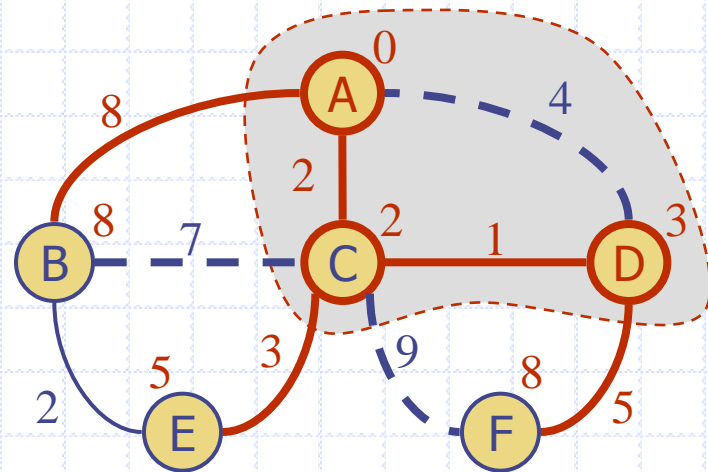
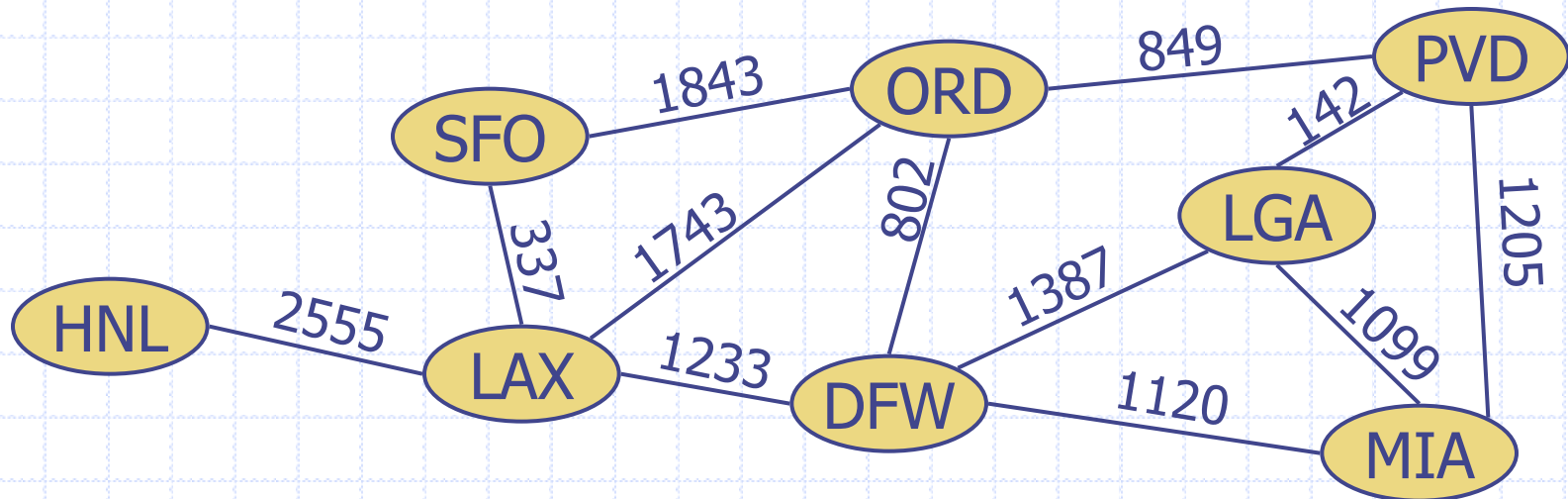


