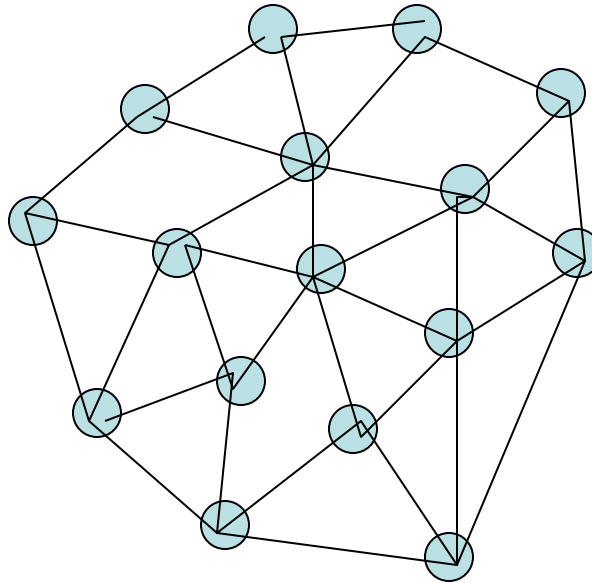


Graphs

Graphs

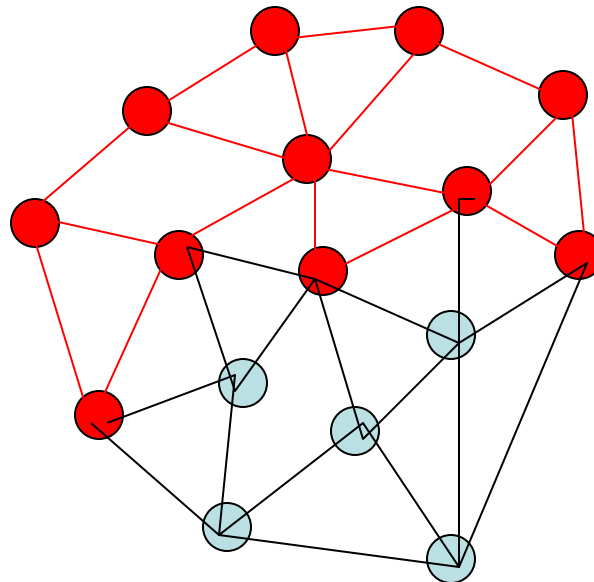
- *Many interesting situations can be modeled by a graph.*

Ex. Mass transportation system, computer network.

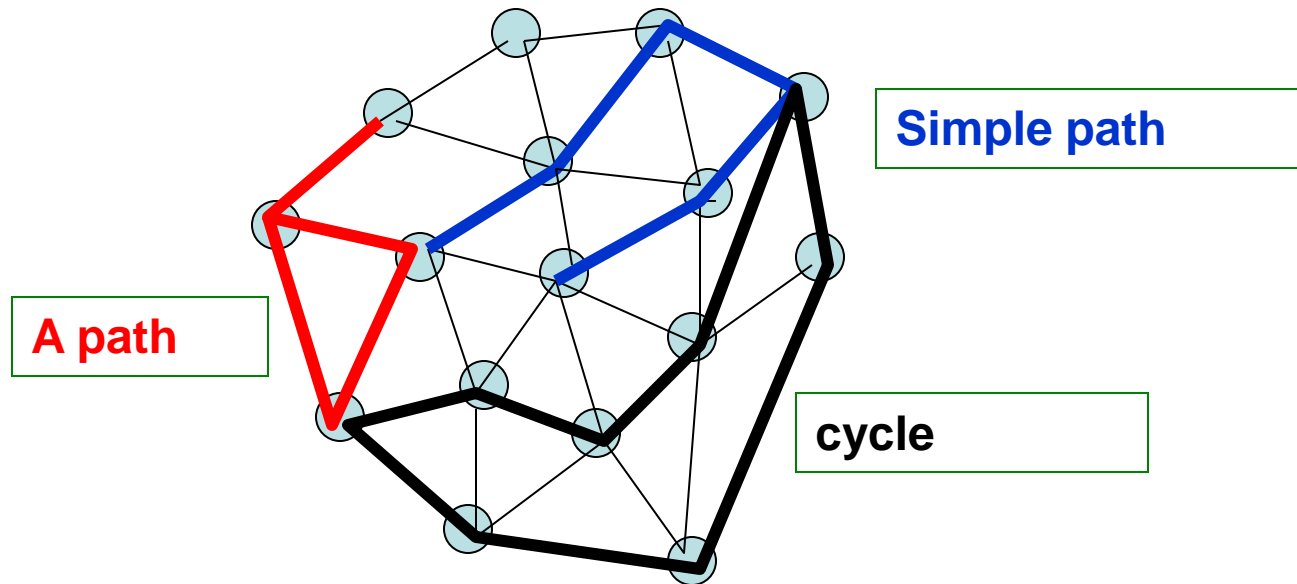


Terminology

- A **graph** consists of a set of nodes (vertices) and a set of edges.
- A **node** is basic component, which usually contains some information.
- An **edge** connects two distinct nodes (vertices).
- A **subgraph** is graph which consists of a subset of nodes (vertices) and a subset of edges of a graph.

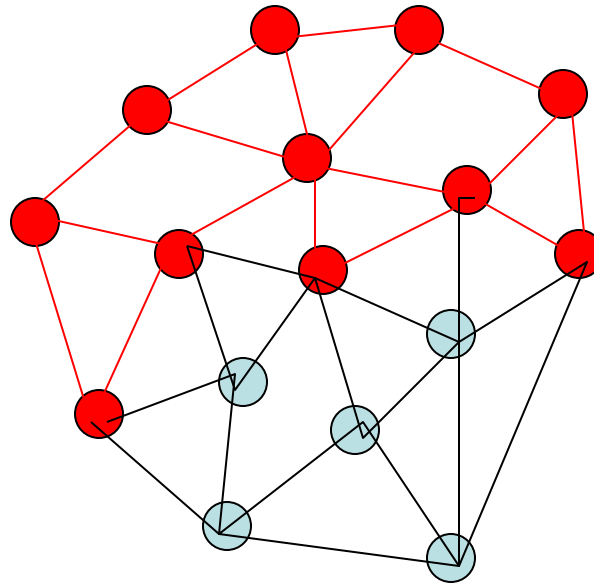


- A **path** is a sequence of nodes such that each successive pair is connected by an edge.
- A path is a **simple path** if each of its nodes occurs once in the sequence.

