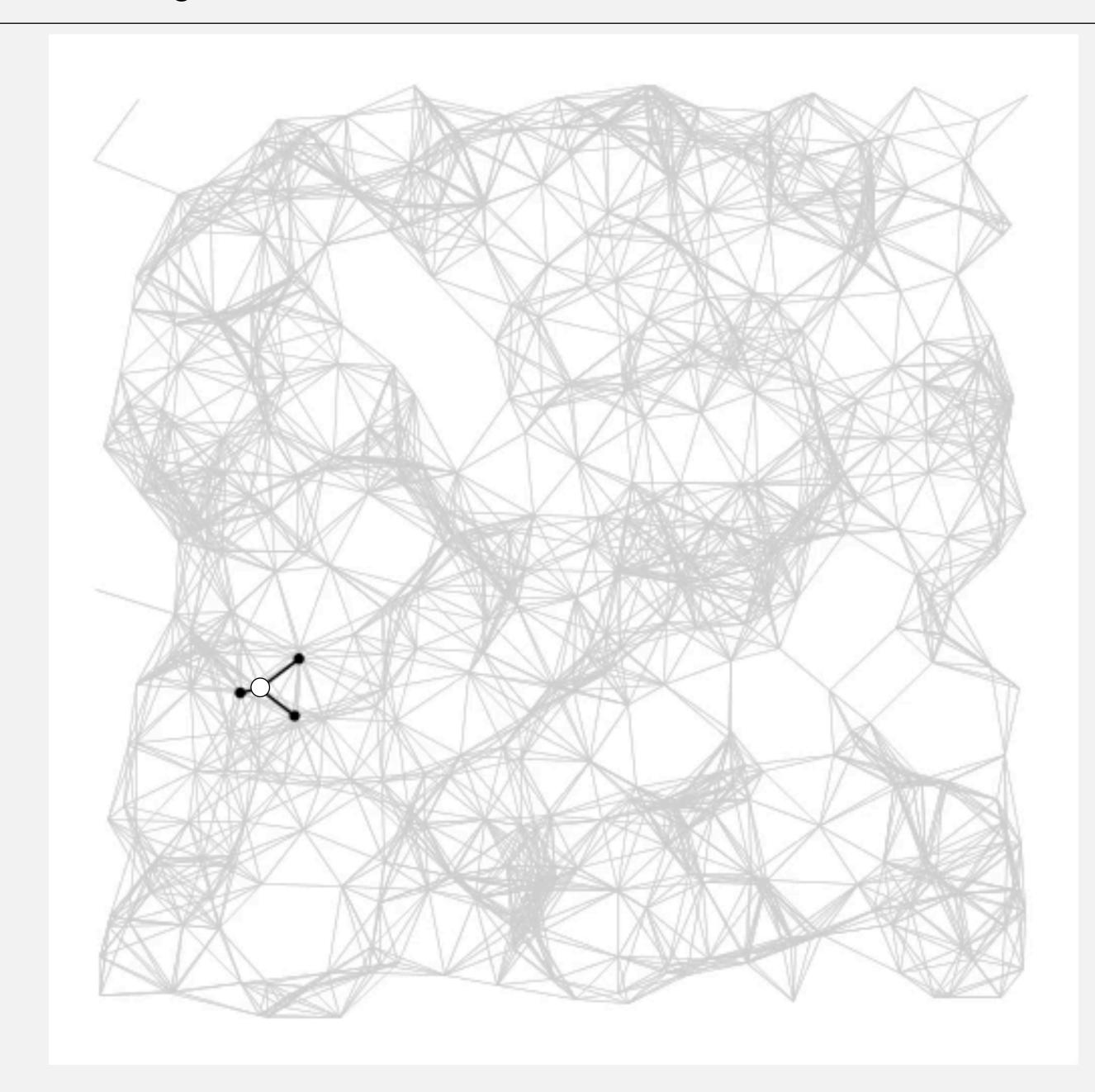
- Consider vertices in increasing order of distance from s (non-tree vertex with the lowest distTo[] value).
- · Add vertex to tree and relax all edges incident from that vertex.



