

Data Structure - Spring 2022

17. Graph – Part 2

Walid Abdullah Al

Computer and Electronic Systems Engineering
Hankuk University of Foreign Studies

TA: **Seong Joo Kim**

Based on:

Goodrich, Chapter 9,11

Karumanchi, Chapter 6-7

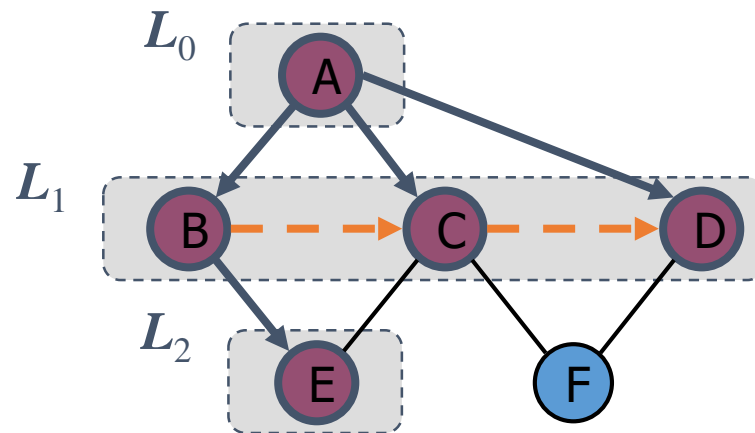
Slides by Prof. Yung Yi, KAIST

Slides by Prof. Chansu Shin, HUFS



Computer Vision Lab
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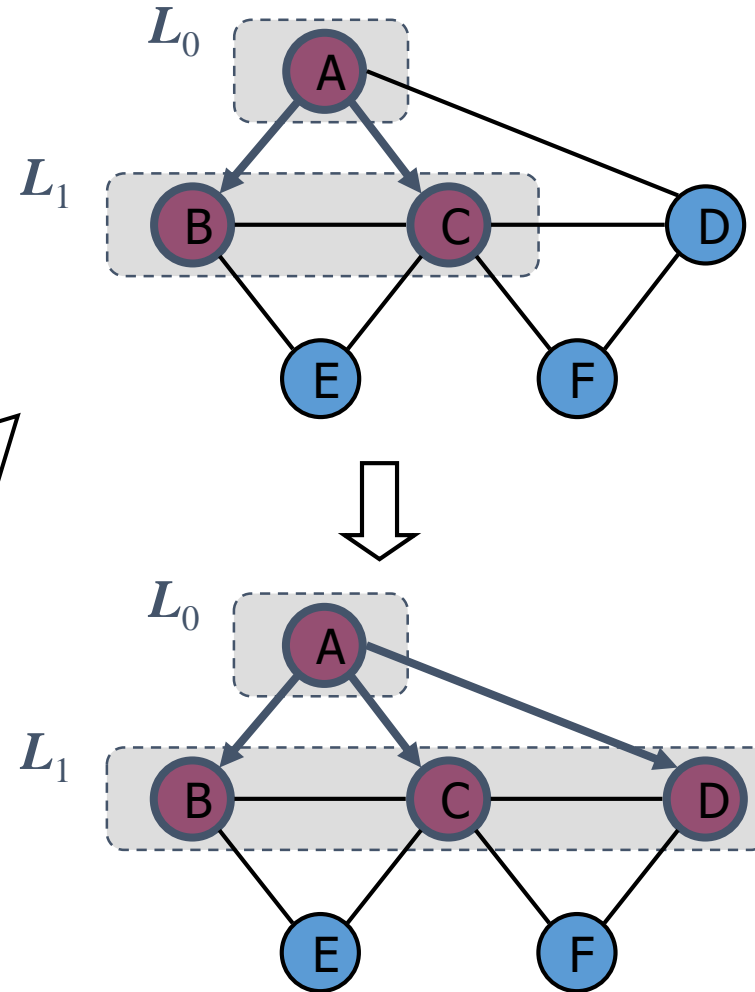
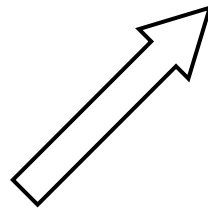
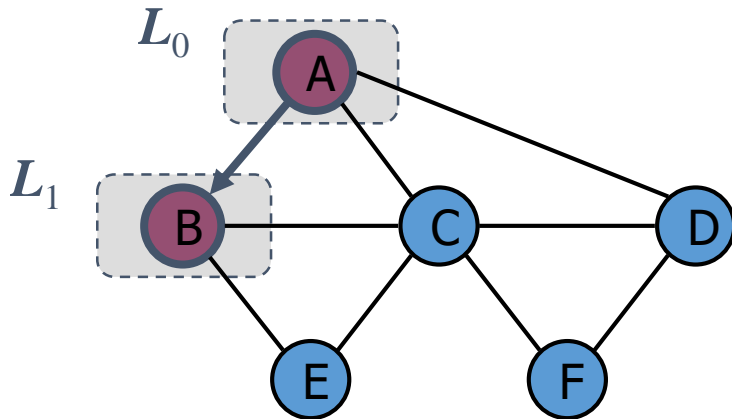
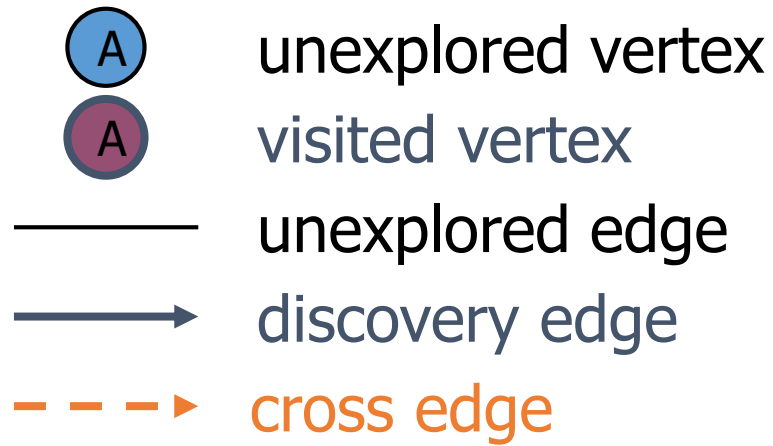
Breadth-First Search



