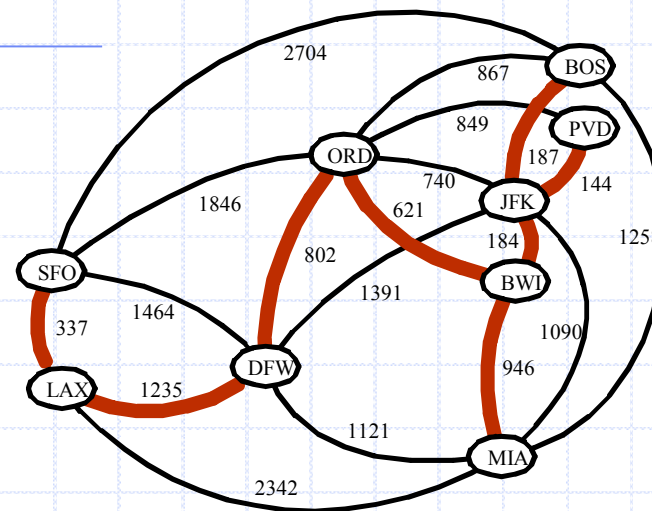


# Minimum Spanning Trees



# Minimum Spanning Trees

## Spanning subgraph

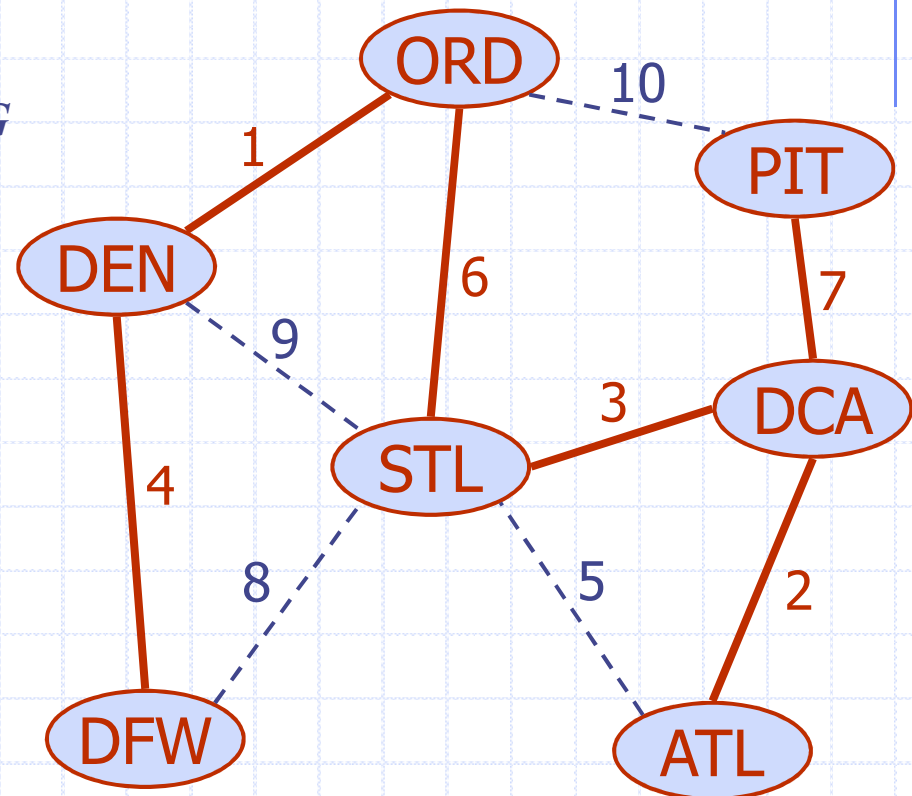
- Subgraph of a graph  $G$  containing all the vertices of  $G$

## Spanning tree

- Spanning subgraph that is itself a (free) tree

## Minimum spanning tree (MST)

- Spanning tree of a weighted graph with minimum total edge weight
- Applications
  - Communications networks
  - Transportation networks



