Elementary Graph Algorithms

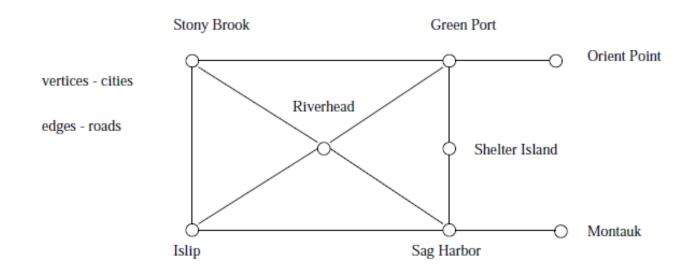
### Graphs

- Graphs are one of the unifying themes of computer science. A graph G = (V, E) is defined by a set of vertices V, and a set of edges E consisting of ordered or unordered pairs of vertices from V.
- Thus a graph G = (V, E)
  - V = set of vertices
  - $\blacksquare$  E = set of edges = subset of V × V
  - Thus  $|E| = O(|V|^2)$

## Examples

#### Road Networks

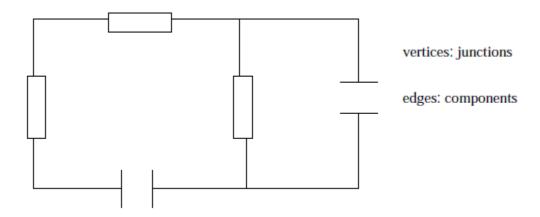
■ In modeling a road network, the vertices may represent the cities or junctions, certain pairs of which are connected by roads/edges.



## Examples

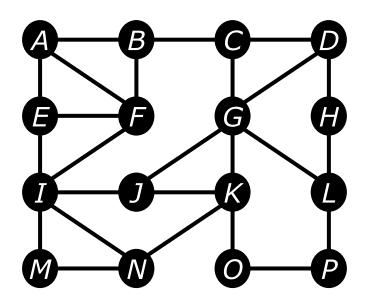
#### Electronic Circuits

■ In an electronic circuit, with junctions as vertices and components as edges.



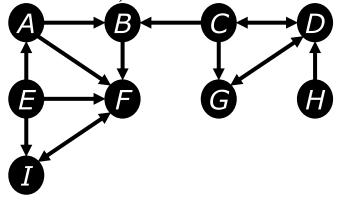
#### **Graph Variations**

- Variations:
  - A connected graph has a path from every vertex to every other
  - In an *undirected graph:* 
    - $\circ$  Edge (u,v) = Edge (v,u)
    - No self-loops



## **Graph Variations**

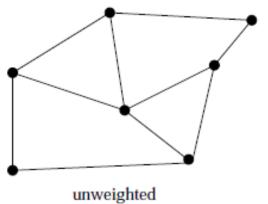
- In a *directed* graph:
  - $\circ$  Edge (u,v) goes from vertex u to vertex v, notated u $\rightarrow$ v
  - o (u,v) is different from (v,u)

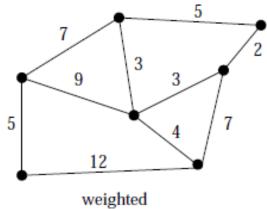


- Road networks between cities are typically undirected.
- Street networks *within* cities are almost always directed because of one-way streets.
- Most graphs of graph-theoretic interest are undirected.

### **Graph Variations**

- More variations:
  - A weighted graph associates weights with either the edges or the vertices
    - E.g., a road map: edges might be weighted with their length, drive-time or speed limit.
  - In *unweighted* graphs, there is no cost distinction between various edges and vertices.

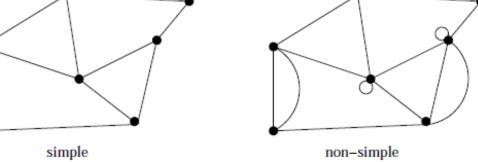




- A *multigraph* allows multiple edges between the same vertices
  - E.g., the call graph in a program (a function can get called from multiple points in another function)

# Simple vs. Non-simple Graphs

- Certain types of edges complicate the task of working with graphs. A *self-loop* is an edge (*x*; *x*) involving only one vertex.
- An edge (x; y) is a *multi-edge* if it occurs more than once in the graph.
- Any graph which avoids these structures is called *simple*.



# Spars Vs Dense Graphs

- Graphs are *sparse* when only a small fraction of the possible number of vertex pairs actually have edges defined between them.
- Graphs are usually sparse due to applicationspecific constraints.
- Road networks must be sparse because of road junctions.

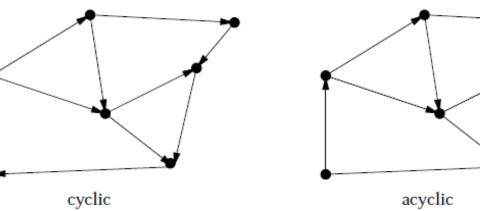
dense

# Spars Vs Dense Graphs

• Typically dense graphs have a quadratic number of edges while sparse graphs are linear in size.

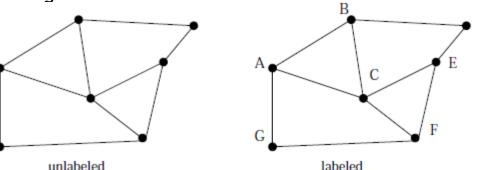
# Cyclic vs. Acyclic Graphs

- An *acyclic* graph does not contain any cycles. *Trees* are connected acyclic *undirected* graphs.
- Directed acyclic graphs are called *DAGs*. They arise naturally in scheduling problems, where a directed edge (x; y) indicates that x must occur before y



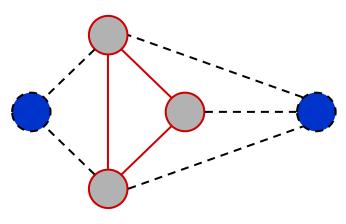
## Labeled vs. Unlabeled Graphs

- In *labeled graphs*, each vertex is assigned a unique name or identifier to distinguish it from all other vertices.
- An important graph problem is *isomorphism testing*, determining whether the topological structure of two graphs are in fact identical if we ignore any labels.

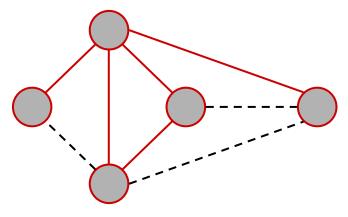


# Subgraphs

- A subgraph S of a graph G
   is a graph such that
  - The edges of S are a subset of the edges of G
- A spanning subgraph of G
  is a subgraph that contains
  all the vertices of G



Subgraph



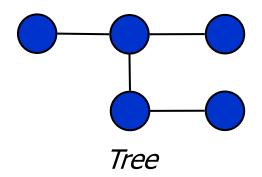
Spanning subgraph

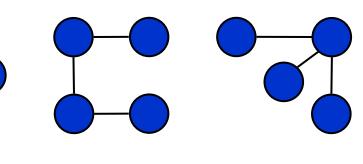
#### **Trees and Forests**

- A tree is an undirected graph T such that
  - T is connected
  - T has no cycles

This definition of tree is different from the one of a rooted tree

- A forest is an undirected graph without cycles
- The connected components of a forest are trees





**Forest** 

## Graph in Applications

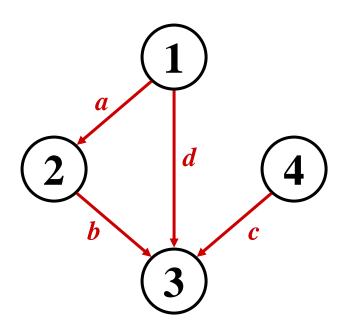
- Internet: web graphs
  - Each page is a vertex
  - Each edge represent a hyperlink
  - Directed
- GPS: highway maps
  - Each city is a vertex
  - Each freeway segment is an undirected edge
  - Undirected
- Graphs are ubiquitous in computer science

### Representing Graphs

- Assume  $V = \{1, 2, ..., n\}$
- An *adjacency matrix* represents the graph as a  $n \times n$  matrix A:
  - A[i, j] = 1 if edge  $(i, j) \in E$  (or weight of edge) = 0 if edge  $(i, j) \notin E$

# Graphs: Adjacency Matrix

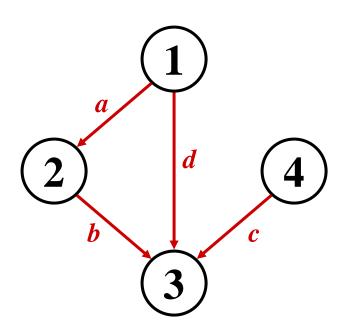
#### • Example:



A	1	2	3	4
1				
2		??		
3				
4				

# Graphs: Adjacency Matrix

#### • Example:



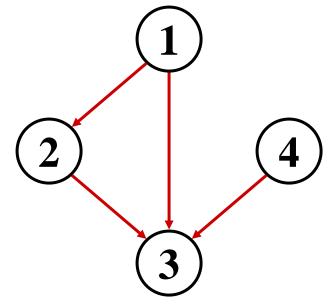
A	1	2	3	4
1	0	1	1	0
2	0	0	1	0
3 1	0	0	0	0
4	0	0	1	0

# Graphs: Adjacency Matrix

- The adjacency matrix is a dense representation
  - Usually too much storage for large graphs
  - But can be very efficient for small graphs
- Most large interesting graphs are sparse
  - E.g., planar graphs, in which no edges cross
  - For this reason the *adjacency list* is often a more appropriate respresentation

# Graphs: Adjacency List

- Adjacency list: for each vertex  $v \in V$ , store a list of vertices adjacent to v
- Example:
  - $\blacksquare$  Adj[1] = {2,3}
  - $Adj[2] = {3}$
  - $Adj[3] = \{\}$
  - $\blacksquare$  Adj[4] = {3}
- Variation: can also keep a list of edges coming *into* vertex



### Graphs: Adjacency List

- How much storage is required?
  - The *degree* of a vertex *v* in an undirected graph = # incident edges
    - o Directed graphs have in-degree, out-degree
  - For directed graphs, # of items in adjacency lists is  $\Sigma$  out-degree(v) = |E| takes  $\Theta(V + E)$  storage
  - For undirected graphs, # items in adj lists is  $\Sigma$  degree(v) = 2 |E| also  $\Theta(V + E)$  storage
- So: Adjacency lists take O(V+E) storage

### **Graph Representation**

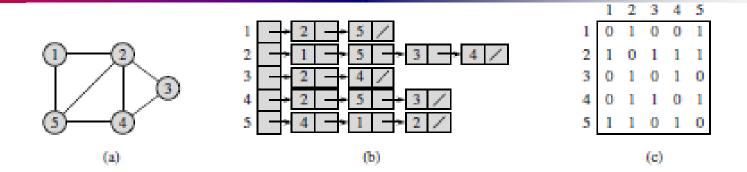


Figure 22.1 Two representations of an undirected graph. (a) An undirected graph G having five vertices and seven edges. (b) An adjacency-list representation of G. (c) The adjacency-matrix representation of G.