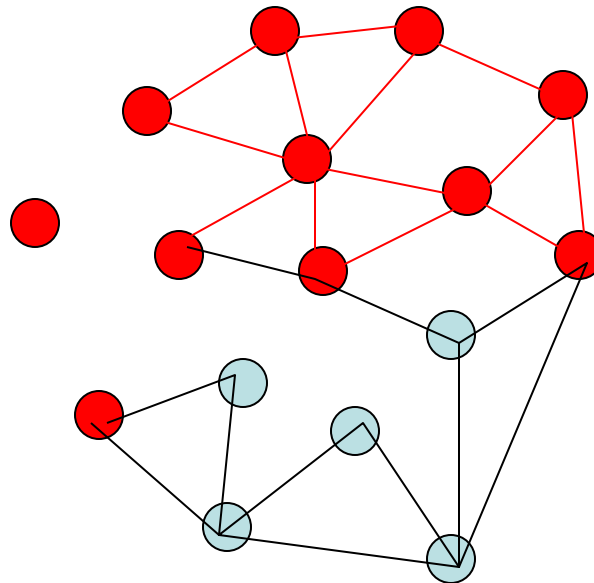


- A **cycle** is path that is a simple path except that the first and last nodes are the same.

- A graph is a **connected graph** if there is a path between every pair of its nodes.

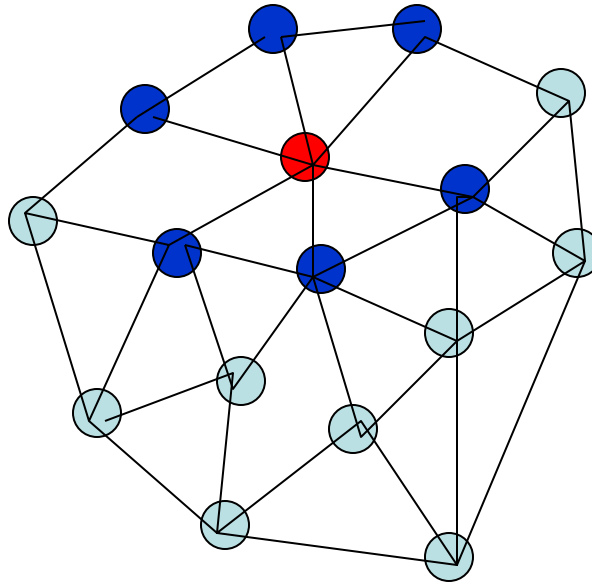


Connected graph



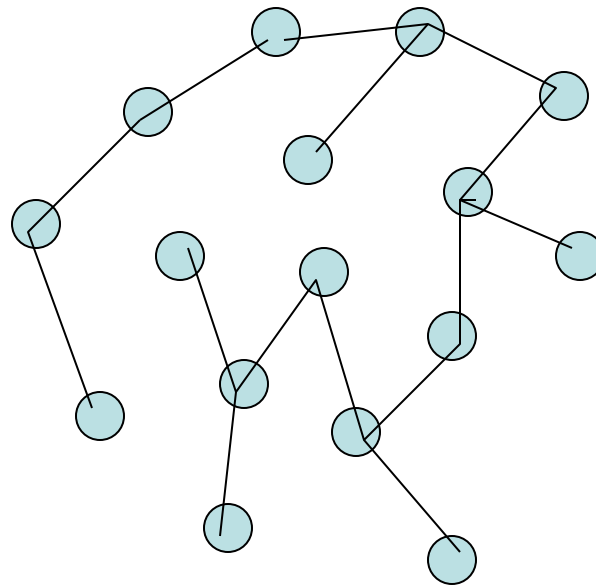
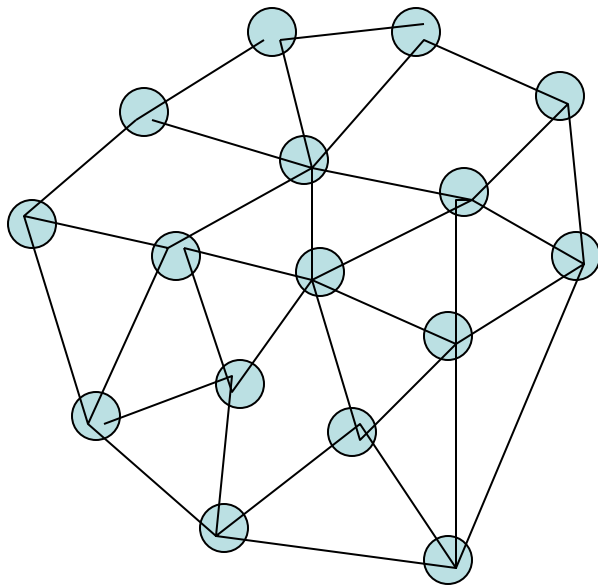
Not a connected graph

- Two nodes are **adjacent nodes** if there is an edge that connects them.



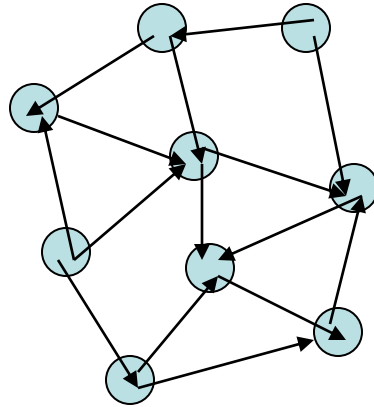
- **Neighbors of a node** are all nodes that are adjacent to it.
- A **Tree** is the special case of a graph that
 - (i) is connected
 - (ii) has no cycle

➤ *If a connected graph has n nodes and $n-1$ edges, then it is a tree. This tree is called **Spanning Tree**.*

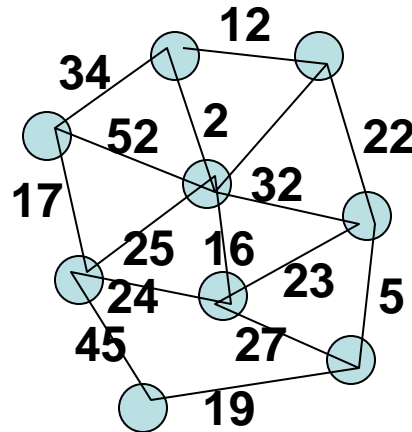


➤ *An **oriented tree** is a tree in which one node has been designated the root node.*

- A *directed graph* or *digraph* is a graph in which each edge has an associated direction.



- A *weighted graph* is a graph in which each edge has an associated value.



Specification of Graph

***Elements:** A graph consists of nodes and edges*

***Structure:** An edge is a one-to-one relationship between a pair of distinct nodes. A pair of nodes can be connected by at most one edge, but any node can be connected to any collection of other nodes.*

***Domain:** The number of nodes (vertices) in a graph is bounded.*

Operations:

***InsertNode** (Type e)*

***Requires:** **G** is not full.*

***Results:** If **G** does not contain an element whose key value is e.key then **e** is inserted in **G** and inserted is true, otherwise inserted is false.*

InsertEdge (Key k_1 , k_2)

Requires: G is not full and $k_1 \neq k_2$.