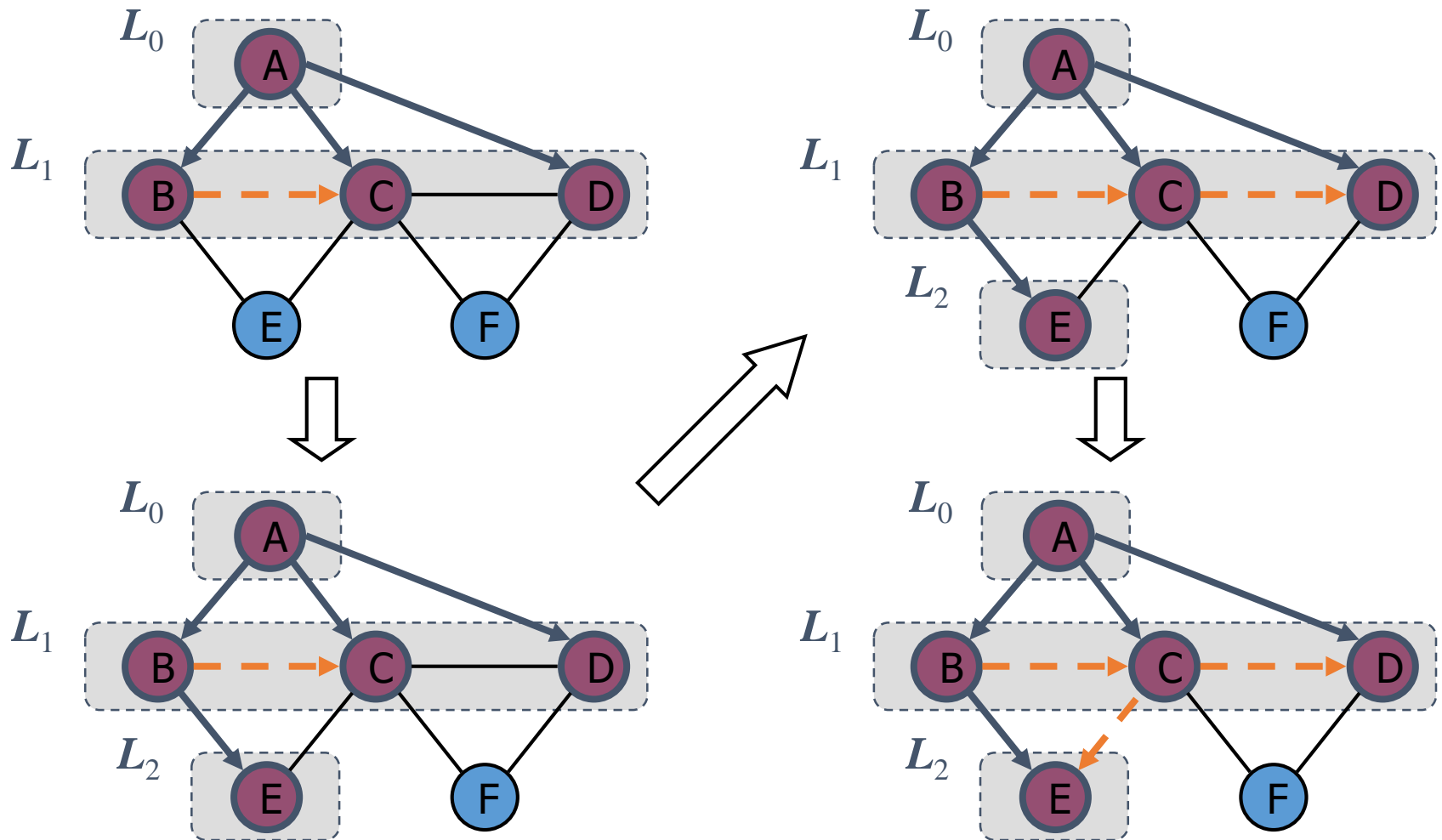
Breadth-First Search

- Breadth-first search (BFS) is another general technique for traversing a graph
- Let's look at the example

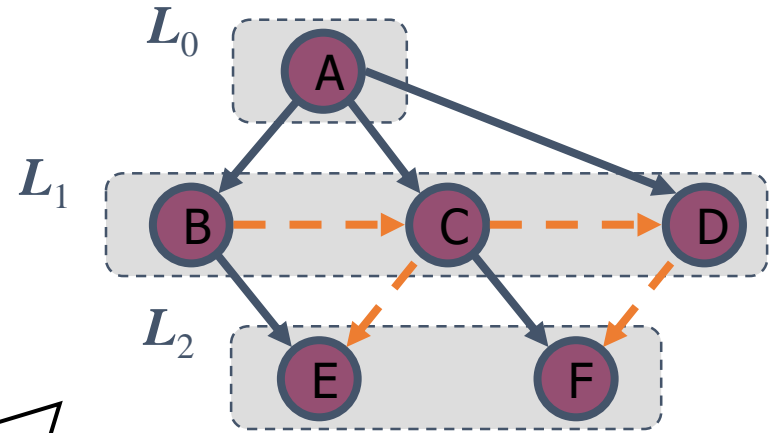
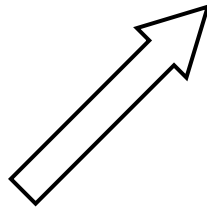
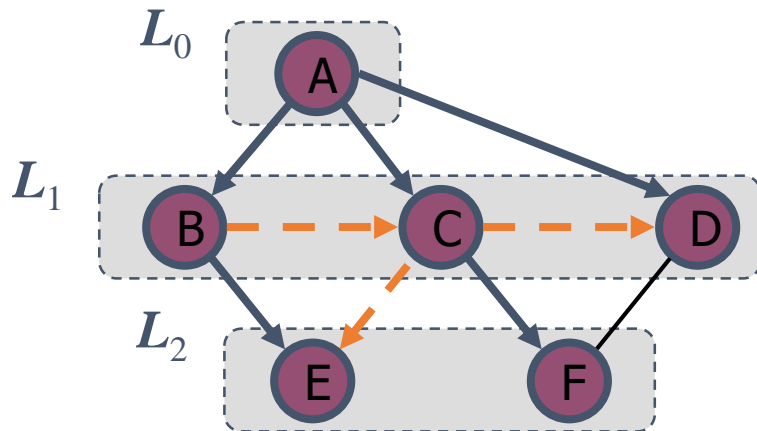
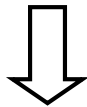
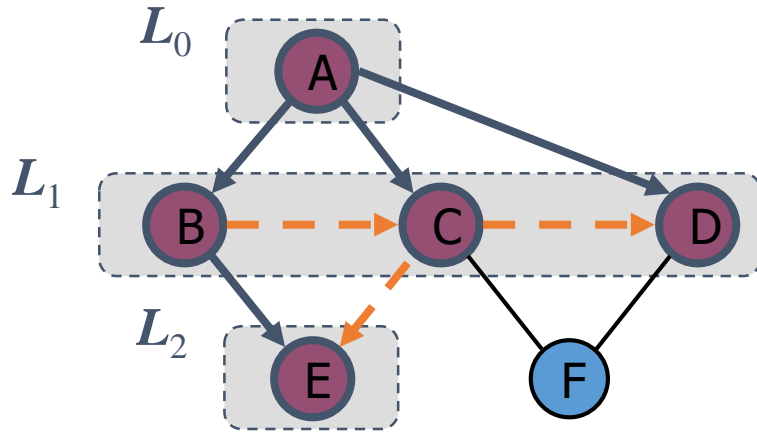
Example



Example (cont.)