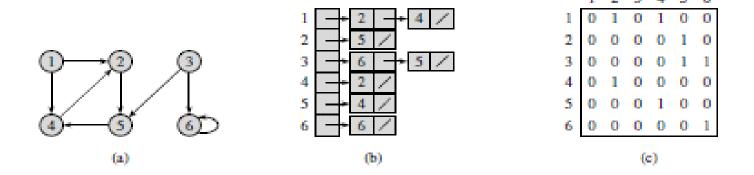


Figure 22.2 Two representations of a directed graph. (a) A directed graph G having six vertices and eight edges. (b) An adjacency-list representation of G. (c) The adjacency-matrix representation of G.

# Google Problem: Graph Searching

- How to search and explore in the Web Space?
- How to find all web-pages and all the hyperlinks?
  - Start from one vertex "www.google.com"
  - Systematically follow hyperlinks from the discovered vertex/web page
  - Build a search tree along the exploration

# General Graph Searching

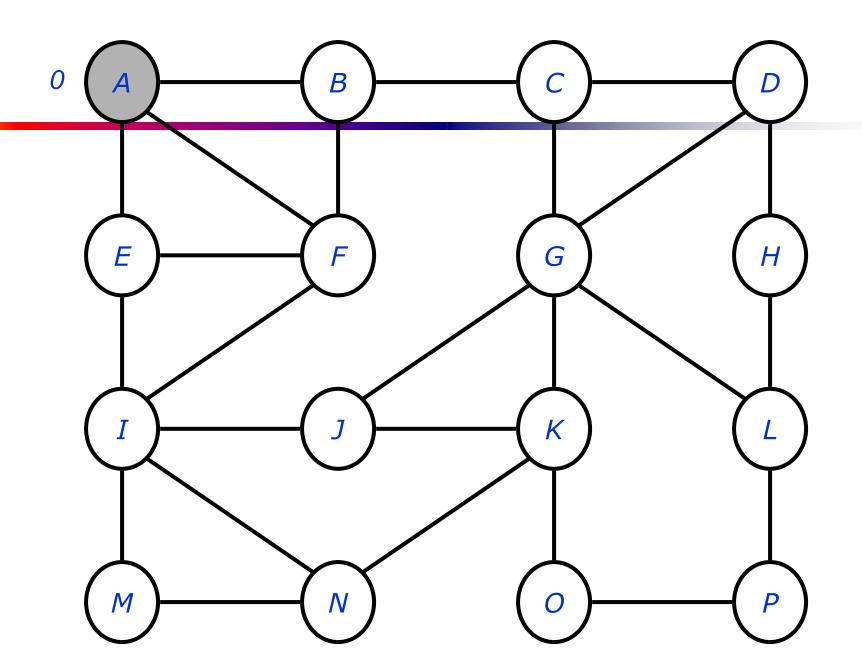
- Given: a graph G = (V, E), directed or undirected
- Goal: methodically explore every vertex and every edge
- Ultimately: build a tree on the graph
  - Pick a vertex as the root
  - Choose certain edges to produce a tree
  - Note: might also build a *forest* if graph is not connected

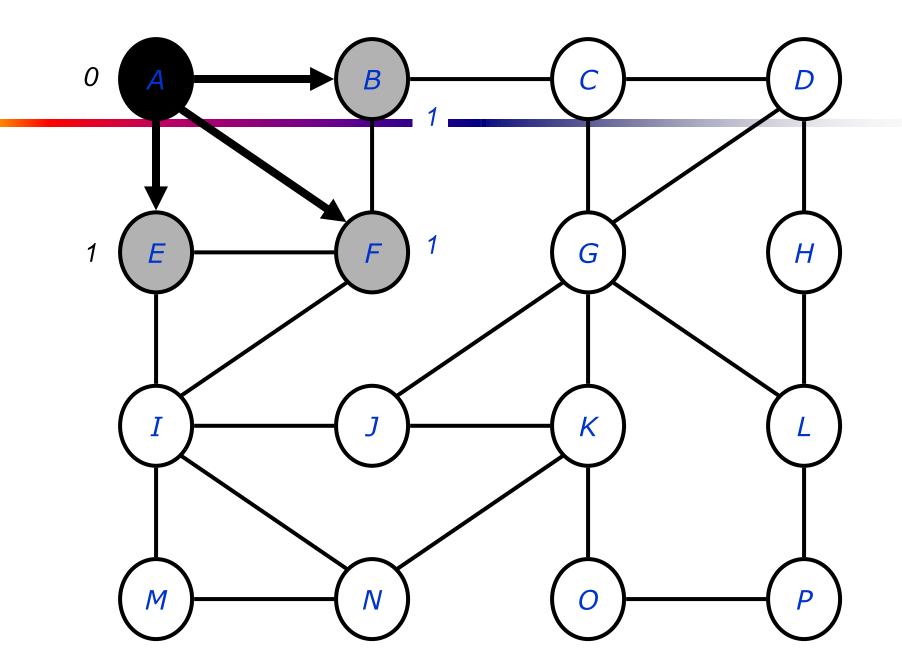
#### **Breadth-First Search**

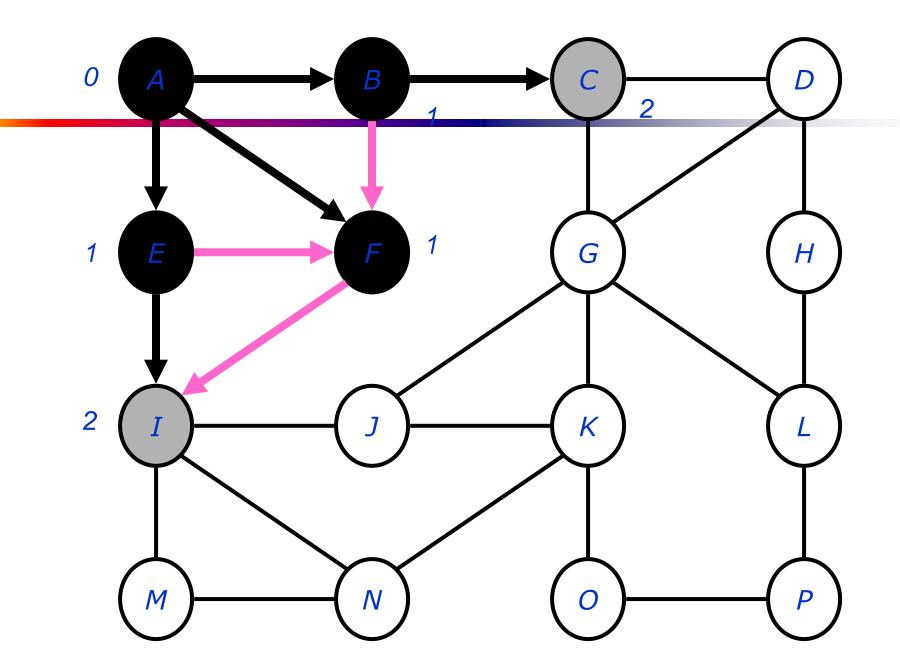
- "Explore" a graph, turning it into a tree
  - One vertex at a time
  - Expand frontier of explored vertices across the breadth of the frontier
- Builds a tree over the graph
  - Pick a *source vertex* to be the root
  - Find ("discover") its children, then their children, etc.

