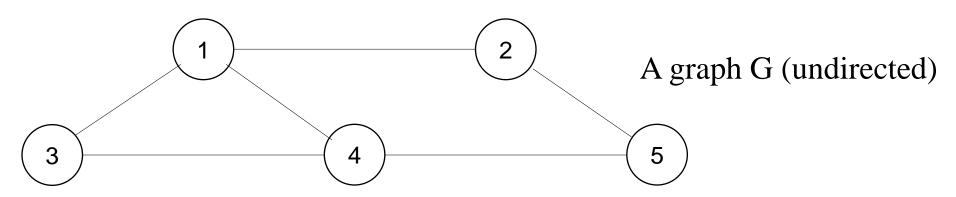
#### **GRAPHS** – Definitions

- A graph G = (V, E) consists of
  - a set of *vertices*, V, and
  - a set of *edges*, E, where each edge is a pair (v,w) s.t.  $v,w \in V$
- Vertices are sometimes called *nodes*, edges are sometimes called *arcs*.
- If the edge pair is ordered then the graph is called a **directed graph** (also called *digraphs*).
- We also call a normal graph (which is not a directed graph) an *undirected graph*.
  - When we say graph we mean that it is an undirected graph.

### **Graph – Definitions**

- Two vertices of a graph are *adjacent* if they are joined by an edge.
- Vertex w is *adjacent to* v iff  $(v,w) \in E$ .
  - In an undirected graph with edge (v, w) and hence (w,v) w is adjacent to v and v is adjacent to w.
- A *path* between two vertices is a sequence of edges that begins at one vertex and ends at another vertex.
  - i.e.  $w_1, w_2, ..., w_N$  is a path if  $(w_i, w_{i+1}) \in E$  for  $1 \le i \le N-1$
- A simple path passes through a vertex only once.
- A *cycle* is a path that begins and ends at the same vertex.
- A *simple cycle* is a cycle that does not pass through other vertices more than once.

#### **Graph – An Example**



The graph G=(V,E) has 5 vertices and 6 edges:

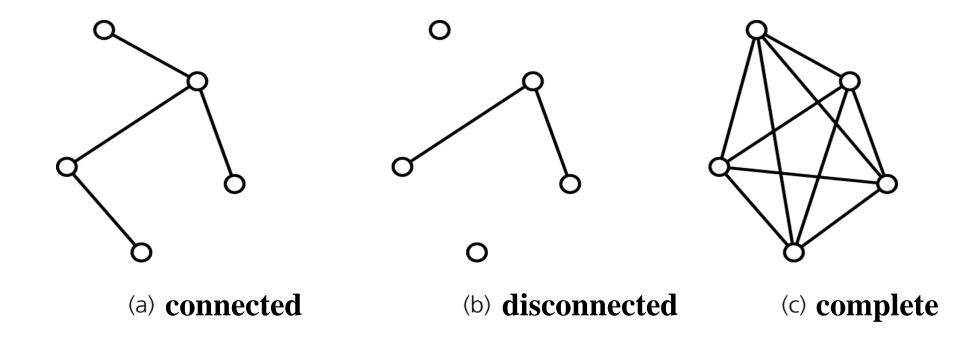
$$V = \{1,2,3,4,5\}$$

$$E = \{ (1,2),(1,3),(1,4),(2,5),(3,4),(4,5), (2,1),(3,1),(4,1),(5,2),(4,3),(5,4) \}$$

- Adjacent:
  - 1 and 2 are adjacent -- 1 is adjacent to 2 and 2 is adjacent to 1
- *Path*:
  - 1,2,5 (a simple path), 1,3,4,1,2,5 (a path but not a simple path)
- *Cycle:* 1,3,4,1 (a simple cycle), 1,3,4,1,4,1 (cycle, but not simple cycle)

#### **Graph -- Definitions**

- A *connected graph* has a path between each pair of distinct vertices.
- A *complete graph* has an edge between each pair of distinct vertices.
  - A complete graph is also a connected graph. But a connected graph may not be a complete graph.