Example (cont.)



Breadth-First Search

- **A BFS traversal of a graph G**

- Visits all the vertices and edges of G
- Determines whether G is connected
- Computes the connected components of G
- Computes a spanning forest of G

- **BFS on a graph with n vertices and m edges takes $O(n + m)$ time**

- **BFS can be further extended to solve other graph problems**

- Find and report a path between two given vertices
- Can label each vertex by the length of a **shortest** path (in terms of # of edges) from the start vertex s
- Find a simple cycle, if there is one

BFS Algorithm

- The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

Algorithm *BFS*(*G*)

Input graph *G*

Output labeling of the edges
and partition of the
vertices of *G*

for all *u* ∈ *G.vertices*()

u.setLabel(UNEXPLORED)

for all *e* ∈ *G.edges*()

e.setLabel(UNEXPLORED)

for all *v* ∈ *G.vertices*()

if *v.getLabel*() = UNEXPLORED

BFS(*G*, *v*)

Algorithm *BFS*(*G*, *s*)

*L*₀ ← new empty sequence

*L*₀.*insertBack*(*s*)

s.setLabel(VISITED)

i ← 0

while ¬*L*_{*i*}.*empty*()

*L*_{*i*+1} ← new empty sequence

for all *v* ∈ *L*_{*i*}.*elements*()

for all *e* ∈ *v.incidentEdges*()

if *e.getLabel*() = UNEXPLORED

w ← *e.opposite*(*v*)

if *w.getLabel*() = UNEXPLORED

e.setLabel(DISCOVERY)

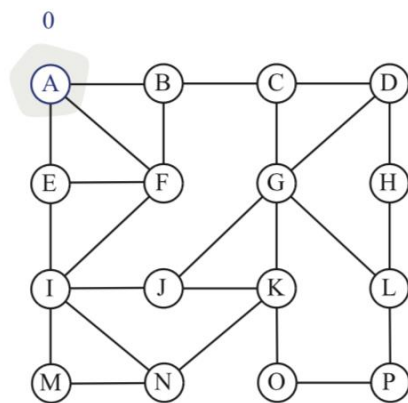
w.setLabel(VISITED)

*L*_{*i*+1}.*insertBack*(*w*)

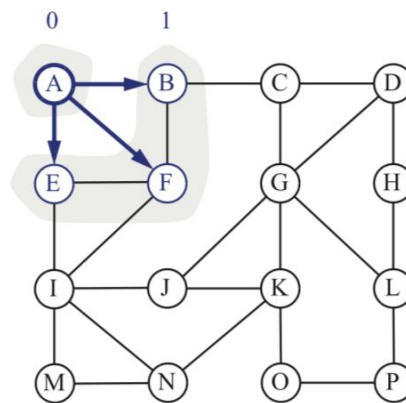
else

e.setLabel(CROSS)

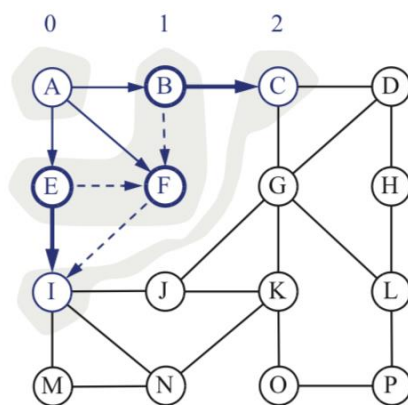
i ← *i* + 1



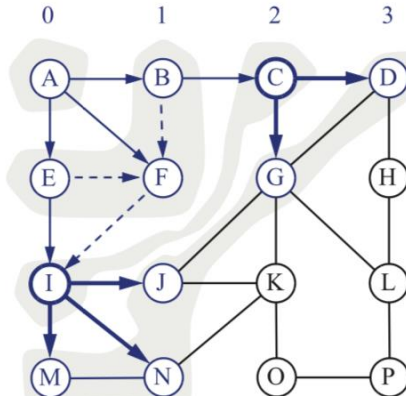
(a)



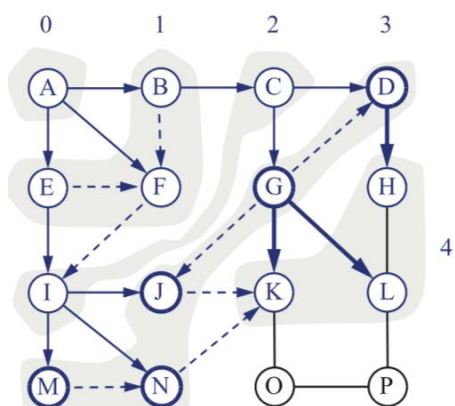
(b)



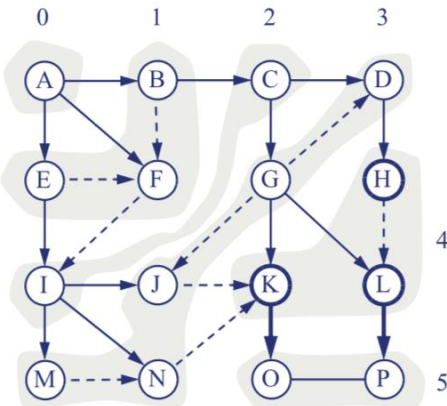
(c)



(d)



(e)



(f)