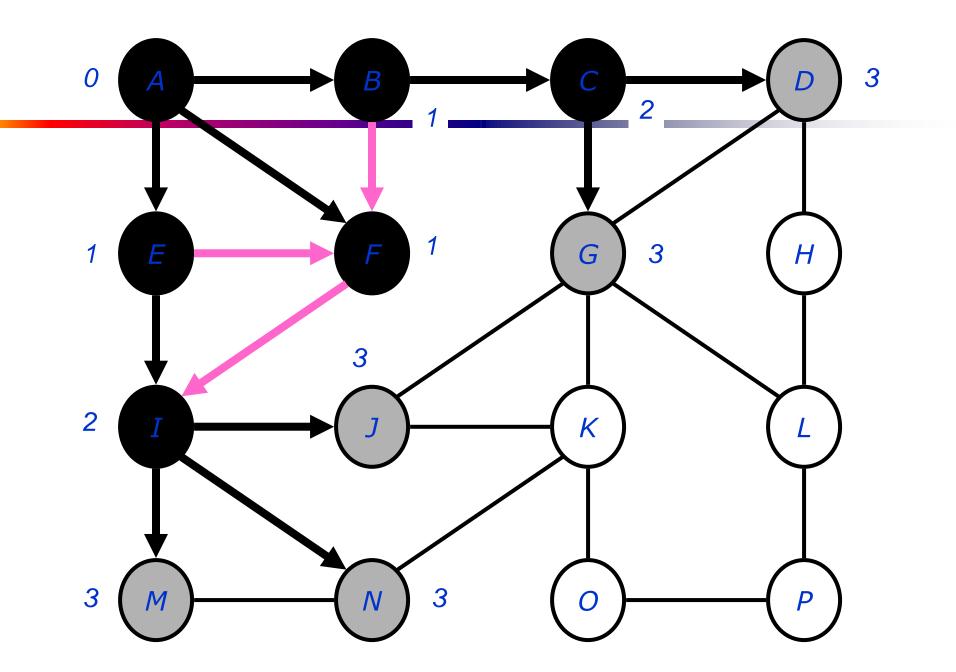
#### **Breadth-First Search**

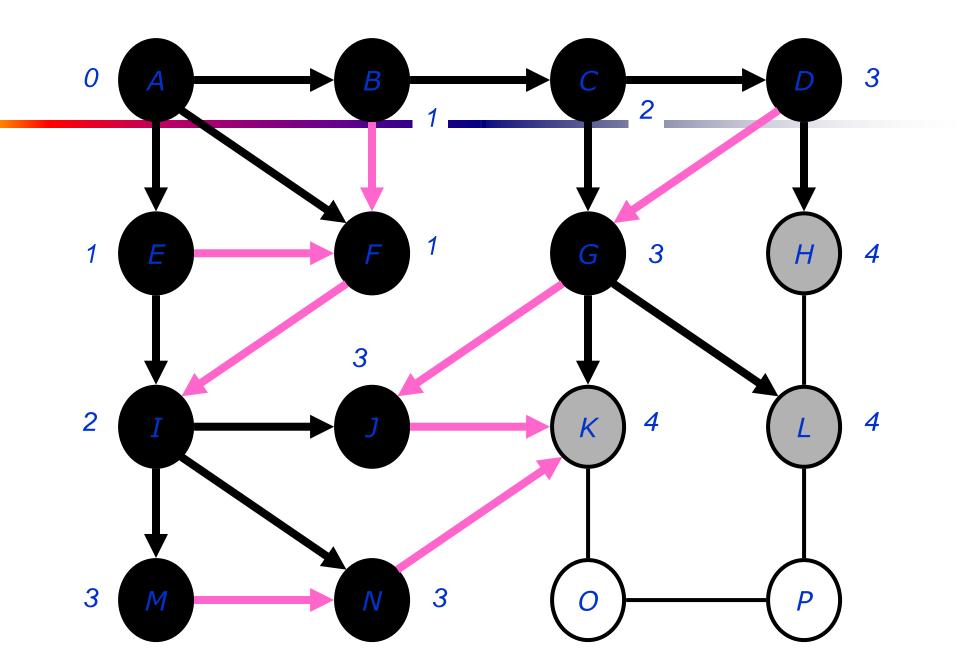
- Again will associate vertex "colors" to guide the algorithm
  - White vertices have not been discovered
    - All vertices start out white
  - Grey vertices are discovered but not fully explored
    - They may be adjacent to white vertices
  - Black vertices are discovered and fully explored
    - They are adjacent only to black and gray vertices
- Explore vertices by scanning adjacency list of grey vertices