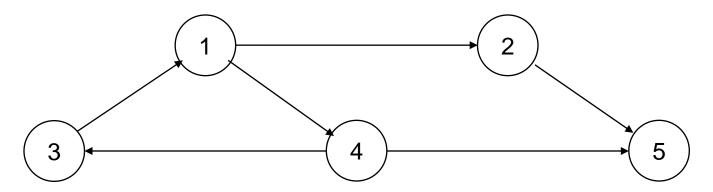
#### **Directed Graphs**

- If the edge pair is ordered then the graph is called a **directed graph** (also called *digraphs*).
- Each edge in a directed graph has a direction, and each edge is called a *directed edge*.
- Definitions given for undirected graphs apply also to directed graphs, with changes that account for direction.
- Vertex w is *adjacent to* v iff  $(v,w) \in E$ .
  - i.e. There is a direct edge from v to w
  - w is *successor* of v
  - v is *predecessor* of w
- A *directed path* between two vertices is a sequence of directed edges that begins at one vertex and ends at another vertex.
  - i.e.  $w_1, w_2, ..., w_N$  is a path if  $(w_i, w_{i+1}) \in E$  for  $1 \le i \le N-1$

### **Directed Graphs**

- A **cycle** in a directed graph is a path of length at least 1 such that  $w_1 = w_N$ .
  - This cycle is simple if the path is simple.
  - For undirected graphs, the edges must be distinct
- A directed acyclic graph (DAG) is a type of directed graph having no cycles.
- An undirected graph is **connected** if there is a path from every vertex to every other vertex.
- A directed graph with this property is called strongly connected.
  - If a directed graph is not strongly connected, but the underlying graph (without direction to arcs) is connected then the graph is weakly connected.

#### **Directed Graph – An Example**



The graph G=(V,E) has 5 vertices and 6 edges:

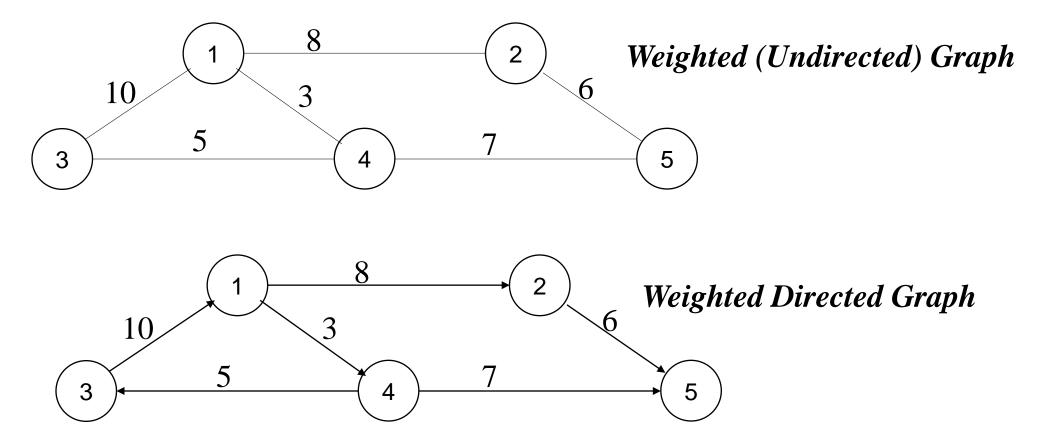
$$V = \{1,2,3,4,5\}$$

$$E = \{ (1,2),(1,4),(2,5),(4,5),(3,1),(4,3) \}$$

- *Adjacent:* 
  - 2 is adjacent to 1, but 1 is NOT adjacent to 2
- *Path*:
  - 1,2,5 (a directed path),
- *Cycle:* 1,4,3,1 (a directed cycle),

#### Weighted Graph

• We can label the edges of a graph with numeric values, the graph is called a *weighted graph*.



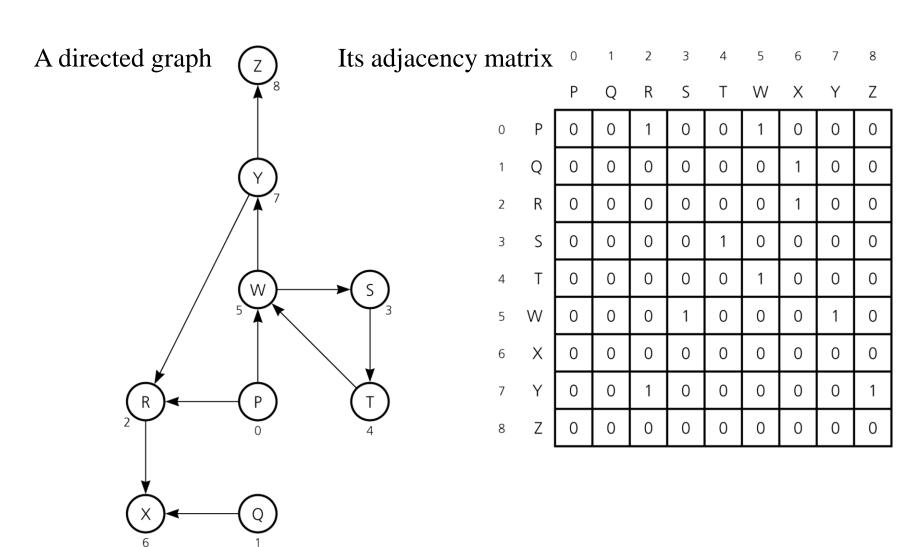
#### **Graph Implementations**

- The two most common implementations of a graph are:
  - Adjacency Matrix
    - A two dimensional array
  - Adjacency List
    - For each vertex we keep a list of adjacent vertices

