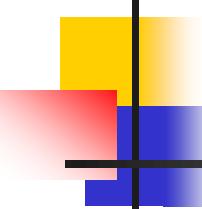


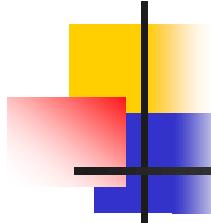
Prim's Algorithm

- Special case of the generic minimum-spanning-tree method.
- A greedy algorithm since at each step it adds to the tree an edge that contributes the minimum amount possible to the tree's weight.
- The edges in the set A always form a single tree.
- Each step adds to the tree A a light edge that connects A to an isolated vertex – one on which no edge of A is incident.
- By Corollary 23.2, this rule adds only edges that are safe for A



Implementation of Prim's algorithm

- Input is a connected Graph G and the root r of the minimum spanning tree.
- During execution of the algorithm, all vertices that are not in the tree reside in a min-priority queue Q based on a key attribute.
- For each v , $v.key$ is the minimum weight of any edge connecting v to a vertex in the tree.
- By convention, $v.key = \infty$ if there is no such edge.
- The attribute $v.\pi$ names the parent of v in the tree.
- The algorithm maintains the set A from Generic-MST as $A=\{(v, v.\pi):v\in V-\{r\}-Q\}$.
- It terminates when the min-priority queue Q is empty.

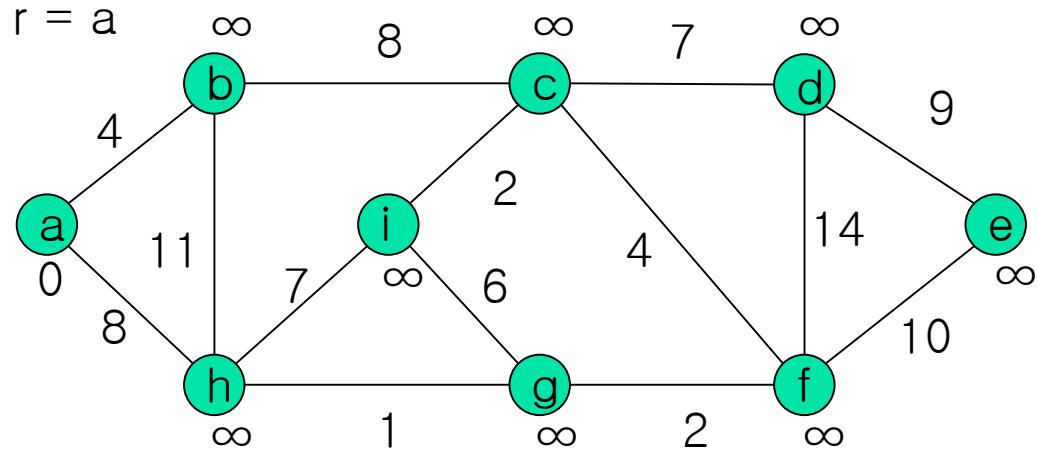


Prim's Algorithm

MST-PRIM(G, w, r)

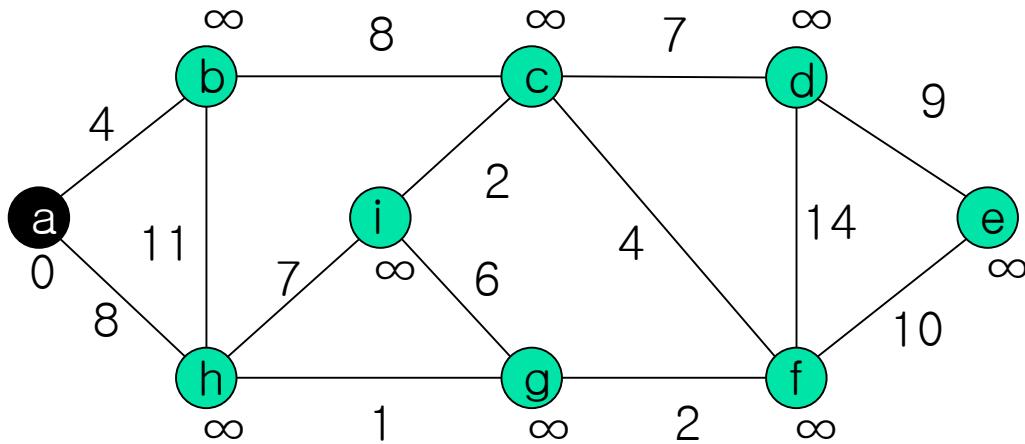
1. **for** each $u \in G.V$
2. $u.key = \infty$
3. $u.\pi = \text{NIL}$
4. $r.key = 0$
5. $Q = G.V$
6. **while** $Q \neq \emptyset$
7. $u = \text{Extract-Min}(Q)$
8. **for** each $v \in G.\text{Adj}[u]$
9. **if** $v \in Q$ and $w(u,v) < v.key$
10. $v.\pi = u$
11. $v.key = w(u,v)$

Prim's Algorithm



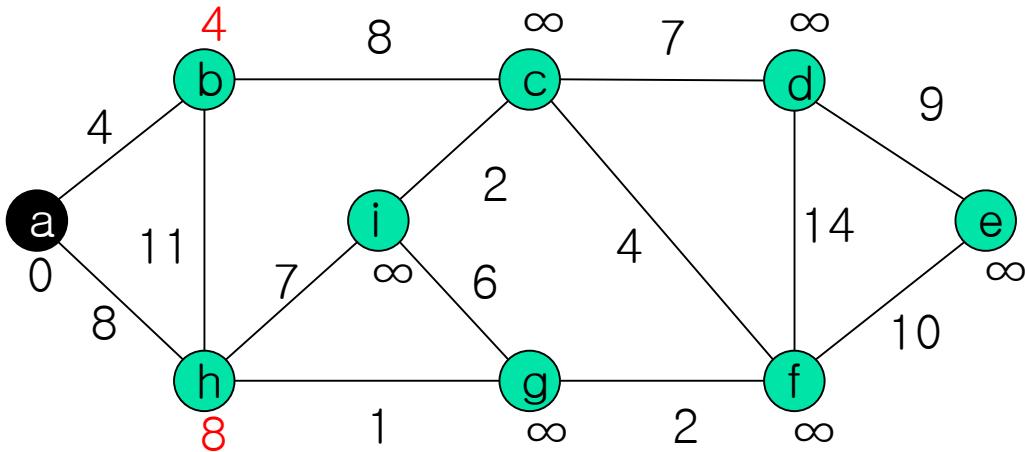
1. for each $u \in G.V$
2. $u.key = \infty$
3. $u.\pi = \text{NIL}$
4. $r.key = 0$
5. $Q = G.V$