# Shortest Paths



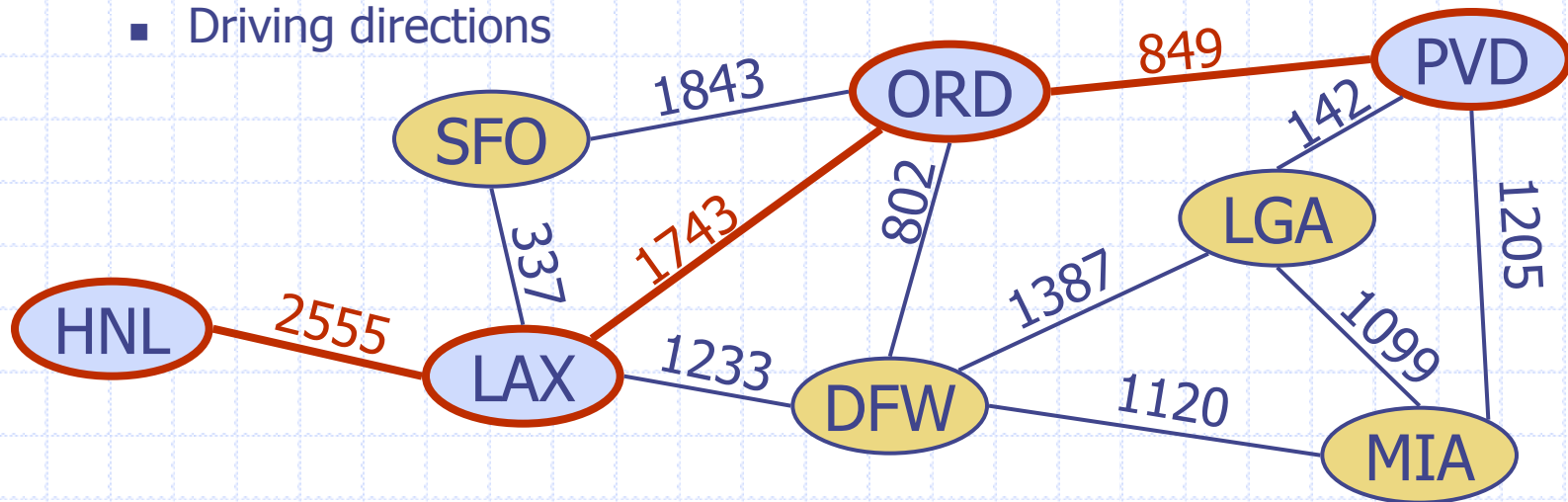
# Weighted Graphs

- 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 Paths

- 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 Properties

## Property 1:

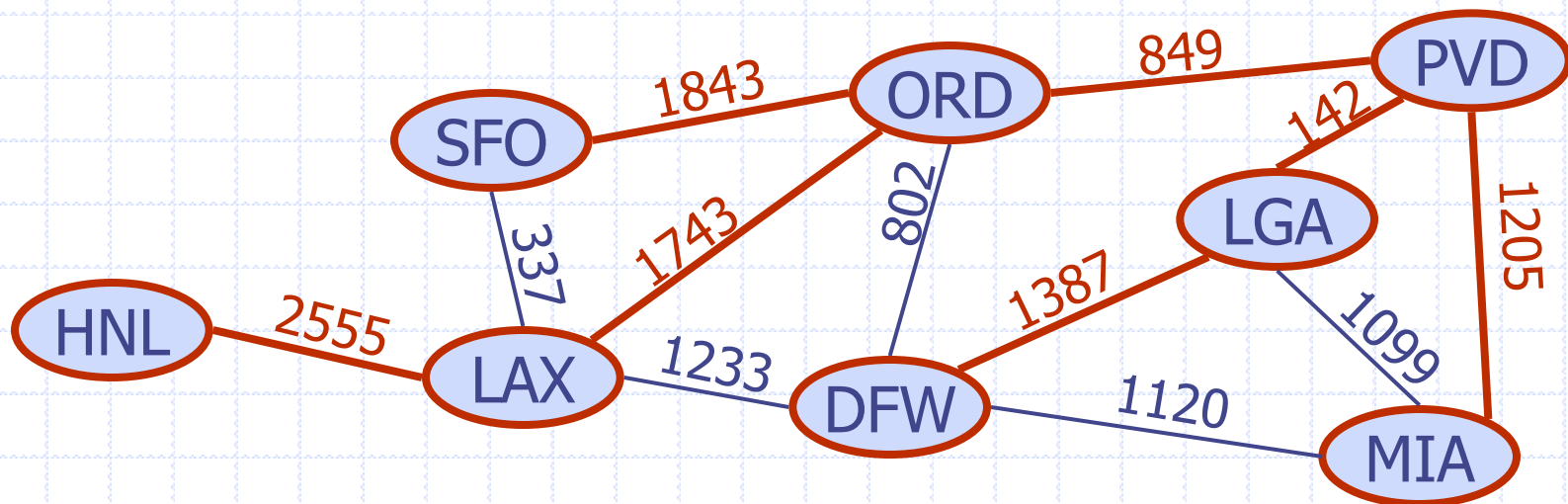
A subpath of a shortest path is itself a shortest path