### **Adjacency Matrix**

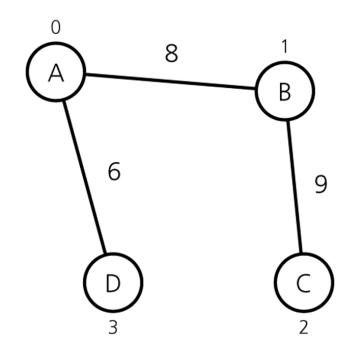
- An *adjacency matrix* for a graph with *n* vertices numbered 0,1,...,n-1 is an *n* by *n* array *matrix* such that *matrix*[*i*][*j*] is 1 (true) if there is an edge from vertex *i* to vertex *j*, and 0 (false) otherwise.
- When the graph is *weighted*, we can let matrix[i][j] be the weight that labels the edge from vertex i to vertex j, instead of simply 1, and let matrix[i][j] equal to  $\infty$  instead of 0 when there is no edge from vertex i to vertex j.
- Adjacency matrix for an undirected graph is symmetrical.
  - i.e. matrix[i][j] is equal to matrix[j][i]
- Space requirement  $O(|V|^2)$
- Acceptable if the graph is dense.

#### **Adjacency Matrix – Example 1**

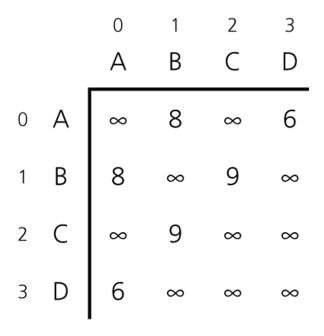


### **Adjacency Matrix – Example 2**

An Undirected Weighted Graph



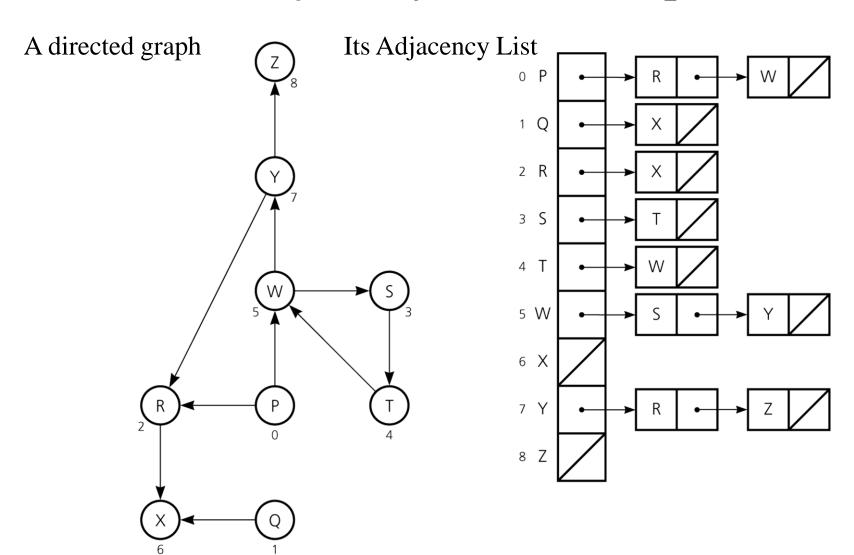
#### Its Adjacency Matrix



### **Adjacency List**

- An *adjacency list* for a graph with n vertices numbered 0,1,...,n-1 consists of n linked lists. The  $i^{th}$  linked list has a node for vertex j if and only if the graph contains an edge from vertex i to vertex j.
- Adjacency list is a better solution if the graph is sparse.
- Space requirement is O(|E| + |V|), which is linear in the size of the graph.
- In an undirected graph each edge (v,w) appears in two lists.
  - Space requirement is doubled.