Prim's Algorithm



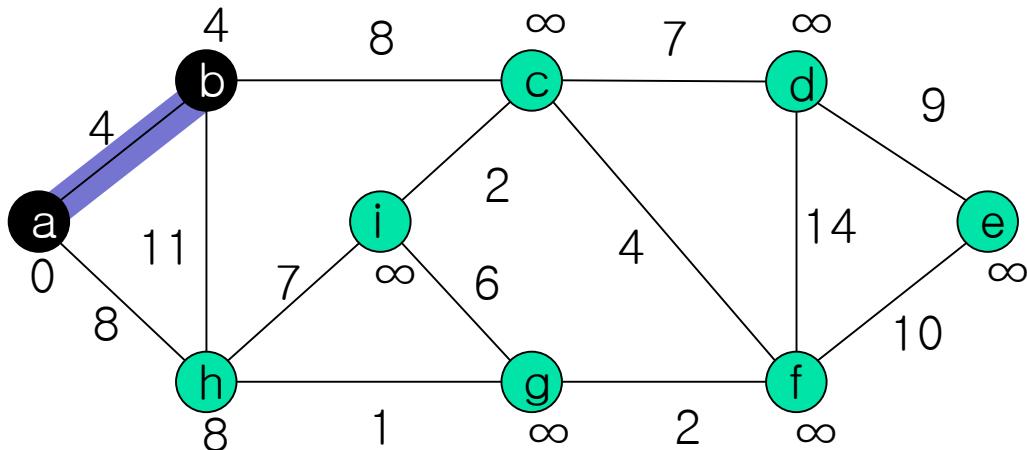
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



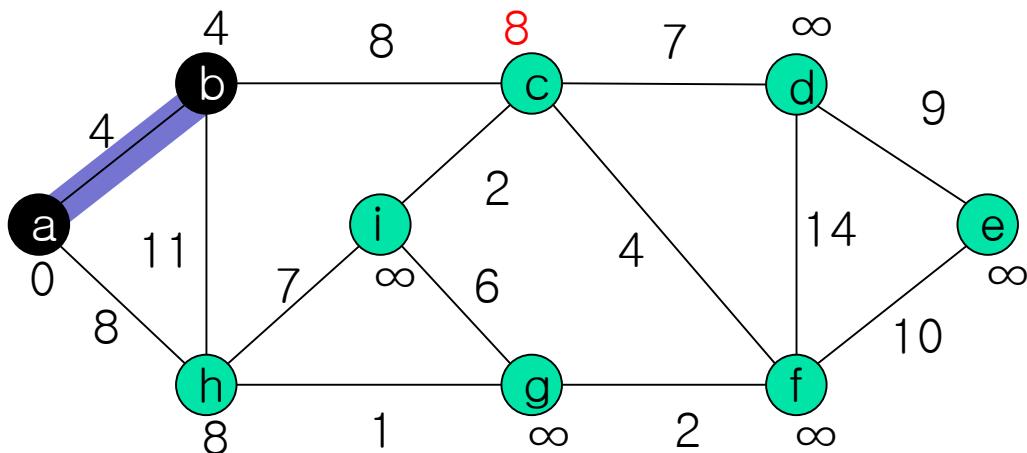
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



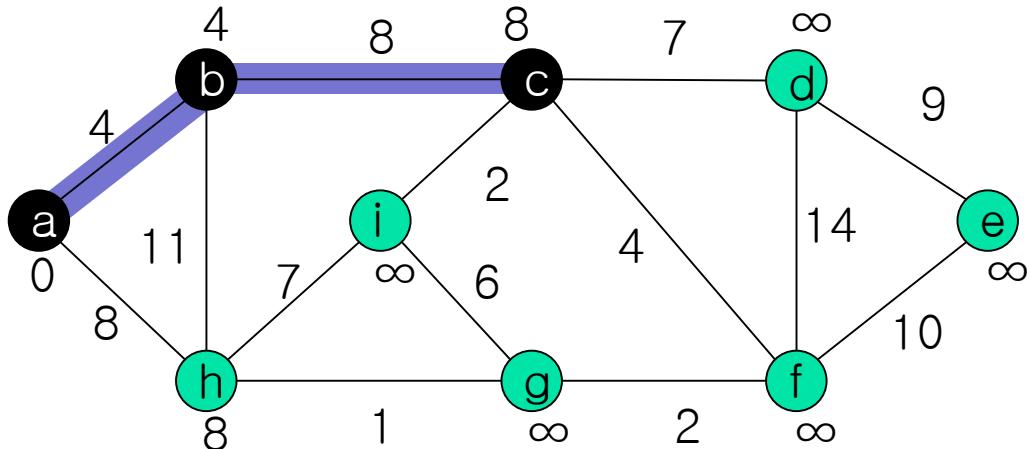
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



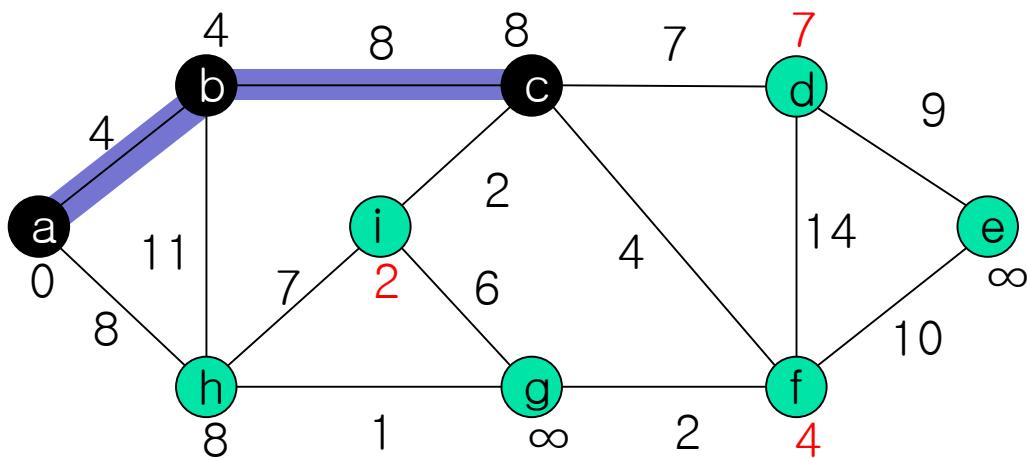
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7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
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```