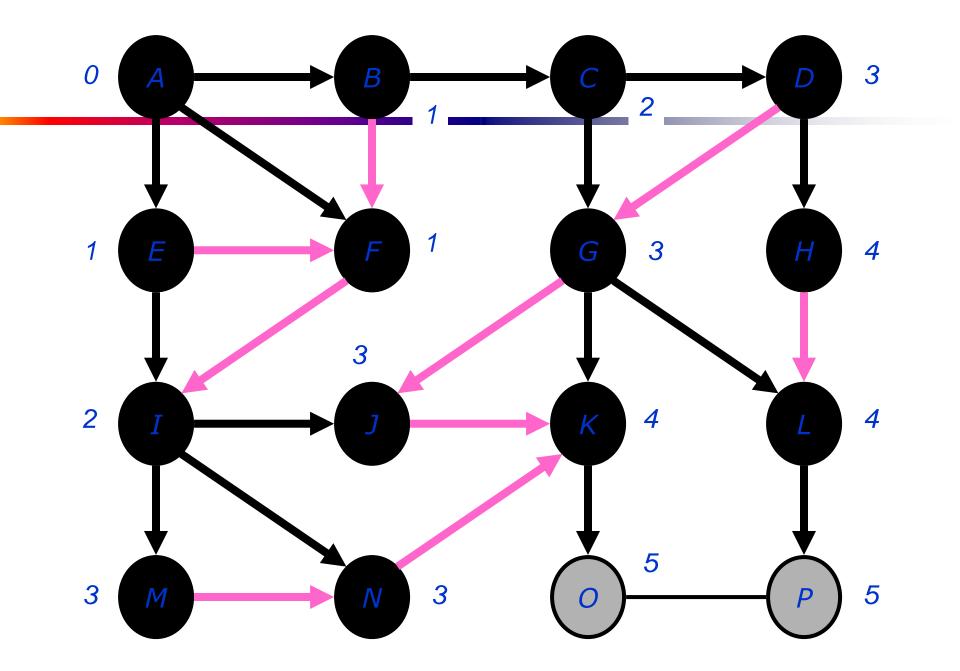


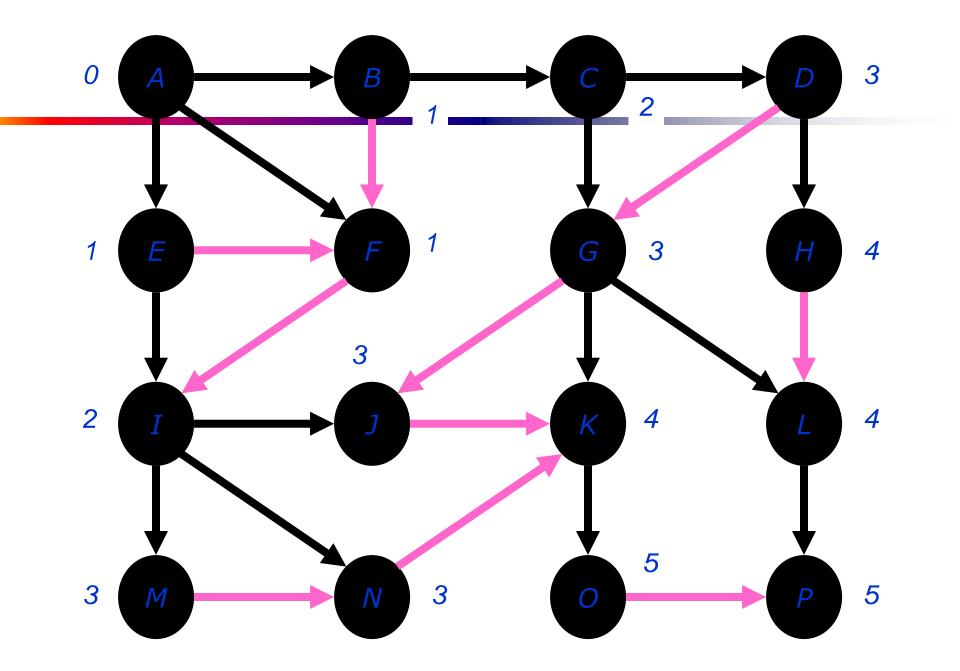


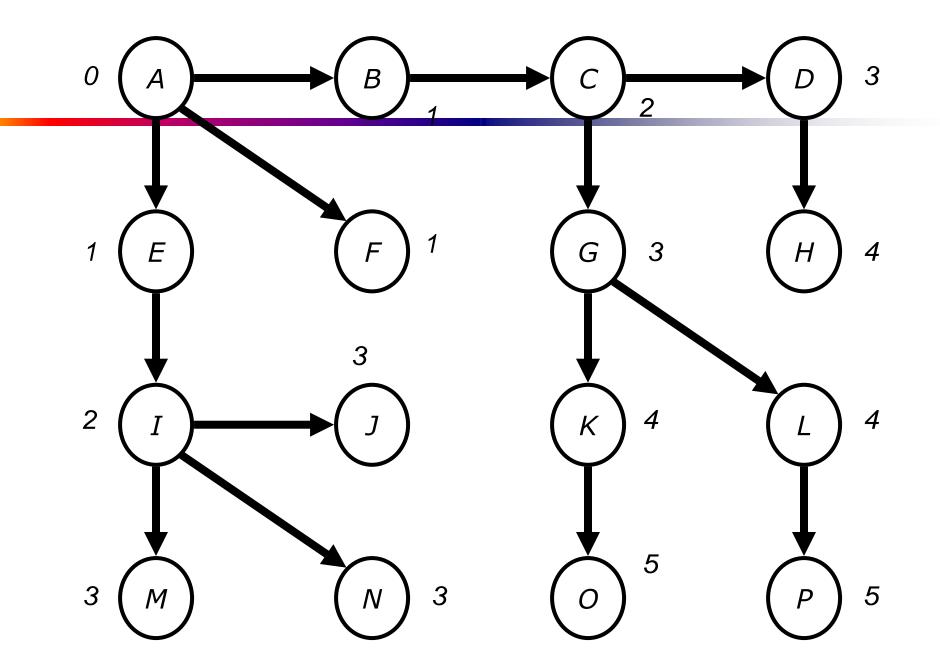






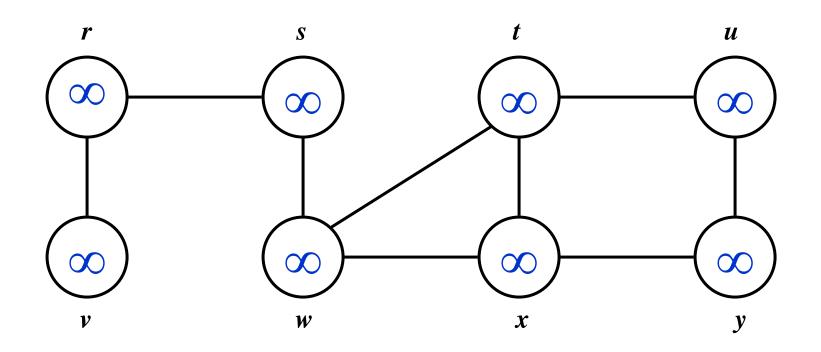


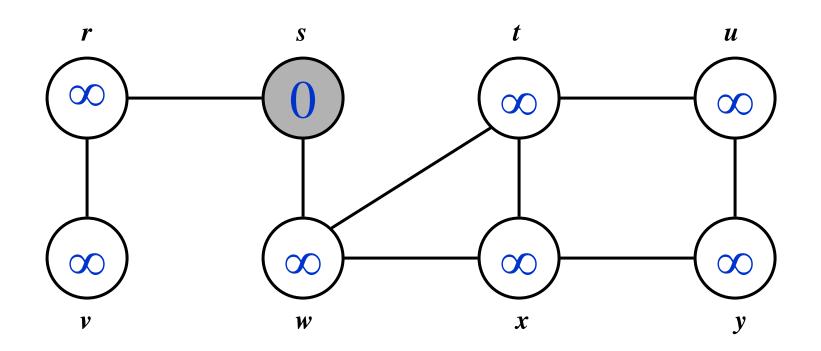




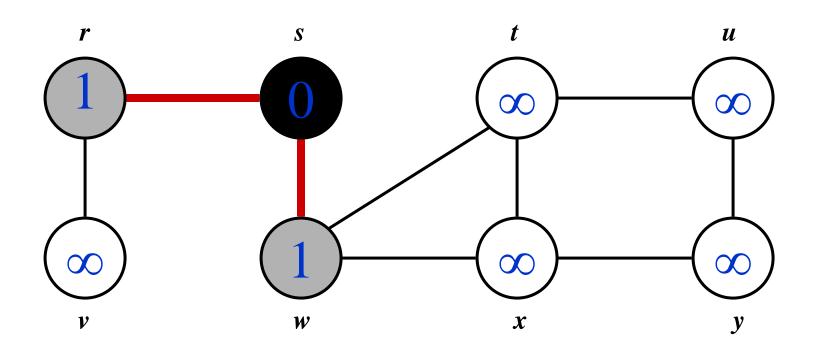
#### **Breadth-First Search**

```
BFS(G, s) {
    initialize vertices:
    Q = \{s\}; // Q is a queue (duh); initialize to s
    while (Q not empty) {
        u = RemoveTop(Q);
        for each v \in u->adj {
            if (v->color == WHITE)
                v->color = GREY;
                v->d = u->d + 1;
                v->p = u;
                Enqueue (Q, v);
        u->color = BLACK;
```

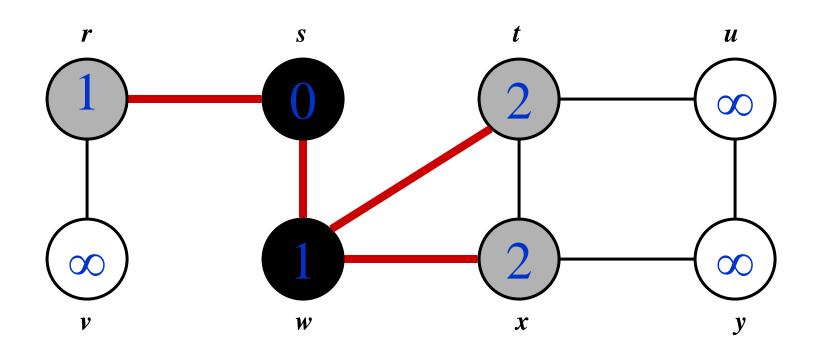




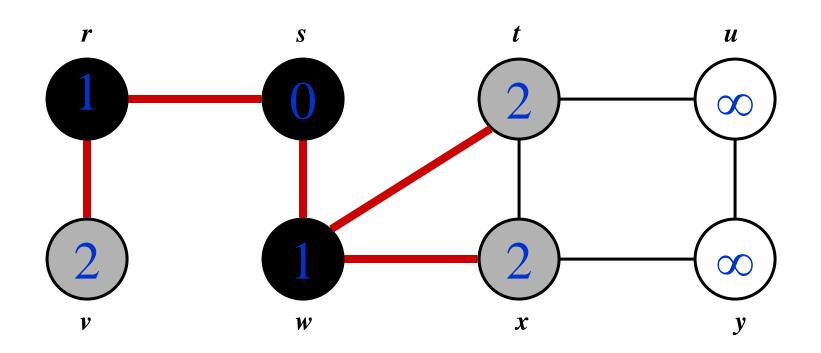
**Q**: s

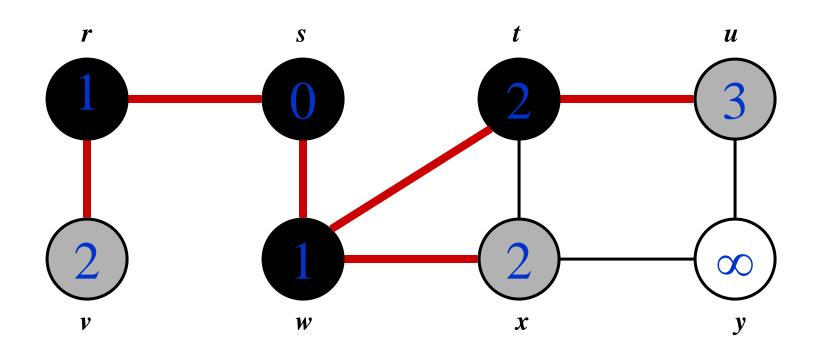


Q: w r

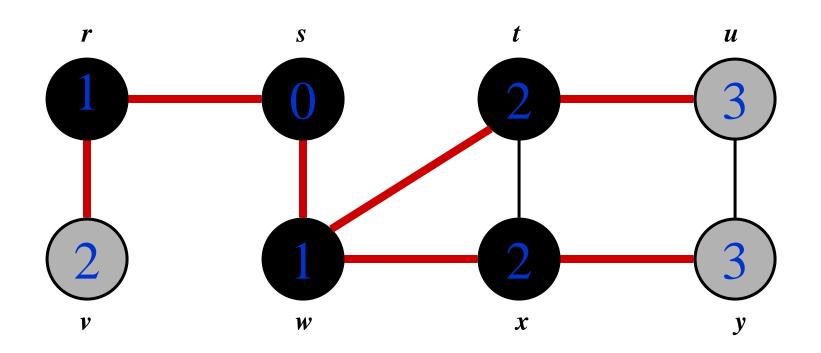


 $Q: \begin{array}{|c|c|c|c|c|} \hline r & t & x \\ \hline \end{array}$ 

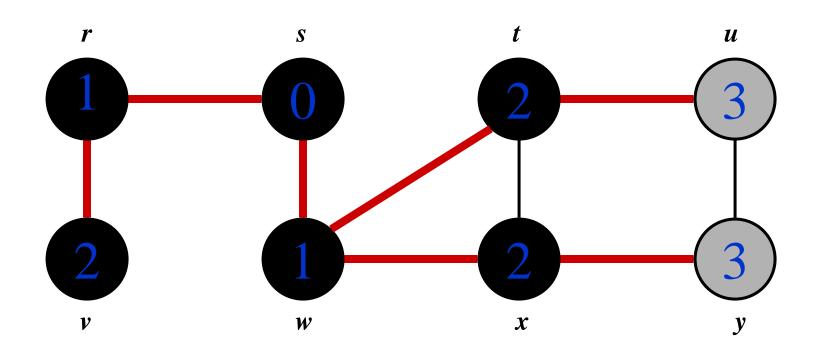




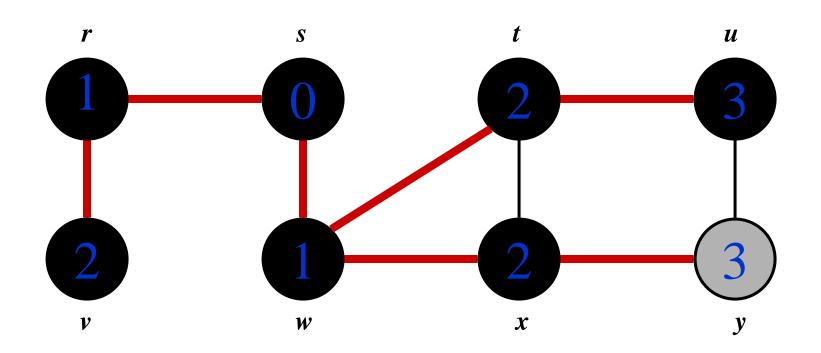
Q: x v u



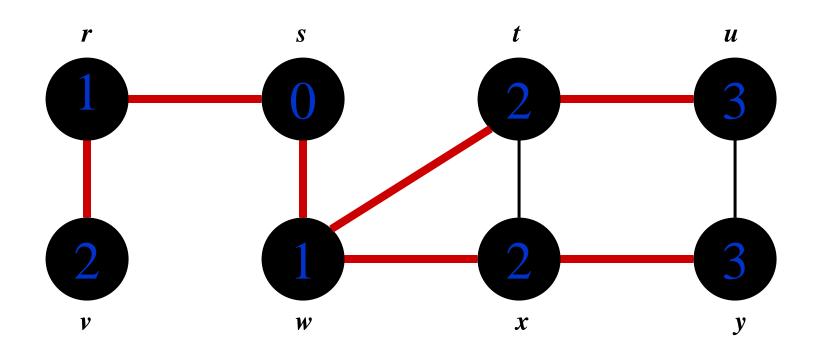
Q: v u y



Q: u y



*Q*: y



Q: Ø

### BFS: The Code Again

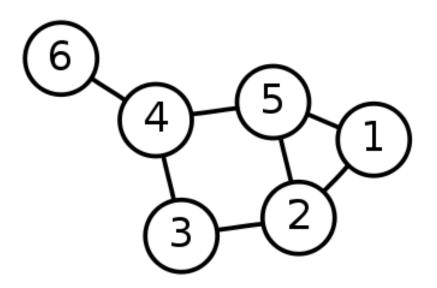
```
BFS(G, s) {
       initialize vertices; \longleftarrow Touch every vertex: O(V)
      Q = \{s\};
       while (Q not empty) {
           u = RemoveTop(Q); \leftarrow u = every vertex, but only once
           for each v \in u-adj \{
                if (v->color == WHITE)
So v = every \ vertex \ v->color = GREY;
                v->d = u->d + 1;
that appears in
some other vert's v->p = u;
                  Enqueue (Q, v);
adjacency list
                                    What will be the running time?
           u \rightarrow color = BLACK;
                                    Total running time: O(V+E)
```