*Results: If G contains two nodes whose key values are k_1 and k_2 then G contains an edge connecting those nodes. If the two nodes were connected by an edge before operation *InsertEdge* then inserted is false; otherwise inserted is true.*

DeleteNode (Key k)

Results: G does not contain an element whose key value is k . if G contained a node before this operation with key value k then deleted is true and no edge that connected this node to an other node is in G ; otherwise deleted is false.

DeleteEdge (Graph g , Key k_1 , k_2)

Results: G does not contain an edge that connects nodes whose key values are k_1 and k_2 . If G contained such an edge before this operation then deleted is true; otherwise deleted is false.

Update (T e)

*Results: If **G** contained a node with key value e.key then the element in the node is e and updated is true. Otherwise updated is false.*

Retrieve (key k, T e)

*Results: If **G** contains a node whose key value is k before this operation then e is that element and retrieved is true; otherwise retrieved is false.*

Full ()

*Results: If **G** is full then Full returns true; otherwise Full returns false.*

Representation of a Graph

➤ *There are two approaches to representing graphs*

(i) Adjacency Matrix

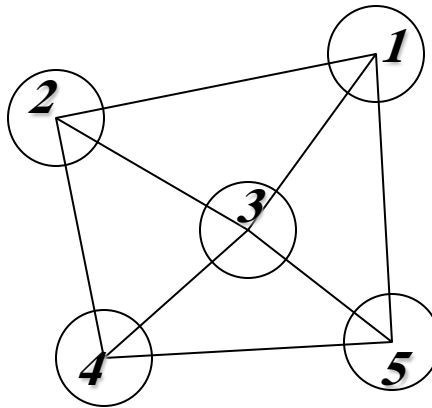
(ii) Adjacency List

Adjacency matrix

A two dimensional array whose components are of type Boolean and whose index values correspond to the nodes.

EX.

	1	2	3	4	5
1	0	1	1	0	1
2	1	0	1	1	0
3	1	1	0	1	1
4	0	1	1	0	1
5	1	0	1	1	0



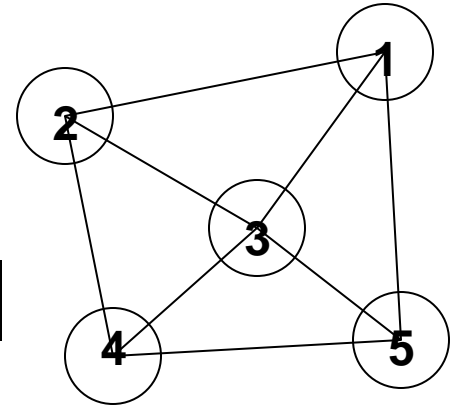
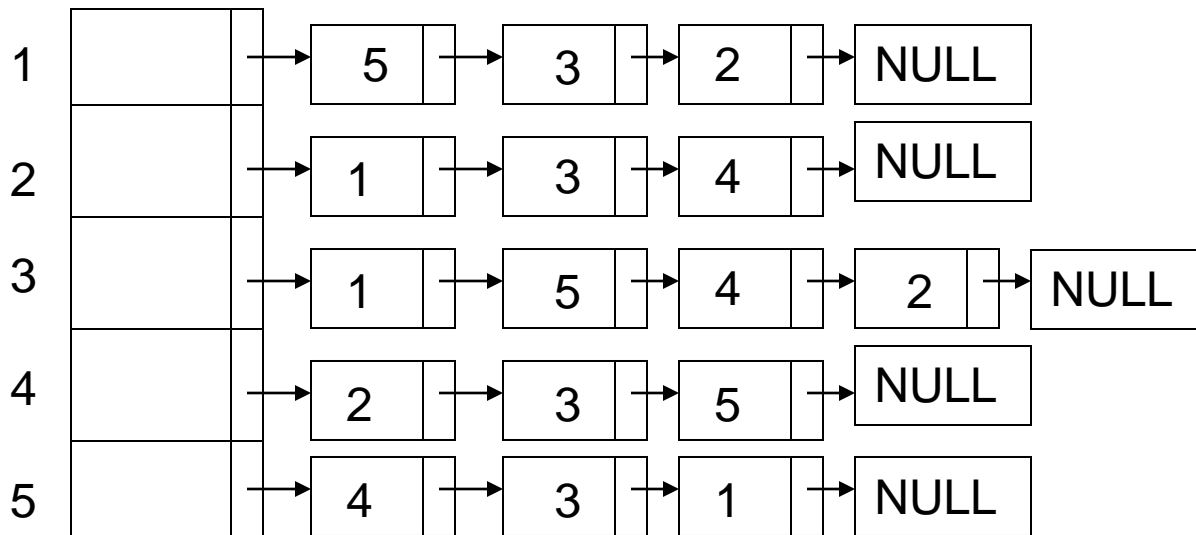
If A is the adjacency matrix corresponding to a graph G , then A_{ij} corresponds to the edge (i, j) , and is true if there is an edge between the vertices i , and j .

- *This representation is very simple, but the space requirement is $O(n^2)$ if the number of vertices is n .*
- *This representation is suitable only if the graph is dense with the number of edges i.e. $O(n^2)$, n being the number of vertices. But in most of the applications this is not true.*

Adjacency list

It is an array where each cell corresponds to a vertex of a graph and stores the header of a list of all adjacent vertices.

Ex. The following is an adjacency list corresponding to the graph on the right.



- *This is an ideal representation if the graph is not dense.*
- *In this case the space requirement is $O(e + n)$ where e is the number of edges and n is the number of vertices.*

Traversal of a Graph

Process each node of the graph only once

➤ *There are two methods for graph traversal*

*(i) **Breadth First Search***

*(ii) **Depth First Search***

Breadth First Search (BFS)