Prim's Algorithm



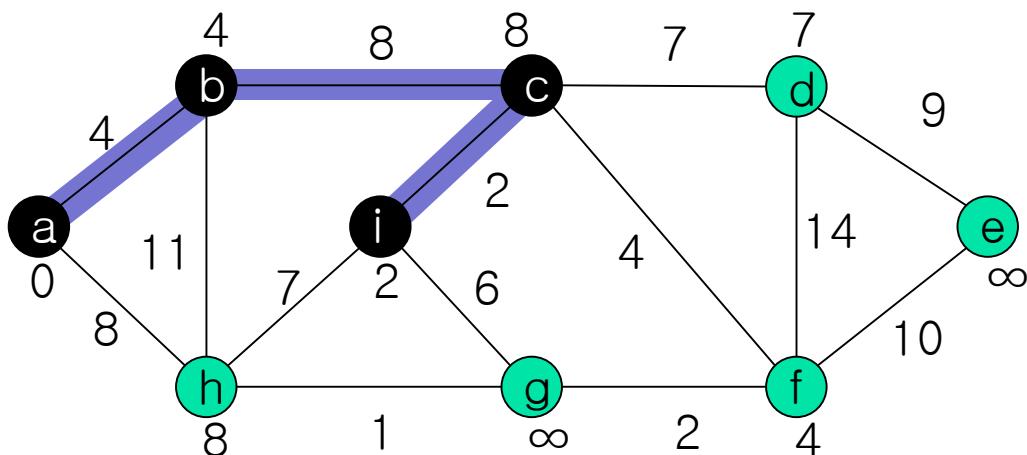
```
7.     u = Extract-Min(Q)
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```

Prim's Algorithm



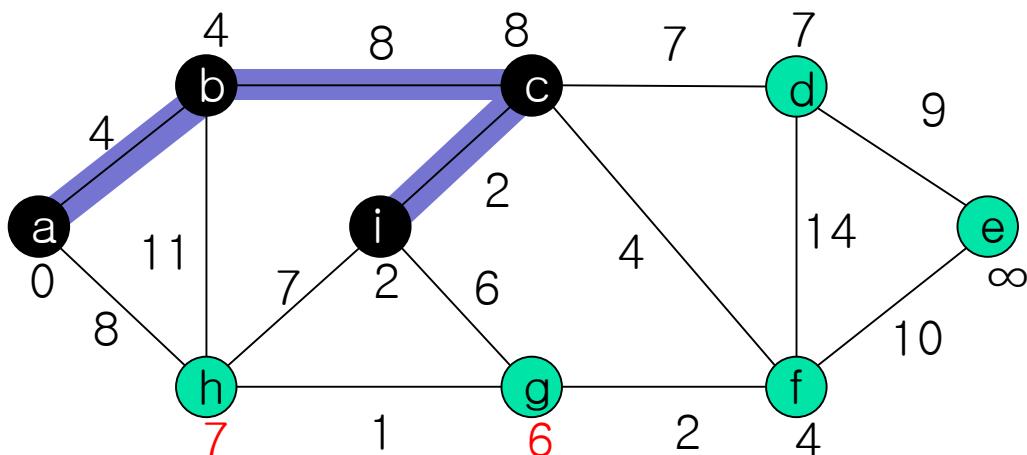
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7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



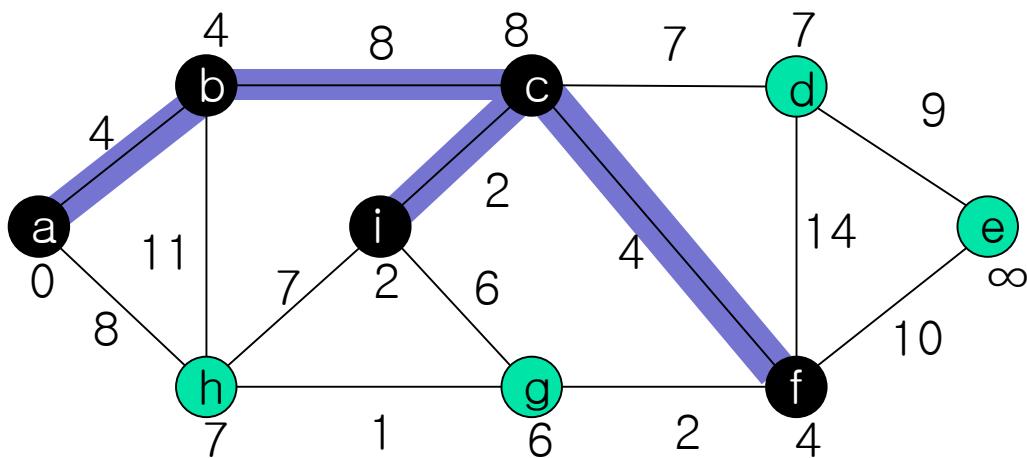
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



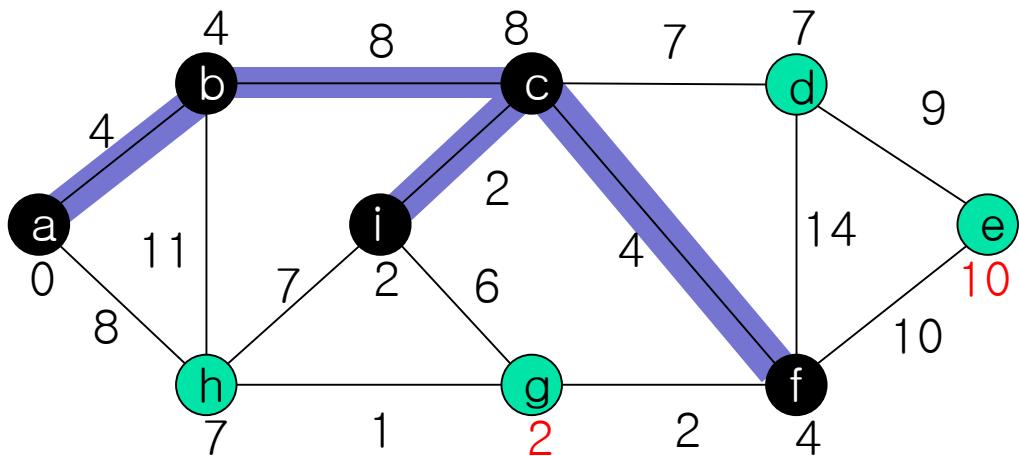
```
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8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



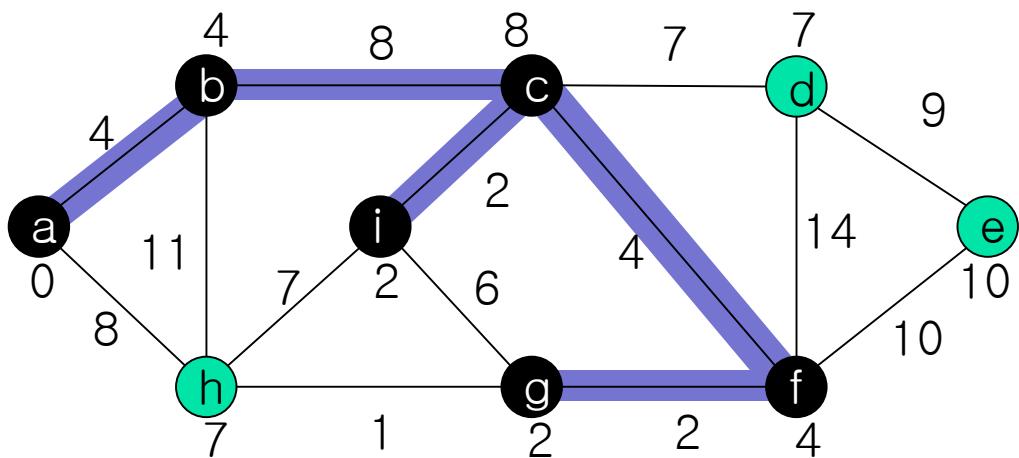
```
7.      u = Extract-Min(Q)
8.      for each v ∈ G.Adj[u]
9.          if v ∈ Q and w(u,v) < v.key
10.             v.π = u
11.             v.key = w(u,v)
```