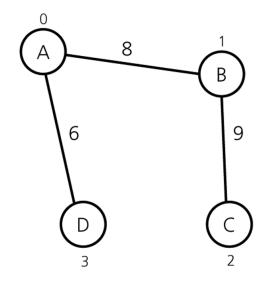
### **Adjacency List – Example 1**

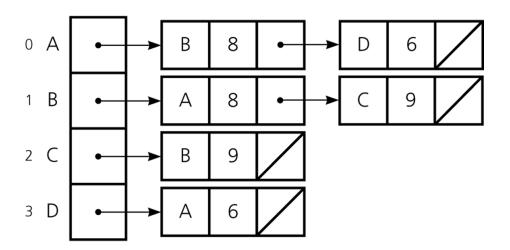


### **Adjacency List – Example 2**

An Undirected Weighted Graph

Its Adjacency List





### Adjacency Matrix vs Adjacency List

- Two common graph operations:
  - 1. Determine whether there is an edge from vertex i to vertex j.
  - 2. Find all vertices adjacent to a given vertex i.
- An adjacency matrix supports operation 1 more efficiently.
- An adjacency list supports operation 2 more efficiently.
- An adjacency list often requires less space than an adjacency matrix.
  - Adjacency Matrix: Space requirement is  $O(|V|^2)$
  - Adjacency List: Space requirement is O(|E| + |V|), which is linear in the size of the graph.
  - Adjacency matrix is better if the graph is dense (too many edges)
  - Adjacency list is better if the graph is sparse (few edges)

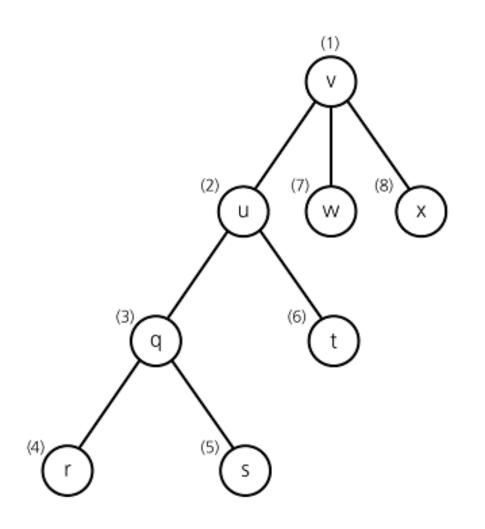
### **Graph Traversals**

- A *graph-traversal* algorithm starts from a vertex v, visits all of the vertices that can be reachable from the vertex v.
- A graph-traversal algorithm visits all vertices if and only if the graph is connected.
- A connected component is the subset of vertices visited during a traversal algorithm that begins at a given vertex.
- A graph-traversal algorithm must mark each vertex during a visit and must never visit a vertex more than once.
  - Thus, if a graph contains a cycle, the graph-traversal algorithm can avoid infinite loop.
- We look at two graph-traversal algorithms:
  - Depth-First Traversal
  - Breadth-First Traversal

#### **Depth-First Traversal**

- For a given vertex v, the *depth-first traversal* algorithm proceeds along a path from v as deeply into the graph as possible before backing up.
- That is, after visiting a vertex v, the *depth-first traversal* algorithm visits (if possible) an unvisited adjacent vertex to vertex v.
- The depth-first traversal algorithm does not completely specify the order in which it should visit the vertices adjacent to v.
  - We may visit the vertices adjacent to v in sorted order.

#### **Depth-First Traversal – Example**



- A depth-first traversal of the graph starting from vertex v.
- Visit a vertex, then visit a vertex adjacent to that vertex.
- If there is no unvisited vertex adjacent to visited vertex, back up to the previous step.

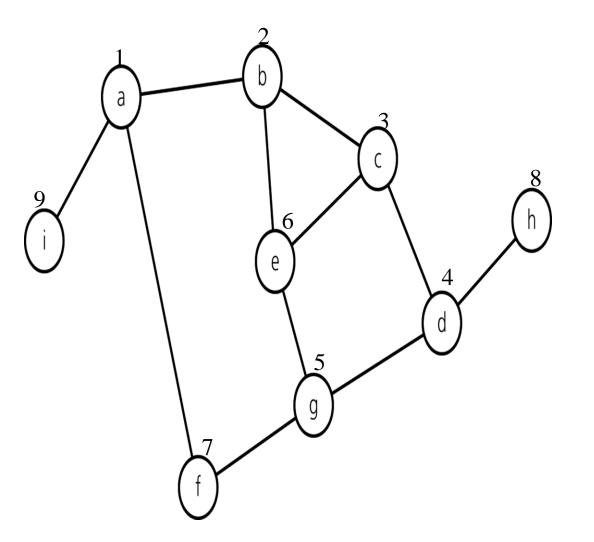
#### **Recursive Depth-First Traversal Algorithm**

```
dft(in v:Vertex) {
// Traverses a graph beginning at vertex v
// by using depth-first strategy
// Recursive Version
 Mark v as visited;
 for (each unvisited vertex u adjacent to v)
     dft(u)
```