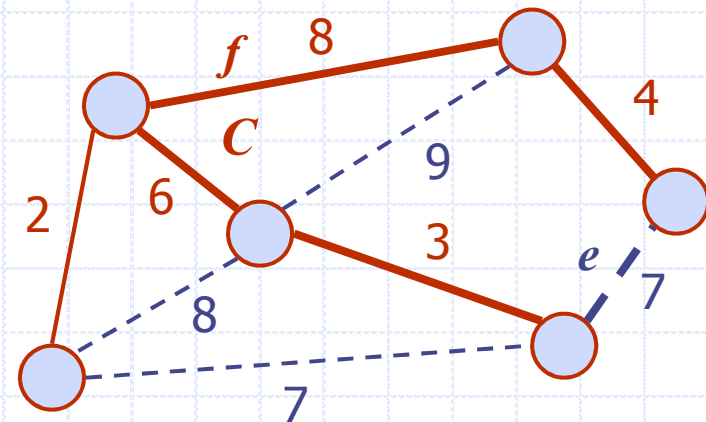
# Cycle Property

## Cycle Property:

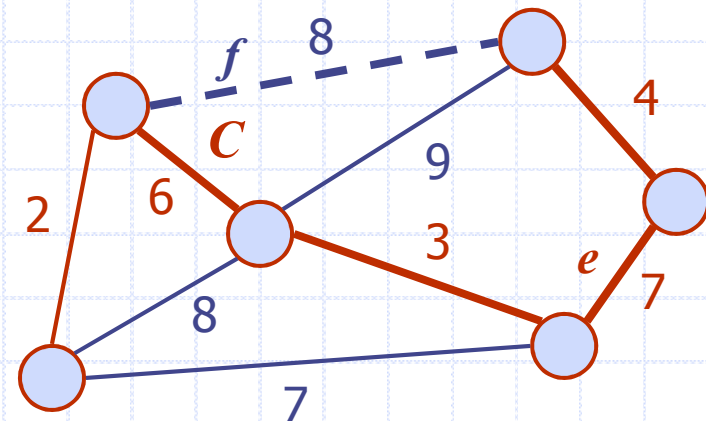
- Let  $T$  be a minimum spanning tree of a weighted graph  $G$
- Let  $e$  be an edge of  $G$  that is not in  $T$  and let  $C$  be the cycle formed by  $e$  with  $T$
- For every edge  $f$  of  $C$ ,  $\text{weight}(f) \leq \text{weight}(e)$

## Proof:

- By contradiction
- If  $\text{weight}(f) > \text{weight}(e)$  we can get a spanning tree of smaller weight by replacing  $e$  with  $f$



Replacing  $f$  with  $e$  yields a better spanning tree



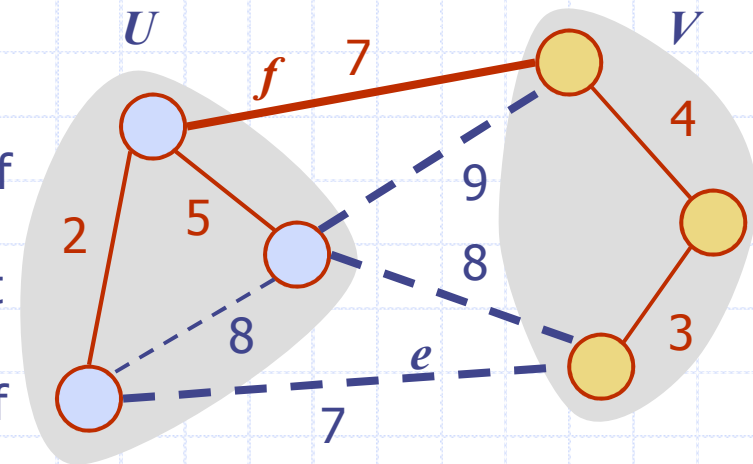
# Partition Property

## Partition Property:

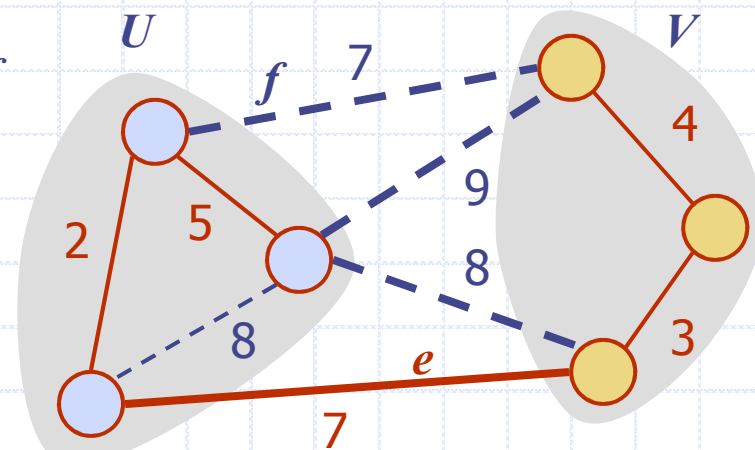
- Consider a partition of the vertices of  $G$  into subsets  $U$  and  $V$
- Let  $e$  be an edge of minimum weight across the partition
- There is a minimum spanning tree of  $G$  containing edge  $e$