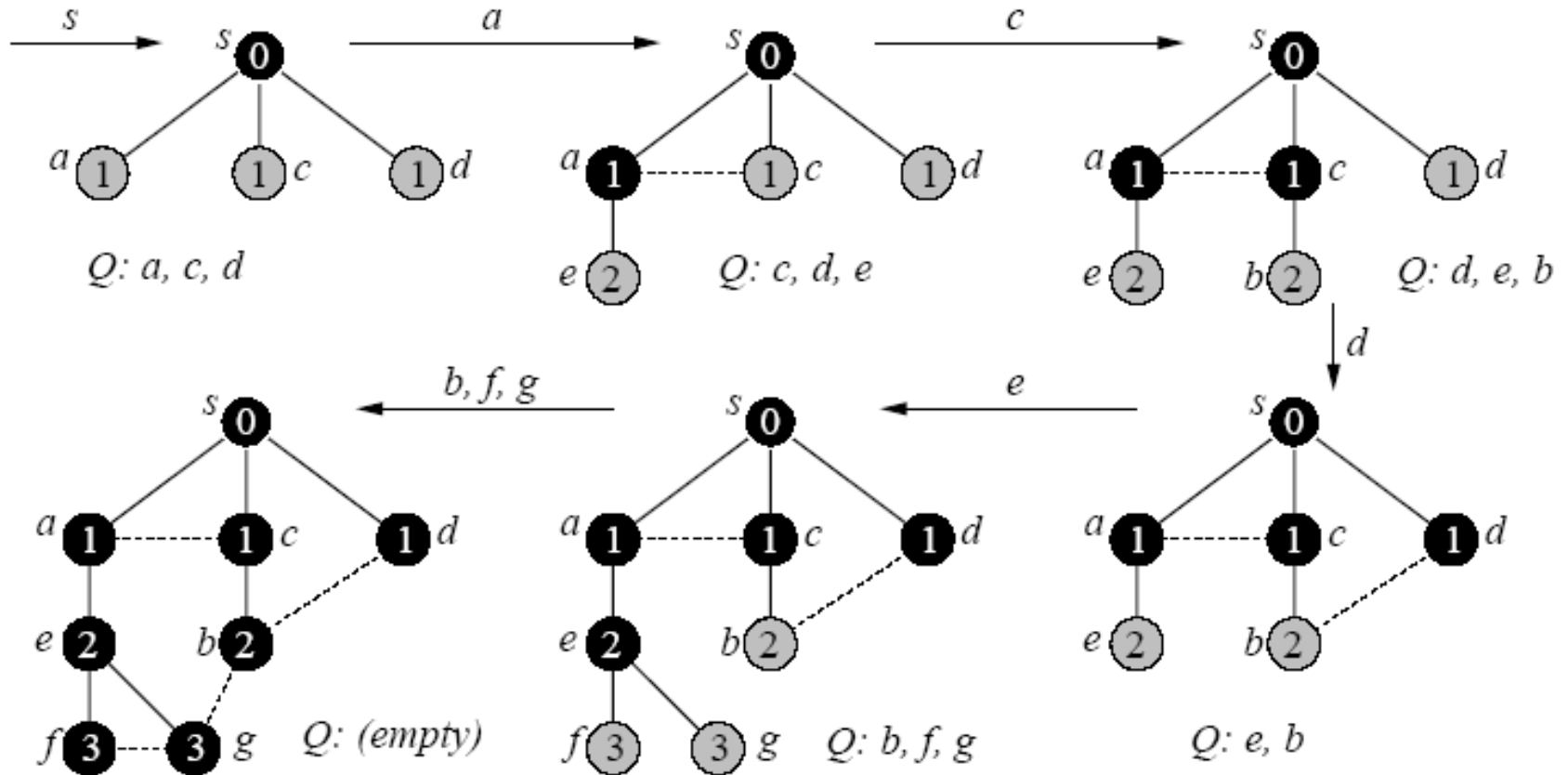
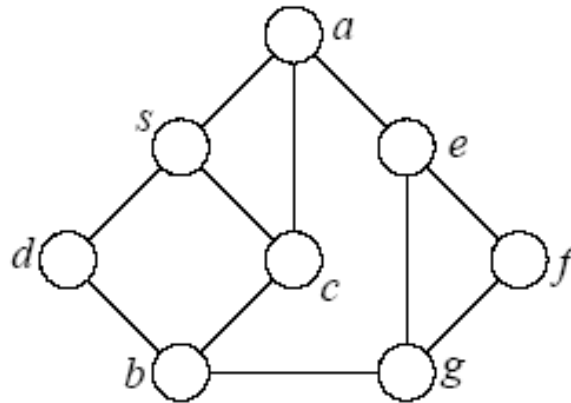
*Start at some source node **s** and visit all its neighbors, then visit the neighbors of each neighbor of s and so on.*

Algorithm

- 1. Assign the status of **waiting** to all nodes of a graph.*
- 2. Start with source node s, and put it in queue*
- 3. Dequeue one node from the queue, assign it the status of **processed**, and assign its neighbors the status of **ready** and put them in the queue.*
- 4. Repeat Step 3 until the queue is empty.*