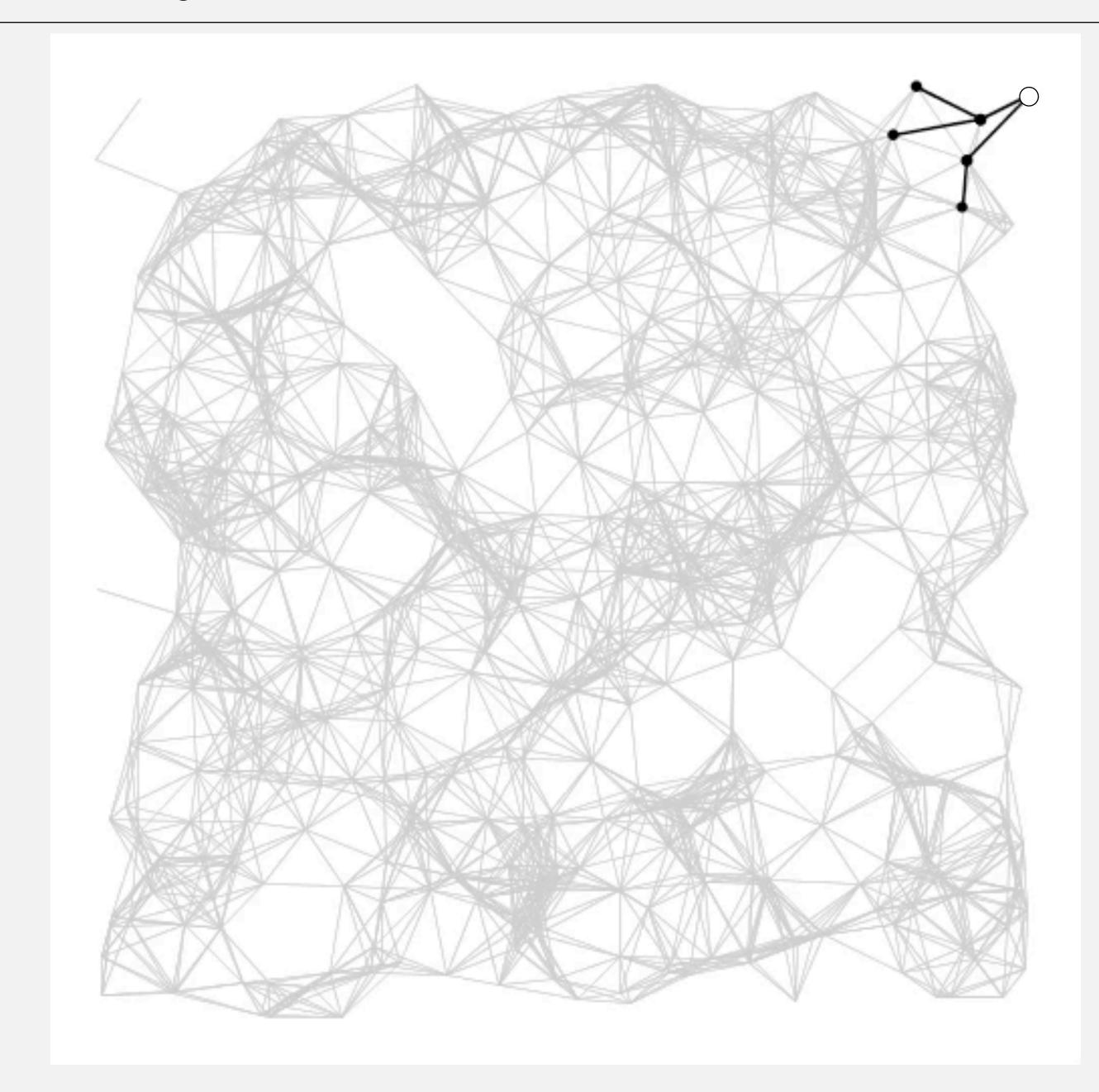
V	distTo[]	edgeTo[]
0	0.0	_
1	5.0	0→1
2	14.0	5→2
3	17.0	2→3
4	9.0	0→4
5	13.0	4→5
6	25.0	2→6
7	8.0	0→7