Properties

Notation

G_s : connected component of s

Property 1

$BFS(G, s)$ visits all the vertices and edges of G_s

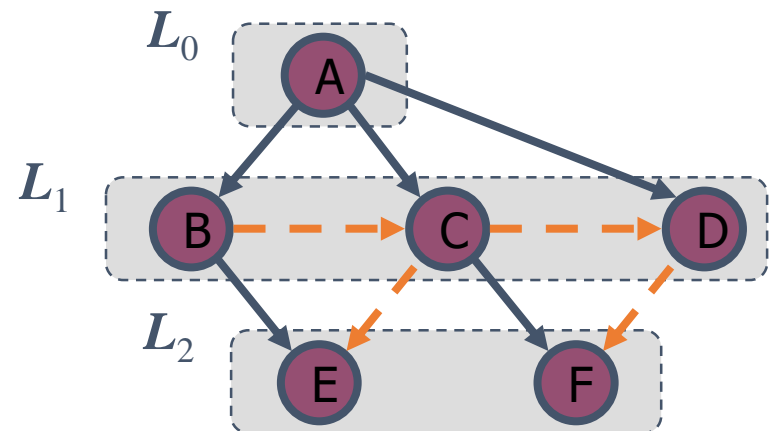
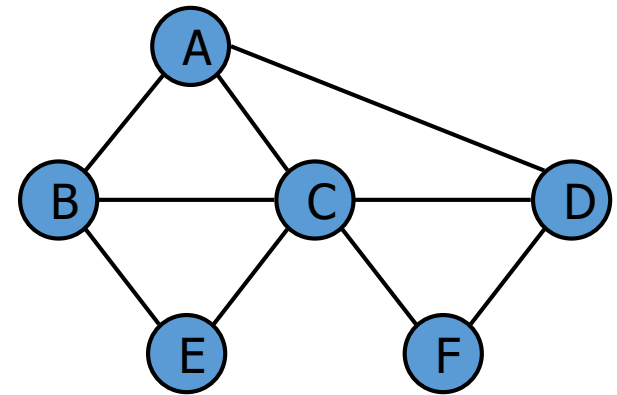
Property 2

The discovery edges labeled by $BFS(G, s)$ form a spanning tree T_s of G_s

Property 3

For each vertex v in L_i

- The path of T_s from s to v has i edges
- Every path from s to v in G_s has at least i edges (i.e., find a shortest path)



Analysis

- **Setting/getting a vertex/edge label takes $O(1)$ time**
- **Each vertex is labeled twice**
 - once as UNEXPLORED
 - once as VISITED
- **Each edge is labeled twice**
 - once as UNEXPLORED
 - once as DISCOVERY or CROSS
- **Each vertex is inserted once into a sequence L_i**
- **Method incidentEdges is called once for each vertex**
- **BFS runs in $O(n + m)$ time provided the graph is represented by the adjacency list structure**
 - Recall that $\sum_v \deg(v) = 2m$

Applications

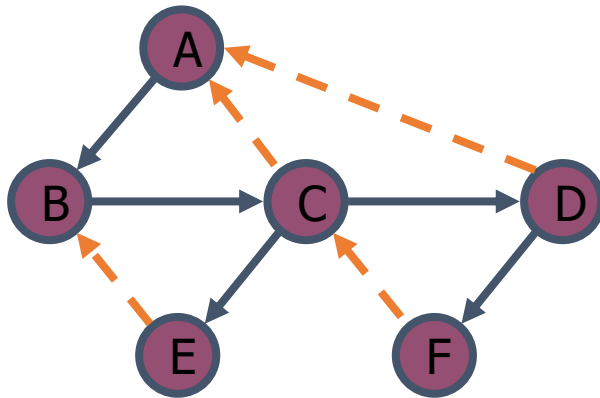
- **Using the template method pattern, we can specialize the BFS traversal of a graph G to solve the following problems in $O(n + m)$ time**
 - Compute the connected components of G
 - Compute a spanning forest of G
 - Find a simple cycle in G , or report that G is a forest
 - Given two vertices of G , find a path in G between them with the minimum number of edges, or report that no such path exists

DFS vs. BFS

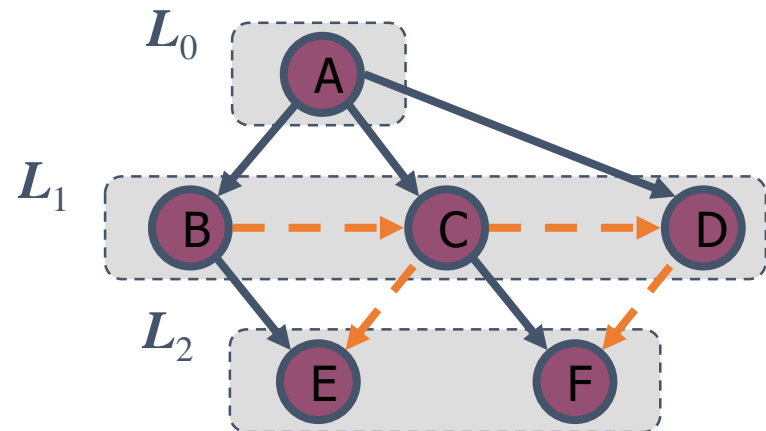
Applications	DFS	BFS
Spanning forest, connected components, paths, cycles	✓	✓
Shortest paths		✓
Biconnected components (how?)	✓	

Biconnected components:

- Connected
- Even after removing any vertex the graph remains connected



DFS

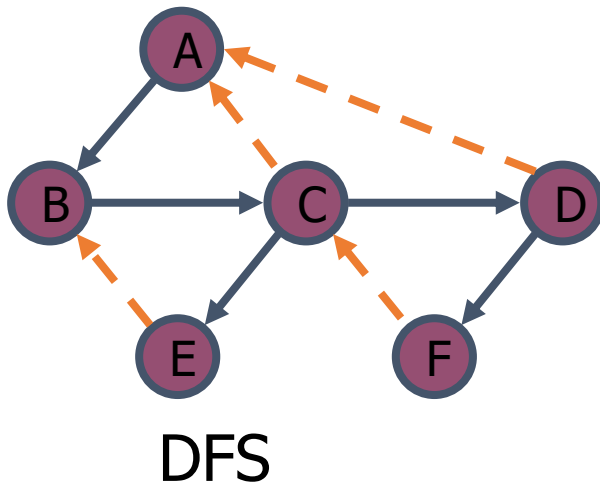


BFS

DFS vs. BFS (cont.)

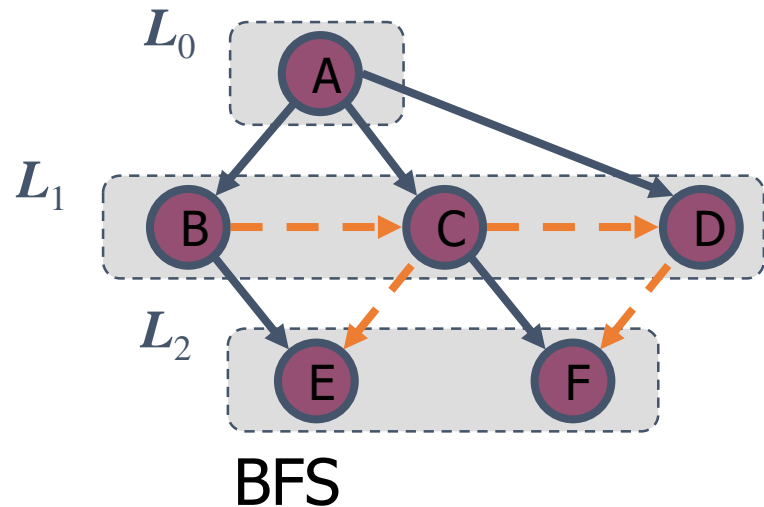
Back edge (v, w)

