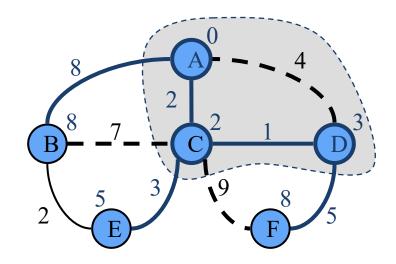
Shortest Paths

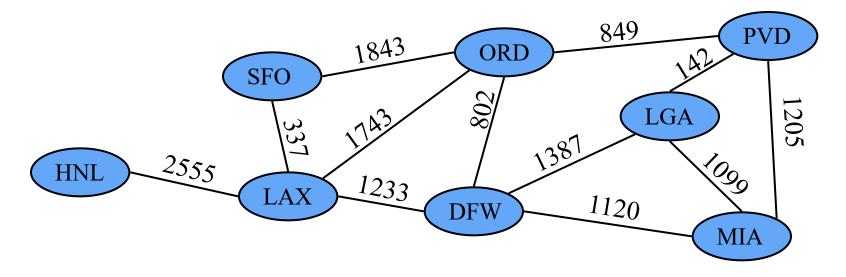


Outline and Reading

- Weighted graphs (7.1)
 - Shortest path problem
 - Shortest path properties
- Dijkstra's algorithm (7.1.1)
 - Algorithm
 - Edge relaxation
- The Bellman-Ford algorithm (7.1.2)
- Shortest paths in DAGs (7.1.3)
- All-pairs shortest paths (7.2.1)

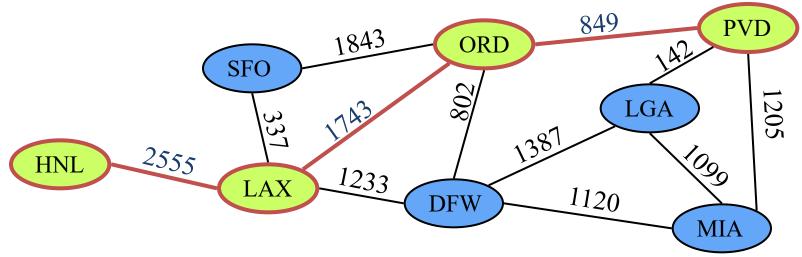
Weighted Graphs

- In a weighted graph, each edge has an associated numerical value, called the weight of the edge
- Edge weights may represent, distances, costs, etc.
- Example:
 - In a flight route graph, the weight of an edge represents the distance in miles between the endpoint airports



Shortest Path Problem

- Given a weighted graph and two vertices *u* and *v*, we want to find a path of minimum total weight between *u* and *v*.
 - Length of a path is the sum of the weights of its edges
- Example: shortest path between Providence and Honolulu
- Applications
 - Internet packet routing
 - Flight reservations
 - Driving directions

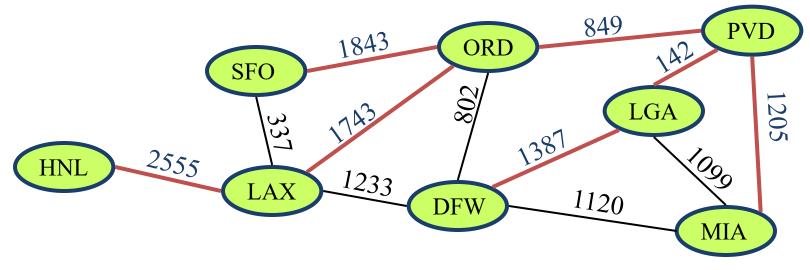


Shortest Path Problem

Property 1. A subpath of a shortest path is itself a shortest path.

Property 2. There is a tree of shortest paths from a start vertex to all other vertices.

• Example: tree of shortest paths from Providence



Dijkstra's Algorithm

The distance of vertex *v* from *s* is the length of a shortest path between *s* and *v*.

Dijkstra's algorithm computes the distances of all the vertices from a given start vertex *s*.