Example

Consider a graph
with nodes *a, b, c, d,*
e, f, g, s



Pseudocode for BFS

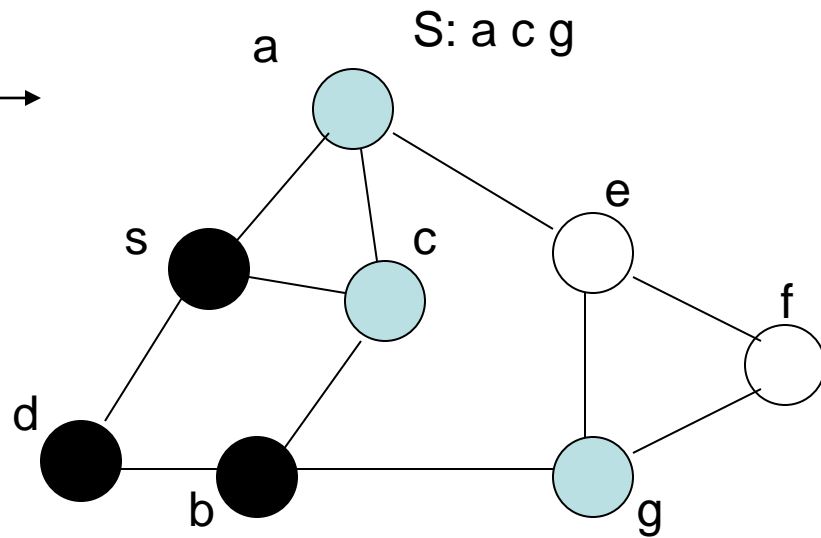
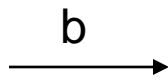
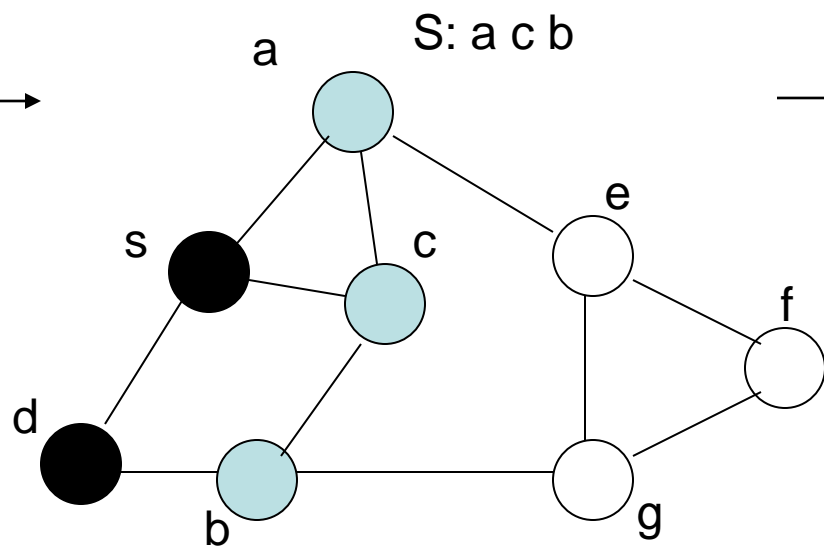
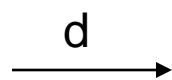
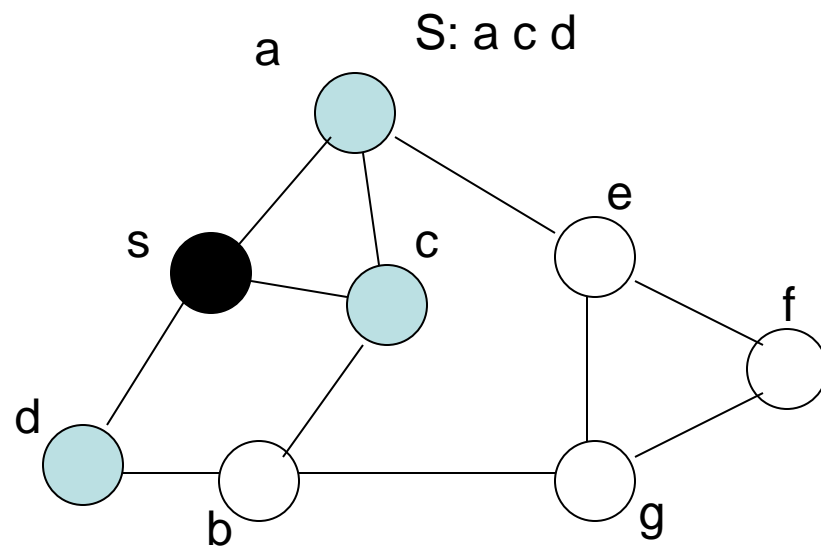
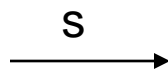
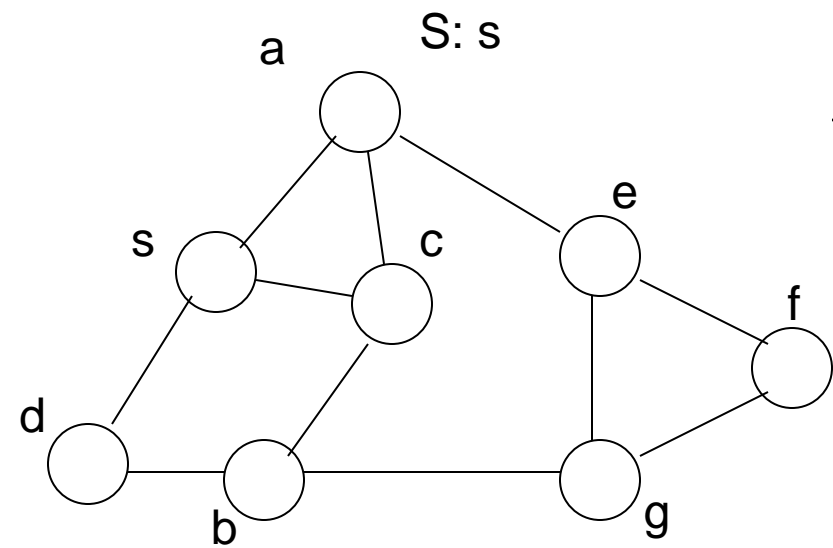
BFS (G, s)

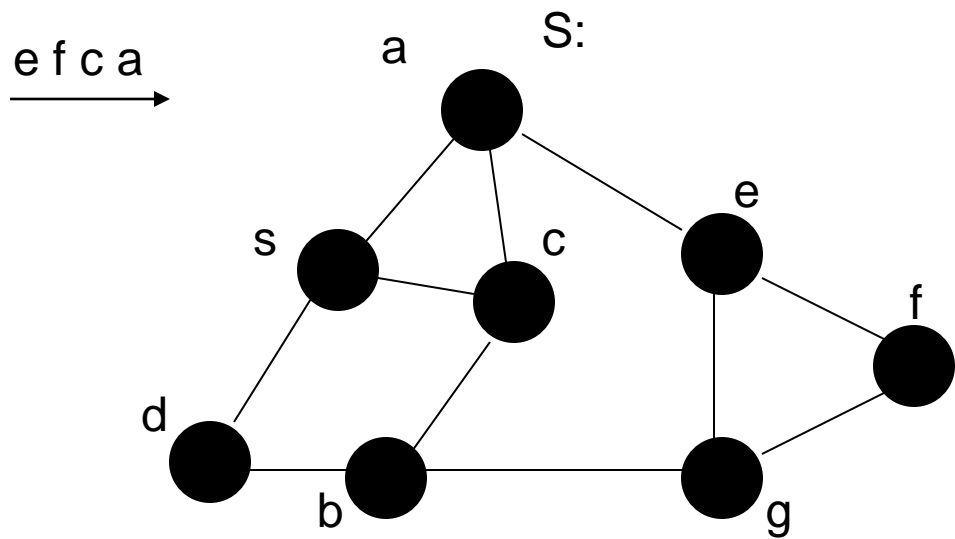
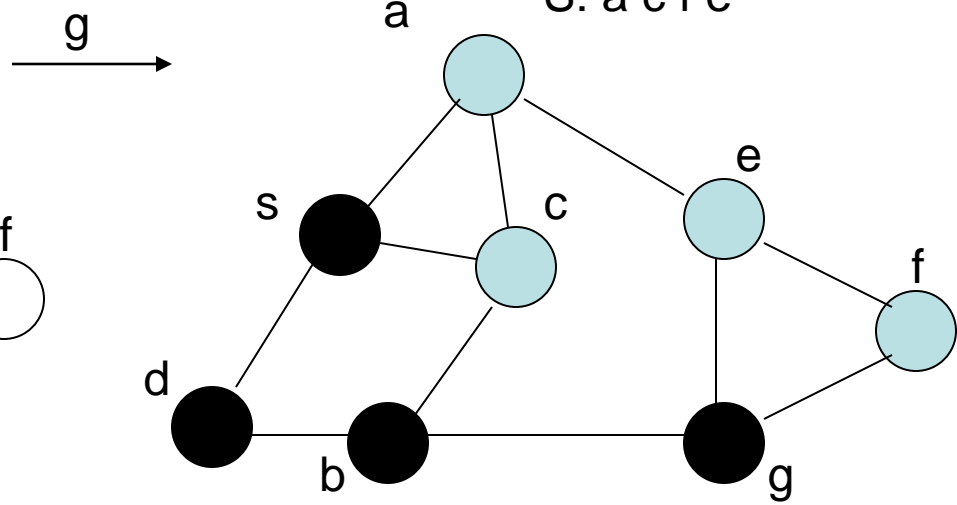
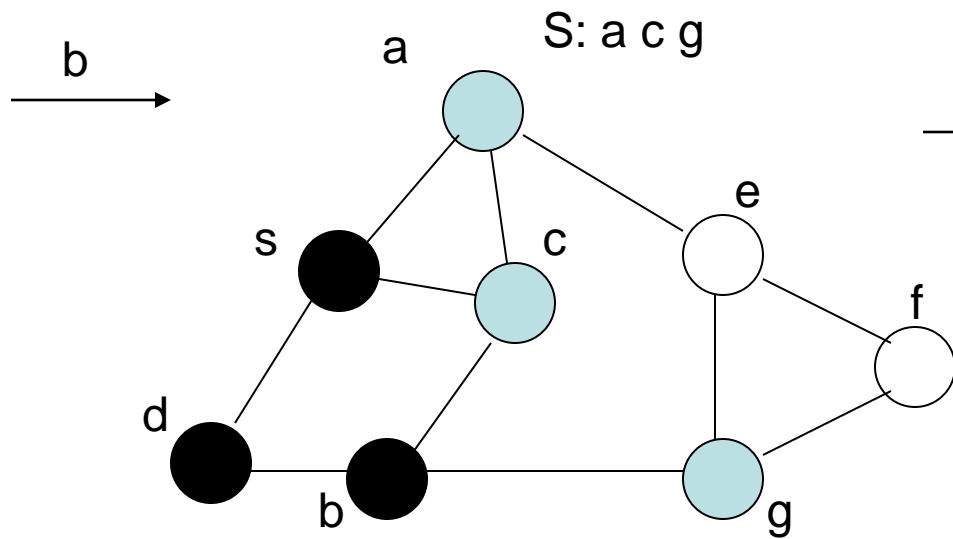
```
{  
    for each node u of G  
        set status[u] = waiting;  
    status[s] = ready;  
    Q = {s};  
    while( !Q.IsEmpty)  
    {  
        u = Q.dequeue();  
        for each v in Neigh[u]  
        {  
            if(status[v] = waiting)  
            {  
                status[v] = ready;  
                Q.enqueue(v);  
            }  
        }  
        status[u] = processed;  
    }  
}
```