## Proof:

- Let  $T$  be an MST of  $G$
- If  $T$  does not contain  $e$ , consider the cycle  $C$  formed by  $e$  with  $T$  and let  $f$  be an edge of  $C$  across the partition
- By the cycle property,  
 $\text{weight}(f) \leq \text{weight}(e)$
- Thus,  $\text{weight}(f) = \text{weight}(e)$
- We obtain another MST by replacing  $f$  with  $e$



Replacing  $f$  with  $e$  yields another MST

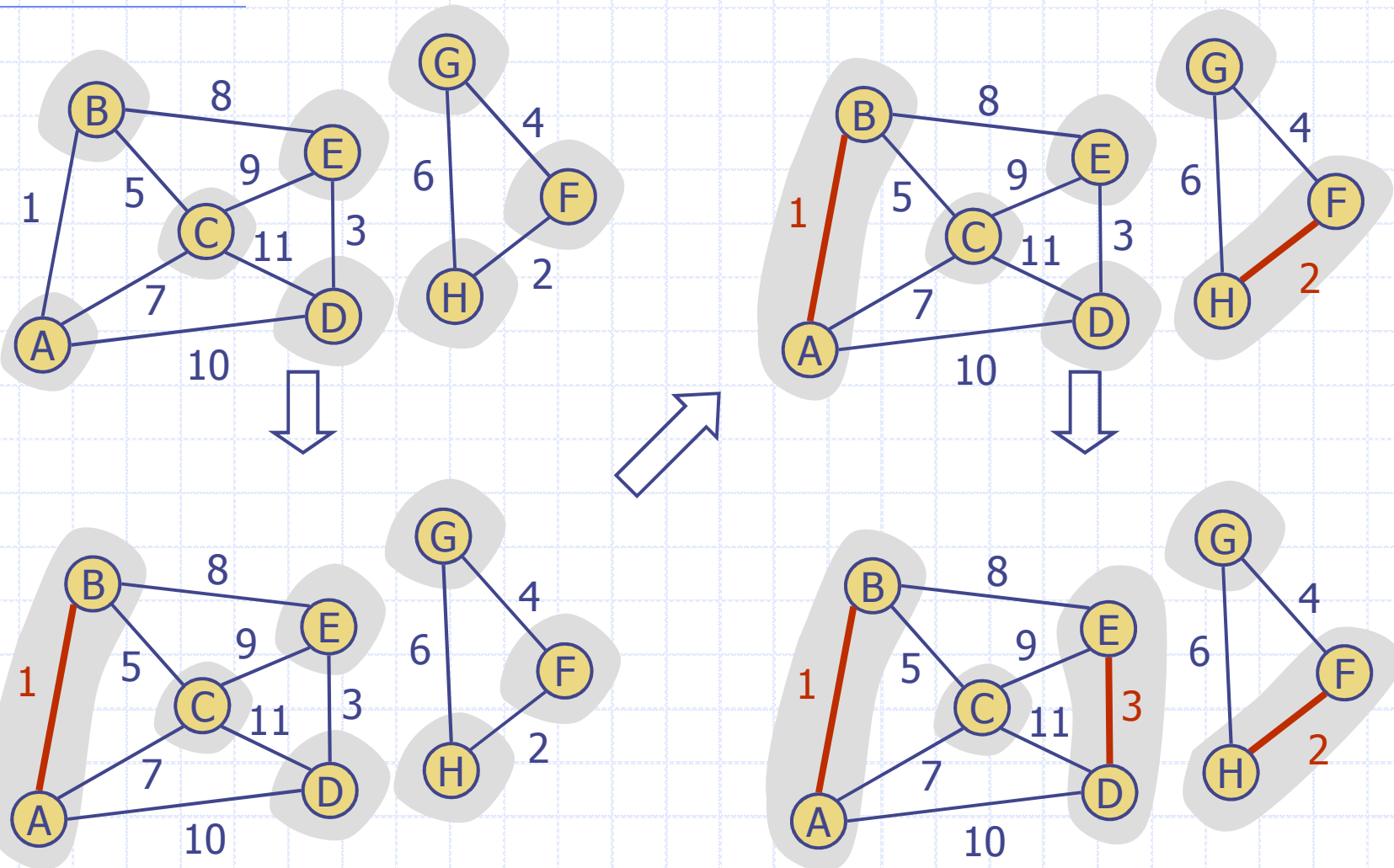


# Kruskal's Algorithm

- Maintain a partition of the vertices into clusters
  - Initially, single-vertex clusters
  - Keep an MST for each cluster
  - Merge “closest” clusters and their MSTs
- A priority queue stores the edges outside clusters
  - Key: weight