- Assumptions:
 - the graph is connected
 - the edges are undirected
 - the edge weights are nonnegative

Idea:

- Grow a "cloud" of vertices, beginning with *s* and eventually covering all vertices
- Store with each vertex v a label d(v) representing the distance of v from s in the subgraph consisting of the cloud and its adjacent vertices
- At each step
 - Add to the cloud the vertex u outside the cloud with the smallest distance label, d(u)
 - Update the labels of the vertices adjacent to *u*

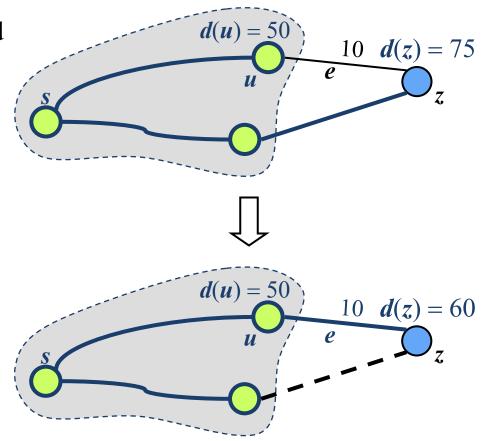
Edge Relaxation

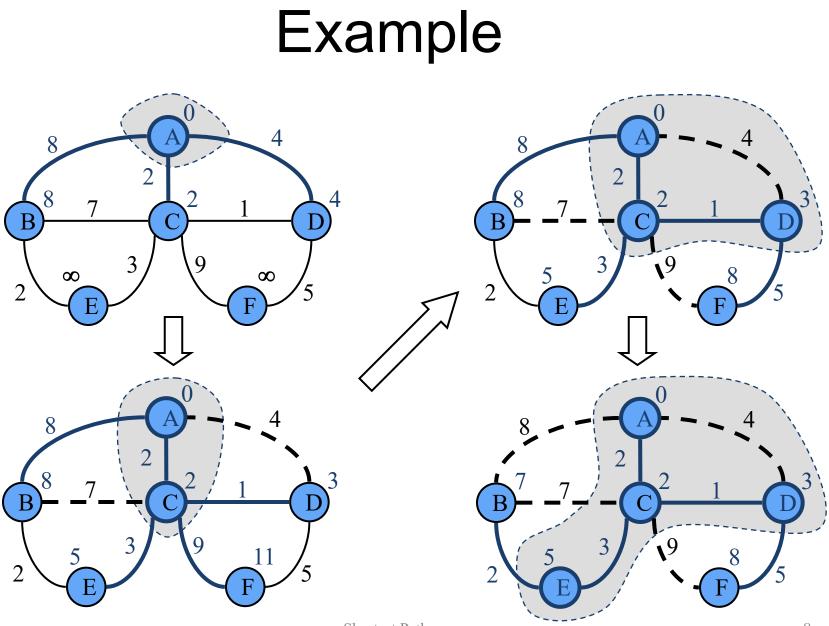
Consider an edge e = (u,z) such that

- *u* is the vertex most recently added to the cloud
- *z* is not in the cloud

The relaxation of edge e updates distance d(z) as follows:

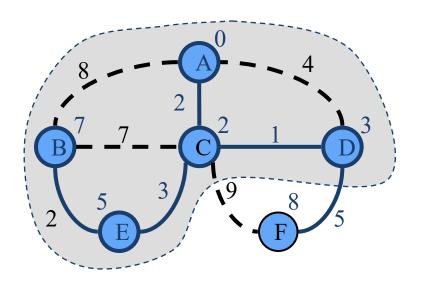
 $d(z) \leftarrow \min\{d(z), d(u) + weight(e)\}$





Shortest Paths

Example (cont.)



Dijkstra's Algorithm

A priority queue stores the vertices outside the cloud

- Key: distance
- Element: vertex

Locator-based methods

- *insert(k,e)* returns a locator
- *replaceKey(l,k)* changes the key of an item

We store two labels with each vertex:

- Distance (d(v) label)
- locator in priority queue

Algorithm *DijkstraDistances*(*G*, *s*) $Q \leftarrow$ new heap-based priority queue for all $v \in G.vertices()$ 2 3 if v = ssetDistance(v, 0) 4 5 else *setDistance*(*v*, ∞) 6 7 $l \leftarrow Q.insert(getDistance(v), v)$ 8 setLocator(v,l) 9 while $\neg Q.isEmpty()$ $u \leftarrow Q.removeMin()$ 10 for all $e \in G.incidentEdges(u)$ 11 { relax edge *e* } 12 $z \leftarrow G.opposite(u,e)$ 13 $r \leftarrow getDistance(u) + weight(e)$ 14 if *r* < *getDistance*(*z*) 15 16 setDistance(z,r) *Q.replaceKey(getLocator(z),r)* 17