Depth First Search (DFS)

DFS visits one node and then visits one of its neighbors and puts the rest of its neighbors into a stack and so on.

- 1. Assign the status of **waiting** to all nodes of a graph.*
- 2. For each node of the graph, follow the following steps,*
 - a) Change the status of the node to be **ready** and put in a stack*
 - b) Repeat the following steps until the stack is empty*
 - c) Get a node from the stack, **process** it, change its status to **processed** and change the status from waiting to ready of its neighbors and put them in the stack*
 - d) Go to step b*





Pseudocode for DFS

BFS (G)

```
{
    for each node u of G
        status[u] = waiting;
    for each node u of G
        if(status[u] == waiting)
            Visit(u)
}
```

Visit(u)

```
{
    stack S;
    status[u] = ready;
    S.push(u)
    while( S.IsEmpty)
    {
        v = S.pop();
        for each w in Neigh[v]
        {
            if(status[w] = waiting)
            {
                status[w] = ready;
                S.push(w);
            }
        }
        status[v] = processed;
    }
}
```

Shortest Path (unweighted)

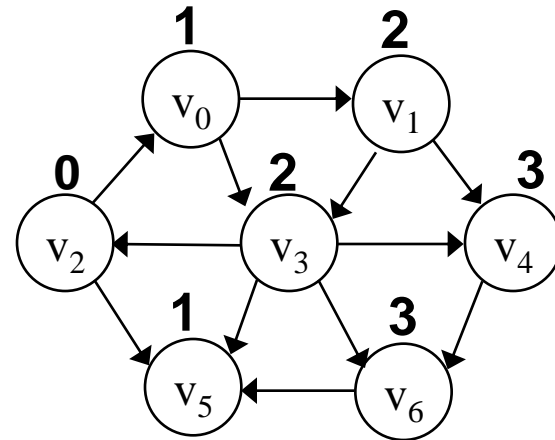
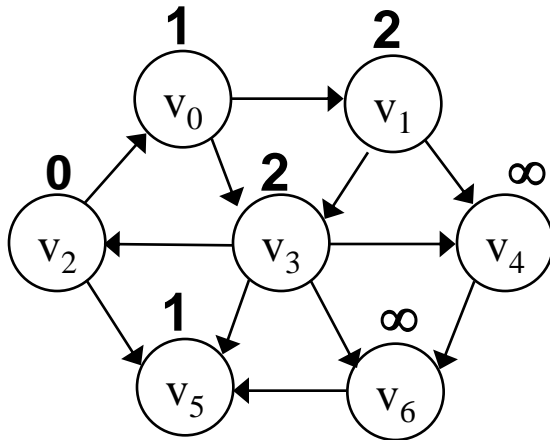
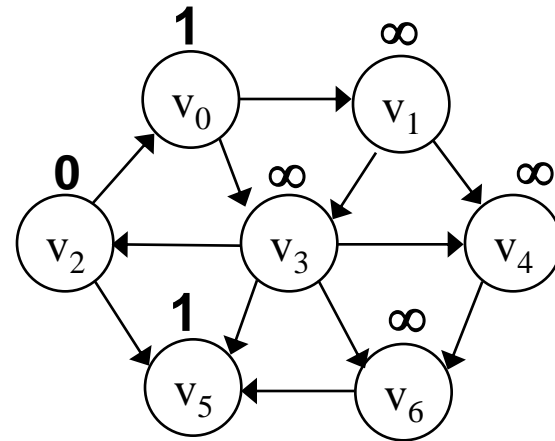
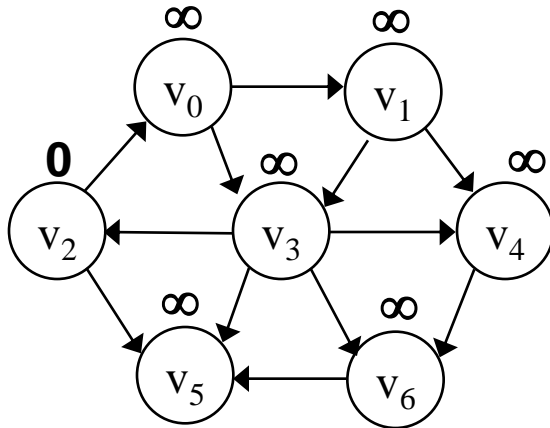
- ❖ *The problem: Find the shortest path from a vertex v to every other vertex in a graph G*
- ❖ *The unweighted path measures the number of edges, ignoring the edge's weights (if any)*

Finding Shortest Unweighted Path (Simple Algorithm)