### **Iterative Depth-First Traversal Algorithm**

```
dft(in v:Vertex) {
// Traverses a graph beginning at vertex v
// by using depth-first strategy: Iterative Version
  s.createStack();
  // push v into the stack and mark it
  s.push(v);
  Mark v as visited;
  while (!s.isEmpty()) {
      if (no unvisited vertices are adjacent to the vertex on
         the top of stack)
         s.pop(); // backtrack
     else {
         Select an unvisited vertex u adjacent to the vertex
            on the top of the stack;
         s.push(u);
        Mark u as visited;
```

### Trace of Iterative DFT – starting from vertex a

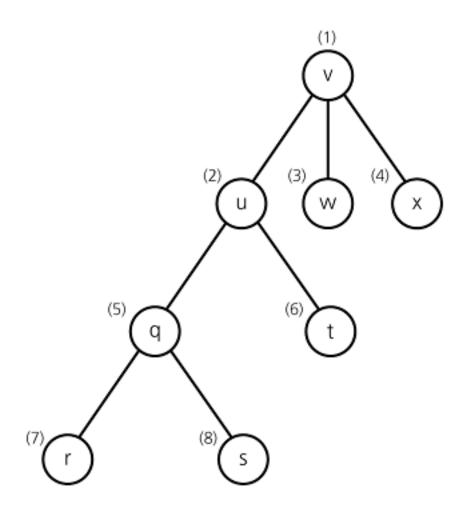


Node visited	Stack (bottom to top)
a	a
b	a b
С	a b c
d	a b c d
g	a b c d g
е	a b c d g e
(backtrack)	a b c d g
f	a b c d g f
(backtrack)	a b c d g
(backtrack)	a b c d
h	a b c d h
(backtrack)	a b c d
(backtrack)	a b c
(backtrack)	a b
(backtrack)	a
i	a i
(backtrack)	a
(backtrack)	(empty)

#### **Breath-First Traversal**

- After visiting a given vertex v, the breadth-first traversal algorithm visits every vertex adjacent to v that it can before visiting any other vertex.
- The breath-first traversal algorithm does not completely specify the order in which it should visit the vertices adjacent to v.
  - We may visit the vertices adjacent to v in sorted order.

### **Breath-First Traversal – Example**

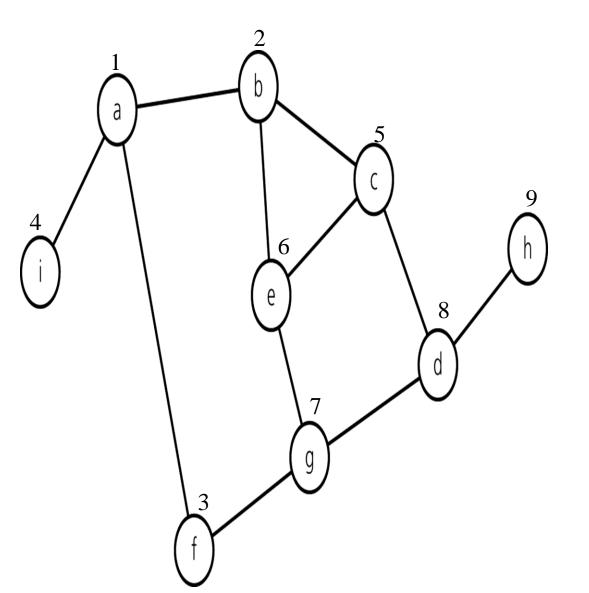


- A breath-first traversal of the graph starting from vertex v.
- Visit a vertex, then visit all vertices adjacent to that vertex.

#### **Iterative Breath-First Traversal Algorithm**

```
bft(in v: Vertex) {
// Traverses a graph beginning at vertex v
// by using breath-first strategy: Iterative Version
  g.createOueue();
  // add v to the queue and mark it
  q.enqueue(v);
  Mark v as visited:
  while (!q.isEmpty()) {
     q.dequeue(w);
     for (each unvisited vertex u adjacent to w) {
        Mark u as visited;
        q.enqueue(u);
```