Prim's Algorithm



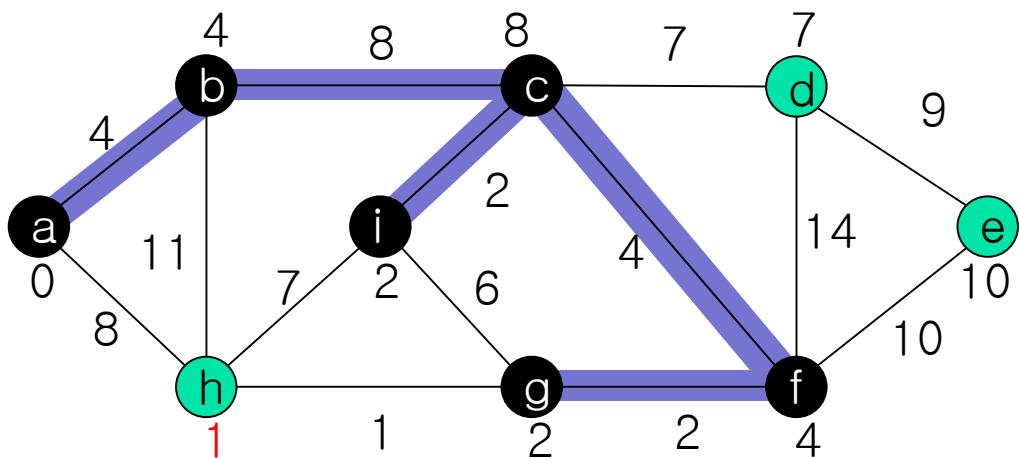
```
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8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



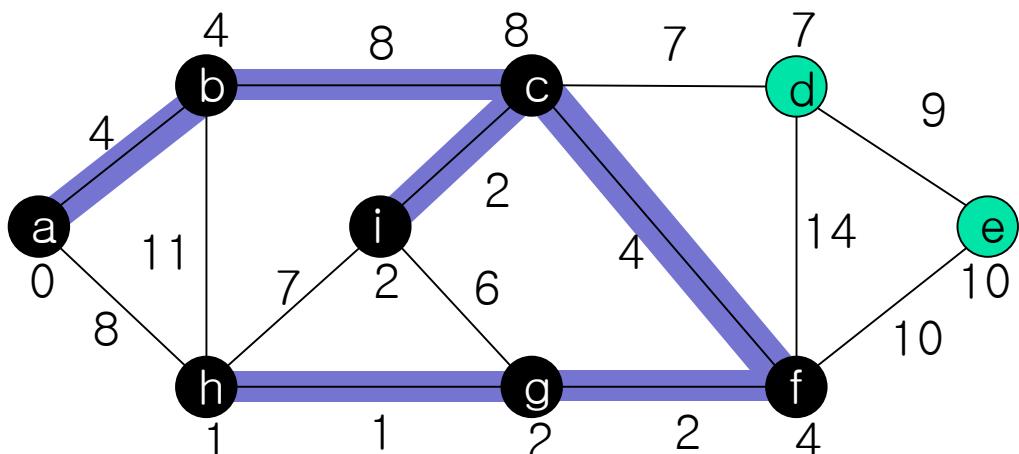
```
7.      u = Extract-Min(Q)
8.      for each v ∈ G.Adj[u]
9.          if v ∈ Q and w(u,v) < v.key
10.             v.π = u
11.             v.key = w(u,v)
```

Prim's Algorithm



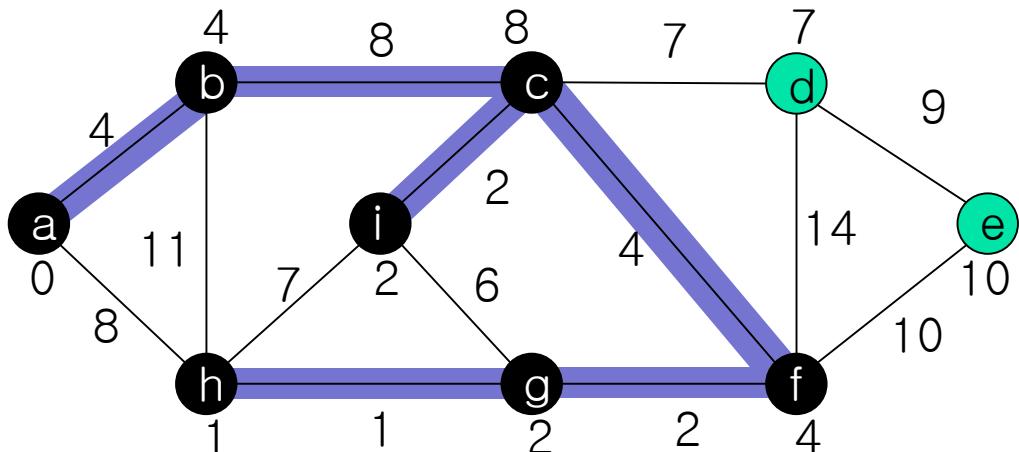
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



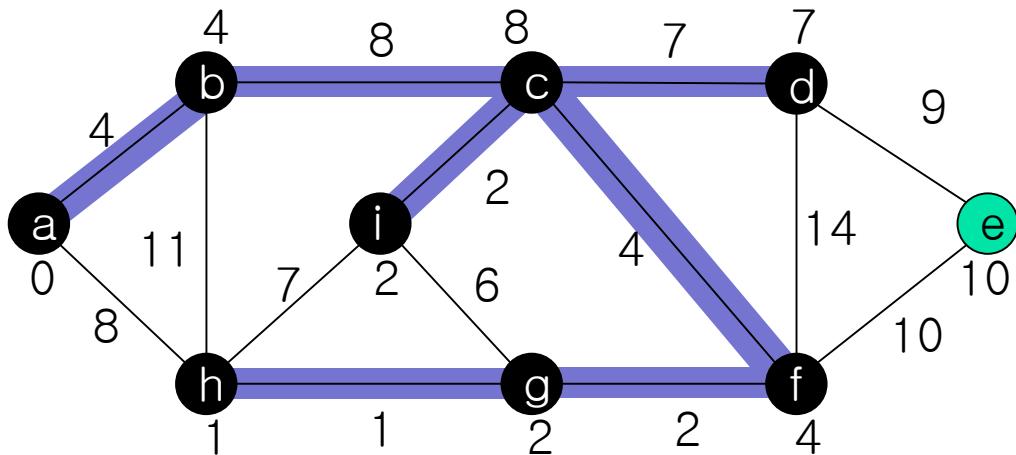
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
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```

Prim's Algorithm



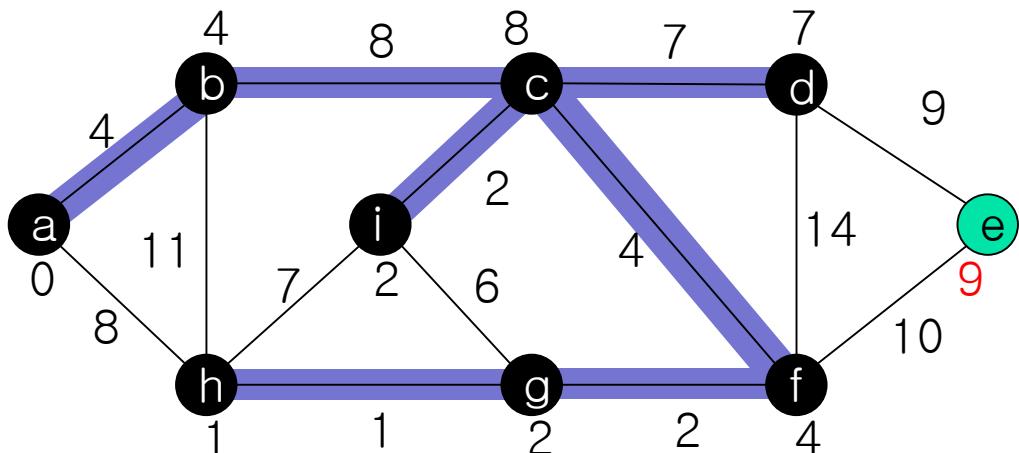
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10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm



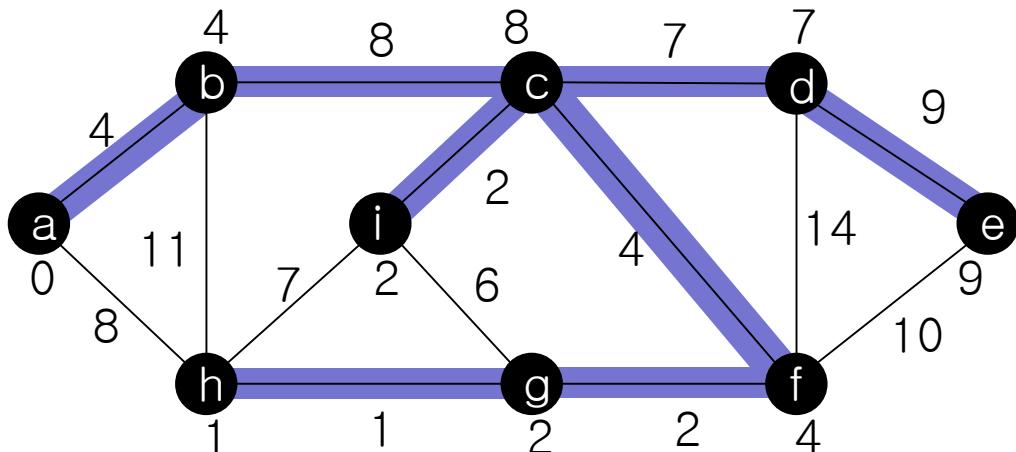
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```

Prim's Algorithm

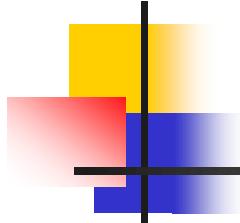


