
UNIT 7 ADVANCED TREES

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7.0 INTRODUCTION

Linked list representations have great advantages of flexibility over the contiguous representation of data structures. But, they have few disadvantages also. Data structures organised as trees have a wide range of advantages in various applications and it is best suited for the problems related to information retrieval.

These data structures allow the searching, insertion and deletion of node in the ordered list to be achieved in the minimum amount of time.

The data structures that we discuss primarily in this unit are Binary Search Trees, AVL trees and B-Trees. We cover only fundamentals of these data structures in this unit. Some of these trees are special cases of other trees and Trees are having a large number of applications in real life.

7.1 OBJECTIVES

After going through this unit, you should be able to

- know the fundamentals of Binary Search trees;
 - perform different operations on the Binary Search Trees;
 - understand the concept of AVL trees;
 - understand the concept of B-trees, and
 - perform various operations on B-trees.
-

7.2 BINARY SEARCH TREES

A Binary Search Tree is a binary tree that is either empty or a node containing a key value, left child and right child.

By analysing the above definition, we note that BST comes in two variants namely empty BST and non-empty BST.

The empty BST has no further structure, while the non-empty BST has three components.

The non-empty BST satisfies the following conditions:

- a) The key in the left child of a node (if exists) is less than the key in its parent node.
- b) The key in the right child of a node (if exists) is greater than the key in its parent node.
- c) The left and right subtrees of the root are again binary search trees.

The following are some of the operations that can be performed on Binary search trees:

- Creation of an empty tree
- Traversing the BST
- Counting internal nodes (non-leaf nodes)
- Counting external nodes (leaf nodes)
- Counting total number of nodes
- Finding the height of tree
- Insertion of a new node
- Searching for an element
- Finding smallest element
- Finding largest element
- Deletion of a node.

7.2.1 Traversing a Binary Search Tree

Binary Search Tree allows three types of traversals through its nodes. They are as follow:

1. Pre Order Traversal
2. In Order Traversal
3. Post Order Traversal

In Pre Order Traversal, we perform the following three operations:

1. Visit the node
2. Traverse the left subtree in preorder
3. Traverse the right subtree in preorder

In Order Traversal, we perform the following three operations:

1. Traverse the left subtree in inorder
2. Visit the root
3. Traverse the right subtree in inorder.

In Post Order Traversal, we perform the following three operations:

1. Traverse the left subtree in postorder
2. Traverse the right subtree in postorder
3. Visit the root

Consider the BST of *Figure 7.1*

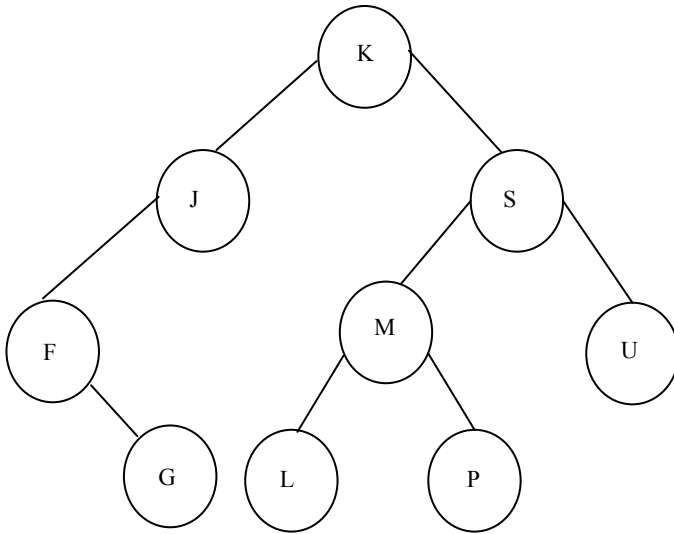


Figure 7.1: A Binary Search Tree(BST)

The following are the results of traversing the BST of *Figure 7.1*:

Preorder : K J F G S M L P U
Inorder : F G J K L M P S U
Postorder: G F J L P M U S K

7.2.2 Insertion of a node into a Binary Search Tree

A binary search tree is constructed by the repeated insertion of new nodes into a binary tree structure.