### Trace of Iterative BFT – starting from vertex a



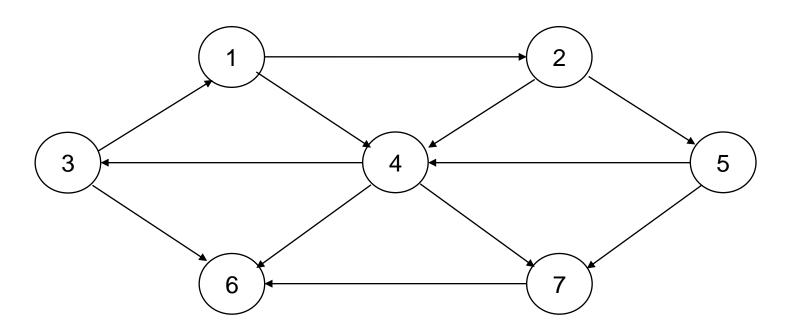
Node visited	Queue (front to back)
a	a
	(empty)
b	b
f	b f
i	bfi
	fi
С	fic
е	fice
	ice
g	i c e g
	c e g
	e g
d	e g d
	g d
	d
	(empty)
h	h
	(empty)

## Some Graph Algorithms

- Shortest Path Algorithms
  - Unweighted shortest paths
  - Weighted shortest paths (Dijkstra's Algorithm)
- Topological sorting
- Network Flow Problems
- Minimum Spanning Tree
- Depth-first search Applications

## **Unweighted Shortest-Path problem**

• Find the shortest path (measured by number of edges) from a designated vertex S to every vertex.

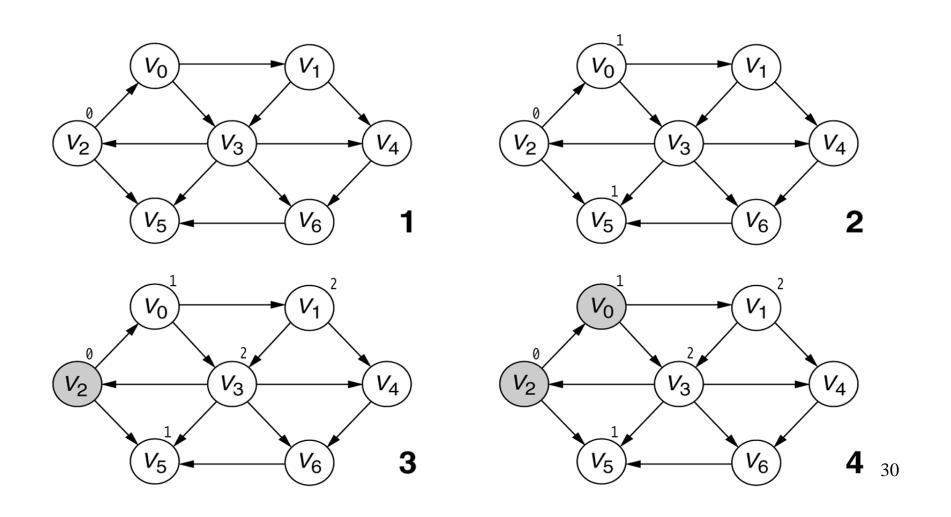


## Algorithm

- 1. Start with an initial node s.
  - Mark the distance of s to s,  $D_s$  as 0.
  - Initially  $D_i = \infty$  for all  $i \neq s$ .
- 2. Traverse all nodes starting from s as follows:
  - 1. If the node we are currently visiting is v, for all w that are adjacent to v:
    - Set  $D_w = D_v + 1$  if  $D_w = \infty$ .
  - 2. Repeat step 2.1 with another vertex u that has not been visited yet, such that  $D_u = D_v$  (if any).
  - 3. Repeat step 2.1 with another unvisited vertex u that satisfies  $D_u = D_v + 1$ .(if any)