```
7.     u = Extract-Min(Q)
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10.            v.π = u
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```

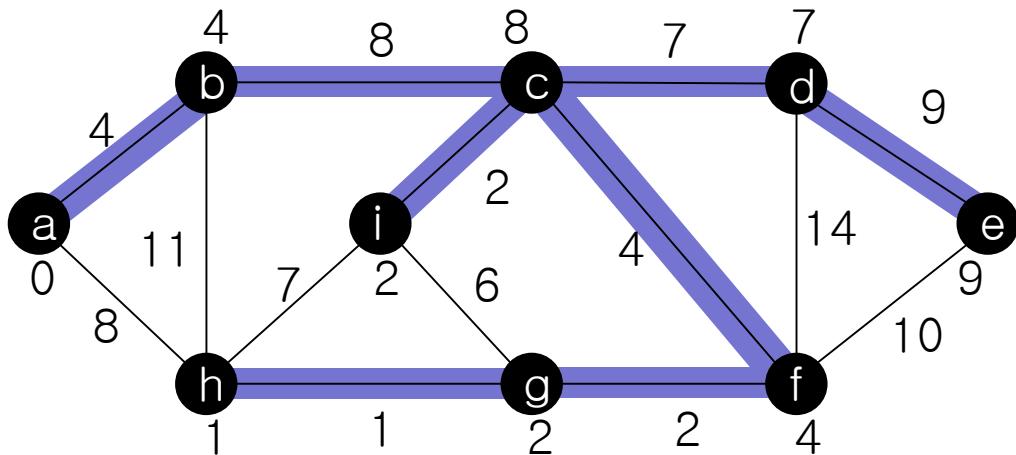
Prim's Algorithm



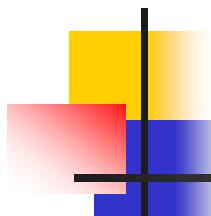
```
7.     u = Extract-Min(Q)
8.     for each v ∈ G.Adj[u]
9.         if v ∈ Q and w(u,v) < v.key
10.            v.π = u
11.            v.key = w(u,v)
```



Prim's Algorithm

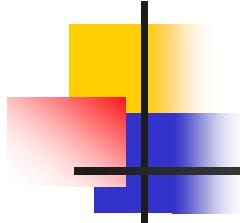


6. **while** $Q \neq \emptyset$
7. $u = \text{Extract-Min}(Q)$



Running Time of Prim's Algorithm

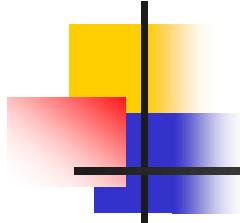
- The running time of Prim's algorithm depends on how we implement the min-priority queue Q.
- If we implement Q as a binary min-heap,
 - EXTRACT-MIN takes $O(\lg |V|)$ time.
 - DECREASE-KEY takes $O(\lg |V|)$ time.
- If we implement Q as a simple array,
 - EXTRACT-MIN takes $O(|V|)$ time.
 - DECREASE-KEY $O(1)$ time.
- If we implement Q as a Fibonacci heap,
 - EXTRACT-MIN takes $O(\lg |V|)$ amortized time.
 - DECREASE-KEY $O(1)$ amortized time.



Prim's Algorithm

MST-PRIM(G, w, r)

1. **for** each $u \in G.V$ $O(|V|)$
2. $u.key = \infty$
3. $u.\pi = \text{NIL}$
4. $r.key = 0$ Q: Implement as a binary min-heap
5. $Q = G.V$
6. **while** $Q \neq \emptyset$ $O(|V| \lg |V|)$
7. $u = \text{Extract-Min}(Q)$
8. **for** each $v \in G.\text{Adj}[u]$
9. **if** $v \in Q$ and $w(u,v) < v.key$
10. $v.\pi = u$
11. $v.key = w(u,v)$ $O(|E| \lg |V|)$
DECREASE-KEY $O(E)$
times



Prim's Algorithm

MST-PRIM(G, w, r)

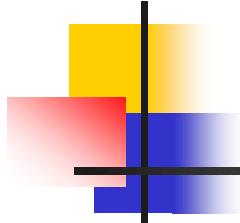
$O(|V|)$

1. **for** each $u \in G.V$
2. $u.key = \infty$
3. $u.\pi = \text{NIL}$
4. $r.key = 0$
5. $Q = G.V$
6. **while** $Q \neq \emptyset$
7. $u = \text{Extract-Min}(Q)$
8. **for** each $v \in G.\text{Adj}[u]$
9. **if** $v \in Q$ and $w(u,v) < v.key$
10. $v.\pi = u$
11. $v.key = w(u,v)$

