

# CSCI 470: Single-Source Shortest Paths, Dijkstra's algorithm, Bellman-Ford algorithm

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1. Shortest-paths problems
2. Dijkstra's algorithm
3. Bellman-Ford algorithm

# Shortest-paths problems

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# Shortest-paths problems

- In a *shortest-paths problem*, we are given a weighted, directed graph  $G = (V, E)$ , with weight function  $w : E \rightarrow \mathbb{R}$  mapping edges to real-valued weights.
- The *weight*  $w(p)$  of path  $p = \langle v_0, v_1, \dots, v_k \rangle$  is the sum of the weights of its constituent edges:

$$w(p) = \sum_{i=1}^k w(v_{i-1}, v_i).$$

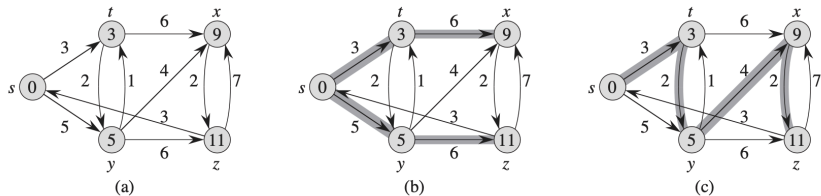
# Shortest-paths problems

- We define the *shortest-path weights*  $\delta(u, v)$  from  $u$  to  $v$  by

$$\delta(u, v) = \begin{cases} \min\{w(p) : u \rightsquigarrow v\} & \exists \text{ a path } u \text{ to } v, \\ \infty & \text{otherwise.} \end{cases}$$

- A *shortest path* from vertex  $u$  to vertex  $v$  is then defined as any path  $p$  with weight  $w(p) = \delta(u, v)$ .
- Edge weights can represent metrics other than distances, such as time, cost, penalties, loss, or any other quantity that accumulates linearly along a path and that we would want to minimize.

# Shortest Paths: Figure

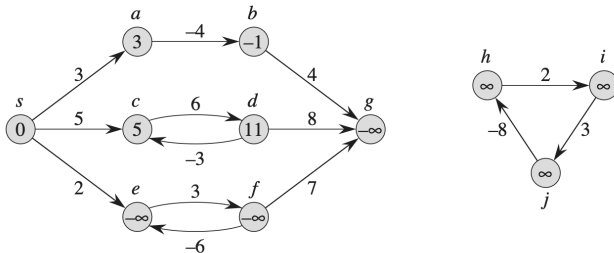


**Figure 24.2** (a) A weighted, directed graph with shortest-path weights from source  $s$ . (b) The shaded edges form a shortest-paths tree rooted at the source  $s$ . (c) Another shortest-paths tree with the same root.

# Negative-weight edges

- If the graph  $G = (V, E)$  contains no negative-weight cycles reachable from the source  $s$ , then for all  $v \in V$ , the shortest-path weight  $\delta(s, v)$  remains well-defined, even if it has a negative value.
- If the graph contains a negative-weight cycle reachable from  $s$ , however, shortest-path weights are not well defined.
- No path from  $s$  to a vertex on the cycle can be a shortest path - we can always find a path with lower weight by following the proposed “shortest” path and then traversing the negative-weight cycle.
- If there is a negative-weight cycle on some path from  $s$  to  $v$ , we define  $\delta(s, v) = -\infty$ .

## Figure: negative-weight edges



**Figure 24.1** Negative edge weights in a directed graph. The shortest-path weight from source  $s$  appears within each vertex. Because vertices  $e$  and  $f$  form a negative-weight cycle reachable from  $s$ , they have shortest-path weights of  $-\infty$ . Because vertex  $g$  is reachable from a vertex whose shortest-path weight is  $-\infty$ , it, too, has a shortest-path weight of  $-\infty$ . Vertices such as  $h, i$ , and  $j$  are not reachable from  $s$ , and so their shortest-path weights are  $\infty$ , even though they lie on a negative-weight cycle.



