

Data Structures

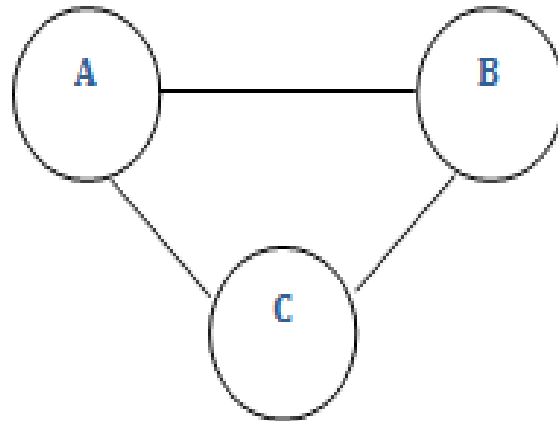
UNIT-IV

- **UNIT-4: Graphs:** Recap of Graphs, Connected Components, Bi connected components,
- **Minimum Cost Spanning Tree:** Prim's Algorithm (recap), Kruskal's Algorithm (recap), Sollin's (Boruvka's) Algorithm
- **Single Source Shortest Path:** Dijkstra's Algorithm, Bellman Ford Algorithm
- **Transitive Closure :** Warshall's Algorithm
- **All Pairs Shortest Path:** Floyd's Algorithm

- A Graph G consists of a set V of vertices (Nodes) and a set of edges (arcs) E . We write $G = (V, E)$.
- V is a finite and non empty set of vertices.
- E is a set of pairs of vertices called edges.
- An edge $e = (v, w)$ is a pair of vertices v and w , and is said to be incident with v and w . Nodes v and w are called as endpoints of e , and v and w are said to be adjacent nodes or neighbors.

Graphs

- Ex: An example graph G



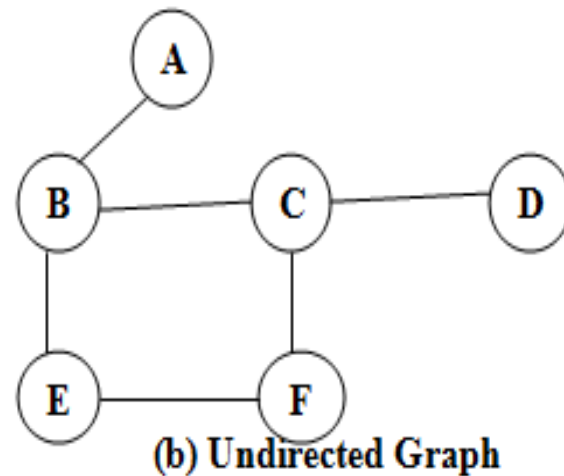
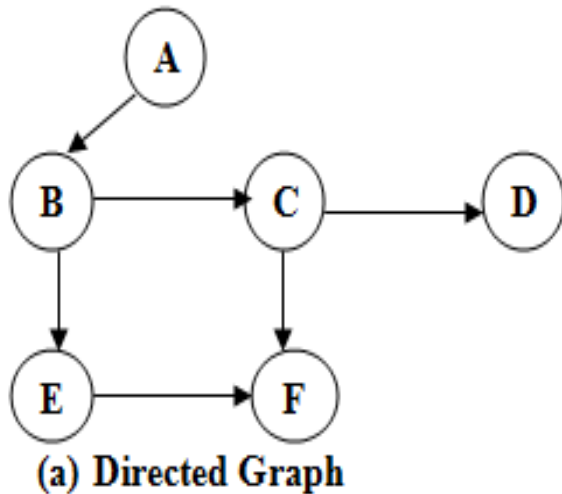
- $V(G) = \{ A, B, C \}$
- $E(G) = \{ (A, B) (A, C) (B, C) \}$

Graph VS Tree

Graph	Tree
Non-linear data structure.	Non-linear data structure.
Collection of vertices and edges.	Collection of vertices and edges.
Each node can have any number of edges.	In binary trees every node can have at the most 2 child nodes.
There is no unique node like trees.	There is a unique node called root node.
A cycle can be formed.	There will not be any cycle.
Applications: used for finding shortest path in networking graph.	Applications: used for expression trees, and decision tree.

Types of Graphs

- **Directed graphs** have edges with direction.
- **Undirected graphs** have edges that do not have a direction.



- A **path** is a sequence of vertices in which each vertex is adjacent to the next one. In directed graph $\{A, B, C, F\}$ is one path and $\{A, B, E, F\}$ is another.
- Two vertices in a graph are said to be **adjacent** vertices (or) neighbors if there is a path of length 1 connecting them.
- A **cycle** is a path consisting of at least three vertices that starts and ends with the same vertex.
- A **loop** is a special case of a cycle in which a single arc begins and ends with the same vertex.

Representation of Graphs

- There are two representation methods for graphs.
 - 1. Adjacency Matrix
 - 2. Adjacency Lists
- **Adjacency Matrix:** The Adjacency Matrix A for a graph $G = (V, E)$ with n vertices, is an $n \times n$ Matrix of bits, such that A

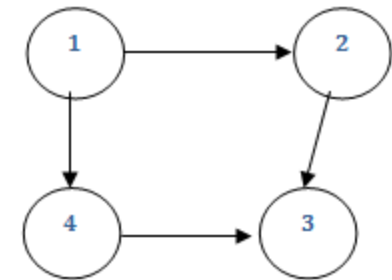
$$A_{ij} = \begin{cases} 1 & \text{If there is an edge } V_i \text{ to } V_j \text{ (un weighted graph)} \\ 0 & \text{If not} \end{cases}$$

- In case of weighted graph we are measuring a weight instead of 1.