## Dijkstra's Algorithm

#### Single-Source Shortest Path Problem

Single-Source Shortest Path Problem - The problem of finding shortest paths from a source vertex *v* to all other vertices in the graph.

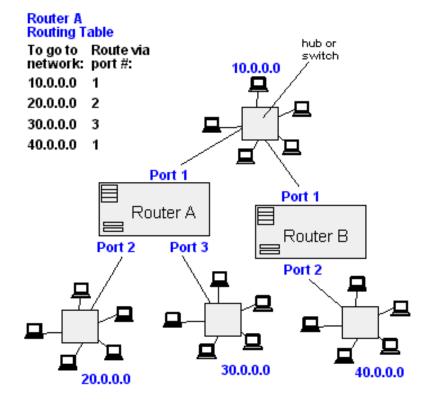


### **Applications**

- Maps (Map Quest, Google Maps)
- Routing Systems



From Computer Desktop Encyclopedia @ 1998 The Computer Language Co. Inc.



### Dijkstra's algorithm

<u>Dijkstra's algorithm</u> - is a solution to the single-source shortest path problem in graph theory.

Works on both directed and undirected graphs. However, all edges must have nonnegative weights.

Input: Weighted graph  $G=\{E,V\}$  and source vertex  $v\in V$ , such that all edge weights are nonnegative

Output: Lengths of shortest paths (or the shortest paths themselves) from a given source vertex *v*∈V to all other vertices

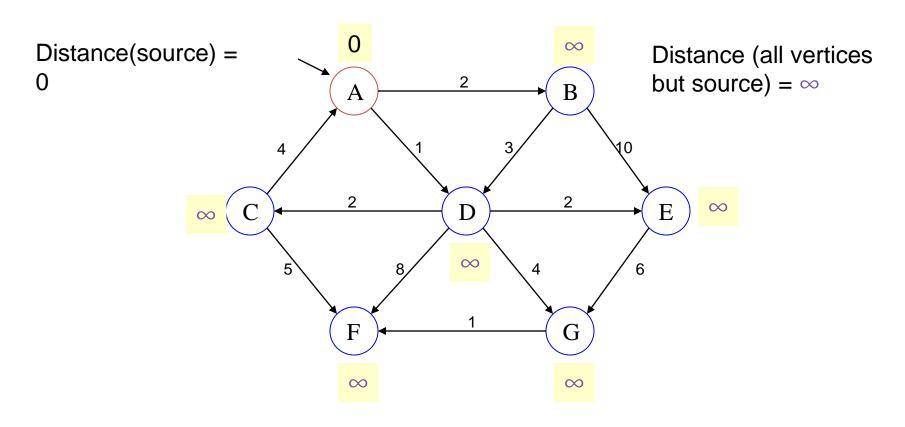
## **Approach**

- The algorithm computes for each vertex u the distance to u
  from the start vertex v, that is, the weight of a shortest path
  between v and u.
- the algorithm keeps track of the set of vertices for which the distance has been computed, called the cloud C
- Every vertex has a label D associated with it. For any vertex u,
  D[u] stores an approximation of the distance between v and
  u. The algorithm will update a D[u] value when it finds a
  shorter path from v to u.
- When a vertex u is added to the cloud, its label D[u] is equal to the actual (final) distance between the starting vertex v and vertex u.

### Dijkstra pseudocode

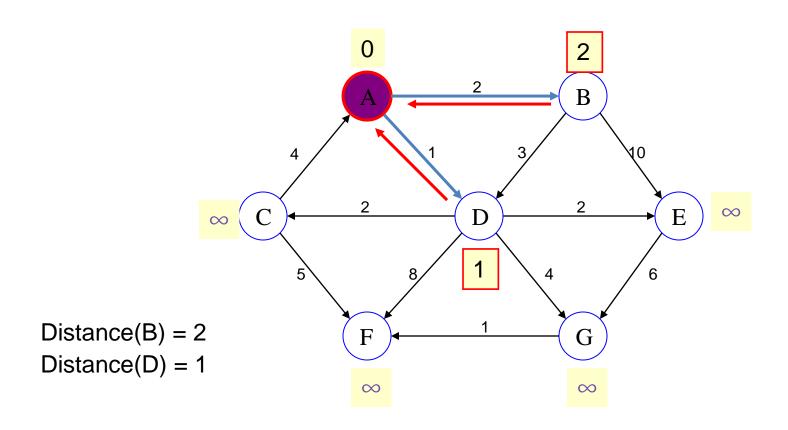
```
Dijkstra(v1, v2):
  for each vertex v:
                                     // Initialization
     v's distance := infinity.
     v's previous := none.
  v1's distance := 0.
  List := {all vertices}.
  while List is not empty:
     v := remove List vertex with minimum distance.
     mark v as known.
     for each unknown neighbor n of v:
       dist := v's distance + edge (v, n)'s weight.
       if dist is smaller than n's distance:
          n's distance := dist.
          n's previous := v.
  reconstruct path from v2 back to v1,
  following previous pointers.
```

## **Example: Initialization**

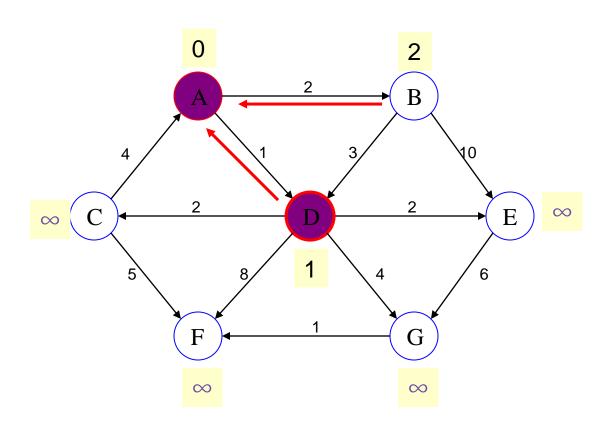


Pick vertex in List with minimum distance.