- w is an ancestor of v in the tree of discovery edges

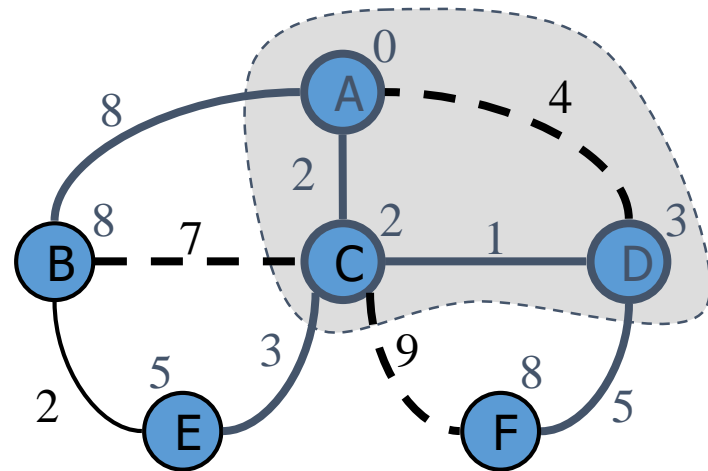


Cross edge (v, w)

- w is in the same level as v or in the next level

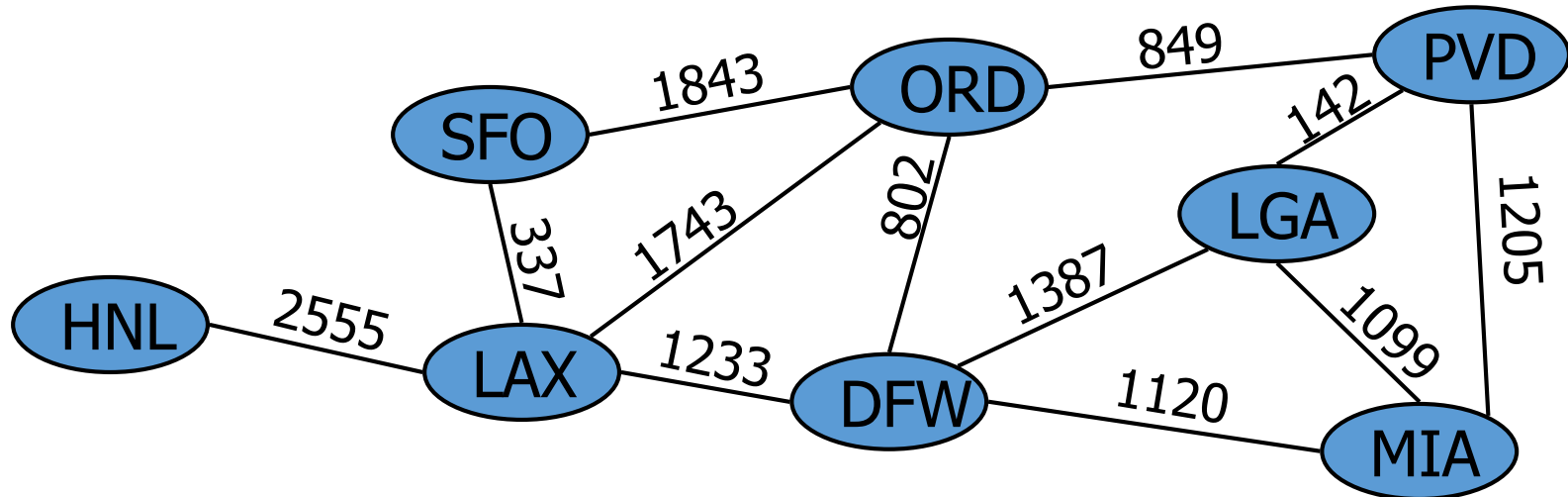


Shortest Paths



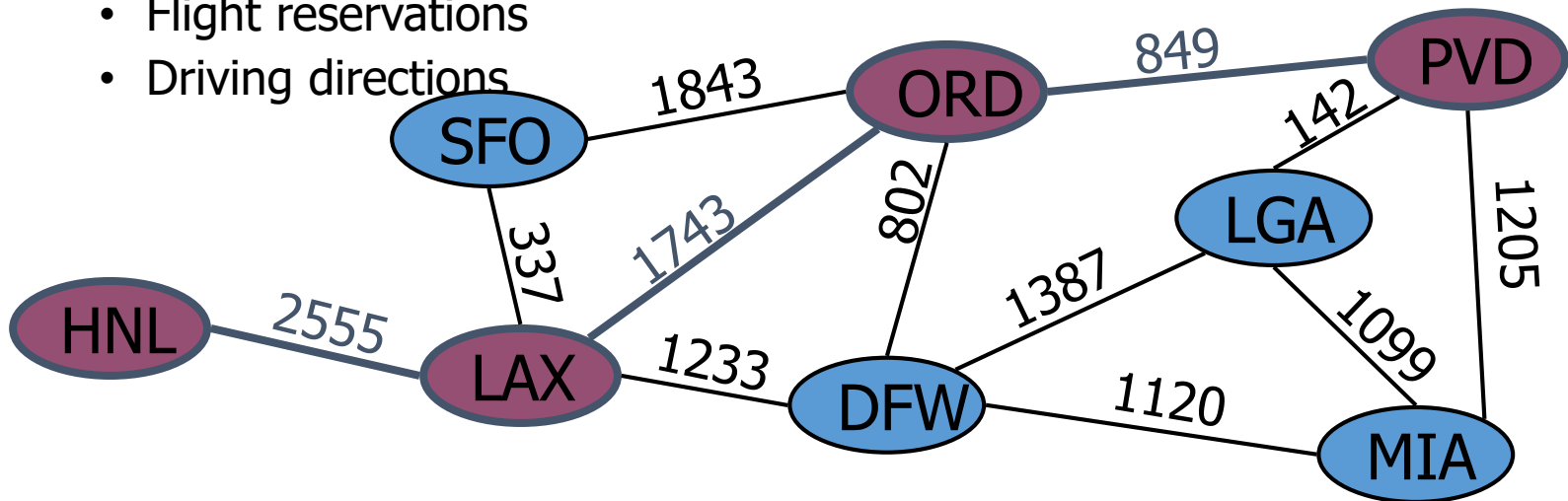
Weighted Graphs

- 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:

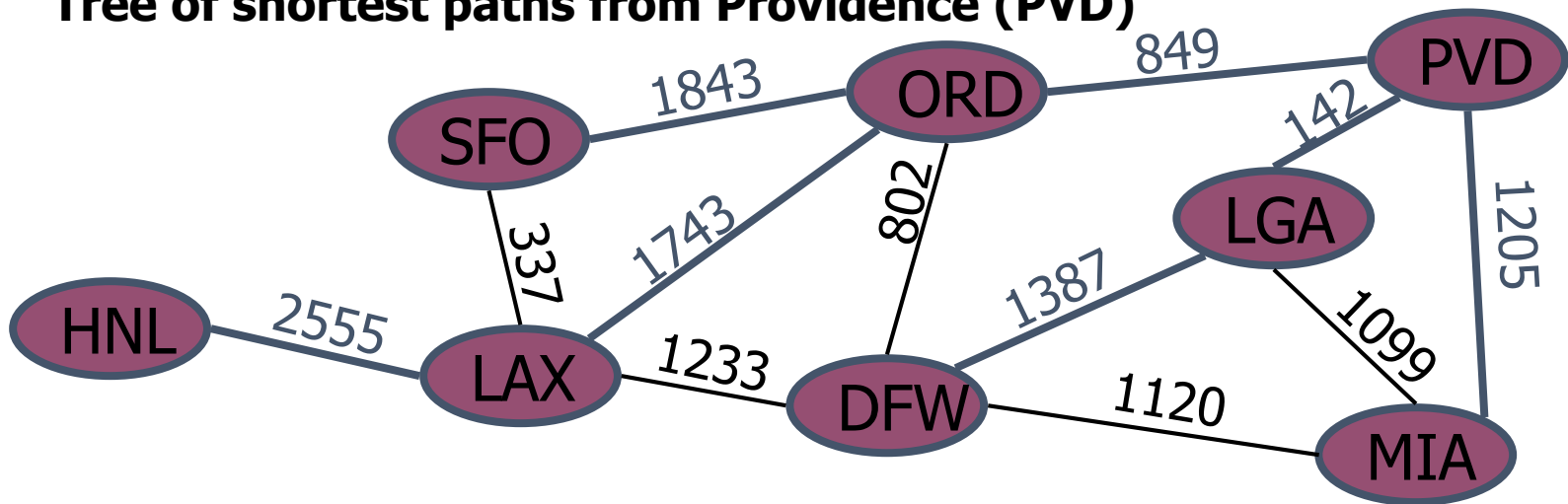
A subpath of a shortest path is itself a shortest path

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 (PVD)



Our goal and Initial Ideas

- **Goal**

- Given a source vertex s , compute the shortest paths to all other vertices

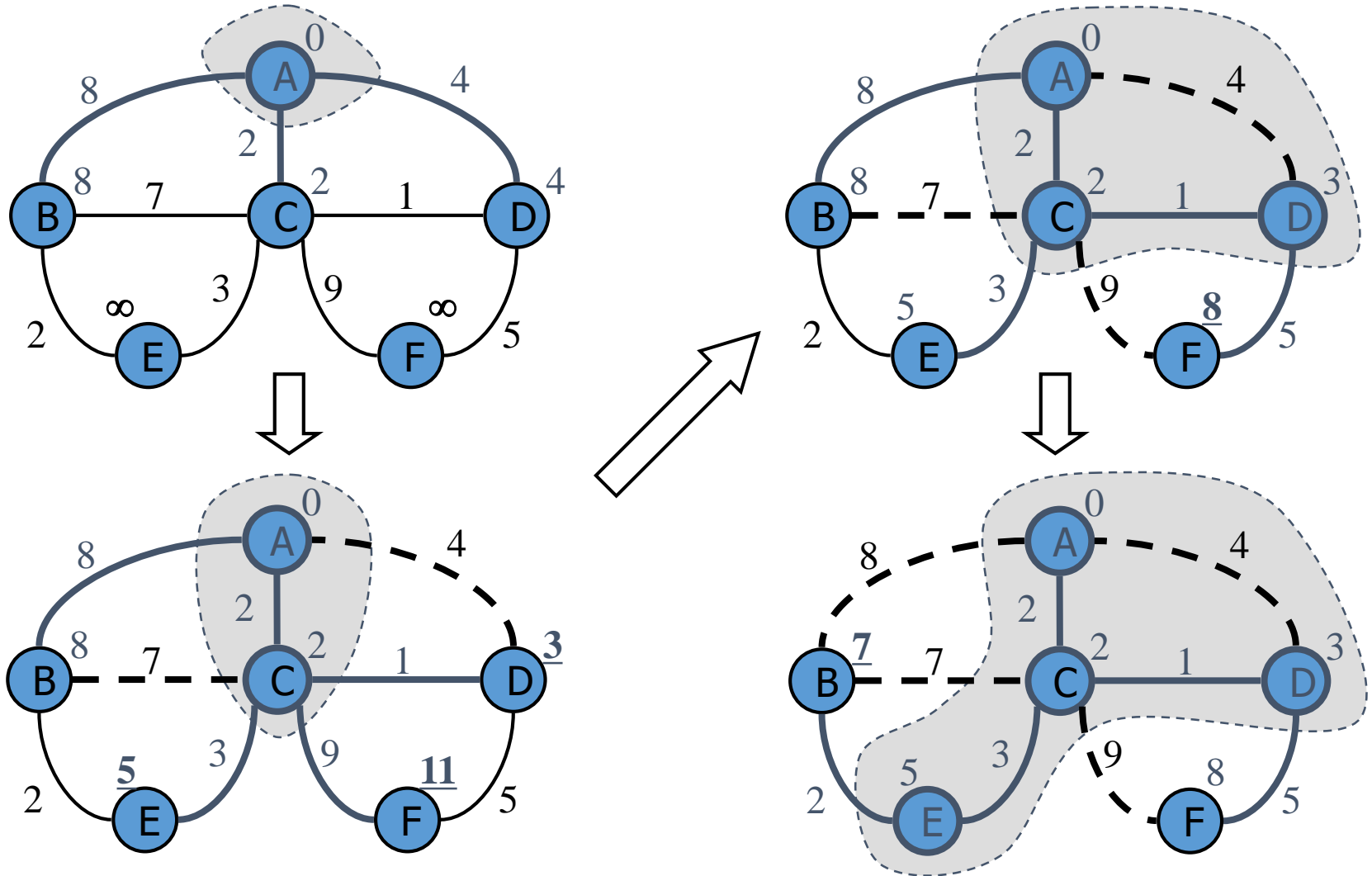
- **Initial Ideas**

- Compute **all** the paths from the source s to other vertices
- Take the minimums
- How much complexity?
 - Exponential (not a polynomial time algorithm)
- Why is this algorithm stupid?
 - Ignore the wisdom from computing the minimum path for computing other minimum paths