Representation of Graphs

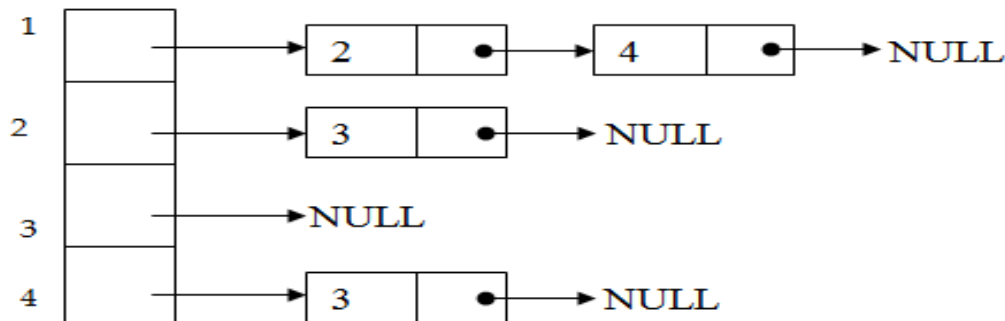
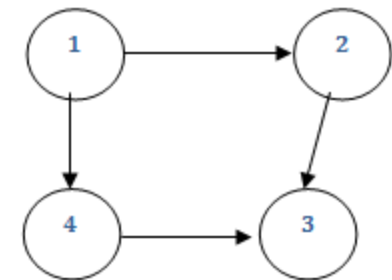
The Adjacency Matrix would be

$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$



Representation of Graphs

- Adjacency List Representation: In this representation, we store a graph as a linked structure. We store all the vertices in a list and then for each vertex, we have a linked list of the adjacent vertices.
- Ex: Let us consider a Graph

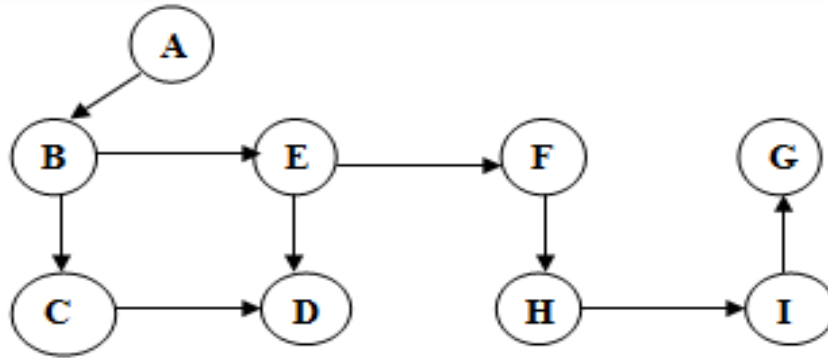


Operations on Graphs

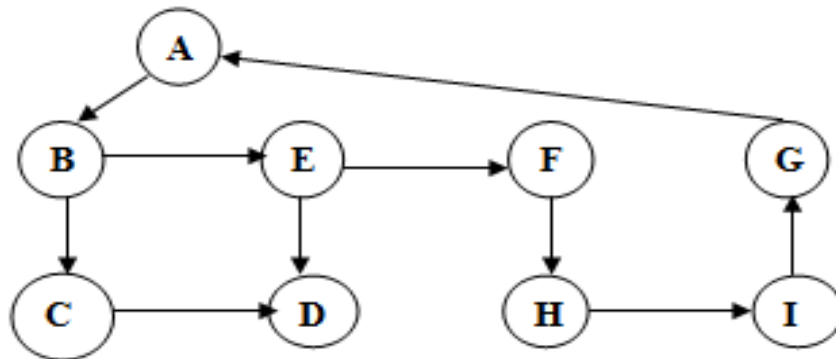
- Insert vertex.
- Delete vertex.
- Insert edge.
- Delete edge.
- Find vertex.

Types of Graphs

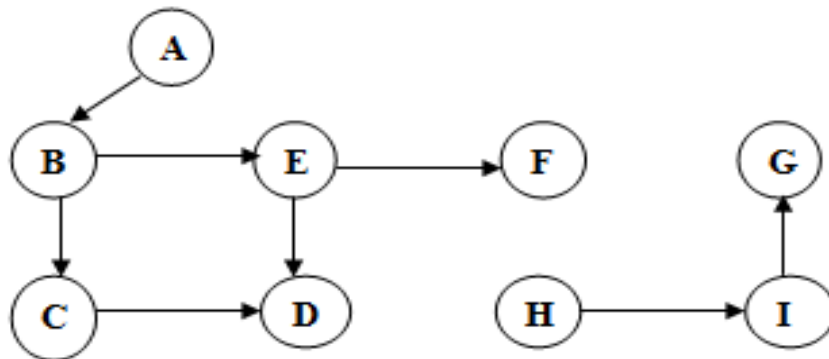
- A directed graph is **strongly connected** if there is a path from each vertex to every other vertex in the digraph.
- A directed graph is **weakly connected** if at least two vertices are not connected.
- A graph is **disconnected** if at least two vertices of the graph are not connected by a path. If a graph G is disconnected, then every maximal connected subgraph of G is called a connected component of the graph G .