## Property 2:

There is a tree of shortest paths from a start vertex to all the other vertices

## Example:

Tree of shortest paths from Providence

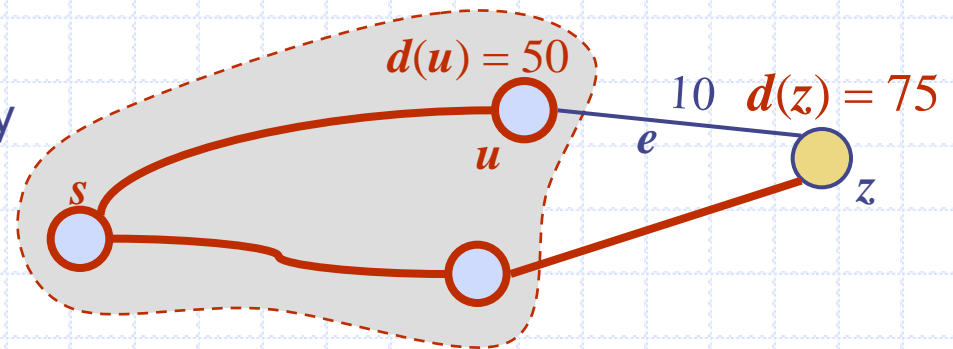


# Dijkstra's Algorithm

- The distance of a vertex  $v$  from a vertex  $s$  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**
- We grow a "**cloud**" of vertices, beginning with  $s$  and eventually covering all the vertices
- We 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
  - We add to the cloud the vertex  $u$  outside the cloud with the smallest distance label,  $d(u)$
  - We update the labels of the vertices adjacent to  $u$

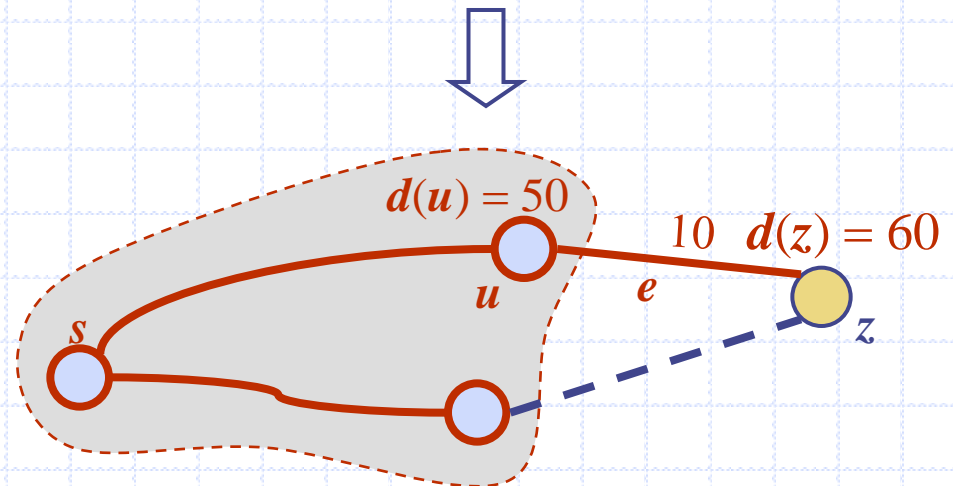
# Edge Relaxation

- Consider an edge  $e = (u, z)$  such that
  - $u$  is the vertex most recently added to the cloud
  - $z$  is not in the cloud