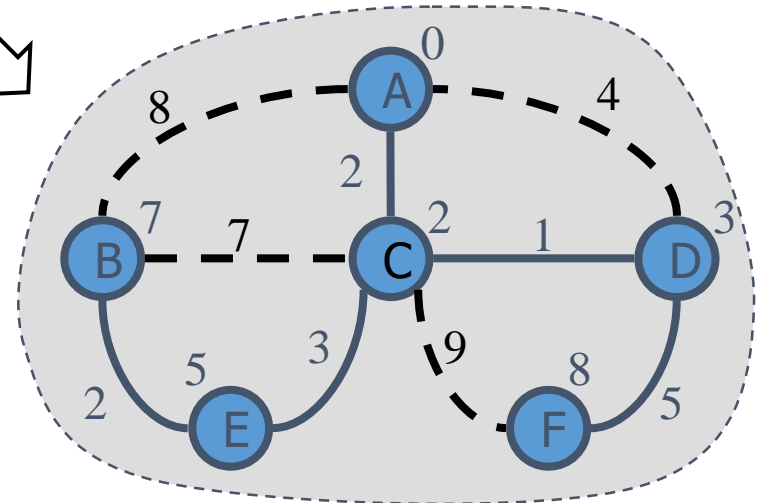
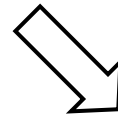
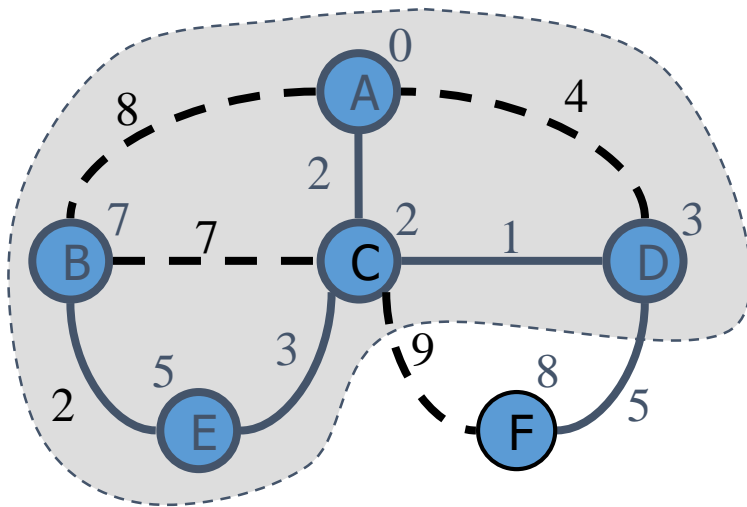
Dijkstra's Algorithm (1)

- 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
 - Remember the “wisdom”
- **Example**
 - What is your distance to “Obama” in facebook? 50
 - Suppose that MoonJaein becomes your friend
 - What is your distance to “Obama” then?
 - Probably much shorter than 50. Maybe 2?
😊

Example first



Example (cont.)



Dijkstra's Algorithm (2)



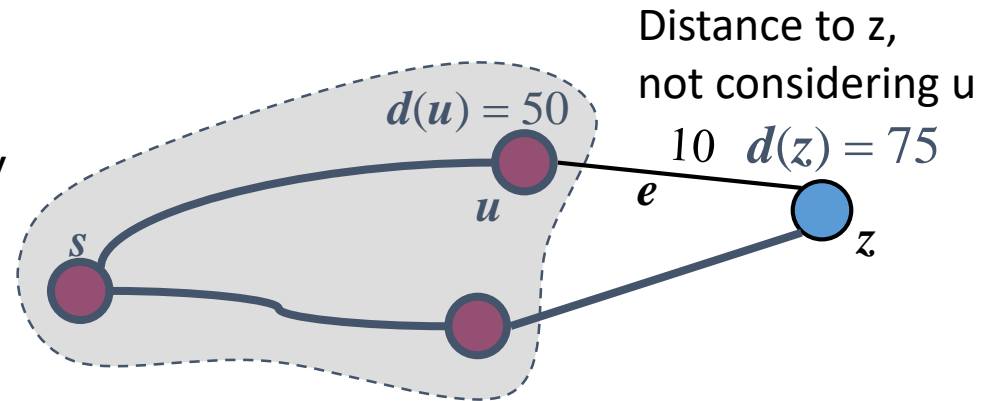
- We store with each vertex v a **label** $d(v)$ representing the distance of v from s in the subgraph consisting of the cloud and its adjacent vertices
- **At each step**
 - We add to the cloud the vertex u outside the cloud with the smallest distance label, $d(u)$
 - We update the labels of the vertices adjacent to u
 - Greedy method: we solve the problem at hand by repeatedly selecting the best choice from among those available in each iteration

Raise your quality standards as high as you can live with, avoid wasting your time on routine problems, and always try to work as closely as possible at the boundary of your abilities. Do this, because it is the only way of discovering how that boundary should be moved forward.

Edsger Dijkstra

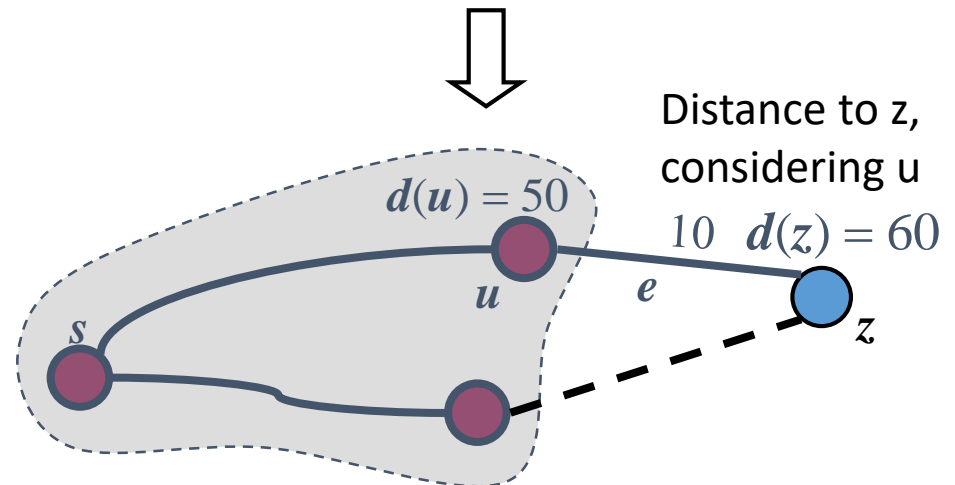
Edge Relaxation

- **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:**

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



Recall: Priority Queue ADT

- **A priority queue stores a collection of entries**
- **Typically, an entry is a pair (key, value), where the key indicates the priority**
- **Main methods of the Priority Queue ADT**
 - `insert(e)` inserts an entry `e`
 - `removeMin()` removes the entry with smallest key
- **Additional methods**
 - `min()` returns, but does not remove, an entry with smallest key
 - `size()`, `empty()`