(d) Weakly Connected



(e) Strongly Connected



(f) Disjoint Graph

Graph Traversals

Depth first Search

Void DFS(Vertex V)

{

visited of all vertices initially false

push v to stack;

while stack not empty

 v=pop();

 if (!visited[v])

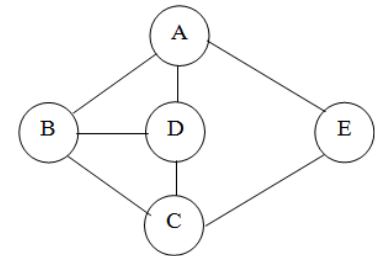
 visited[v]=true;

 for each Vertex W adjacent to V

 if(!Visited[W])

 push(W);

}



Breadth first Search

Void BFS(Vertex V)

{

visited of all vertices initially false

initialize Q to be Empty //Q is a Queue

Enqueue(V);

visited[V]=true;

while(Q not empty)

{

V=Dequeue()

for each vertex W adjacent to V

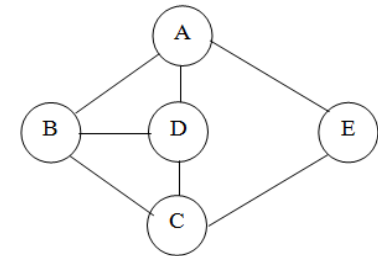
if(visited[W] = false)

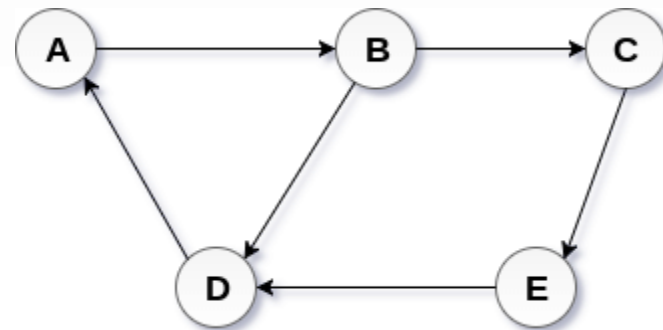
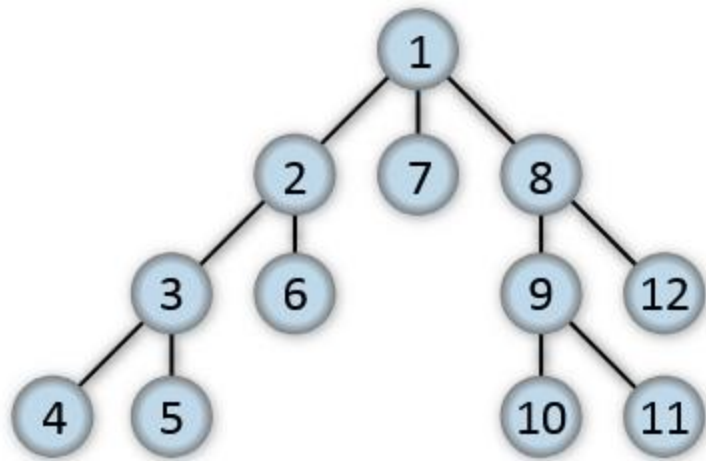
{

Enqueue(W);

Visited[W] = True;

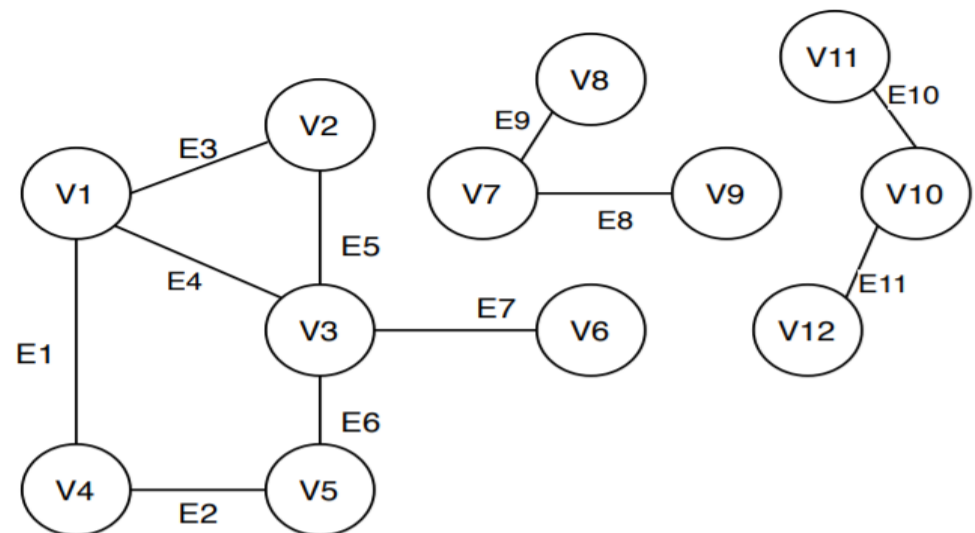
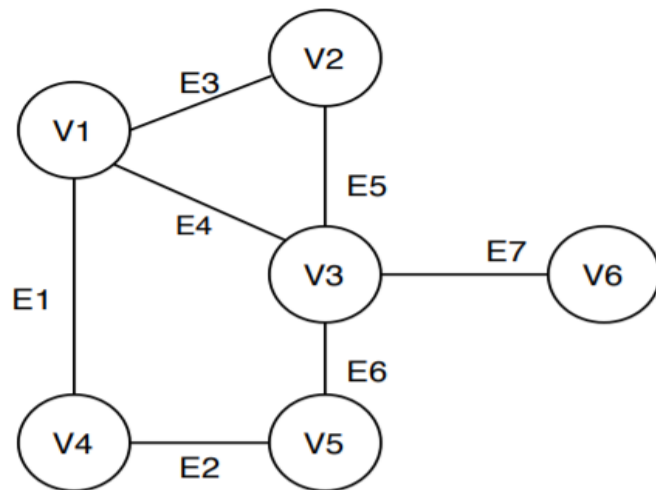
} }