For a vertex v , d_v is the distance between a starting vertex and v

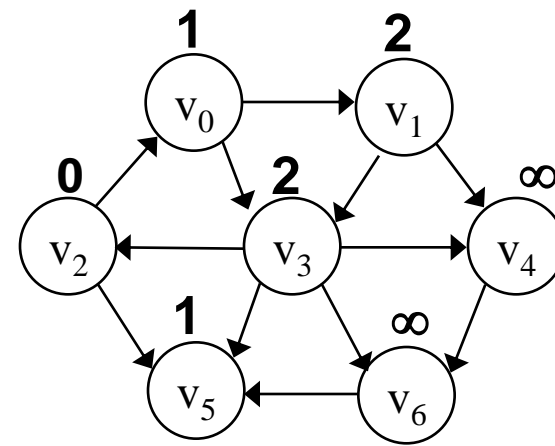
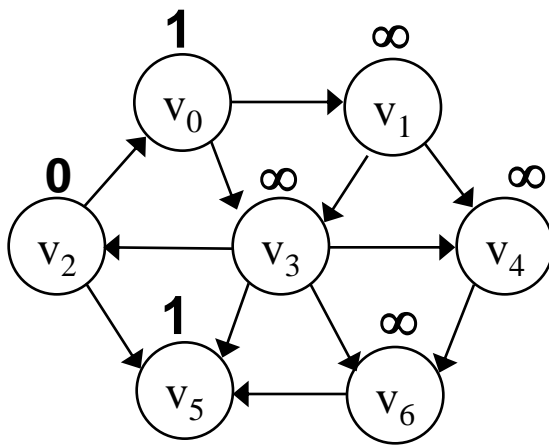
- 1 Mark all vertices with $d_v = \text{infinity}$*
- 2 Select a starting vertex s , and set $d_s = 0$, and set $\text{shortest} = 0$*
- 3 For all vertices v with $d_v = \text{shortest}$, scan their adjacency lists for vertices w where d_w is infinity
 - For each such vertex w , set d_w to $\text{shortest}+1$**
- 4 Increment shortest and repeat step 3, until there are no vertices w*

Simple algorithm: example



Simple algorithm: using queues

- ❖ *In step 3 of the simple algorithm, we search through the entire graph for vertices v with $d_v = \text{shortest}$*
- ❖ *Instead, because we deal with exactly these vertices in the previous iteration of step 3, we can queue them*



Shortest weighted path: Dijkstra's (Dike-Struh) algorithm

- ❖ *What is the lowest-cost path between two vertices in a weighted graph?*
- ❖ *Only works on graphs all of whose weights are positive*
- ❖ *Works in a similar way to the unweighted algorithm, except we possibly adjust the values in the vertices more than once*

Dijkstra's (Dike-Struh) Algorithm

Initialise all costs to infinity

Select a start vertex

current_v = start vertex

Repeat

*For every adjacent vertex **v** of **current_v***

*If (cost of **current_v** + weight of edge to **v**) < cost of **v***

*Reduce cost of **v** to (cost of **current_v** + weight of edge **v**)*

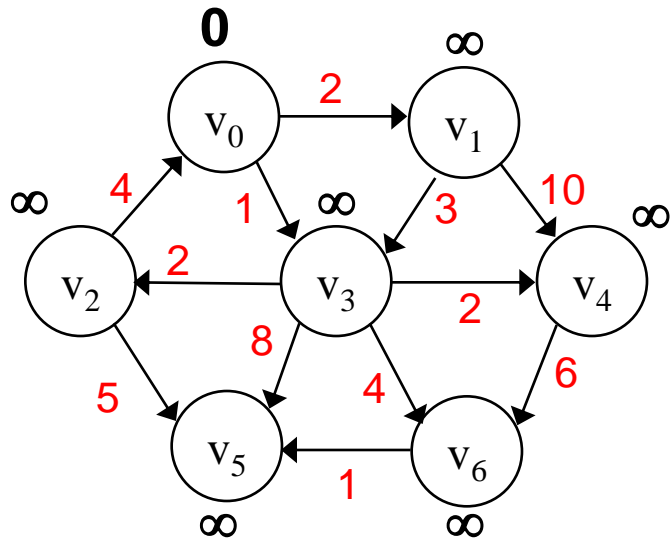
*Select lowest cost vertex as **current_v** and examine adjacent vertices*

Until all vertices have been processed

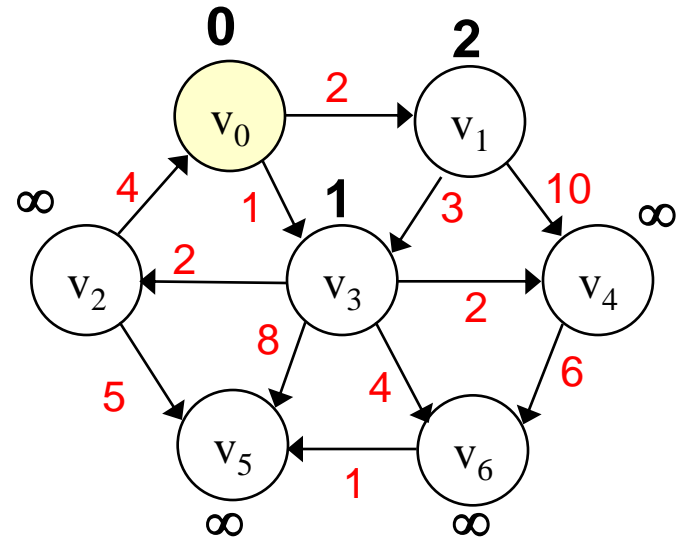
NOTE:

- ❖ *We can use an 'inverse' priority queue, in order to select the vertex with lowest cost*
- ❖ *Algorithm based on breadth-first traversal*