Q: Implement as an array

$O(|V|^2)$

$O(|E|)$



Prim's Algorithm

MST-PRIM(G, w, r)

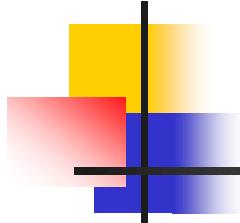
$O(|V|)$

1. **for** each $u \in G.V$
2. $u.key = \infty$
3. $u.\pi = \text{NIL}$
4. $r.key = 0$
5. $Q = G.V$
6. **while** $Q \neq \emptyset$
7. $u = \text{Extract-Min}(Q)$
8. **for** each $v \in G.\text{Adj}[u]$
9. **if** $v \in Q$ and $w(u,v) < v.key$
10. $v.\pi = u$
11. $v.key = w(u,v)$

Q: Implement as a Fibonacci mean-
heap

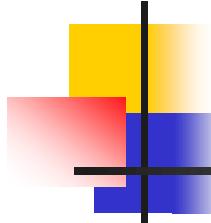
$O(|V| \lg |V|)$

$O(|E|)$



Minimum-Cost Spanning Trees

- Cost of a spanning tree
 - Sum of the costs (weights) of the edges in the spanning tree
- Min-cost spanning tree
 - A spanning tree of least cost
- Greedy method
 - At each stage, make the best decision possible at the time
 - Based on either a least cost or a highest profit criterion
 - Make sure the decision will result in a feasible solution
 - Satisfy the constraints of the problem
- To construct min-cost spanning trees
 - Best decision : least-cost
 - Constraints
 - Use only edges within the graph
 - Use exactly $n-1$ edges
 - May not use edges that produce a cycle

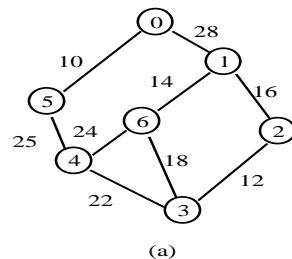


Kruskal's Algorithm

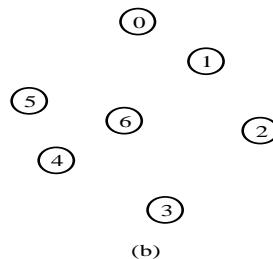
- Procedure

- Build a min-cost spanning tree T by adding edges to T one at a time
- Select edges for inclusion in T in nondecreasing order of their cost
- Edge is added to T if it does not form a cycle

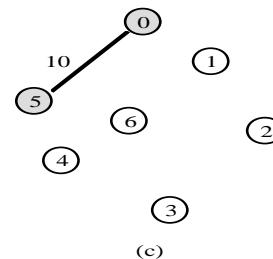
Kruskal's Algorithm (Cont.)



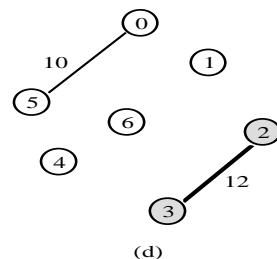
(a)



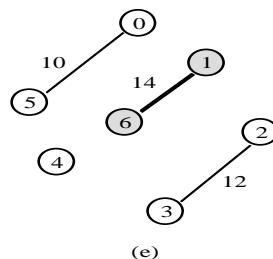
(b)



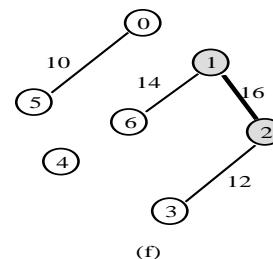
(c)



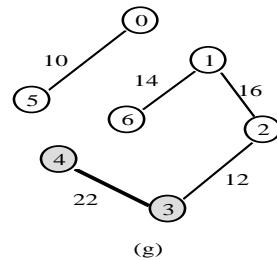
(d)



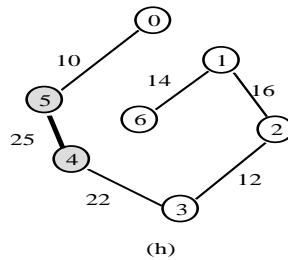
(e)



(f)

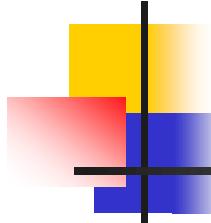


(g)



(h)

Figure 6.23 : Stages in Kruskal's algorithm

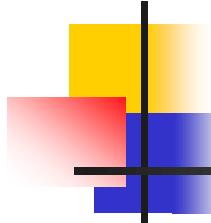


Kruskal's Algorithm (Cont.)

```
1. T = Φ;  
2. while( (T contains less than n-1 edges) && (E not empty) ) {  
3.     choose an edge (v,w) from E of the lowest cost;  
4.     delete (v,w) from E;  
5.     if( (v,w) does not create a cycle in T ) add (v,w) to T;  
6.     else discard (v,w);  
7. }  
8. if (T contains fewer than n-1 edge) cout << "no spanning tree" << endl;
```

Program 6.6: Kruskal's algorithm

- Time Complexity
 - When we use a min heap to determine the lowest cost edge , $O(e \log e)$



Prim's Algorithm

- Property
 - At all times during the algorithm the set of selected edges forms a tree
- Procedure
 - Begin with a tree T that contains a single vertex
 - Add a least-cost edge (u,v) to T such that $T \cup \{(u,v)\}$ is also a tree
 - Repeat until T contains $n-1$ edges

Prim's Algorithm (Cont.)