  - Element: edge
- At the end of the algorithm
  - One cluster and one MST

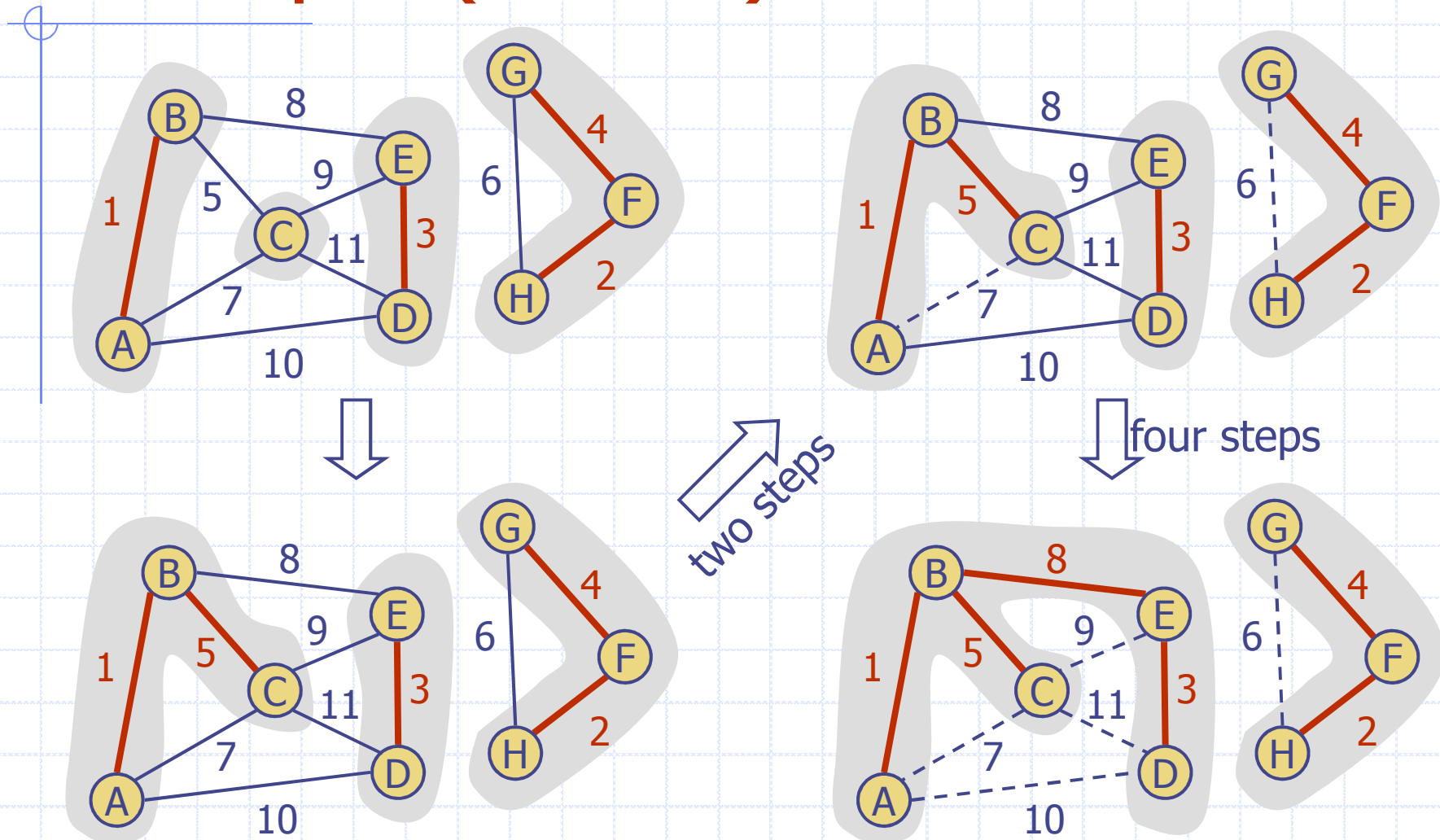
```
Algorithm KruskalMST( $G$ )  
  for each vertex  $v$  in  $G$  do  
    Create a cluster consisting of  $v$   
  let  $Q$  be a priority queue.  
  Insert all edges into  $Q$   
   $T \leftarrow \emptyset$   
  { $T$  is the union of the MSTs of the clusters}  
  while  $T$  has fewer than  $n - 1$  edges do  
     $e \leftarrow Q.removeMin().getValue()$   
     $[u, v] \leftarrow G.endVertices(e)$   
     $A \leftarrow getCluster(u)$   
     $B \leftarrow getCluster(v)$   
    if  $A \neq B$  then  
      Add edge  $e$  to  $T$   
      mergeClusters( $A, B$ )  
  return  $T$ 
```