Insertion must maintain the order of the tree. The value to the left of a given node must be less than that node and value to the right must be greater.

In inserting a new node, the following two tasks are performed :

- Tree is searched to determine where the node is to be inserted.
- On completion of search, the node is inserted into the tree

Example: Consider the BST of *Figure 7.2* After insertion of a new node consisting of value 5, the BST of *Figure 7.3* results.

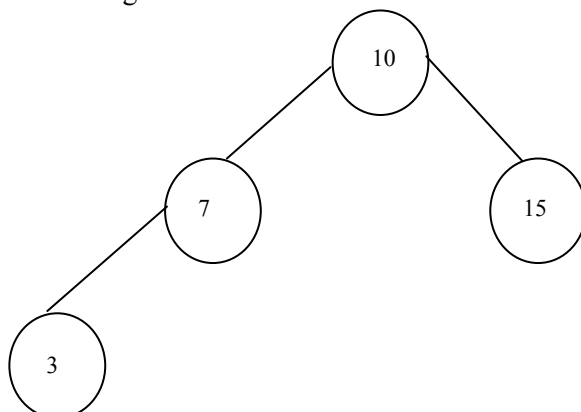


Figure 7.2: A non-empty

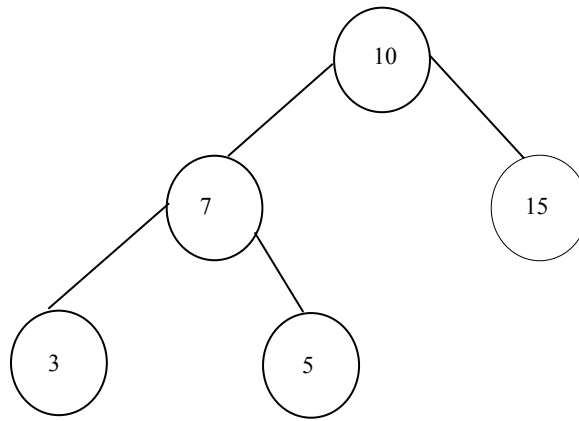


Figure 7.3: Figure 7.2 after insertion of 5

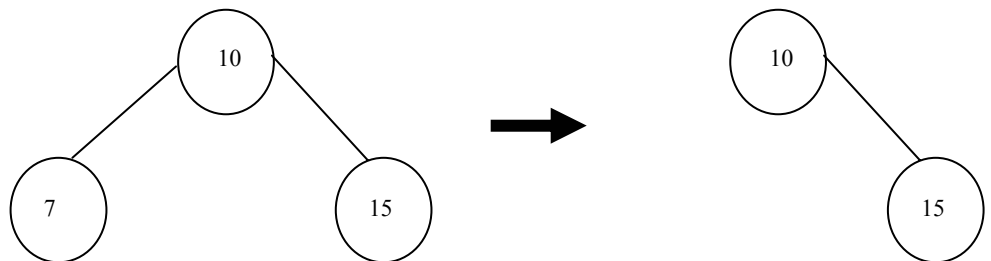
7.2.3 Deletion of a node from a Binary Search Tree

The algorithm to delete a node with key from a binary search tree is not simple where as many cases needs to be considered.