Connected Components

- A connected component or simply component of an undirected graph is a subgraph in which each pair of nodes is connected with each other via a path.
- A set of nodes forms a connected component in an undirected graph if any node from the set of nodes can reach any other node by traversing edges.



Connected Components

Algorithm 1: Finding Connected Components using DFS

Data: Given an undirected graph $G(V, E)$

Result: Number of Connected Components

Component_Count = 0;

for each vertex $k \in V$ **do**

 | Visited[k] = False;

end

for each vertex $k \in V$ **do**

 | **if** Visited[k] == False **then**

 | DFS(V,k);

 | Component_Count = Component_Count + 1;

 | **end**

end

Print Component_Count;

Procedure DFS(V,k)

Visited[k] = True;

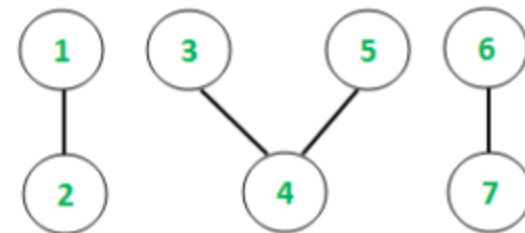
for each vertex $p \in V.Adj[k]$ **do**

 | **if** Visited[p] == False **then**

 | DFS(V,p);

 | **end**

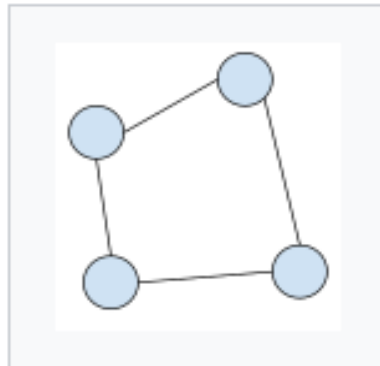
end



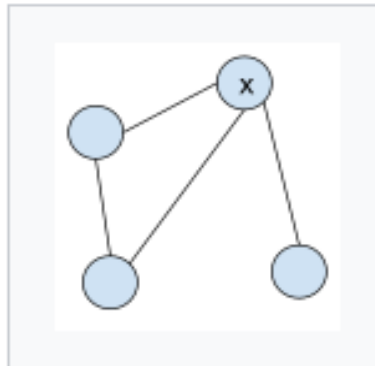
BiConnected Components

3A105

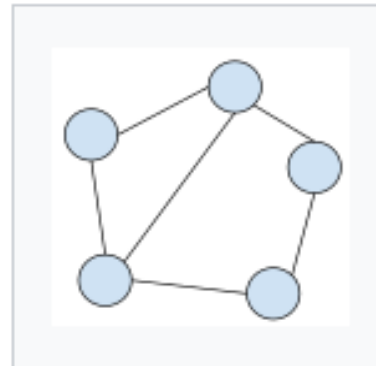
- A graph is said to be Biconnected if It is connected, i.e. it is possible to reach every vertex from every other vertex, by a simple path, even after removing any vertex the graph remains connected.



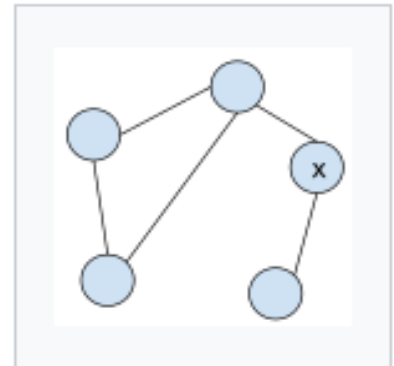
A biconnected graph on four vertices and four edges



A graph that is not biconnected. The removal of vertex x would disconnect the graph.



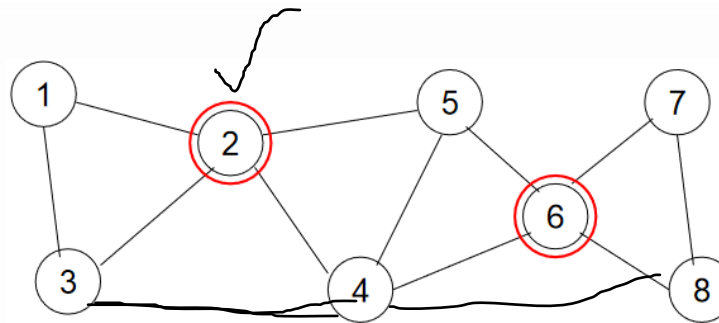
A biconnected graph on five vertices and six edges



A graph that is not biconnected. The removal of vertex x would disconnect the graph.