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- It cannot contain a positive-weight cycle, since removing the cycle from the path produces a path with the same source and destination vertices and a lower path weight.
- How about 0-weight cycles?
- We can always remove a 0-weight cycle to produce an alternative path whose path weight will still be the same.

- Therefore, without loss of generality we can assume that when we are finding shortest paths, they have no cycles, i.e., they are simple paths.
- Since any acyclic path in a graph  $G = (V, E)$  contains at most  $|V|$  distinct vertices, it also contains at most  $|V| - 1$  edges. Thus, we can restrict our attention to shortest paths of at most  $|V| - 1$  edges.

INITIALIZE-SINGLE-SOURCE( $G, s$ )

```
1  for each vertex  $v \in G.V$ 
2       $v.d = \infty$ 
3       $v.\pi = \text{NIL}$ 
4   $s.d = 0$ 
```

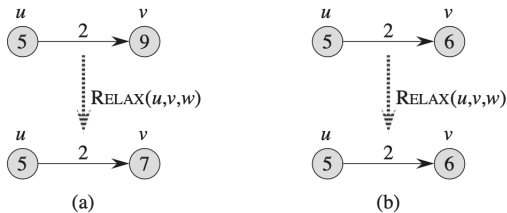
- After initialization, we have  $v.\pi = \text{NIL}$  for all  $v \in V$ ,  $s.d = 0$ , and  $v.d = \infty$  for  $v \in V - \{s\}$ .

RELAX( $u, v, w$ )

- 1    **if**  $v.d > u.d + w(u, v)$
- 2         $v.d = u.d + w(u, v)$
- 3         $v.\pi = u$



# Relaxation: Figure



**Figure 24.3** Relaxing an edge  $(u, v)$  with weight  $w(u, v) = 2$ . The shortest-path estimate of each vertex appears within the vertex. (a) Because  $v.d > u.d + w(u, v)$  prior to relaxation, the value of  $v.d$  decreases. (b) Here,  $v.d \leq u.d + w(u, v)$  before relaxing the edge, and so the relaxation step leaves  $v.d$  unchanged.

## Dijkstra's algorithm

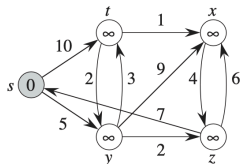
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# Dijkstra's algorithm

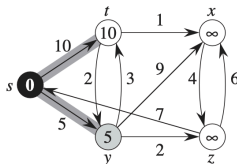
DIJKSTRA( $G, w, s$ )