# Example: Update neighbors' distance

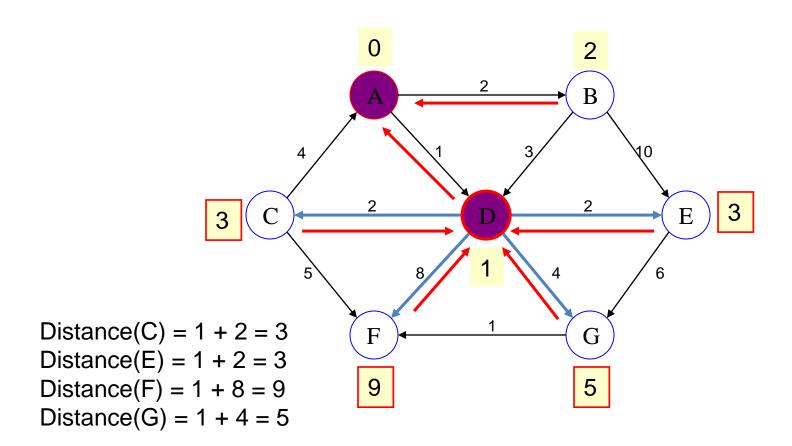


# Example: Remove vertex with minimum distance

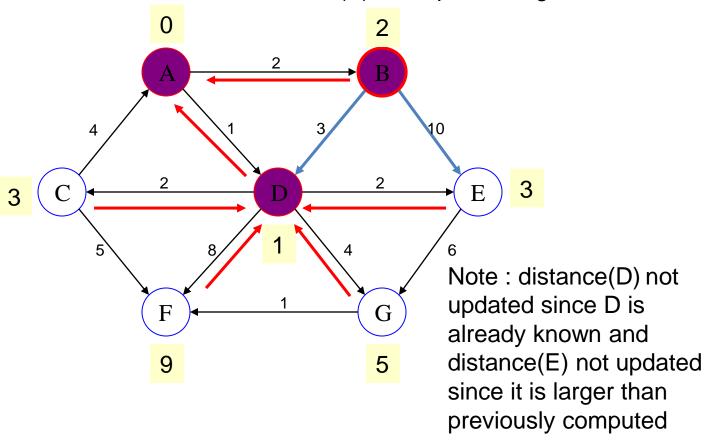


Pick vertex in List with minimum distance, i.e., D

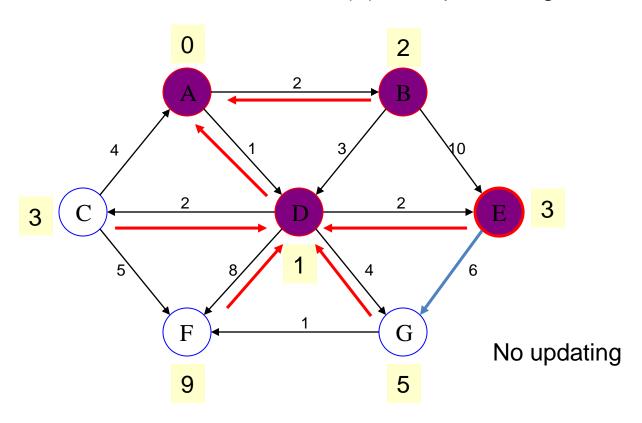
### Example: Update neighbors



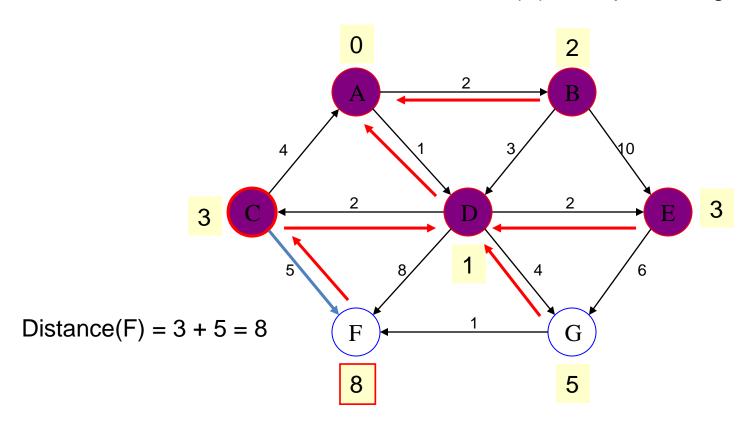
Pick vertex in List with minimum distance (B) and update neighbors



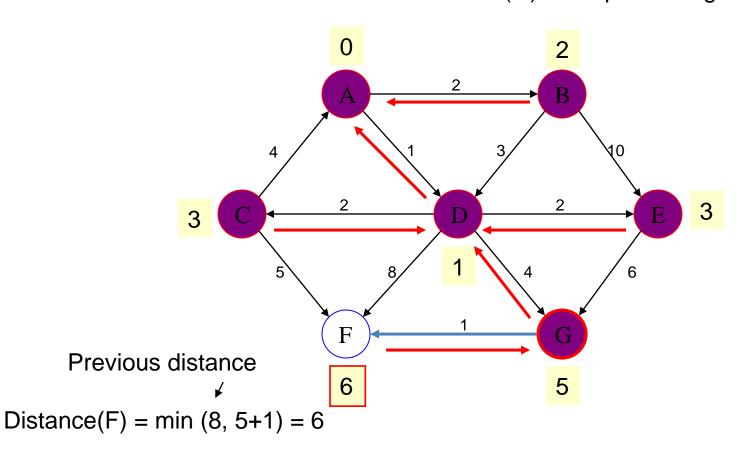
Pick vertex List with minimum distance (E) and update neighbors



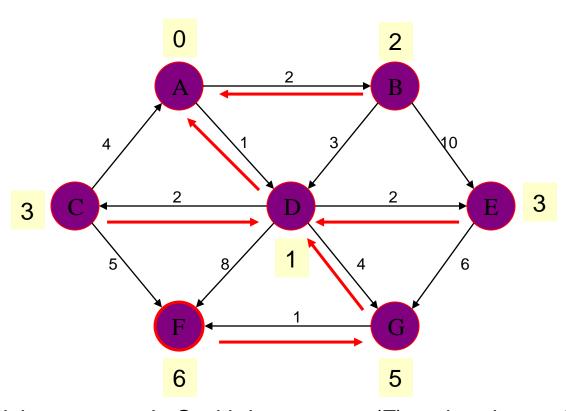
Pick vertex List with minimum distance (C) and update neighbors



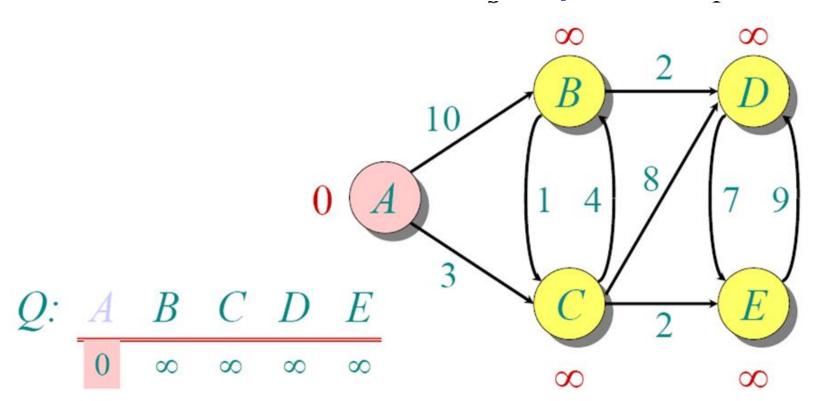
Pick vertex List with minimum distance (G) and update neighbors

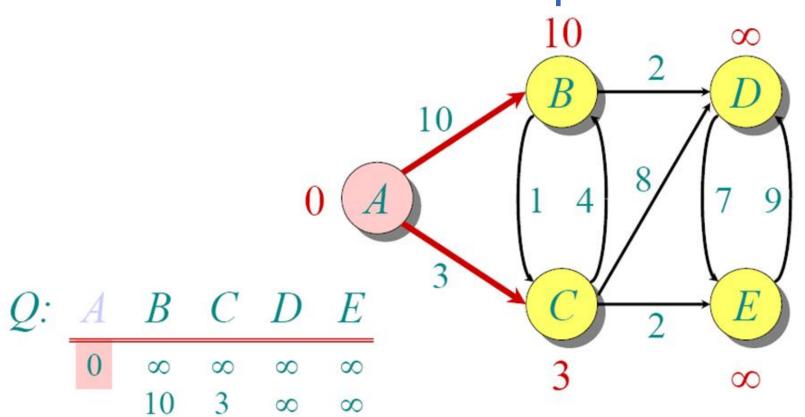


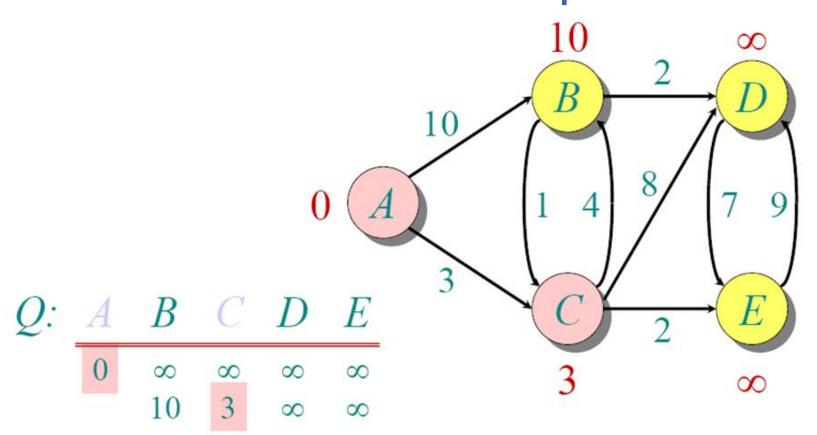
## Example (end)



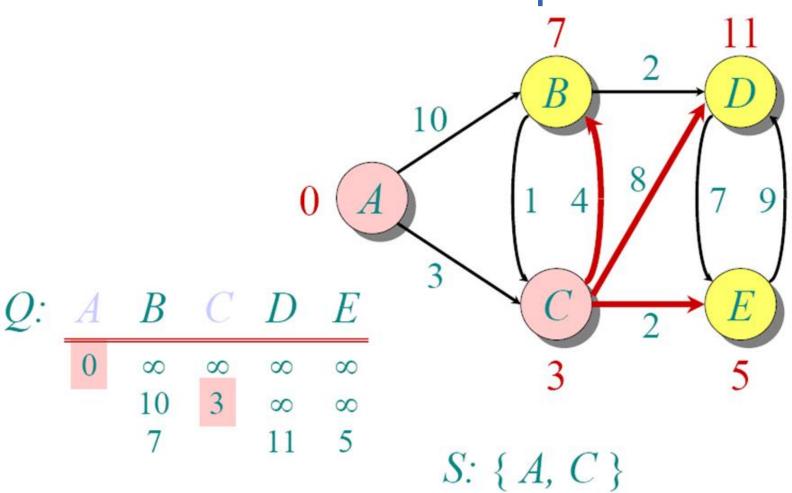
Pick vertex not in S with lowest cost (F) and update neighbors