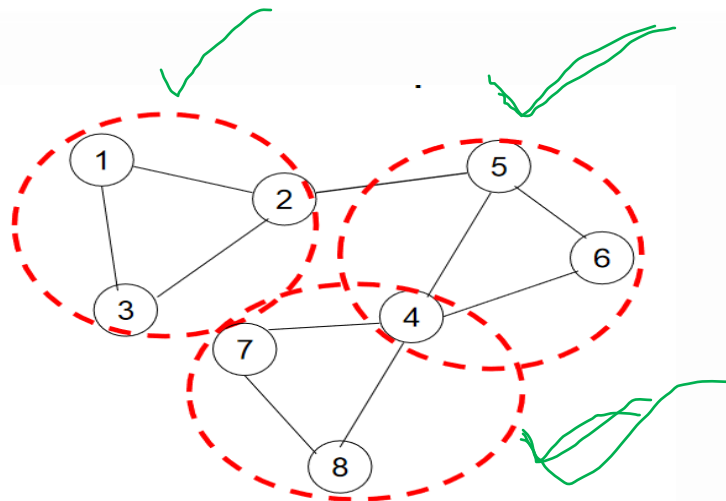
Articulation Points

- An articulation point of a graph is a vertex v such that the removal of v will not break the graph into two or more pieces.
- A connected graph with no articulation points is said to be biconnected.



BiConnected Components

- *Biconnected components*: A biconnected component of a graph is a maximal biconnected subgraph. Two bi-coonected componets can have one vertex in common that point is articulation point. Hence no edge can be in two bi-connected components.
- ✓ Depth-first search is particularly useful in finding the biconnected components of a graph.



Algorithm for finding Articulation Points

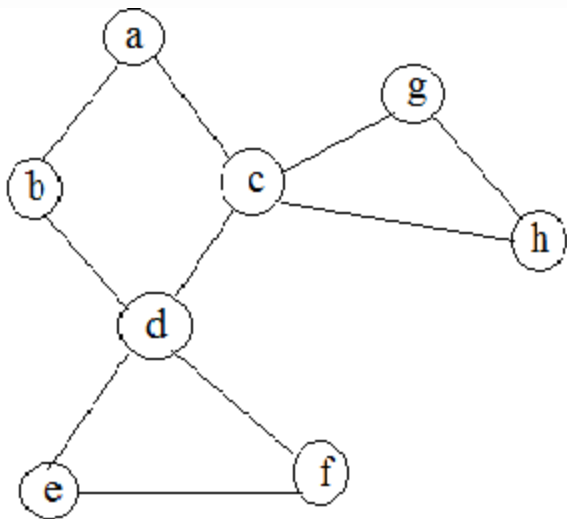
- **Step 1:** Find DFS and assign discovery number(dfn) to every vertex.
- **Step 2:** Find Low() for every vertex.
- **Step 3:** Check if $dfn(u) \leq Low(v)$ [u parent, v child] then u is articulation point.

dfn: The number at which the vertex is discovered in DFS

BackEdges: are the edges which are present in the original graph but not present in the DFS Spanning tree

Low(): The lowest discovery number that can be reached from the current vertex by taking one back edge

$Low(u) = \min(dfn \text{ of } u, \min \text{ of lowest value of children, min of dfn value of back edges})$



Vertex	h	g	e					
dfn	2	7	6					
Low	2	2	1					

Spanning Tree

SPANNING TREES:

- A *spanning tree* of a graph is just a subgraph that contains all the vertices and is a tree. A graph may have many spanning trees.

Minimum cost spanning tree

- Minimum cost spanning tree of a weight connected graph is a spanning tree with minimum cost or weight.
- There are three ways to find minimum cost spanning trees:
 - Prim's algorithm
 - Kruskal's algorithm
 - Sollin's algorithm((Boruvka's)

Prim's Algorithm

- procedurePrim(G: weighted connected graph with n vertices)**

$T :=$ a minimum-weight edge

for $i = 1$ to $n - 2$

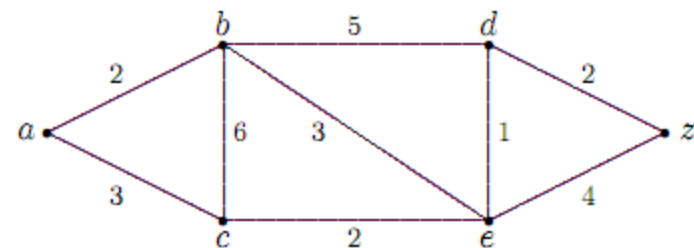
begin

$e :=$ an edge of minimum weight adjacent to a vertex in T and not forming a cycle in T if added to T

$T := T$ with e added

end

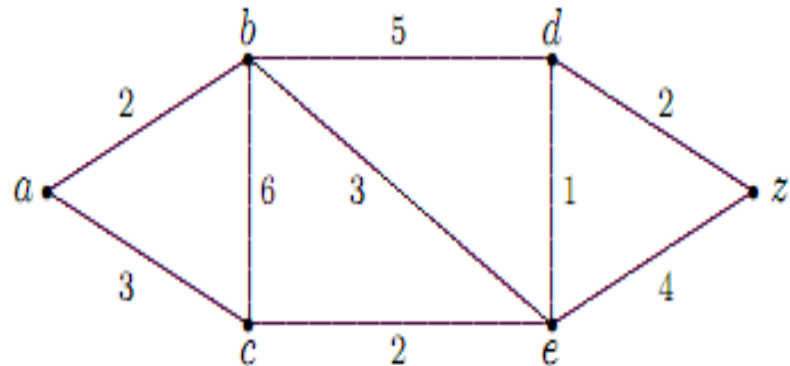
return(T).