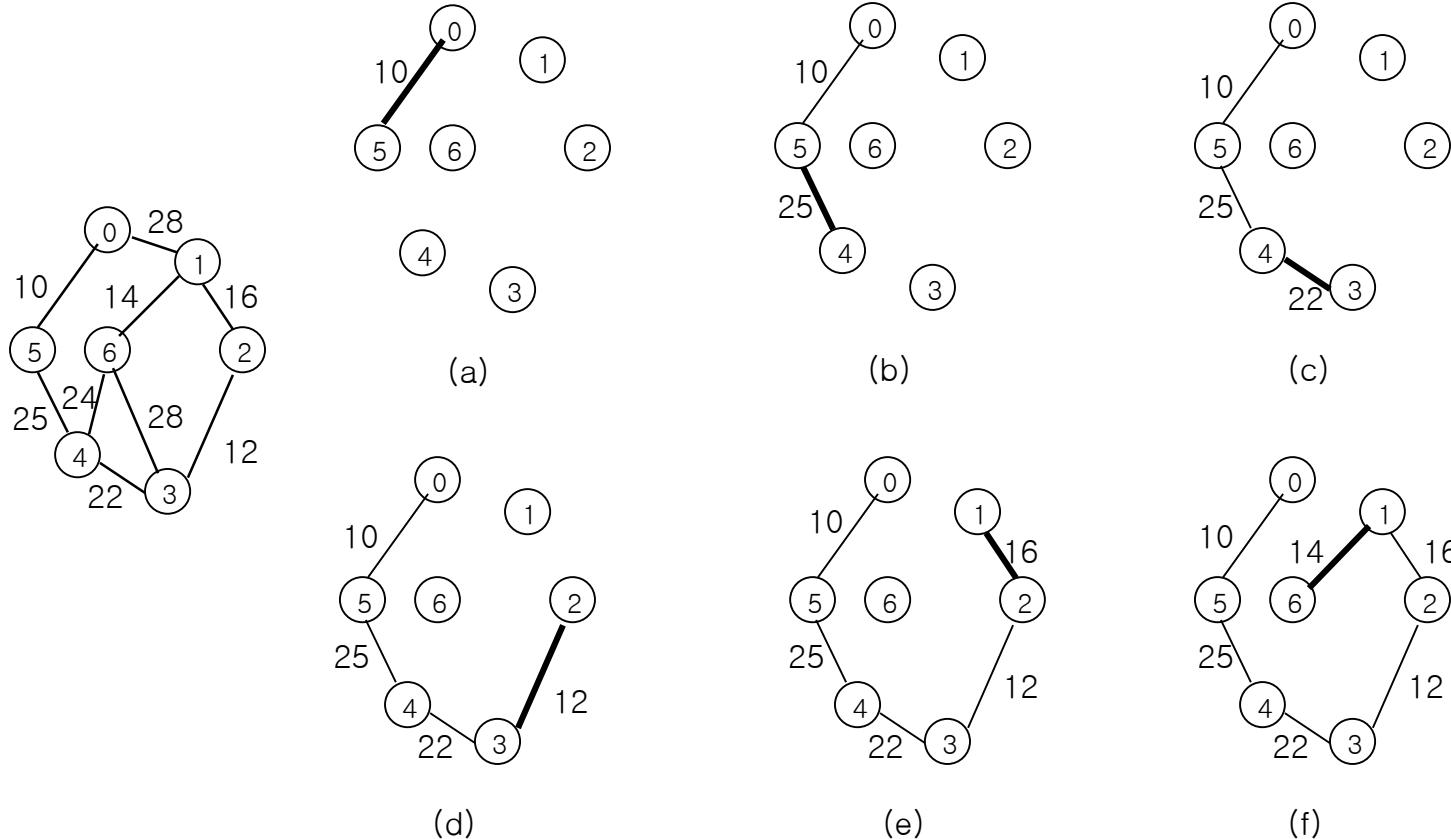
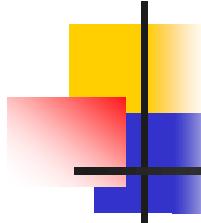


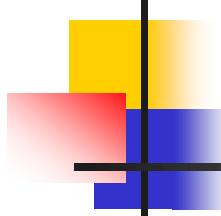
Figure 6.24: Stages in Prim's algorithm



Prim's Algorithm (Cont.)

```
1. // Assume that G has at least one vertex.  
2. TV = { 0 }; // start with vertex 0 and no edges  
3. for(T = Φ; T contains fewer than n-1 edges; add (u,v) to T)  
4. {  
5.     Let (u,v) be a least-cost edge such that u ∈ TV and v !∈ TV;  
6.     if(there is no such edge) break;  
7.     add v to TV;  
8. }  
9. if(T contains fewer than n-1 edges) cout << "no spanning tree" << endl;
```

Program 6.7: Prim's algorithm



Sollin's Algorithm

- Select several edges at each stage
- At the start of a stage, the selected edges, together with all n graph vertices, form a spanning forest at each stage
- During a stage, select one minimum cost edge for each tree in this forest
- The selected edges are added to the spanning tree being constructed

Sollin's Algorithm (Cont.)

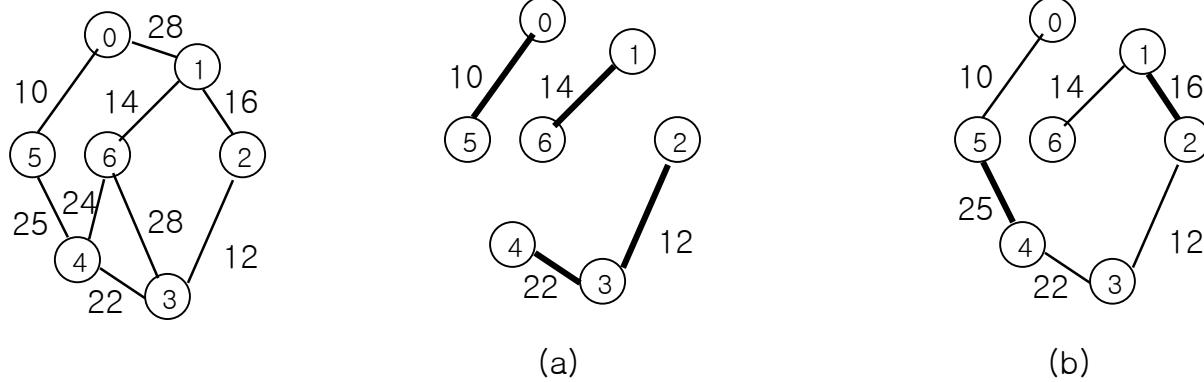
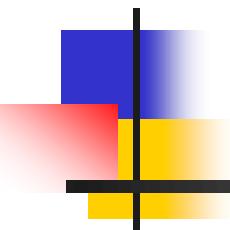


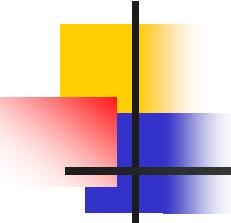
Figure 6.25: Stages in Sollin's algorithm



Single-source Shortest Paths

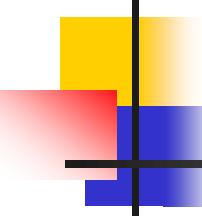
Single-source Shortest-Paths Problem

- Given a weighted directed graph $G=(V,E)$ with weight function $w:E \rightarrow \mathbb{R}$ mapping edges to real-valued weights, find the minimum-weight path from a given source vertex s to another vertex v .
 - The weight $w(p)$ of a path $p = \langle v_0, v_1, \dots, v_k \rangle$ is the sum of the weights of its constituent edges: $w(p) = \sum_{i=1}^k w(v_{i-1}, v_i)$
 - The shortest-path weight $\delta(s,v)$ from s to v by
$$\delta(s, v) = \begin{cases} \min\{w(p) : s \leadsto v\} & \text{if there is a path } p \text{ from } s \text{ to } v \\ \infty & \text{otherwise} \end{cases}$$



Optimal Substructure Property

- A shortest path between two vertices contains other shortest paths within it.
- Lemma 24.1 (Subpaths of shortest paths are shortest paths)
 - Consider a weighted graph G , with weight function $w:E \rightarrow R$,
 - Let $p = \langle v_0, v_1, v_2, v_3, \dots, v_{k-1}, v_k \rangle$ be a shortest path from vertex v_0 to vertex v_k .
 - For any i and j such that $1 \leq i \leq j \leq k$, let $p_{ij} = \langle v_i, v_{i+1}, \dots, v_j \rangle$ be the subpath of p from vertex v_i to vertex v_j .
 - Then, p_{ij} is a shortest path from v_i to v_j .
- Proof is straightforward.