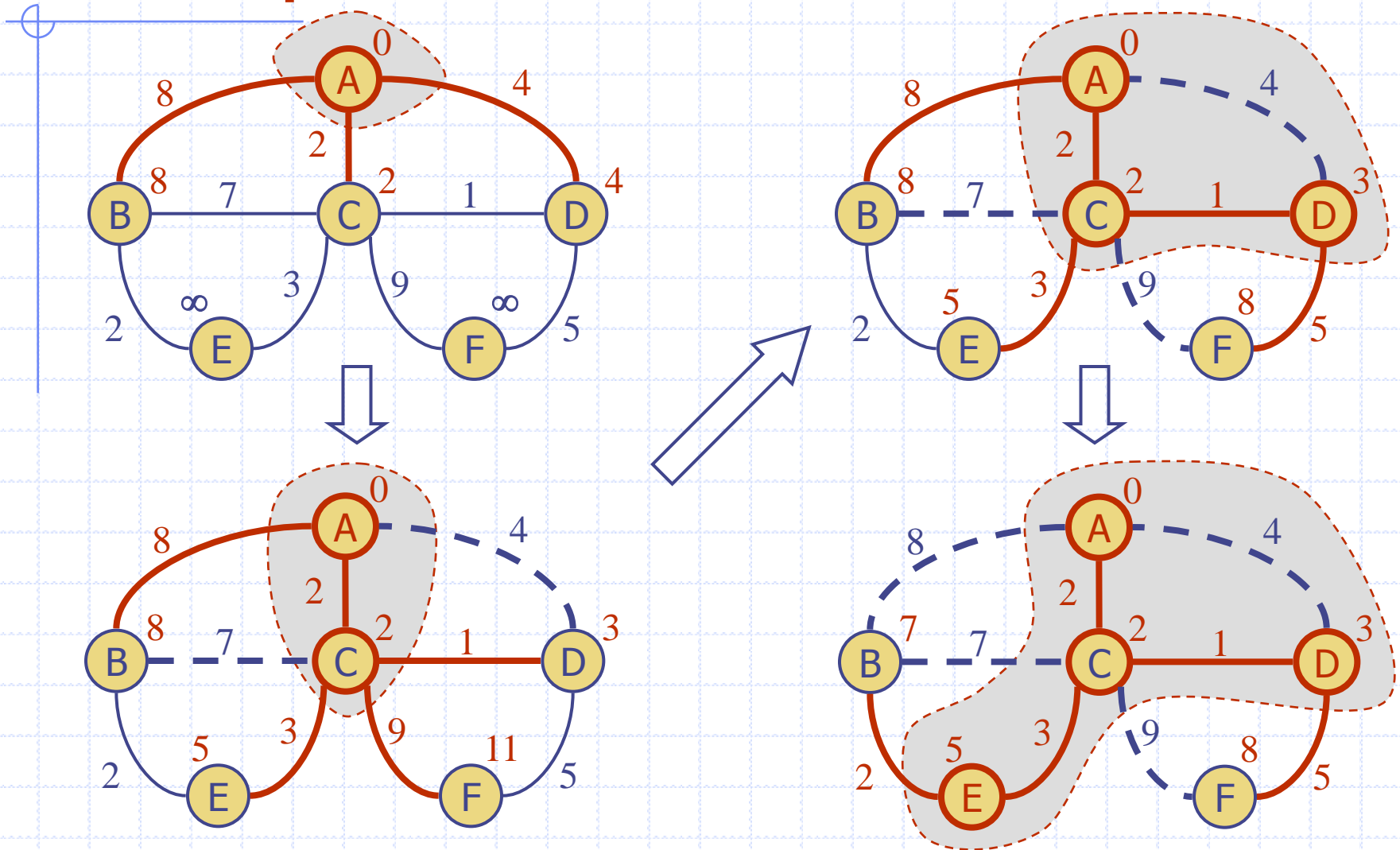


- The relaxation of edge  $e$  updates distance  $d(z)$  as follows:

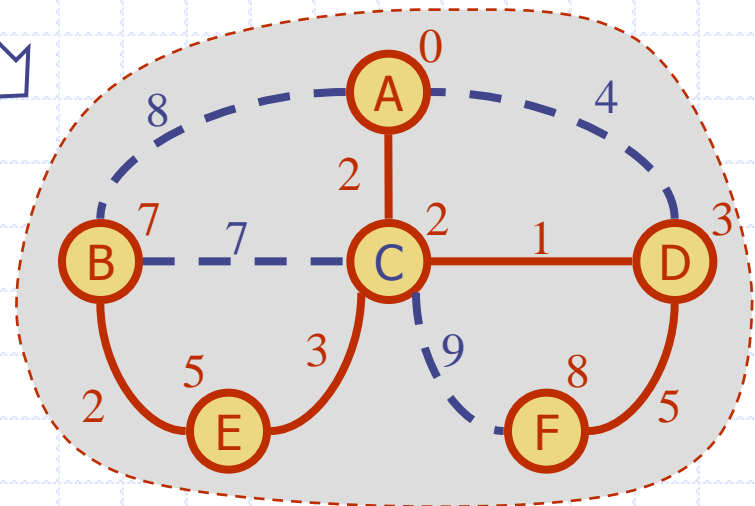
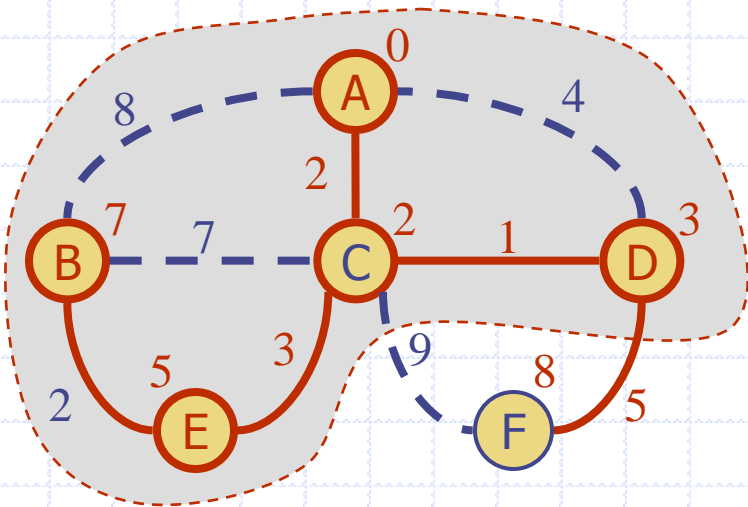
$$d(z) \leftarrow \min\{d(z), d(u) + \text{weight}(e)\}$$



# Example



# Example (cont.)





# Dijkstra's Algorithm

- A heap-based adaptable priority queue with location-aware entries stores the vertices outside the cloud
  - Key: distance
  - Value: vertex
  - Recall that method *replaceKey(l,k)* changes the key of entry *l*
- We store two labels with each vertex:
  - Distance
  - Entry in priority queue

**Algorithm** *DijkstraDistances*(*G*, *s*)

```
Q ← new heap-based priority queue
for all v ∈ G.vertices()
    if v = s
        setDistance(v, 0)
    else
        setDistance(v, ∞)
    l ← Q.insert(getDistance(v), v)
    setEntry(v, l)
while ¬Q.isEmpty()
    l ← Q.removeMin()
    u ← l.getValue()
    for all e ∈ G.incidentEdges(u) { relax e }
        z ← G.opposite(u, e)
        r ← getDistance(u) + weight(e)
        if r < getDistance(z)
            setDistance(z, r)
            Q.replaceKey(getEntry(z), r)
```

# Analysis of Dijkstra's Algorithm

- 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  $\sum_v \deg(v) = 2m$
- The running time can also be expressed as  $O(m \log n)$  since the graph is connected

# Shortest Paths Tree

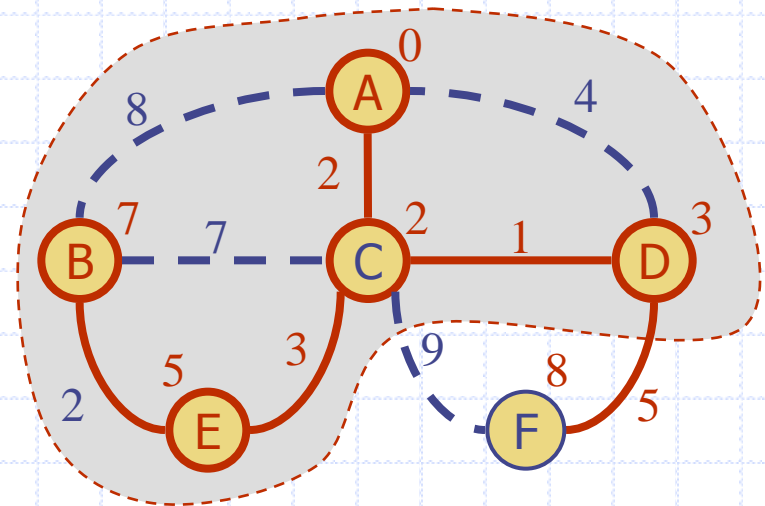
- Using the template method pattern, we can extend Dijkstra's algorithm to return a **tree of shortest paths** from the start vertex to all other vertices
- We store with each vertex a third label:
  - parent edge in the shortest path tree
- In the edge relaxation step, we update the parent label