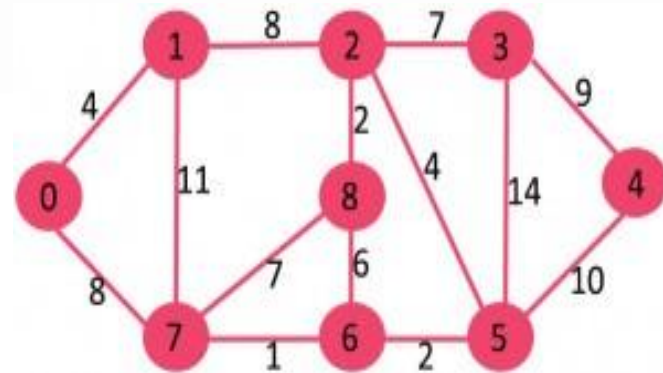
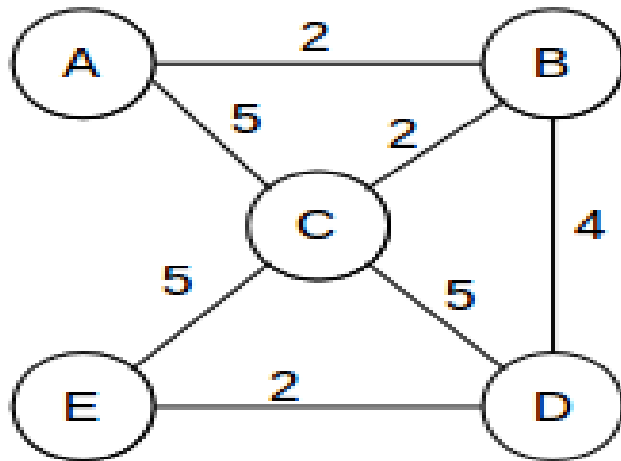
Kruskal's Algorithm

Kruskal's algorithm:

- sort the edges of G in increasing order by length
- keep a subgraph S of G , initially empty
- for each edge e in sorted order
- if the endpoints of e are disconnected in S
- add e to S
- return S



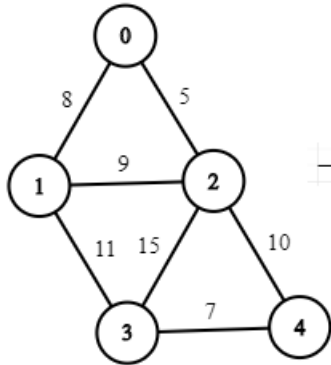
Kruskal's Algorithm



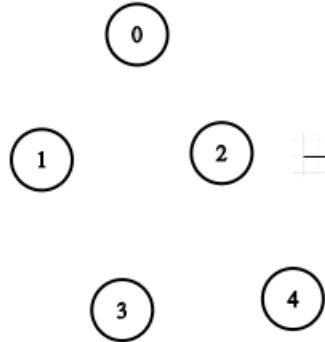
Sollin's(Boruvka's) Algorithm

- 1) Input is a connected, weighted and un-directed graph.
- 2) Initialize all vertices as individual components (or sets).
- 3) Initialize MST as empty.
- 4) While there are more than one components, do following for each component.
 - a) Find the closest weight edge that connects this component to any other component.
 - b) Add this closest edge to MST if not already added.
- 5) Return MST.

Sollin's(Boruka's) Algorithm

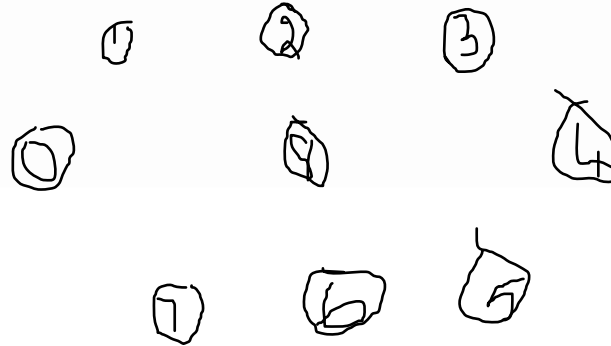
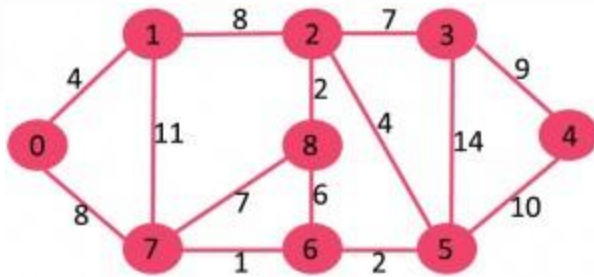