shortest-paths tree from vertex s

# Dijkstra's algorithm visualization



# Dijkstra's algorithm visualization



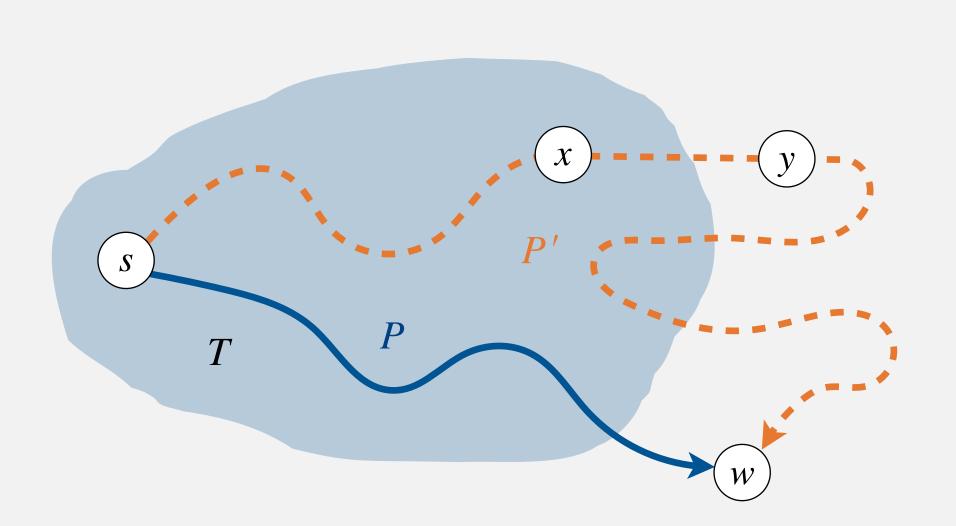
### Dijkstra's algorithm: correctness proof

Invariant. For each vertex v in T, distTo[v] =  $d^*(v)$ .

length of shortest path from s to v

### Pf. [by induction on |T|]

- Let w be next vertex added to T.
- Let P be the path from s to w of length distTo[w].
- Consider any other path P' from s to w.
- Let  $x \rightarrow y$  be first edge in P' that leaves T.
- *P'* is no shorter than *P*:



$$length(P) = distTo[w]$$

$$Dijkstra\ chose \\ w\ instead\ of\ y \longrightarrow \leq distTo[y]$$

$$relax\ vertex\ x \longrightarrow \leq distTo[x] + weight(x,y)$$

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

$$weights\ are \\ non-negative \longrightarrow \leq length(P') \quad \blacksquare$$

### Dijkstra's algorithm: correctness proof

Invariant. For each vertex v in T, distTo[v] =  $d^*(v)$ .

length of shortest path from s to v

Corollary. Dijkstra's algorithm computes shortest path distances.

**Pf.** Upon termination, *T* contains all vertices (reachable from *s*).

### Dijkstra's algorithm: Java implementation

```
public class DijkstraSP
   private DirectedEdge[] edgeTo;
   private double[] distTo;
                                                               PQ that supports
   private IndexMinPQ<Double> pq;
                                                              decreasing the key
                                                                 (stay tuned)
   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())
                                                               relax vertices in order
          int v = pq.delMin();
                                                                of distance from s
          for (DirectedEdge e : G.adj(v))
             relax(e);
```