# Example





## Example (contd.)

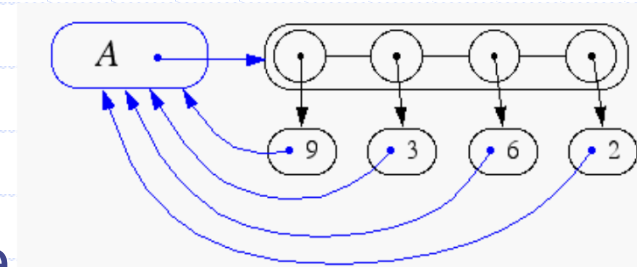


# Data Structure for Kruskal's Algorithm

- ❑ The algorithm maintains a forest of trees
- ❑ A priority queue extracts the edges by increasing weight
- ❑ An edge is accepted if it connects distinct trees
- ❑ We need a data structure that maintains a **partition**, i.e., a collection of **disjoint sets**, with operations:
  - **makeSet**(u): create a set consisting of u
  - **find**(u): return the set storing u
  - **union**(A, B): replace sets A and B with their union



# Recall of List-based Partition



- Each set is stored in a sequence
- Each element has a reference back to the set
  - operation **find**(u) takes  **$O(1)$**  time, and returns the set of which u is a member.
  - in operation **union**(A,B), we move the elements of the smaller set to the sequence of the larger set and update their references
  - the time for operation **union**(A,B) is  **$\min(|A|, |B|)$**
- Whenever an element is processed, it goes into a set of size at least double, hence each element is processed at most  $\log n$  times

# Partition-Based Implementation

- Partition-based version of Kruskal's Algorithm
  - Cluster merges as unions
  - Cluster locations as finds
- Running time
  - $O((n + m) \log n)$ 
    - PQ operations  $O(m \log n)$
    - UF operations  $O(n \log n)$

## Algorithm *KruskalMST*( $G$ )

Initialize a partition  $P$

for each vertex  $v$  in  $G$  do

$P.makeSet(v)$

let  $Q$  be a priority queue.

Insert all edges into  $Q$

$T \leftarrow \emptyset$

{ $T$  is the union of the MSTs of the clusters}

while  $T$  has fewer than  $n - 1$  edges do

$e \leftarrow Q.removeMin().getValue()$

$[u, v] \leftarrow G.endVertices(e)$

$A \leftarrow P.find(u)$

$B \leftarrow P.find(v)$

if  $A \neq B$  then

Add edge  $e$  to  $T$

$P.union(A, B)$

return  $T$

# Prim-Jarnik's Algorithm

- Similar to Dijkstra's algorithm
- We pick an arbitrary vertex  $s$  and we grow the MST as a cloud of vertices, starting from  $s$
- We store with each vertex  $v$  label  $d(v)$  representing the smallest weight of an edge connecting  $v$  to a vertex in the cloud
- At each step:
  - We add to the cloud the vertex  $u$  outside the cloud with the smallest distance label
  - We update the labels of the vertices adjacent to  $u$