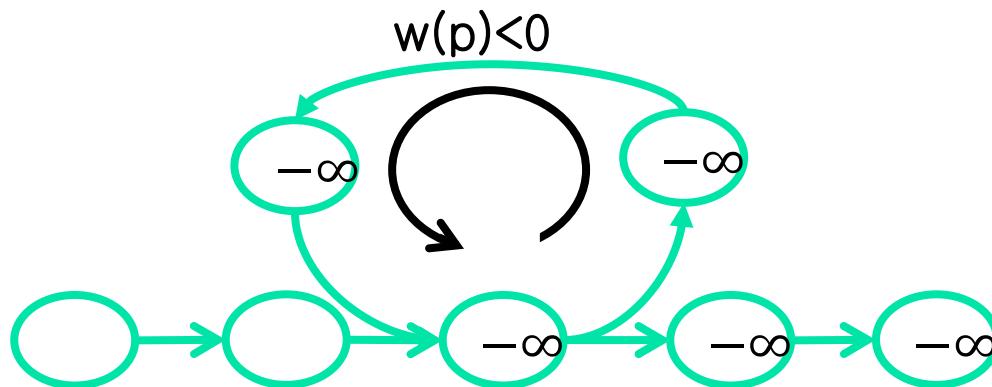


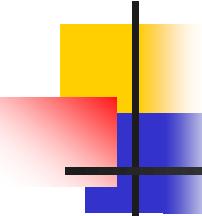
Shortest Path Properties

- Some instances of the single-source shortest-paths problem may include edges whose weights are negative.
- 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.

Shortest Path Properties

- 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 $\delta(s,v) = -\infty$.



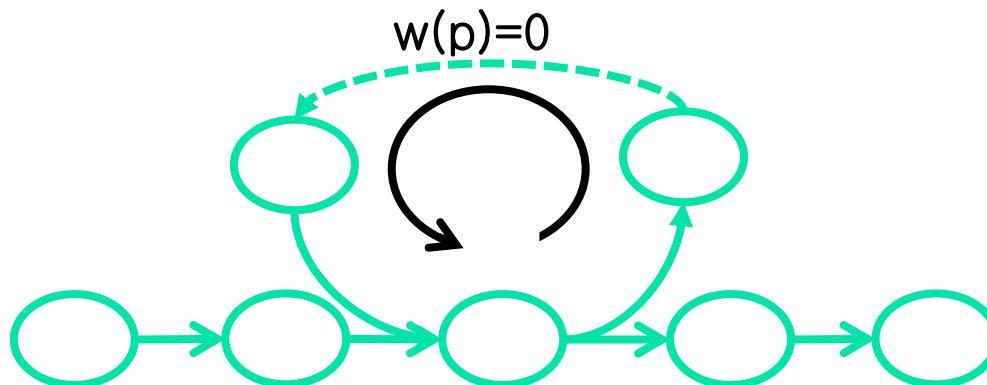


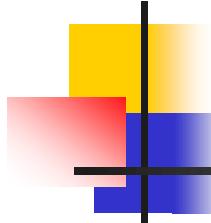
Can a Shortest Path Contain a Cycle?

- As we have just seen, it cannot contain a negative-weight cycle.
- Nor can it 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.
- We can remove a 0-weight cycle from any path to produce another path whose weight is the same.
- Thus, if there is a shortest path from a source vertex s to a destination vertex that contains a 0-weight cycle, then there is another shortest path from s to without this cycle.

Shortest Path Properties

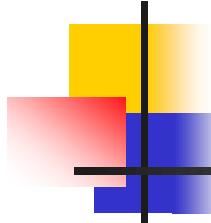
- Without loss of generality, we can assume that when we are finding shortest paths, they have no cycles.
- Since any acyclic path in a graph $G=(V,E)$, contains at most $|V|$ distinct vertices, it also contains at most $|V|-1$ edges.
- If have only 0-weight cycles, we can restrict our attention to shortest paths of at most $|V|-1$ edges.





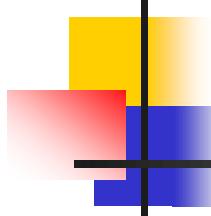
Representing Shortest Paths

- We represent shortest paths similarly to how we represented breadth-first trees.
- Shortest-paths tree:
 - Predecessor subgraph $G_\pi = (V_\pi, E_\pi)$
 - $V_\pi = \{v \in V : v.\pi \neq \text{NIL}\} \cup \{s\}$
 - $E_\pi = \{(v.\pi, v) \in E : v \in V_\pi - \{s\}\}$
 - V_π is the set of vertices of G with non-NIL predecessors, plus the source s .
 - E_π is the set of edges induced by the π values for vertices in V_π .
 - G_π forms a rooted tree with root s containing a shortest path from the source s to every vertex that is reachable from s .



A Shortest-path Tree

- Let $G=(V, E)$ be a weighted, directed graph with weight function $w:E \rightarrow \mathbb{R}$.
- Assume that G contains no negative-weight cycles reachable from the source vertex $s \in V$, so that shortest paths are well defined.
- A shortest-path tree rooted at s is a directed subgraph $G'=(V', E')$, where $V' \subset V$ and $E' \subset E$, such that
 - V' is the set of vertices reachable from s in G .
 - G' forms a rooted tree with root s .
 - For all $v \in V'$, the unique simple path from s to v in G' is a shortest path from s to v in G .

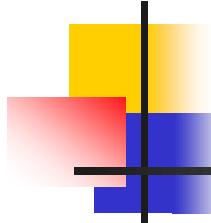


Relaxation

- The algorithms in this chapter use the technique of **relaxation**.
- For all $v \in V$, we maintain an attribute $v.d$ which is an upper bound on the weight of a shortest path from source s to v .
- We call $v.d$ a **shortest-path estimate**.

INITIALIZE-SINGLE-SOURCE(G, s)

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

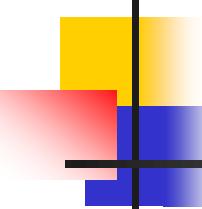


Relaxation

- The process of **relaxing** an edge (u,v) consists of testing whether we can improve the shortest path to v found so far by going through u and, if so, updating $v.d$ and $v.\pi$.
- The following code performs a relaxation step on edge (u,v) in $O(1)$ time.

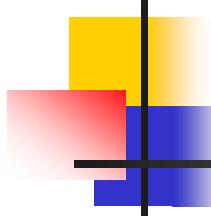
RELAX(u, v, w)

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



Relaxation

- Each algorithm in this chapter calls INITIALIZE-SINGLE-SOURCE and then repeatedly relaxes edges.
- Moreover, relaxation is the only means by which shortest path estimates and predecessors change.
- The algorithms in this chapter differ in how many times they relax each edge and the order in which they relax edges.
 - The Bellman-Ford algorithm relaxes each edge $|V|-1$ times.
 - Dijkstra's algorithm for directed acyclic graphs relaxes each edge exactly once.



Initialize

INITIALIZE-SINGLE-SOURCE(G, s)

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

Relaxation

RELAX(u, v, w)

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