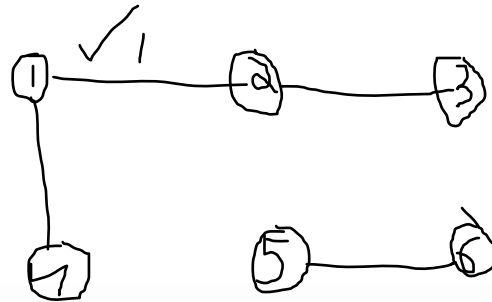
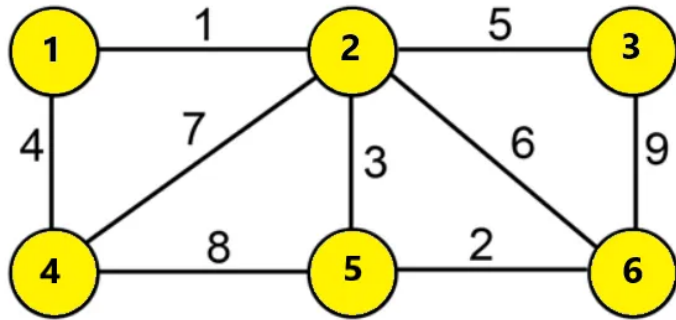
Step 0: The Graph



Step 1: Bunch of unconnected trees

Sollin's(Boruka's) Algorithm





Single-source shortest path

For a weighted graph $G=(V,E,w)$, the single source shortest path problem is to find the shortest paths from a vertex $v \in V$ to all other vertices in V .

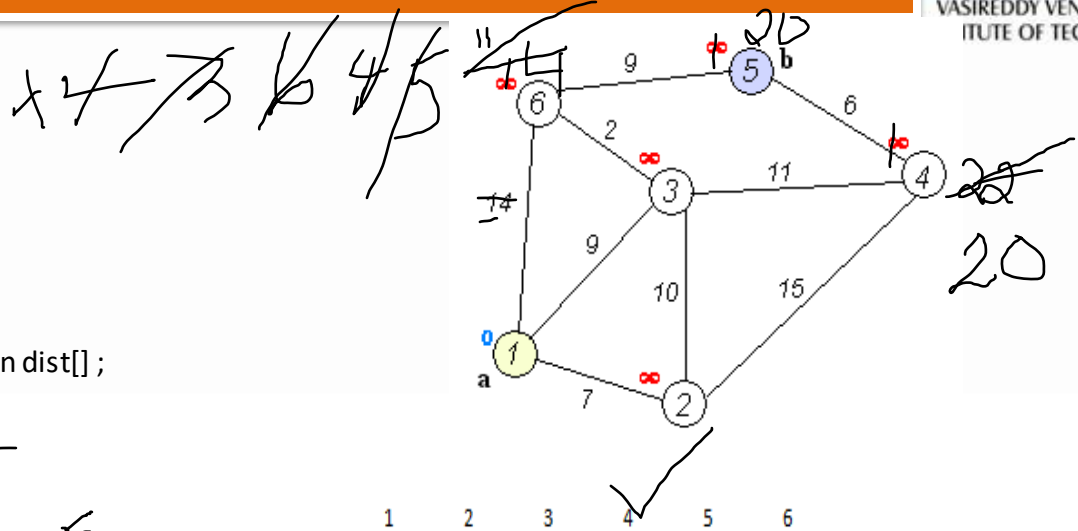
There are two algorithm to solve this problem.

- Dijkstra's algorithm
- Belleman Ford algorithm

Dijkstra's algorithm

```

function Dijkstra(Graph, source):
2  for each vertex v in Graph:
3    dist[v] := infinity;
4    previous[v] := undefined;
5  end for;
6  dist[source] := 0;
7  Q := the set of all nodes in Graph;
8  while Q is not empty
9    u := vertex in Q with smallest distance in dist[];
10   if dist[u] = infinity:
11     break;
12   end if;
13   remove u from Q;
14   for each neighbor v of u:
15     alt := dist[u] + dist_between(u, v);
16     if alt < dist[v]:
17       dist[v] := alt;
18       previous[v] := u;
19       decrease-key v in Q;
20     end if;
21   end for;
22 end while;
23 Return dist[];
24 end Dijkstra.
    
```