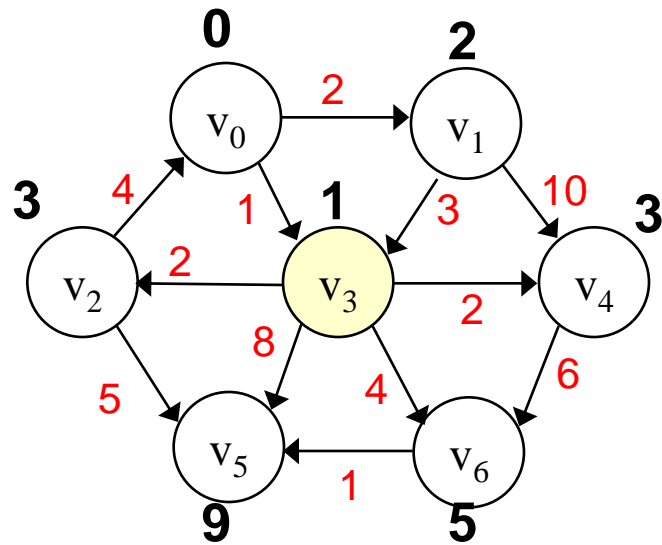
Example



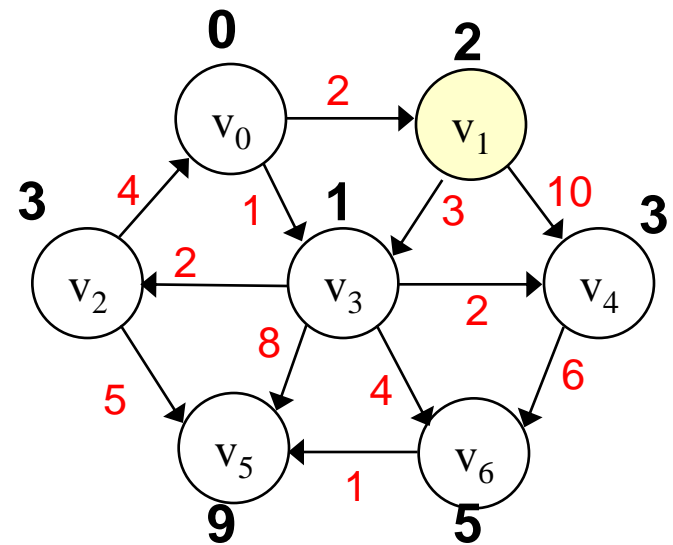
Queue: v_0



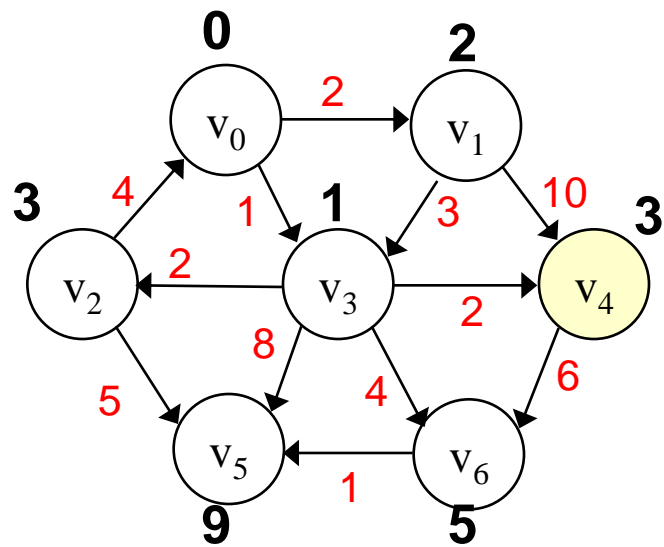
Queue: v_3, v_1



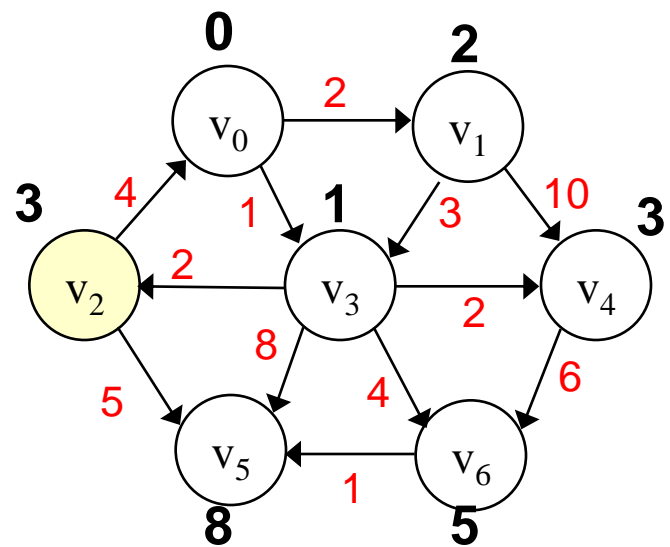
Queue: v_1, v_4, v_2, v_6, v_5



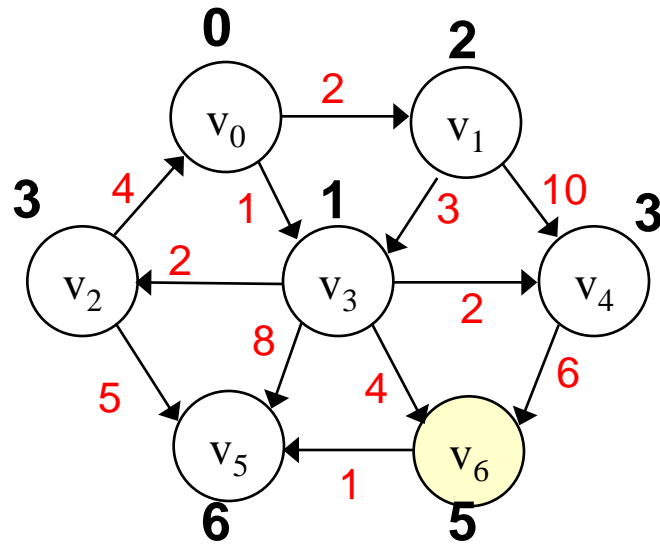
Queue: v_4, v_2, v_6, v_5



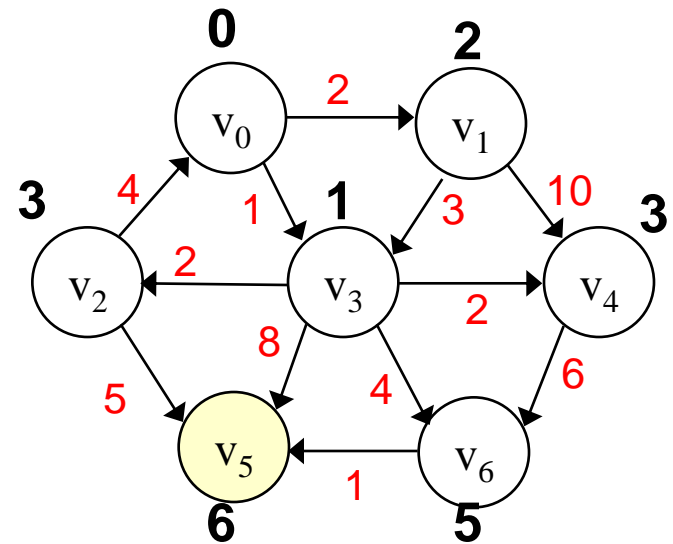
Queue: v_2, v_6, v_5



Queue: v_6, v_5



Queue: v_5



Queue:

The shortest path from v_0 to v_6 is $v_0v_3v_6$ and the shortest path from v_0 to v_4 is $v_0v_3v_4$

Complexity of Dijkstra's algorithm

- The time complexity is $O(|V|^2 + |E|)$
- For dense graphs where $|E| = O(|V|^2)$, the algorithm is optimal and the time complexity is $O(|V|^2)$
- For sparse graphs, it can be improved to give a time complexity of $O(|E| \log |V|)$