### 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.

### 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,

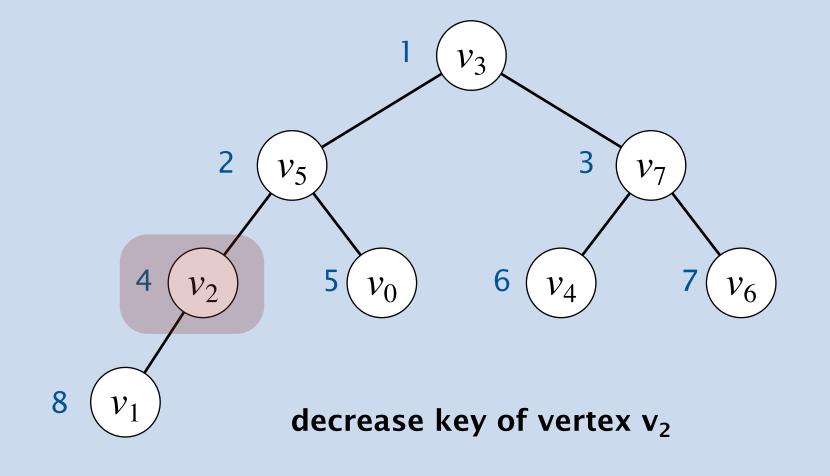
key = distance from s
```

```
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.



# DECREASE-KEY IN A BINARY HEAP

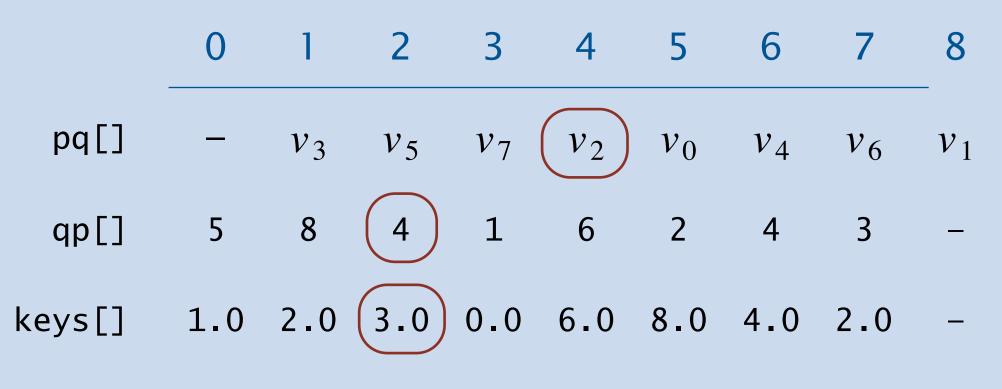


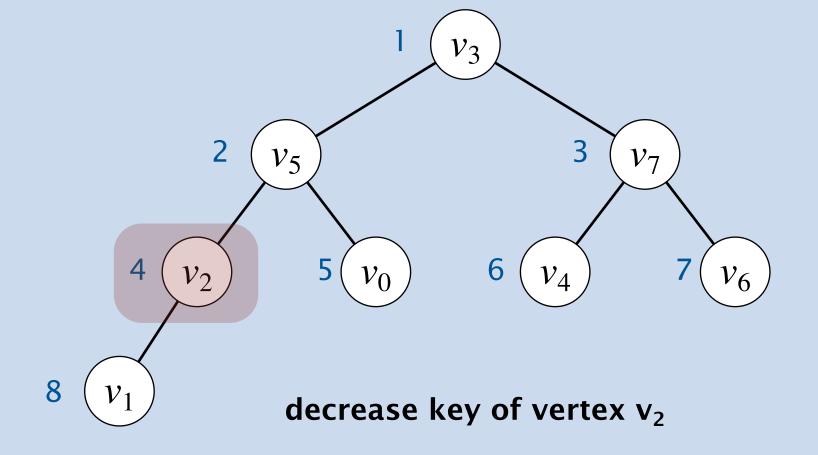
Goal. Implement Decrease-Key operation in a binary heap.

#### Solution.

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

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





### 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	$\log V^{\dagger}$	1	$E + V \log V$

† 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.