1	2	3	4	5	6
0	7	9	20	20	11

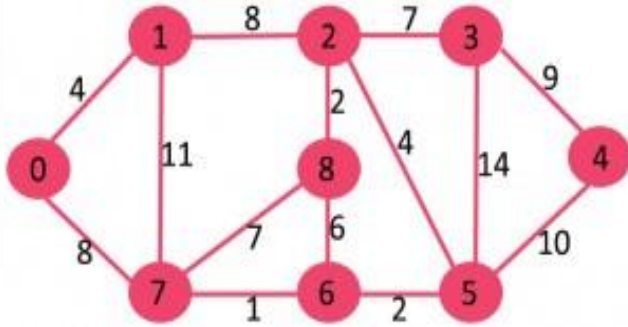
1	2	3	4	5	6
-1	1	1	3	6	3

If we are only interested in a shortest path between vertices *source* and *target*, we can terminate the search at line 13 if $u = \text{target}$. Now we can read the shortest path from *source* to *target* by iteration:

```

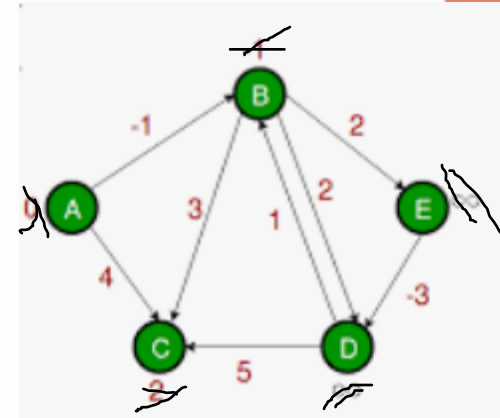
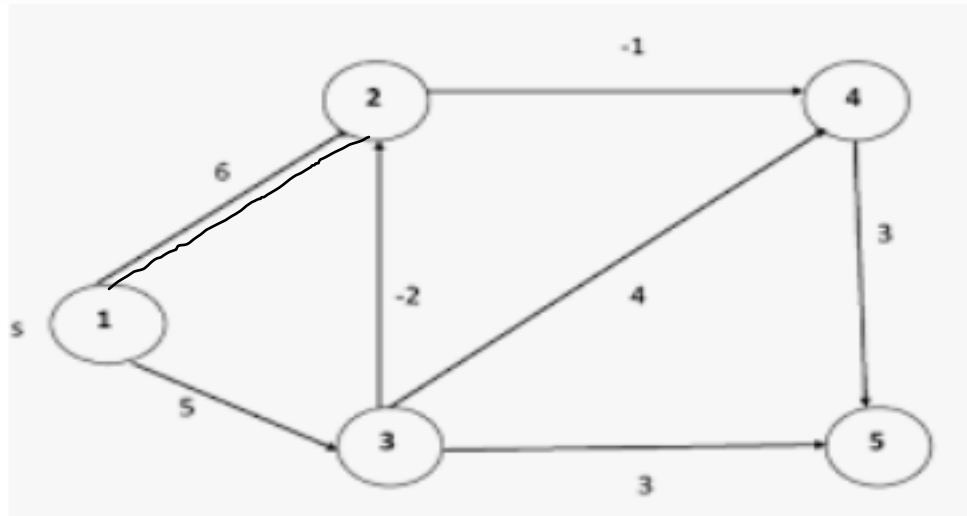
1 S := empty sequence
2 u := target
3 while previous[u] is defined:
4   insert u at the beginning of S
5   u := previous[u]
6 end while;
    
```

Dijkstra's algorithm



BellmanFord Algorithm

- **function** BellmanFord(*list* vertices, *list* edges, *vertex* source)
is
- distance := *list* of size n
- predecessor := *list* of size n
- **for each** vertex v **in** vertices **do**
 - distance[v] := **inf** predecessor[v] := null
- distance[source] := 0
- **repeat** $|V| - 1$ **times**:
 - **for each** edge (u, v) **with** weight w **in** edges **do**
 - **if** distance[u] + w < distance[v] **then**
 - distance[v] := distance[u] + w
 - predecessor[v] := u
- **for each** edge (u, v) **with** weight w **in** edges **do**
 - **if** distance[u] + w < distance[v] **then error** "Graph contains a negative-weight cycle"
- **return** distance, predecessor



A	B	C	D	E		
1	2	3	4	5	6	7
0	-1	2	-2	1		

(1 2)

vis

vis

1	2	3	4	5	6	7

Transitive closure

- Consider a directed graph $G=(V,E)$, where V is the set of vertices and E is the set of edges. The transitive closure of G is a graph $G^+ = (V,E^+)$ such that for all v,w in V there is an edge (v,w) in E^+ if and only if there is a non-null path from v to w in G .
- We have seen Adjacency Matrix, which tells you whose directly connected to whom, but there can be many other paths which may exist and might not be visible from the Adjacency matrix, as these paths may have intermediate vertices. To find all such round about or via-paths there is a procedure to be followed, known as the transitive closure, which results in a matrix showing all those via-paths, if ever exist.
- We can find the transitive closure using Floyd Warshall Algorithm

Floyd Warshall Algorithm

transColsure(graph)

Input: The given graph.

Output: Transitive Closure matrix.

Begin

copy the adjacency matrix into another matrix named transMat

for any vertex k in the graph, do

for each vertex i in the graph, do

for each vertex j in the graph, do

$\text{transMat}[i, j] := \text{transMat}[i, j] \text{ OR } (\text{transMat}[i, k]) \text{ AND } \text{transMat}[k, j]$

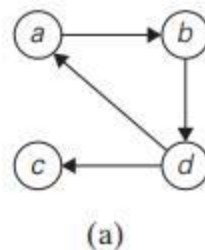
End-for

End-for

End-for

Display the transMat

End



$$A = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

(b)

$$T = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

(c)

All pairs shortest path

floydWarshal(cost)

Input – The cost matrix of given Graph.

Output – Matrix to for shortest path between any vertex to any vertex.

Begin

for k := 0 to n, do

for i := 0 to n, do

for j := 0 to n, do

if $\text{cost}[i,k] + \text{cost}[k,j] < \text{cost}[i,j]$, then

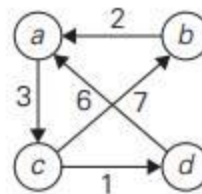
$\text{cost}[i,j] := \text{cost}[i,k] + \text{cost}[k,j]$

End-for

End-for

End-for

display the current cost matrix End



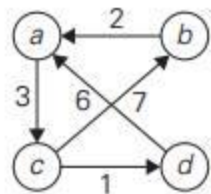
(a)

$$W = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \infty & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

(b)

$$D = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 10 & 3 & 4 \\ 2 & 0 & 5 & 6 \\ 7 & 7 & 0 & 1 \\ 6 & 16 & 9 & 0 \end{bmatrix} \end{matrix}$$

(c)



(a)

$$W = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \infty & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

(b)

$$D = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 10 & 3 & 4 \\ 2 & 0 & 5 & 6 \\ 7 & 7 & 0 & 1 \\ 6 & 16 & 9 & 0 \end{bmatrix} \end{matrix}$$

(c)