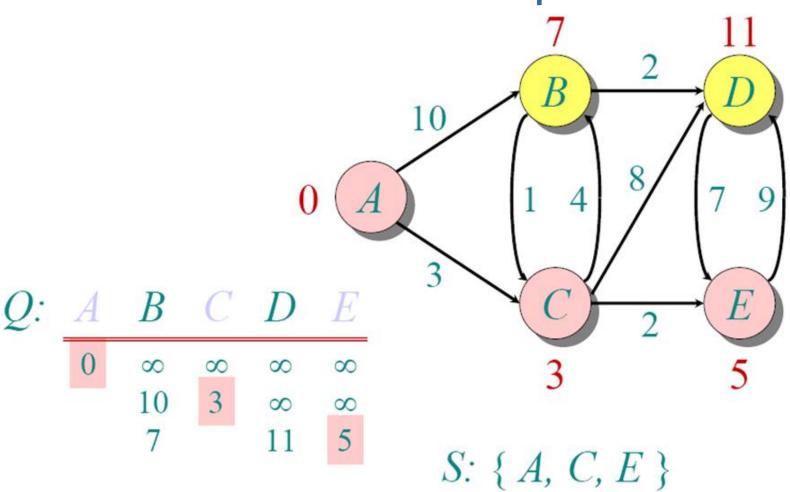


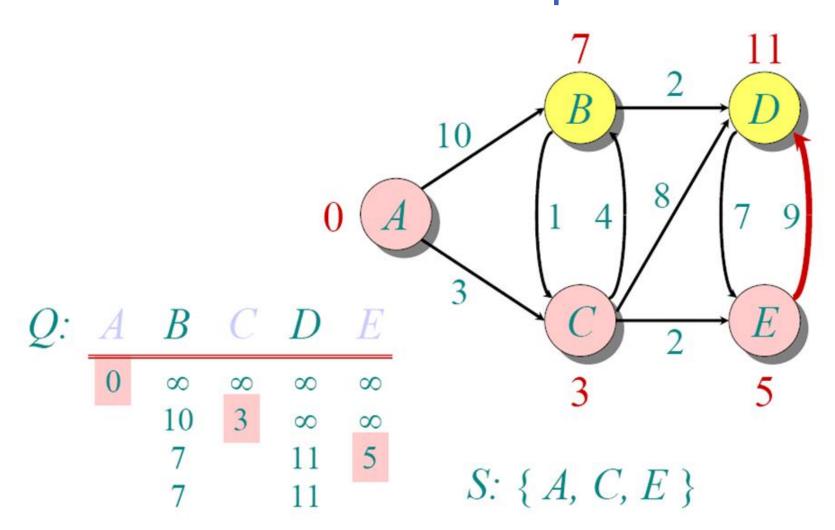


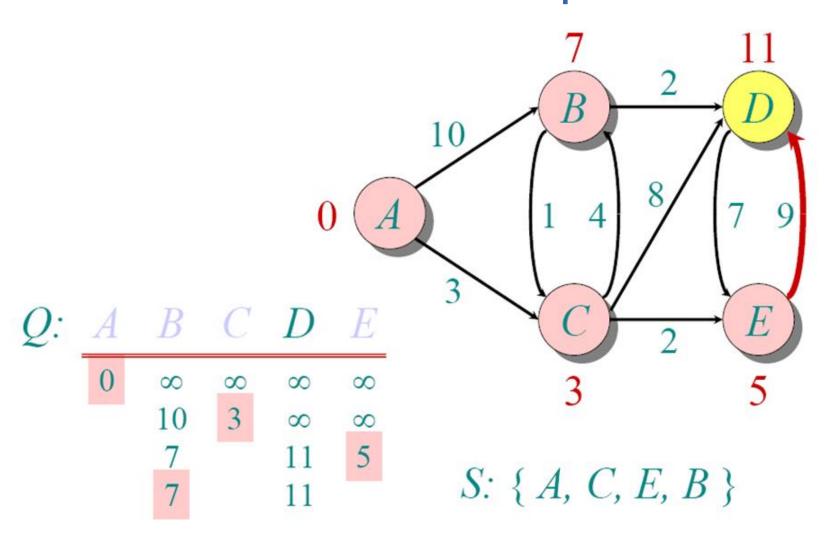


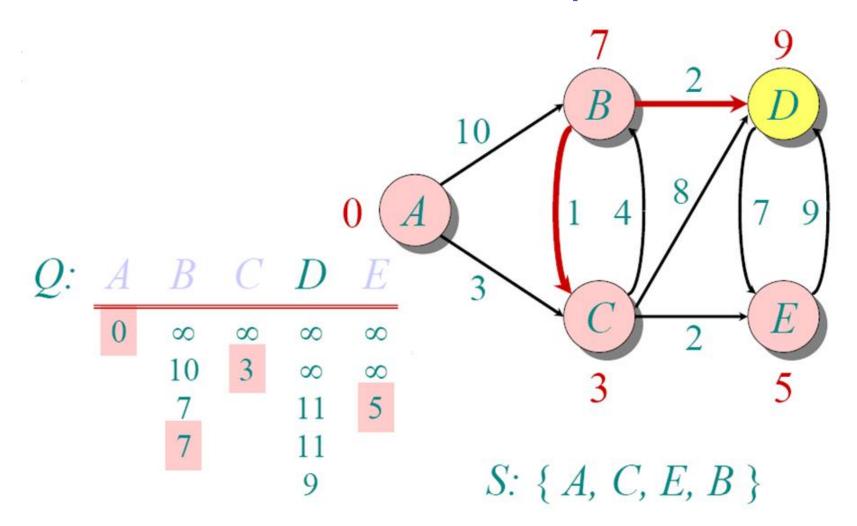
S: { A, C }

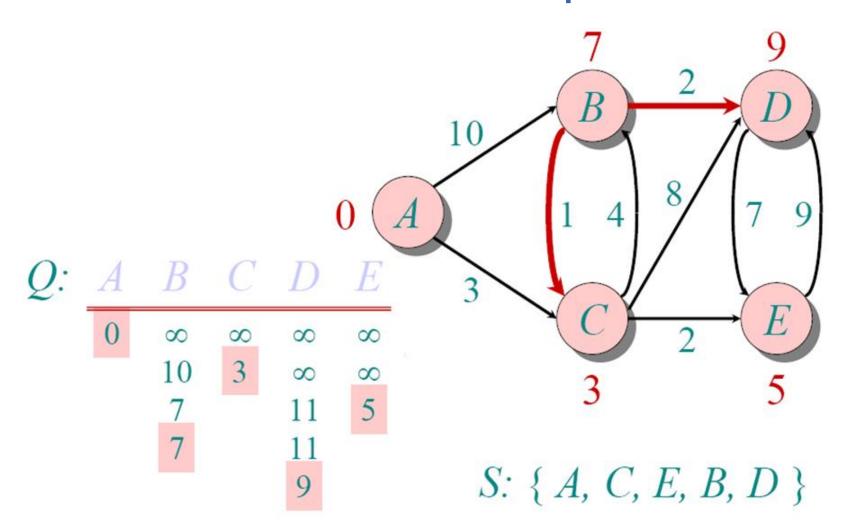






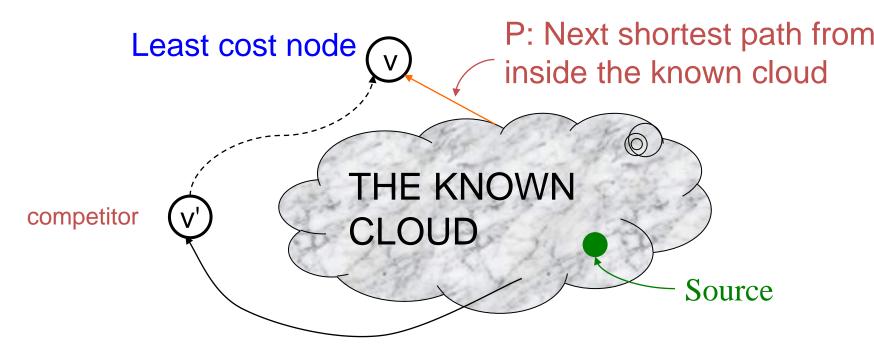






## Correctness: "Cloudy" Proof

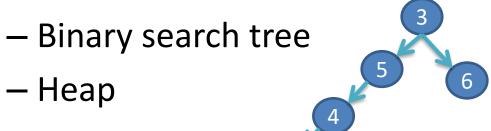
When a vertex is added to the cloud, it has shortest distance to source.



 If the path to v is the next shortest path, the path to v' must be at least as long. Therefore, any path through v' to v cannot be shorter! Minimum Spanning trees

#### What are trees in undirected graphs?

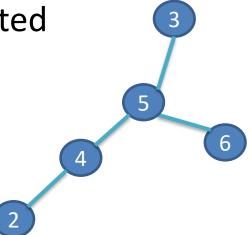
 We are familiar with a rooted directed tree, such as:



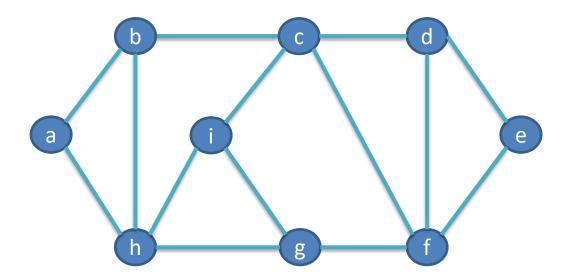
- They look like:
  - You know which one the root is
  - Who the parent is
  - Who the children are

#### What are trees in undirected graphs?

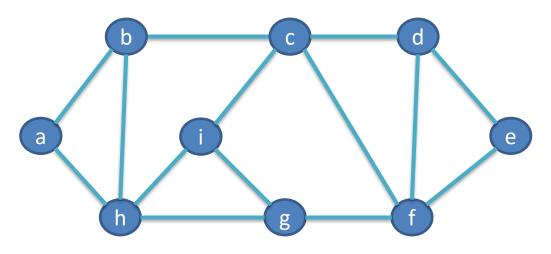
- However, in undirected graphs, there is another definition of trees
- Tree
  - A undirected graph (V, E), where E is the set of undirected edges
  - All vertices are connected
  - -|E|=|V|-1



- Unless otherwise specified, spanning tree is about a tree of a undirected graph
- A spanning tree of graph G=(V, E) is:
  - A graph G'=(V', E'), where  $V' = V, E' \subseteq E$
  - G' is a tree
- Given the following graph, are they spanning trees?



 Given the following graph, are they spanning trees?

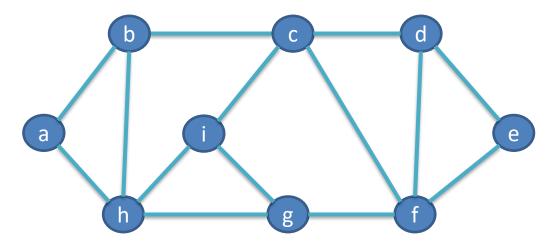


Is the red sub graph a tree?