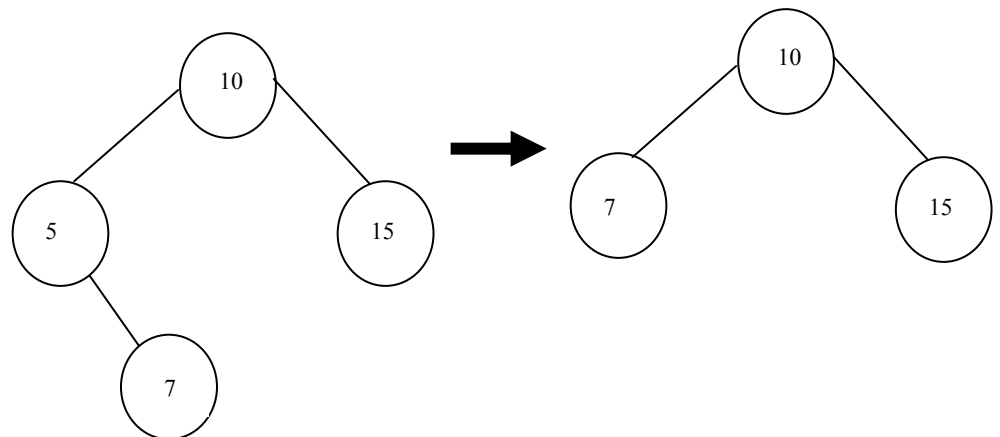
- If the node to be deleted has no sons, then it may be deleted without further adjustment to the tree.
- If the node to be deleted has only one subtree, then its only son can be moved up to take its place.
- The node p to be deleted has two subtrees, then its inorder successor s must take its place. The inorder successor cannot have a left subtree. Thus, the right son of s can be moved up to take the place of s .

Example: Consider the following cases in which node 5 needs to be deleted.

1. The node to be deleted has no children.



2. The node has one child



3. The node to be deleted has two children. This case is complex. The order of the binary tree must be kept intact.

☞ Check Your Progress 1

1) What are the different ways of traversing a Binary Search Tree?

.....
.....

2) What are the major features of a Binary Search Tree?

.....
.....

7.3 AVL TREES

An AVL tree is a binary search tree which has the following properties:

- The sub-tree of every node differs in height by at most one.
- Every sub tree is an AVL tree.

Figure 7.4 depicts an AVL tree.

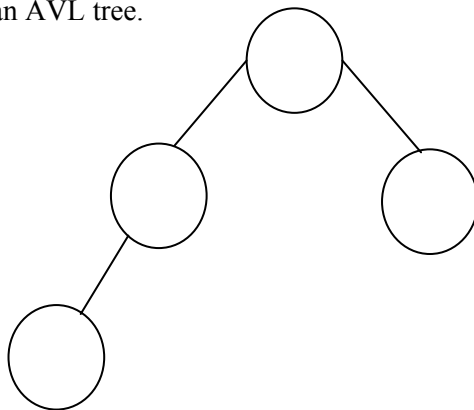


Figure 7.4 : Balance requirement for an AVL tree: the left and right subtree differ by at most one in height

AVL stands for the names of G.M. Adelson – Velskii and E.M. Landis, two Russian mathematicians, who came up with this method of keeping the tree balanced.

An AVL tree is a binary search tree which has the balance property and in addition to its key, each node stores an extra piece of information: the current balance of its subtree. The three possibilities are:

- Left – HIGH (balance factor -1)
The left child has a height that is greater than the right child by 1.
- BALANCED (balance factor 0)
Both children have the same height
- RIGHT – HIGH (balance factor +1)
The right child has a height that is greater by 1.

An AVL tree which remains balanced guarantees $O(\log n)$ search time, even in the worst case. Here, n is the number of nodes. The AVL data structure achieves this property by placing restrictions on the difference in heights between the subtrees of a given node and rebalancing the tree even if it violates these restrictions.

7.3.1 Insertion of a node into an AVL tree

Nodes are initially inserted into an AVL tree in the same manner as an ordinary binary search tree.

However, the insertion algorithm for an AVL tree travels back along the path it took to find the point of insertion and checks the balance at each node on the path.

If a node is found that is unbalanced (if it has a balance factor of either -2 or +2) then rotation is performed, based on the inserted nodes position relative to the node being examined (the unbalanced node).