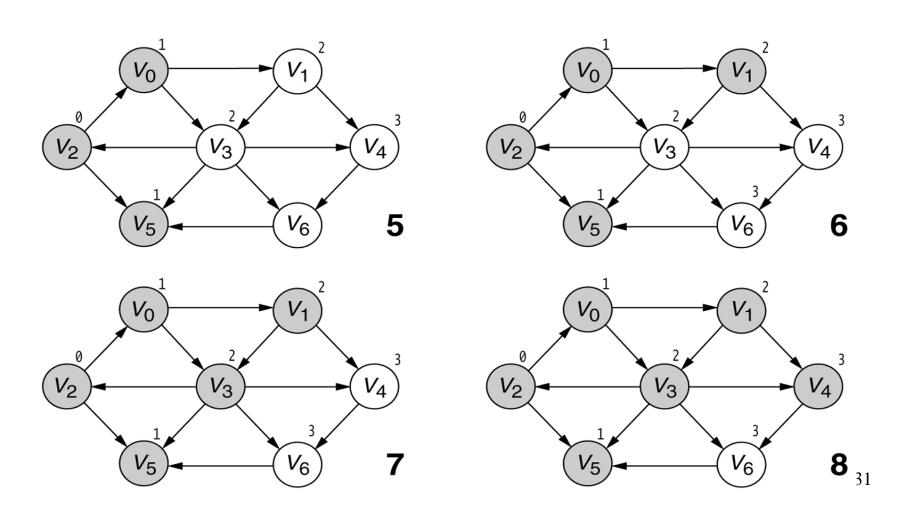
#### **Figure 14.21A**

Searching the graph in the unweighted shortest-path computation. The darkest-shaded vertices have already been completely processed, the lightest-shaded vertices have not yet been used as v, and the medium-shaded vertex is the current vertex, v. The stages proceed left to right, top to bottom, as numbered (continued).



#### **Figure 14.21B**

Searching the graph in the unweighted shortest-path computation. The darkest-shaded vertices have already been completely processed, the lightest-shaded vertices have not yet been used as v, and the medium-shaded vertex is the current vertex, v. The stages proceed left to right, top to bottom, as numbered.

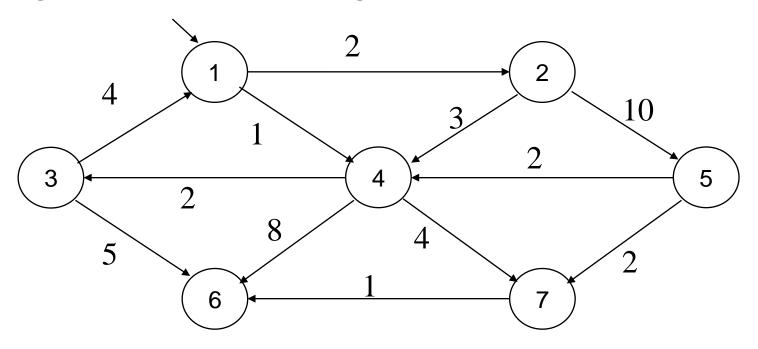


# Unweighted shortest path algorithm

```
void Graph::unweighted shortest paths(vertex s)
  Queue q(NUM VERTICES);
  Vertex v,w;
  q.enqueue(s);
  s.dist = 0;
  while (!q.isEmpty())
      v= q.dequeue();
      v.known = true; // not needed anymore
       for each w adjacent to v
              if (w.dist == INFINITY)
                    w.dist = v.dist + 1;
                    w.path = v;
                    q.enqueue(w);
```

## Weighted Shortest-path Problem

• Find the shortest path (measured by total cost) from a designated vertex S to every vertex. All edge costs are nonnegative.

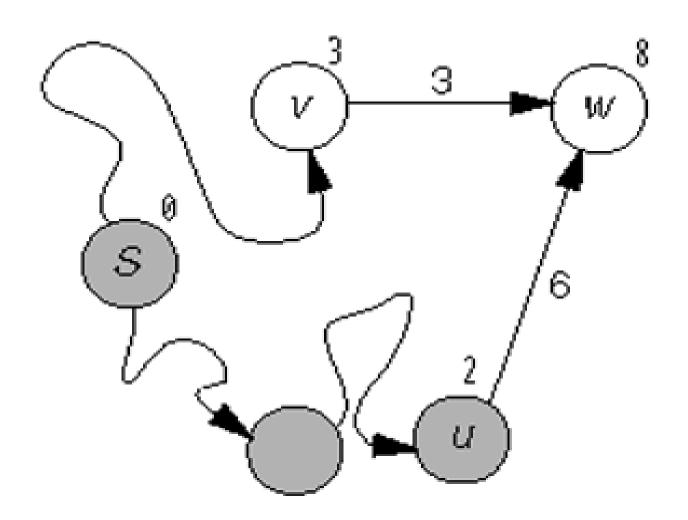


## Weighted Shortest-path Problem

- The method used to solve this problem is known as Dijkstra's algorithm.
  - An example of a greedy algorithm
  - Use the local optimum at each step
- Solution is similar to the solution of unweighted shortest path problem.
- The following issues must be examined:
  - How do we adjust D<sub>w</sub>?
  - How do we find the vertex v to visit next?

#### **Figure 14.23**

The eyeball is at v and w is adjacent, so  $D_w$  should be lowered to 6.