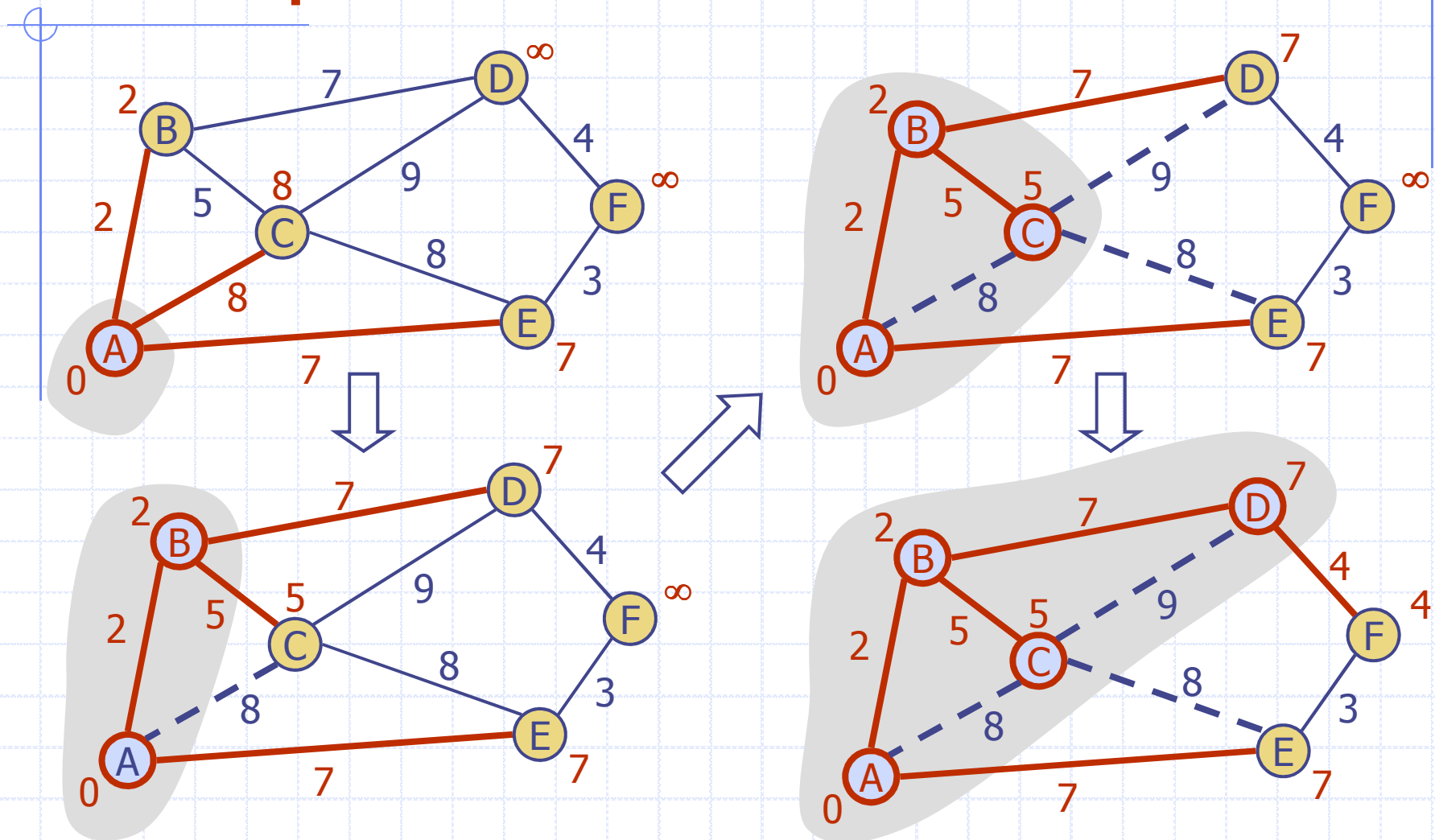
# Prim-Jarnik's Algorithm (cont.)