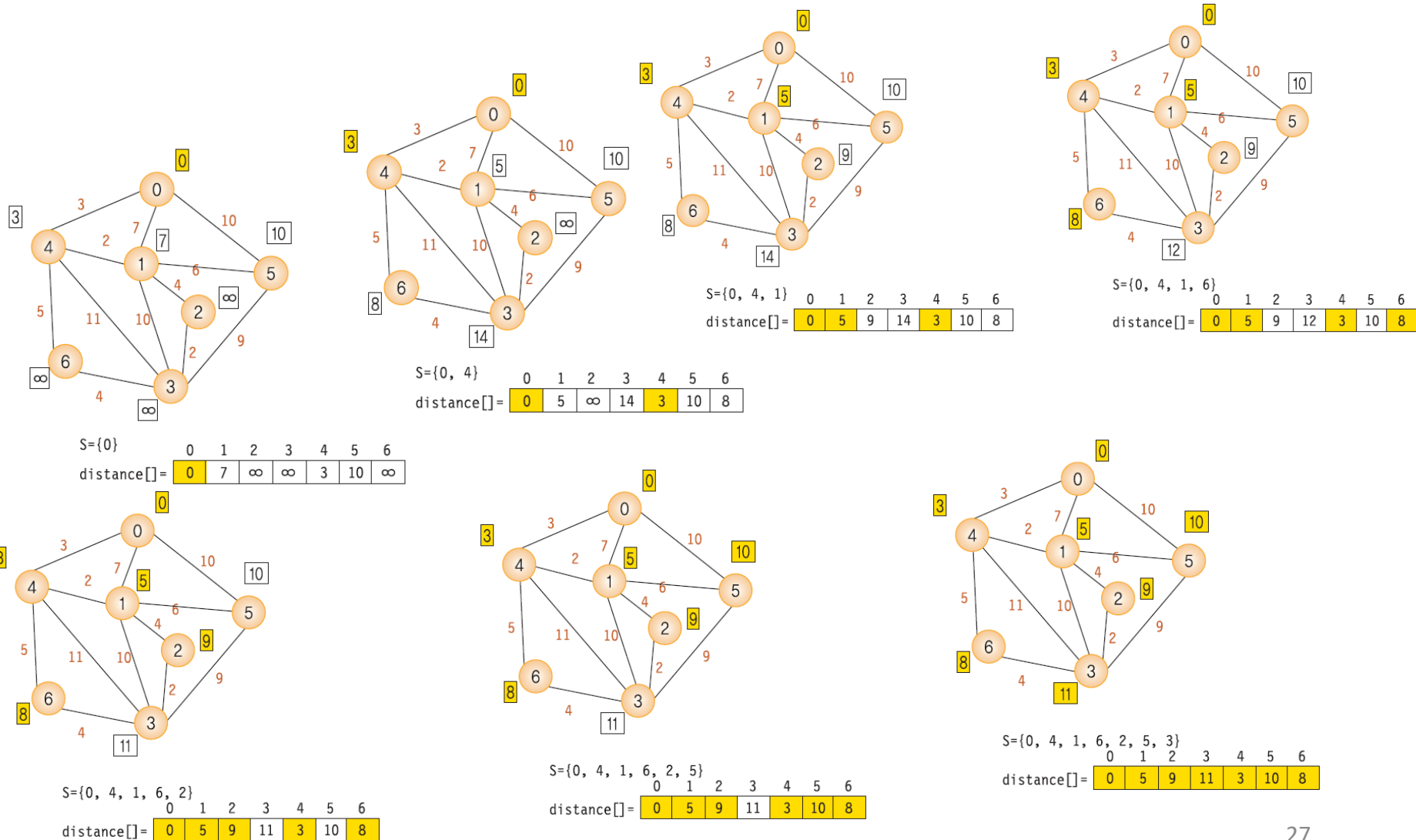
Dijkstra's Algorithm

- **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* with *k*
- **We store two labels with each vertex:**
 - Distance
 - Entry in priority queue
- **We take out the vertex with the minimum distance so far**

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

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

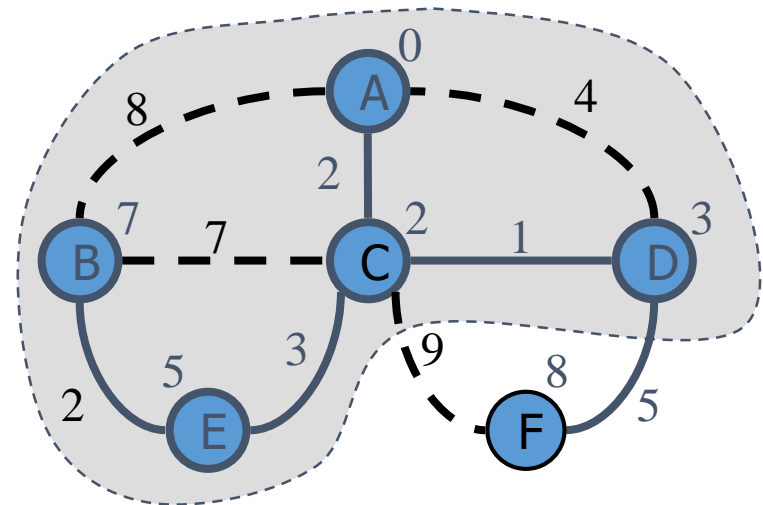
Dijkstra's alg. using distance array



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
- (Question) Why not working for non-negative weight?



Analysis of Dijkstra's Algorithm

- **Graph operations**
 - incidentEdges is called once for each v
- **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 v is inserted once into and removed once from the PQ, where each insertion or removal takes $O(\log n)$ time \rightarrow total $nO(\log n)$
 - The key of a vertex in the PQ is modified at most $\deg(v)$ 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**

Algorithm *DijkstraDistances*(G, s)

```
 $Q \leftarrow$  new heap-based priority queue
for all  $v \in G.vertices()$ 
    if  $v = s$ 
         $v.setDistance(0)$ 
    else
         $v.setDistance(\infty)$ 
 $l \leftarrow Q.insert(v.getDistance(), v)$ 
 $v.setEntry(l)$ 
while  $\neg Q.empty()$ 
     $l \leftarrow Q.removeMin()$ 
     $u \leftarrow l.getValue()$  // take out the closest node
    for all  $e \in u.incidentEdges()$  { relax  $e$  }
         $z \leftarrow e.opposite(u)$ 
         $r \leftarrow u.getDistance() + e.weight()$ 
        if  $r < z.getDistance()$ 
             $z.setDistance(r)$ 
             $Q.replaceKey(z.getEntry(), r)$ 
```