Analysis

- Graph operations
 - Method incidentEdges is called once for each vertex
- Label operations
 - We set/get the distance and locator labels of vertex $z O(\deg(z))$ times
 - Setting/getting a label takes O(1) time
- Priority queue operations
 - Each vertex is inserted once into and removed once from the priority queue, where each insertion or removal takes $O(\log n)$ time
 - The key of a vertex in the priority queue is modified at most deg(w) times, where each key change takes O(log n) time
- Dijkstra's algorithm runs in $O((n + m) \log n)$ time provided the graph is represented by the adjacency list structure
 - Recall that $\Sigma_v \deg(v) = 2m$
- The running time can also be expressed as $O(m \log n)$ since the graph is connected.

Extension

Using the template method pattern, we can extend Dijkstra's algorithm to return a tree of shortest paths from the start vertex to all other vertices

- Store with each vertex a third label:
 - parent edge in the shortest path tree
- In the edge relaxation step, update the parent label

```
Algorithm DijkstraShortestPathsTree(G, s)
      . . .
  for all v \in G.vertices()
     setParent(v, Ø)
     . . .
     for all e \in G.incidentEdges(u)
        { relax edge e }
        z \leftarrow G.opposite(u,e)
        r \leftarrow getDistance(u) + weight(e)
        if r < getDistance(z)
           setDistance(z,r)
```

Q.replaceKey(getLocator(z),r)

setParent(z,e)

Why Dijkstra's Algorithm Works

Dijkstra's algorithm is based on the greedy method. It adds vertices by increasing distance.

<u>Claim</u>: Whenever a vertex *u* is pulled into the cloud, D[u] = d(v, u).

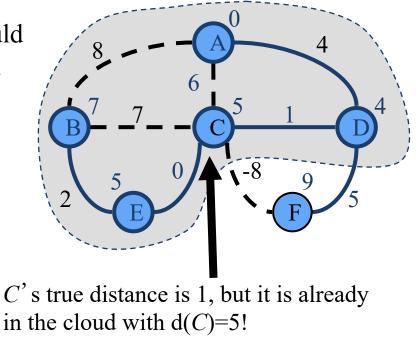
Outline of Proof (by contradiction):

- Suppose *u* is the first vertex such that D[u] > d(v, u).
- Let *z* be the first vertex on the shortest *v*-*u* path *P* which hasn't been pulled into the cloud yet, and let *y* be the vertex before *z* on *P*.
 - Then, D[z] = d(v, z).
 - Since z is on shortest v-u path, d(v, z) + d(z, u) = d(v, u).
 - Since *u* is processed before *z*, $D[u] \le D[z]$.
- $D[u] \le D[z] = d(v, z) \le d(v, z) + d(z, u) = d(v, u)$, a contradiction.

Why It Doesn't Work for Negative-Weight Edges

Dijkstra's algorithm is based on the greedy method. It adds vertices by increasing distance.

- If a node with a negative incident edge were to be added late to the cloud, it could mess up distances for vertices already in the cloud.
- This violates the greedy property.



Bellman-Ford Algorithm

- Works even with negative-weight edges
- Must assume directed edges (otherwise we would have negative-weight cycles)
- Iteration *i* finds all shortest paths that use *i* edges beginning at *s*
- Running time: *O*(*nm*).
- Can be extended to detect a negative-weight cycle if it exists

- How?

```
Algorithm BellmanFord(G, s)

for all v \in G.vertices()

if v = s

setDistance(v, 0)

else

setDistance(v, \infty)

for i \leftarrow 1 to n-1 do

for each (directed) edge e=(u,z) \in G.edges()

{ relax edge e }

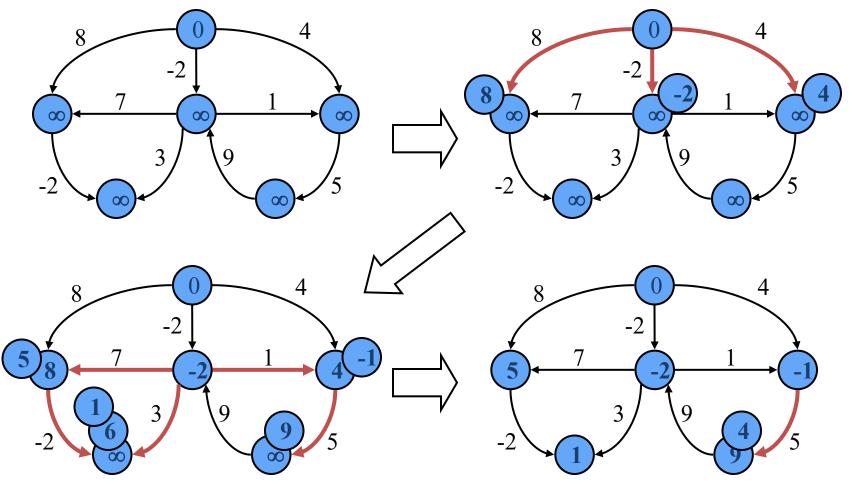
r \leftarrow getDistance(u) + weight(e)

if r < getDistance(z)

setDistance(z,r)
```