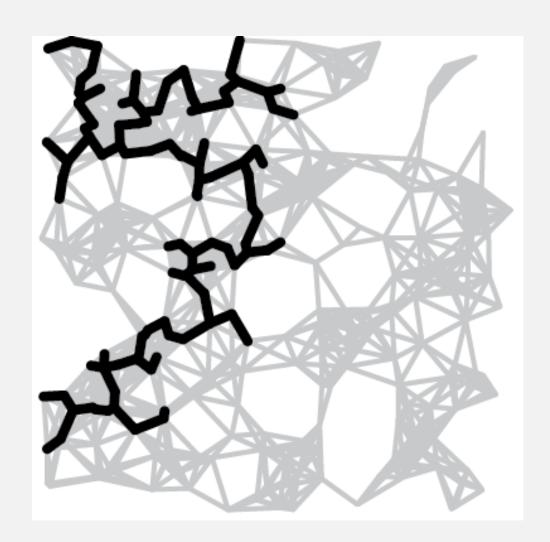
### Priority-first search

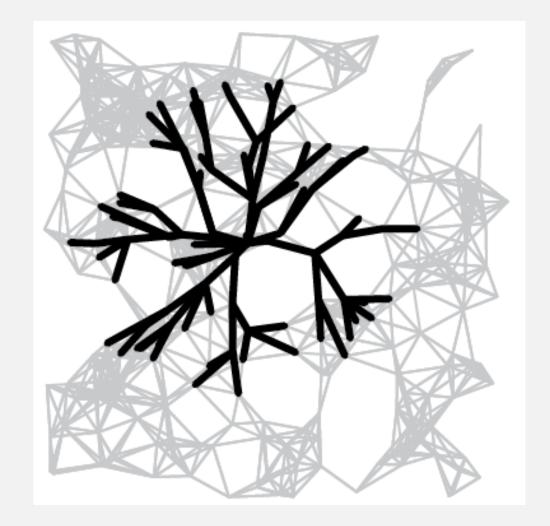
#### Dijkstra's algorithm seems familiar?

- Prim's algorithm is essentially the same algorithm.
- Both in same family of algorithms.

Main distinction: rule used to choose next vertex for the tree.

- Prim: Closest vertex to the tree (via an undirected edge).
- Dijkstra: Closest vertex to the source (via a directed path).





Note: DFS and BFS are also in same family.

## 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	
Bellman-Ford	E V			
Dijkstra	$E \log V$			
topological sort	$\boldsymbol{E}$			

### 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			

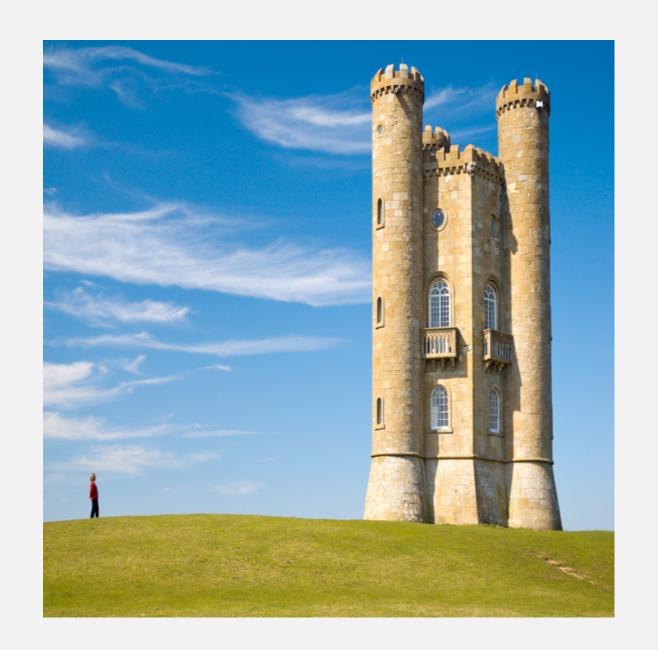


Seam carving. [Avidan-Shamir] Resize an image without distortion for display on cell phones and web browsers.