- 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 three labels with each vertex:
  - Distance
  - Parent edge in MST
  - Entry in priority queue

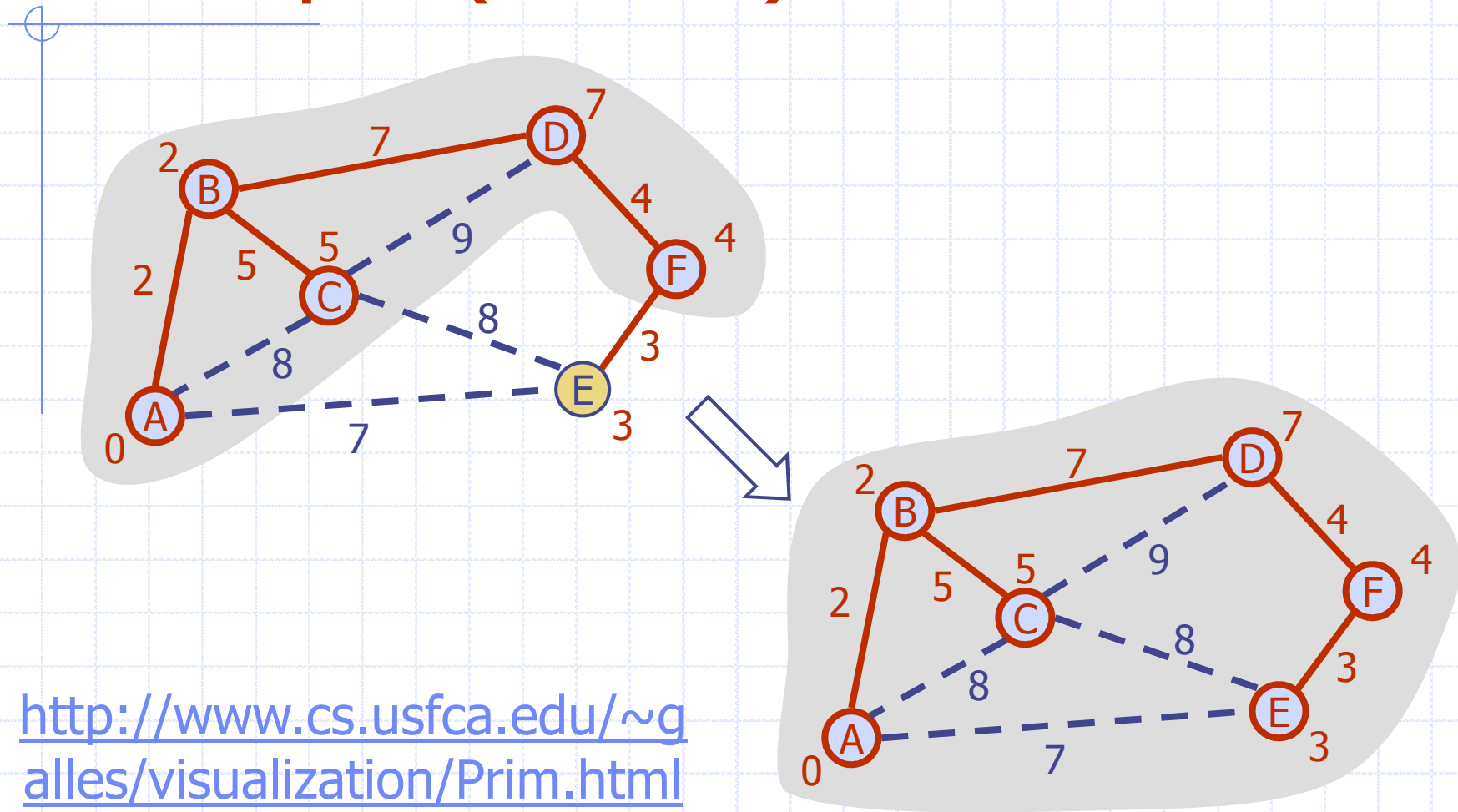
## Algorithm *PrimJarnikMST(G)*

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

# Example



# Example (contd.)



<http://www.cs.usfca.edu/~galles/visualization/Prim.html>



# Analysis

- Graph operations
  - Method `incidentEdges` is called once for each vertex
- Label operations
  - We set/get the distance, parent 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  $w$  in the priority queue is modified at most  $\deg(w)$  times, where each key change takes  $O(\log n)$  time
- Prim-Jarnik'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 is  $O(m \log n)$  since the graph is connected