Bellman-Ford Example

Nodes are labeled with their d(v) values



DAG-based Algorithm

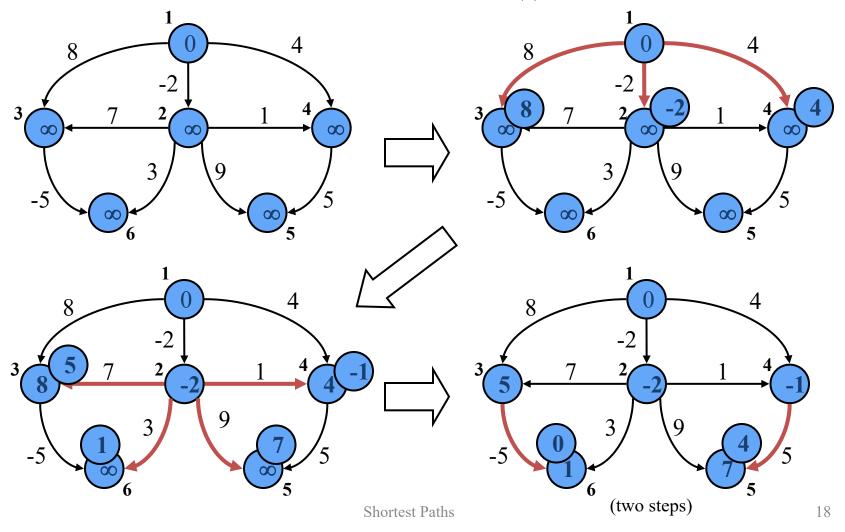
- Assumes *G* is a DAG
- Works even with negative-weight edges
- Uses topological order
- Much faster than Dijkstra's algorithm

• Running time: O(n+m).

Algorithm *DagDistances*(*G*, *s*) for all $v \in G.vertices()$ if v = ssetDistance(v, 0) else *setDistance*(*v*, ∞) **Perform a topological sort of the vertices** for $u \leftarrow 1$ to *n* do {in topological order} for each edge $e=(u,z) \in G.edges()$ { relax edge *e* } $r \leftarrow getDistance(u) + weight(e)$ if *r* < *getDistance*(*z*) setDistance(z,r)

DAG Example

Nodes are labeled with their d(v) values



All-Pairs Shortest Paths

Find the distance between every pair of vertices in a weighted directed graph G.

- We can make *n* calls to Dijkstra's algorithm (if no negative edges), which takes O(*nm*log *n*) time.
- Likewise, *n* calls to Bellman-Ford would take $O(n^2m)$ time.

We can achieve $O(n^3)$ time using the Floyd-Warshall dynamic programming algorithm. Algorithm *AllPair*(G) {assumes vertices 1,...,n} for all *vertex pairs (i,j)* if i = j $D_{\theta}[i,i] \leftarrow \theta$ else if (*i*,*j*) is an edge in G $D_{\theta}[i,j] \leftarrow weight of edge(i,j)$ else $D_{\theta}[i,j] \leftarrow +\infty$ for $k \leftarrow 1$ to *n* do for $i \leftarrow 1$ to *n* do for $i \leftarrow 1$ to *n* do $D_k[i,j] \leftarrow \min\{D_{k-1}[i,j], D_{k-1}[i,k] + D_{k-1}[k,j]\}$ return **D**_n

Uses only vertices numbered 1,...,k Uses only vertices numbered 1,...,k-1 Uses only vertices numbered 1,...,k-1

Shortest Paths