**Algorithm** *DijkstraShortestPathsTree*( $G, s$ )

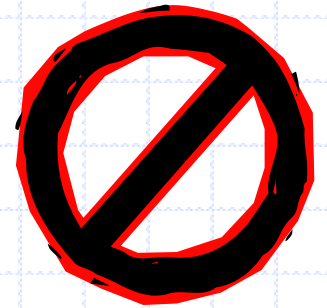
```
...  
for all  $v \in G.vertices()$   
...  
setParent( $v, \emptyset$ )  
...  
  
for all  $e \in G.incidentEdges(u)$   
  { relax edge  $e$  }  
   $z \leftarrow G.opposite(u, e)$   
   $r \leftarrow getDistance(u) + weight(e)$   
  if  $r < getDistance(z)$   
    setDistance( $z, r$ )  
    setParent( $z, e$ )  
    Q.replaceKey(getEntry( $z, r$ )
```

# Why Dijkstra's Algorithm Works

- Dijkstra's algorithm is based on the greedy method. It adds vertices by increasing distance.
  - Suppose it didn't find all shortest distances. Let F be the first wrong vertex the algorithm processed.
  - When the previous node, D, on the true shortest path was considered, its distance was correct
  - But the edge (D,F) was **relaxed** at that time!
  - Thus, so long as  $d(F) \geq d(D)$ , F's distance cannot be wrong. That is, there is no wrong vertex
- 
- The diagram shows a graph with nodes A, B, C, D, and E. Node A is at the top with a distance of 0. Node B is on the left with a distance of 7. Node C is in the middle with a distance of 2. Node D is on the right with a distance of 9. Node E is at the bottom with a distance of 5. Edges are shown with weights: A-B (8), A-C (2), B-C (7), B-E (2), C-E (3), and C-D (9). A dashed blue line indicates the shortest path from A to C to D. A dashed red line indicates the shortest path from A to C to E. A solid red line indicates the shortest path from A to C to E to D. A dashed red line also connects A to B to E to D.

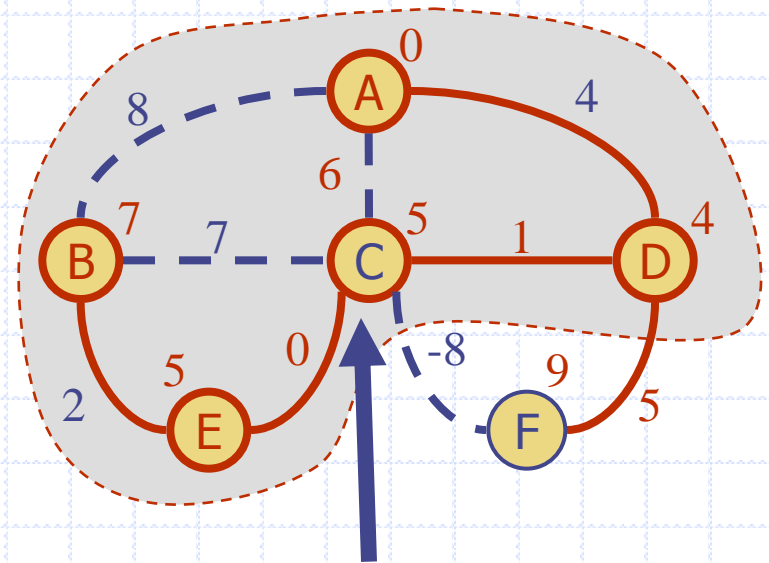


# Why It Doesn't Work for Negative-Weight Edges



◆ 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.



C's true distance is 1, but it is already in the cloud with  $d(C)=5$ !