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
2   $S = \emptyset$ 
3   $Q = G.V$ 
4  while  $Q \neq \emptyset$ 
5       $u = \text{EXTRACT-MIN}(Q)$ 
6       $S = S \cup \{u\}$ 
7      for each vertex  $v \in G.Adj[u]$ 
8          RELAX( $u, v, w$ )
```

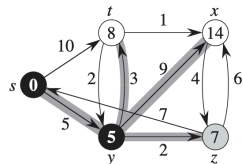
# Dijkstra: Figure



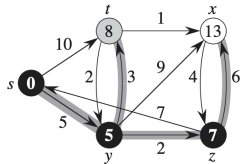
(a)



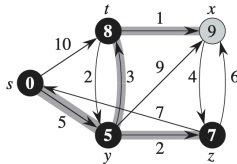
(b)



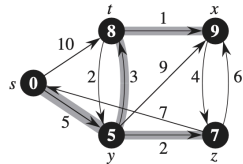
(c)



(d)



(e)



(f)

- There are three priority-queue operations involved here: INSERT (line 3), EXTRACT-MIN (line 5), DECREASE-KEY (implicit in RELAX, which is called in line 8).
- The algorithm calls both INSERT and EXTRACT-MIN once per vertex.
- Because each vertex  $u \in V$  is added to set  $S$  exactly once, each edge in the adjacency list  $Adj[u]$  is examined in the **for** loop of lines 7-8 exactly once during the course of algorithm.
- Since the total number of edges in all the adjacency lists is  $|E|$ , this **for** loop iterates a total of  $|E|$  times, and thus DECREASE-KEY gets called at most  $|E|$  times overall.

If we implement priority-queue as an array, without a min-heap counterpart.

- Consider a case in which we maintain the min-priority queue by taking advantage of the vertices being numbered 1 to  $|V|$ .
- We simply store  $v.d$  in the  $v$ th entry of an array.
- Each INSERT and DECREASE-KEY operation takes  $O(1)$  time, and each EXTRACT-MIN takes  $O(V)$  time.
- This amounts to  $O(V^2 + E) = O(V^2)$ .

If we implement priority-queue with a binary min-heap.

- The time to build the binary min-heap is  $O(V)$ .
- Each EXTRACT-MIN takes  $O(\lg(V))$  time, and there are  $|V|$  such operations.
- Each DECREASE-KEY takes  $O(\lg(V))$  time, and there are  $|E|$  such operations.
- The total running time is therefore  $O(V \lg V + E \lg V) = O((V + E) \lg V)$ .

# Bellman-Ford algorithm

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# Bellman-Ford algorithm

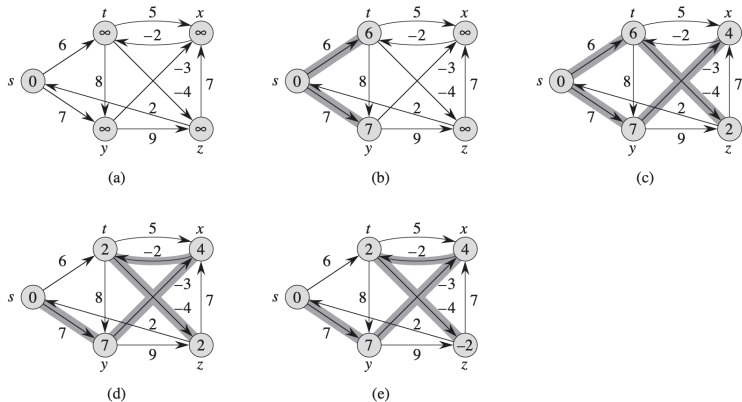
- The ***Bellman-Ford algorithm*** solves the single-source shortest-paths problem in the general case in which edge weights may be negative.
- Given a weighted, directed graph  $G = (V, E)$  with source  $s$  and weight function  $w : E \rightarrow \mathbb{R}$ , the Bellman-Ford algorithm returns a boolean value indicating whether or not there is a negative-weight cycle that is reachable from the source.
- If there is such a cycle, the algorithm indicates that no solution exists. If there is no such cycle, the algorithm produces the shortest paths and their weights.
- The algorithm relaxes edges, progressively decreasing an estimate  $v.d$  on the weight of a shortest path from the source  $s$  to each vertex  $v \in V$  until it achieves the actual shortest-path weight  $\delta(s, v)$ .

# Bellman-Ford algorithm: pseudocode

BELLMAN-FORD( $G, w, s$ )

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
2  for  $i = 1$  to  $|G.V| - 1$ 
3      for each edge  $(u, v) \in G.E$ 
4          RELAX( $u, v, w$ )
5  for each edge  $(u, v) \in G.E$ 
6      if  $v.d > u.d + w(u, v)$ 
7          return FALSE
8  return TRUE
```

# Bellman-Ford: illustration



**Figure 24.4** The execution of the Bellman-Ford algorithm. The source is vertex  $s$ . The  $d$  values appear within the vertices, and shaded edges indicate predecessor values: if edge  $(u, v)$  is shaded, then  $v.\pi = u$ . In this particular example, each pass relaxes the edges in the order  $(t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)$ . (a) The situation just before the first pass over the edges. (b)–(e) The situation after each successive pass over the edges. The  $d$  and  $\pi$  values in part (e) are the final values. The Bellman-Ford algorithm returns TRUE in this example.

# Bellman-Ford runtime

- The Bellman-Ford algorithm runs in time  $O(VE)$ , since the initialization in line 1 takes  $\Theta(V)$  time, each of the  $|V| - 1$  passes over the edges in lines 2-4 takes  $\Theta(E)$  time, and the **for** loop of lines 5-7 takes  $O(E)$  time.