Yes

Is the red sub graph a spanning No.  $V' \neq V$  tree?

 Given the following graph, are they spanning trees?

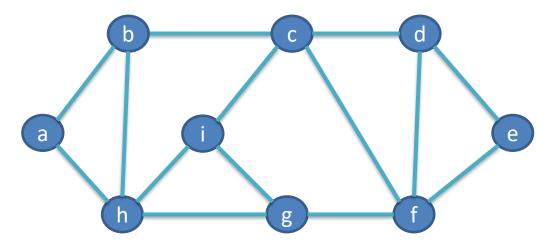


Is the red sub graph a tree?

No. 
$$|E| \neq |V'| - 1$$

Is the red sub graph a spanning No. It is not a tree. tree?

 Given the following graph, are they spanning trees?

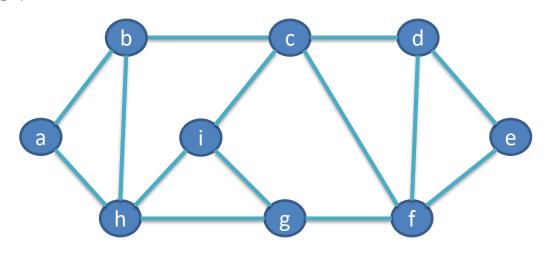


Is the red sub graph a tree?

No. Vertices are not all connected

Is the red sub graph a spanning No. It is not a tree. tree?

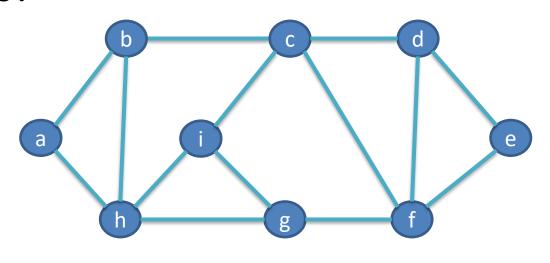
 Given the following graph, are they spanning trees?



Is the red sub graph a tree? Yes

Is the red sub graph a spanning Yes tree?

 Given the following graph, are they spanning trees?



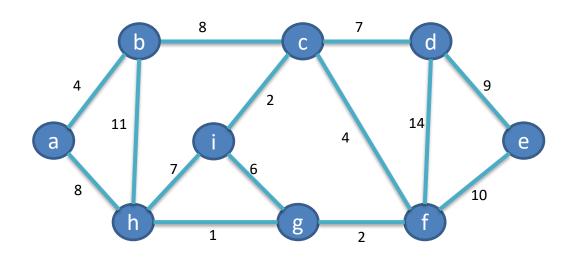
Is the red sub graph a tree? Yes

Is the red sub graph a spanning Yes tree?

There could be more than one spanning trees

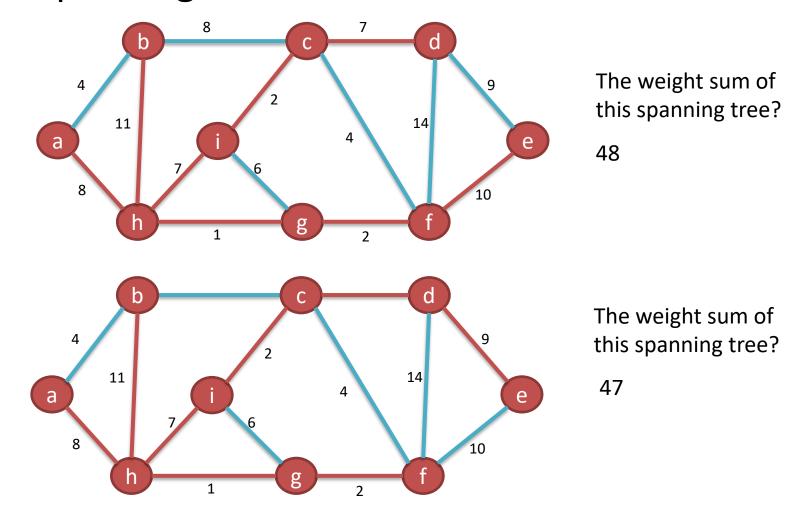
# What is a minimum spanning tree?

- When edges are weighted
- A minimum spanning tree of G is:
  - A spanning tree of G
  - With a minimum sum of a edge weights sum among all spanning trees



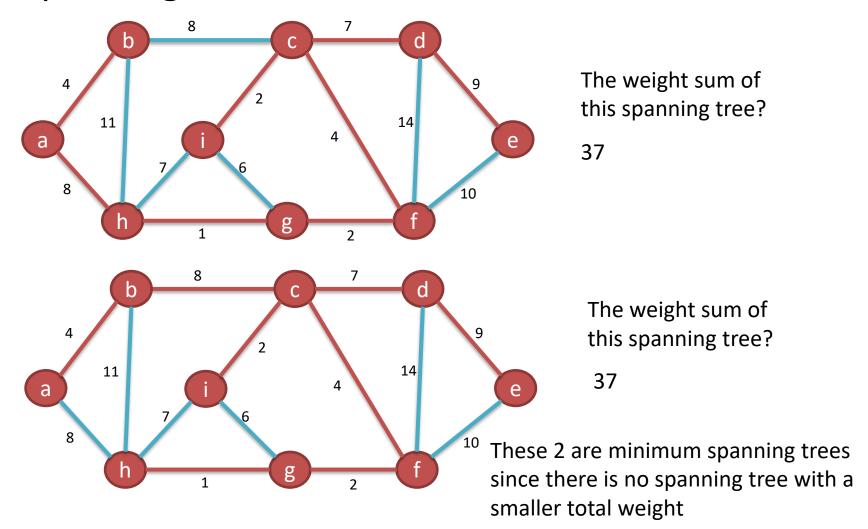
# What is a minimum spanning tree?

#### Spanning trees:



### What is a minimum spanning tree?

#### Spanning trees:



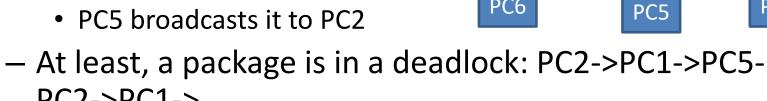
## Minimum spanning tree problem

- Minimum spanning tree (MST) problem: given a undirected graph G, find a spanning tree with the minimum total edge weights.
- There are 2 classic greedy algorithms to solve this problem
  - Kruskal's algorithm
  - Prim's algorithm

### Minimum spanning tree problem

- A application example of MST in computer network
  - What if PC3 wants to communicate with PC7?
  - By a routing protocol:
    - PC3 broadcasts this requirement to PC2, PC4
    - PC2 broadcasts it to PC1, PC5
    - PC1 broadcasts it to PC5, PC6

- PC5 broadcasts it to PC2
- PC2->PC1-> .....



PC6

PC4

Find and use a MST can solve this problem, since there is no cycle

### Minimum spanning tree problem

- The MST algorithm can be considered as choose |V|-1 edges from E to form a minimum spanning tree
- Both Kruskal's algorithm and Prim's algorithm proposed a greedy choice about how to choose a edge
- They have greedy choice property because of textbook Corollary 23.2

#### Kruskal's algorithm

- Input: G=(V, E)
  Output: a minimum spanning tree (V, A)
  Kruskal (G=(V, E))
  Set each vertices in V as a vertex set;
  Set A as an empty set;
  while (there are more than 1 vertex set){
  Add a smallest edge (u, v) to A where u, v are not in the same set;
  Merge the set contains u with the set contains v;
  }
- In most of the implementations, edges are sorted on weights at the beginning, and then scan them once
- The running time = O(|E|lg|V|)

#### Kruskal's algorithm

#### Kruskal (G=(V, E))

Set each vertices in V as a vertex set;

Set A as an empty set;

while (there are more than 1 vertex set){

Add a smallest edge (u, v) to A where u,

v are not in the same set;

Merge the set contains u with the set

contains v;

4 11 8	8 7 6	7 C	d 9 14	е
h	1	<b>g</b> 2	f	

Total weight: 37

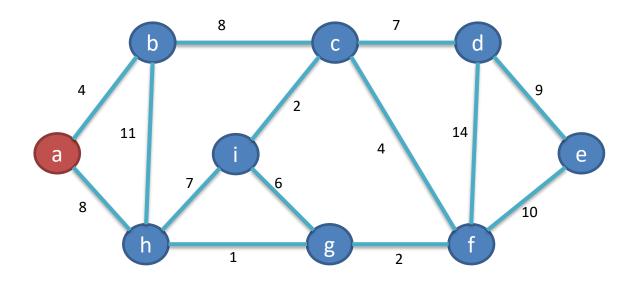
Edge	Weight
(h, g)	1
(i, c)	2
<b>(g, f)</b>	2
(c, f)	4
(a, b)	4
(i, g)	6
(h, i)	7
o (c, d)	7
o (a, h)	8
(b, c)	8
o (d, e)	9
(e, f)	10
(b, h)	11
(d, f)	14

#### Prim's algorithm

- Input: G=(V, E) Output: a minimum spanning tree (V, A) Prim(G=(V, E)) Set A as an empty set; Choose a vertex as the current MST; while (not all the vertices are added into the MST){ Choose the smallest edge crossing current MST vertices and other vertices, and put it to MST;
- The running time = O(|E|Ig|V|)

#### Prim's algorithm

```
Prim(G=(V, E))
    Set A as an empty set;
    Choose a vertex as the current MST;
    while (not all the vertices are added into the MST){
        Choose the smallest safe edge, and put it to MST;
    }
```



Total weight: 37

#### Summary

 Prim's algorithm and Kruskal's algorithm have the same asymptotically running time