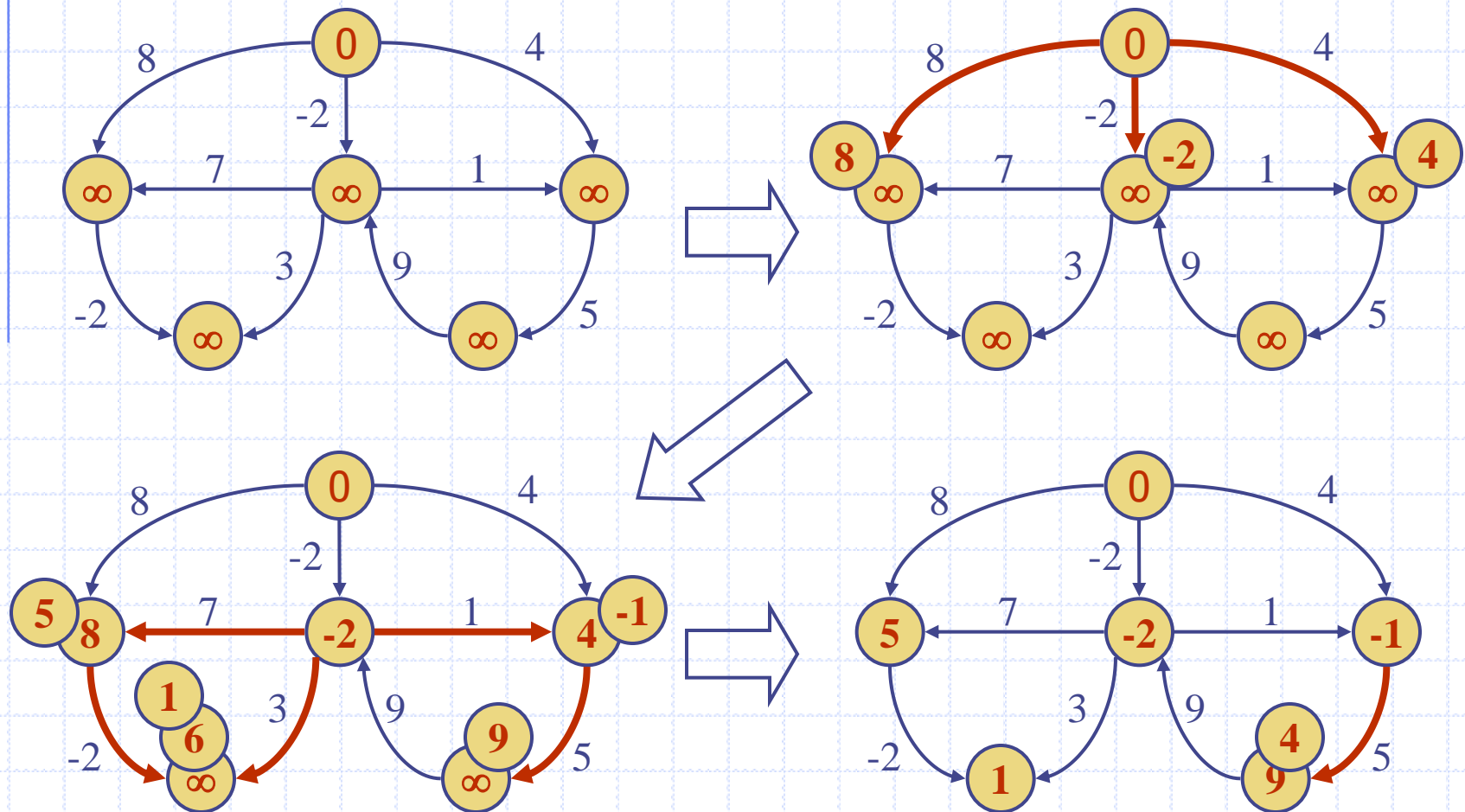
# Bellman-Ford Algorithm (not in book)

- Works even with negative-weight edges
- Must assume directed edges (for otherwise we would have negative-weight cycles)
- Iteration  $i$  finds all shortest paths that use  $i$  edges.
- 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  $e \in G.edges()$   
      { relax edge  $e$  }  
       $u \leftarrow G.origin(e)$   
       $z \leftarrow G.opposite(u, e)$   
       $r \leftarrow getDistance(u) + weight(e)$   
      if  $r < getDistance(z)$   
        setDistance( $z, r$ )
```

# Bellman-Ford Example

Nodes are labeled with their  $d(v)$  values





# DAG-based Algorithm (not in book)

- ❑ Works even with negative-weight edges
- ❑ Uses topological order
- ❑ Doesn't use any fancy data structures
- ❑ Is much faster than Dijkstra's algorithm
- ❑ Running time:  $O(n+m)$ .

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



# DAG Example

Nodes are labeled with their  $d(v)$  values