Seam carving. [Avidan-Shamir] Resize an image without distortion for display on cell phones and web browsers.







In the wild. Photoshop, Imagemagick, GIMP, ...

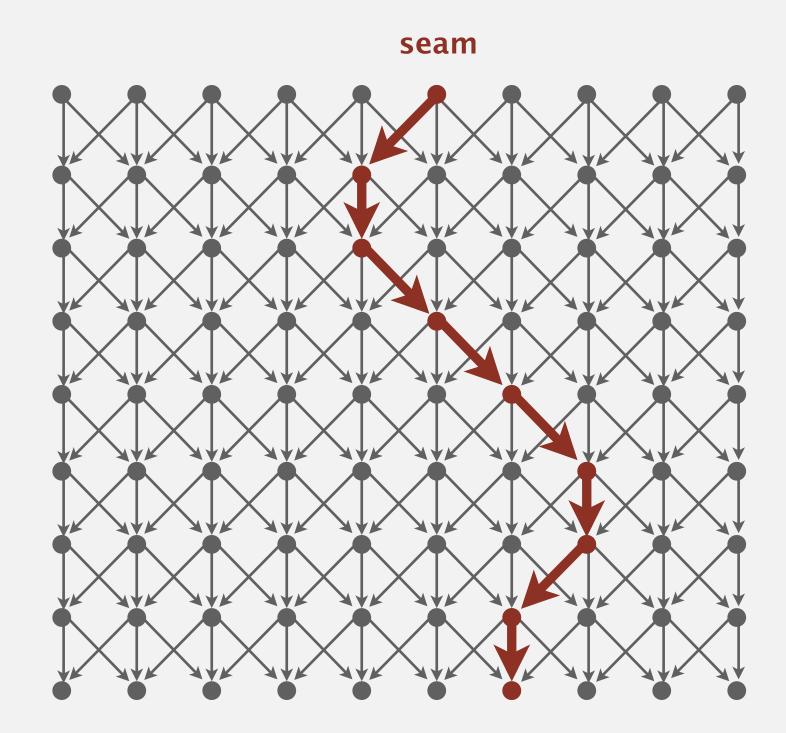
#### To find vertical seam:

- Grid graph: vertex = pixel; edge = from pixel to 3 downward neighbors.
- Weight of pixel = "energy function" of 8 neighboring pixels.
- Seam = shortest path (sum of vertex weights) from top to bottom.

•	•	•	•	•	•	•	•	•	•
•	•	•	•	•	•	•	•	•	•
•	•	•	•	•	•	•	•	•	
•	•	•	•	•	•	•	•	•	•
•	•	•	•	•	•	•	•	•	
•	•	•	•	•	•	•	•	•	•
•	•	•	•	•	•	•	•	•	•
•	•	•	•	•	•	•	•	•	
•	•	•	•	•	•	•	•	•	•

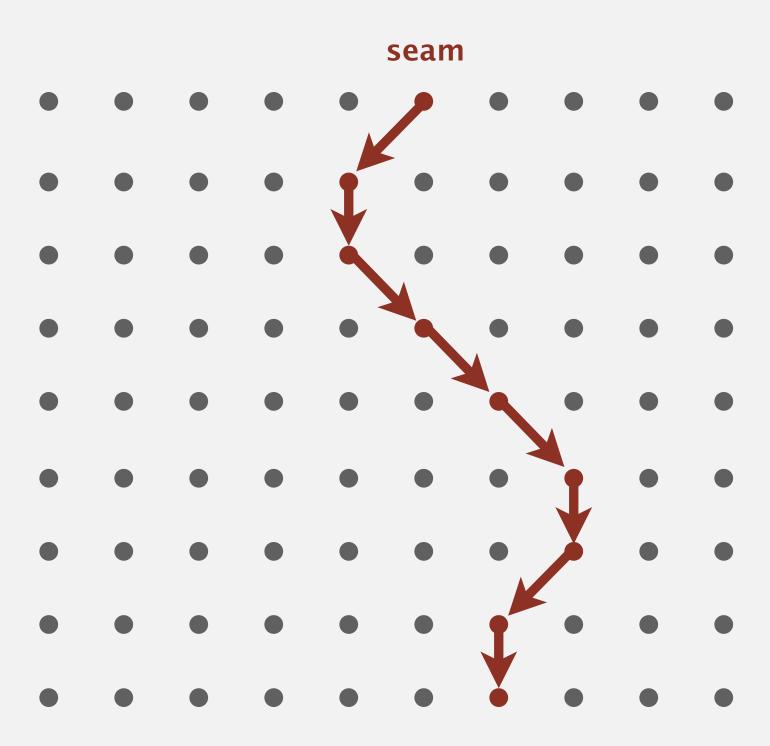
#### To find vertical seam:

- Grid graph: vertex = pixel; edge = from pixel to 3 downward neighbors.
- Weight of pixel = "energy function" of 8 neighboring pixels.
- Seam = shortest path (sum of vertex weights) from top to bottom.



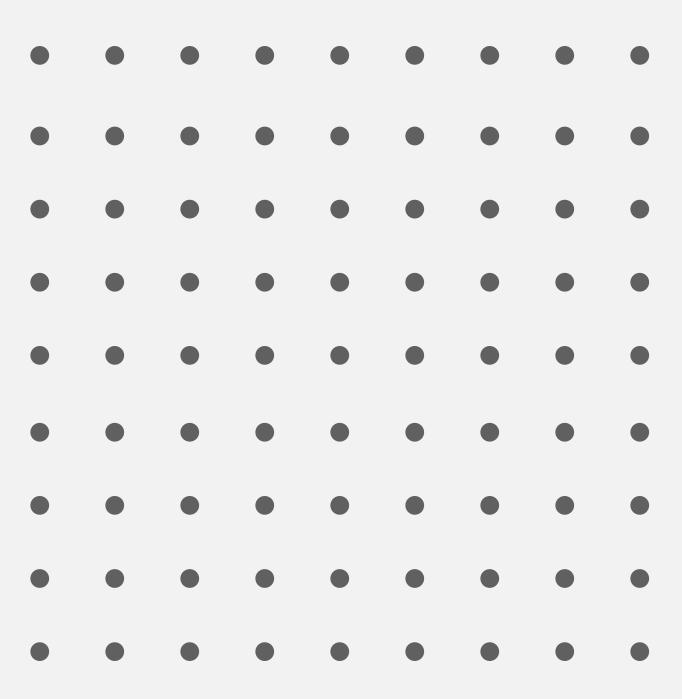
#### To remove vertical seam:

• Delete pixels on seam (one in each row).



#### To remove vertical seam:

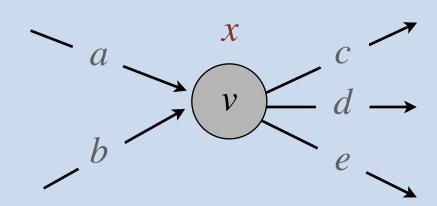
• Delete pixels on seam (one in each row).



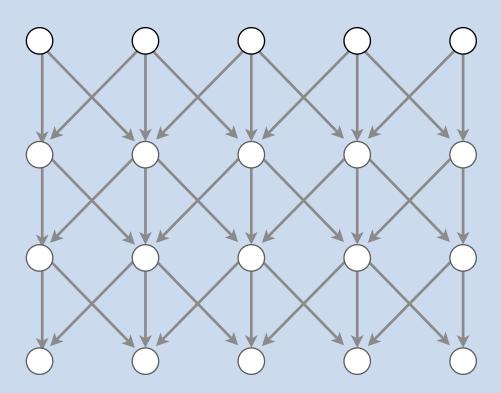
# SHORTEST PATH VARIANTS IN A DIGRAPH



Q1. How to model vertex weights (along with edge weights)?



Q2. How to model multiple sources and sinks?



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