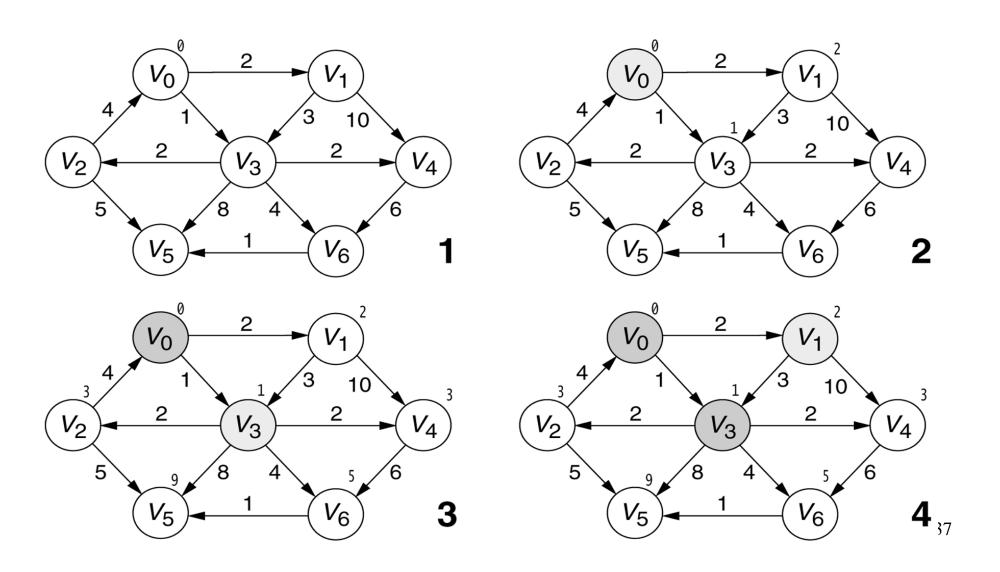


## Dijkstra's algorithm

- The algorithm proceeds in stages.
- At each stage, the algorithm
  - selects a vertex v, which has the smallest distance  $D_v$  among all the *unknown* vertices, and
  - declares that the shortest path from s to v is *known*.
  - then for the adjacent nodes of v (which are denoted as w)  $D_w$  is updated with new distance information
- How do we change D<sub>w</sub>?
  - If its current value is larger than  $D_v + c_{v,w}$  we change it.

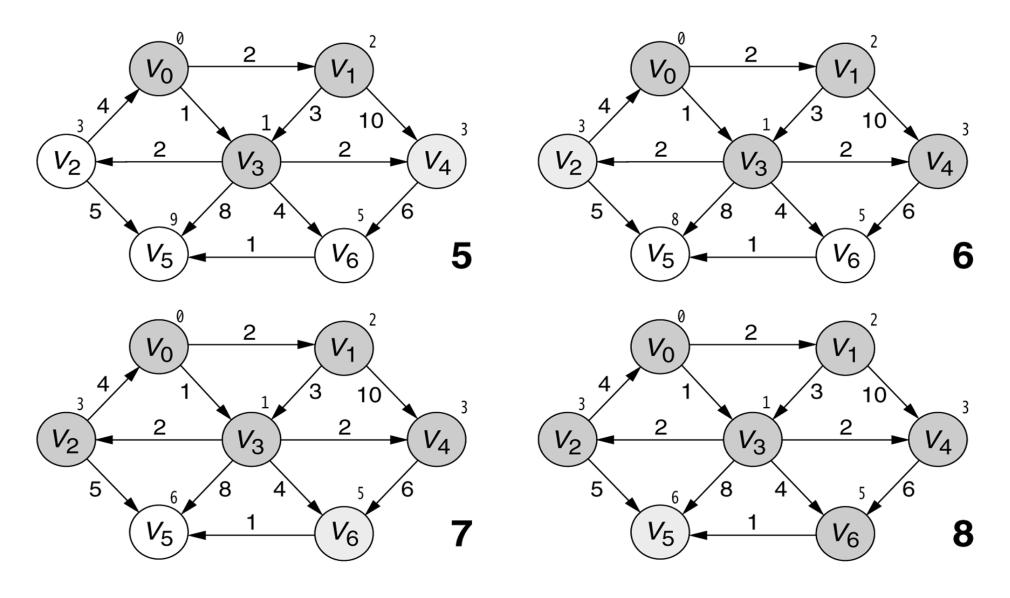
#### **Figure 14.25A**

Stages of Dijkstra's algorithm. The conventions are the same as those in Figure 14.21 (*continued*).



#### **Figure 14.25B**

Stages of Dijkstra's algorithm. The conventions are the same as those in Figure 14.21.



## **Implementation**

- A queue is no longer appropriate for storing vertices to be visited.
- The priority queue is an appropriate data structure.
- Add a new entry consisting of a vertex and a distance, to the priority queue every time a vertex has its distance lowered.