7.3.2 Deletion of a node from an AVL tree

The deletion algorithm for AVL trees is a little more complex as there are several extra steps involved in the deletion of a node. If the node is not a leaf node, then it has at least one child. Then the node must be swapped with either its in-order successor or predecessor. Once the node has been swapped, we can delete it.

If a deletion node was originally a leaf node, then it can simply be removed.

As done in insertion, we traverse back up the path to the root node, checking the balance of all nodes along the path. If unbalanced, then the respective node is found and an appropriate rotation is performed to balance that node.

7.3.3 AVL tree rotations

AVL trees and the nodes it contains must meet strict balance requirements to maintain $O(\log n)$ search time. These balance restrictions are maintained using various rotation functions.

The four possible rotations that can be performed on an unbalanced AVL tree are given below. The before and after status of an AVL tree requiring the rotation are shown (refer to *Figures 7.5, 7.6, 7.7 and 7.8*).

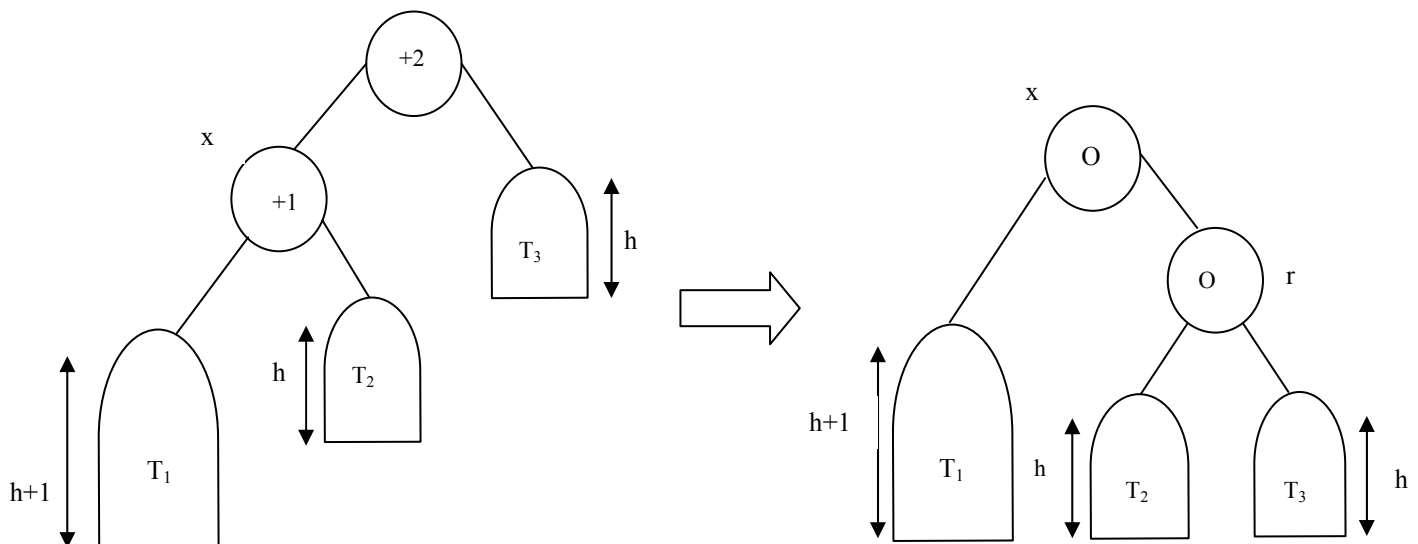


Figure 7.5: LL Rotation

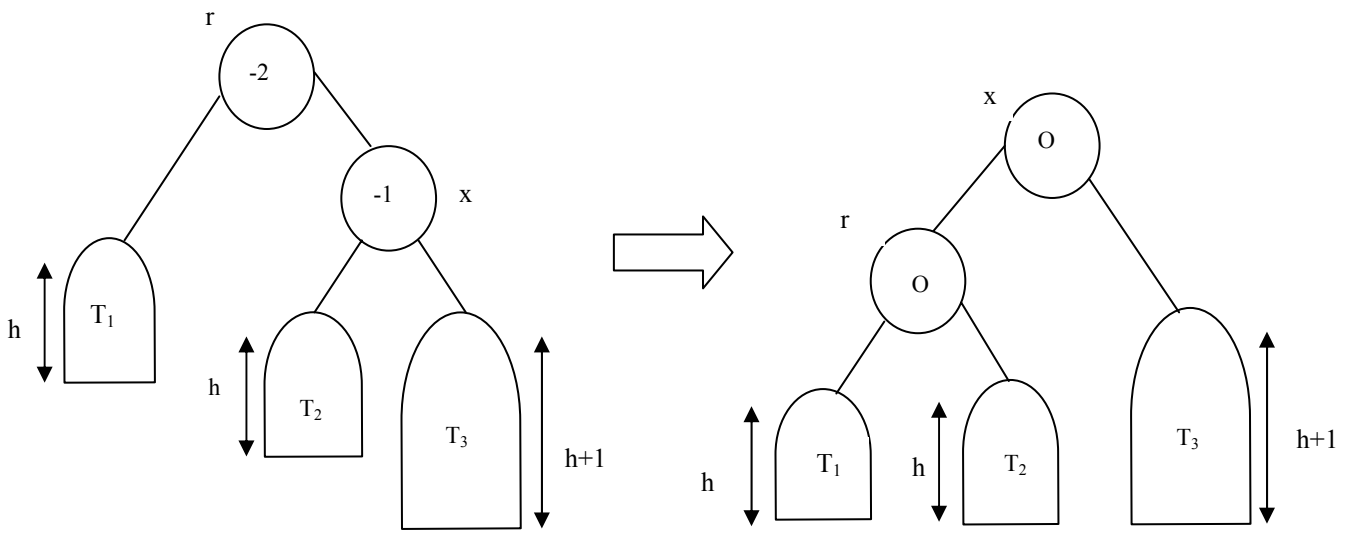


Figure 7.6: RR Rotation

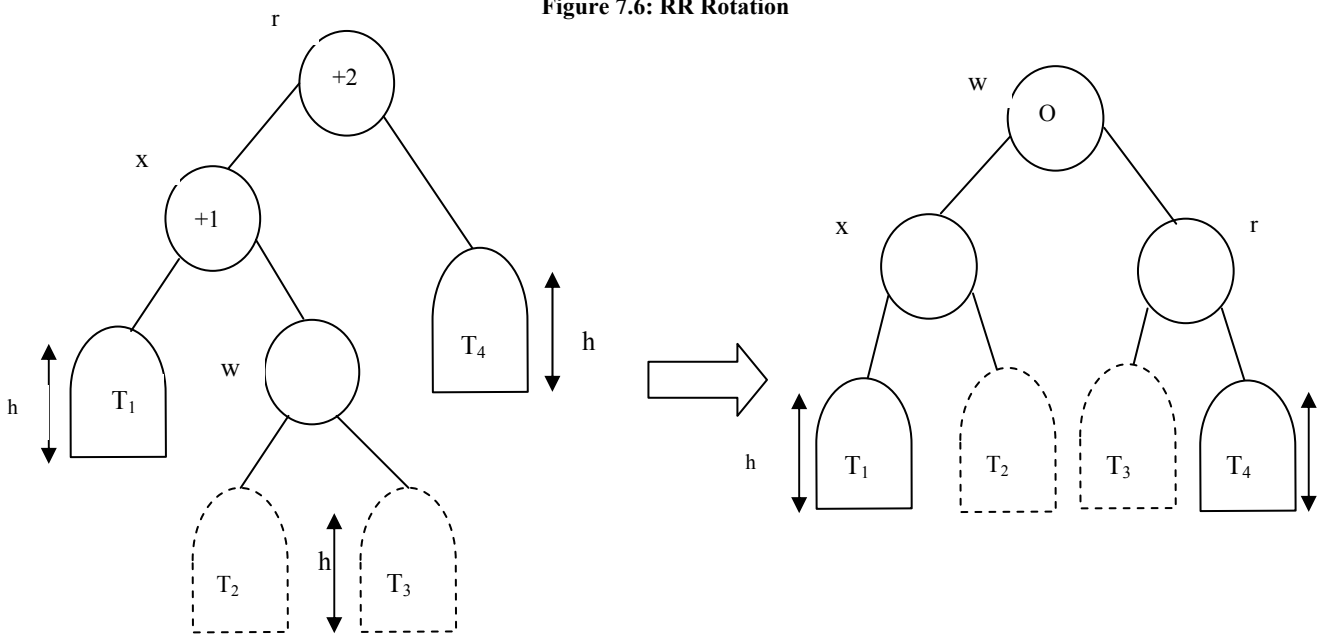


Figure 7.7: LR Rotation

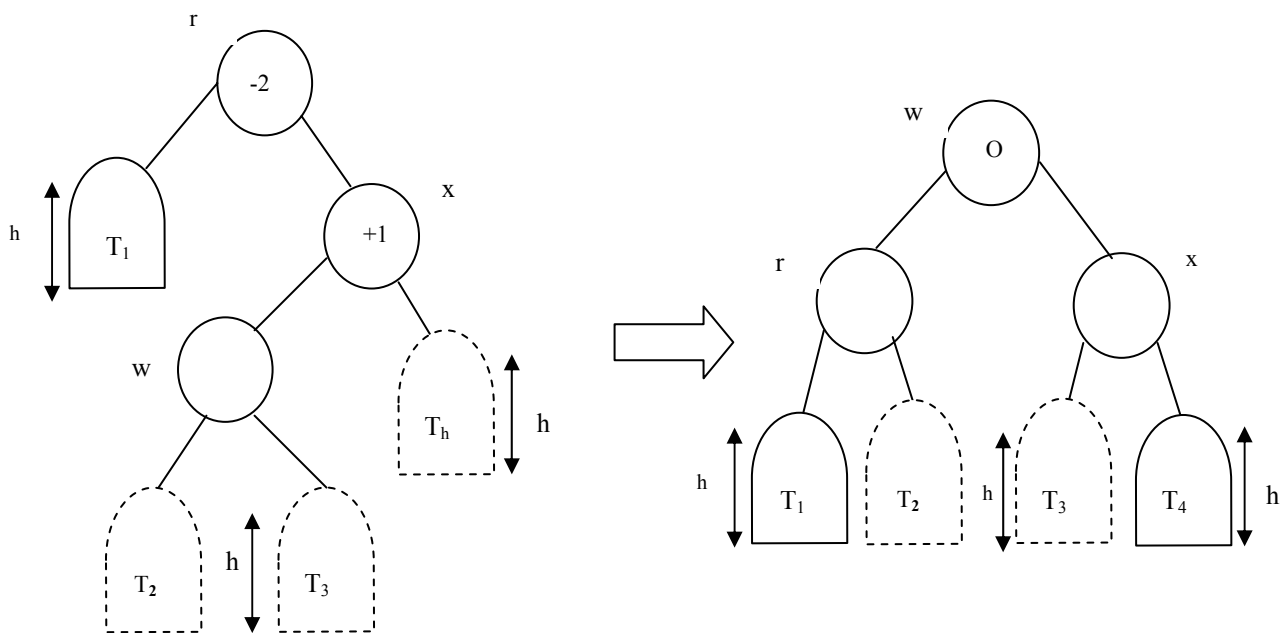


Figure 7.8: RL Rotation

Example: (Single rotation in AVL tree, when a new node is inserted into the AVL tree (LL Rotation)) (refer to *Figure 7.9*).

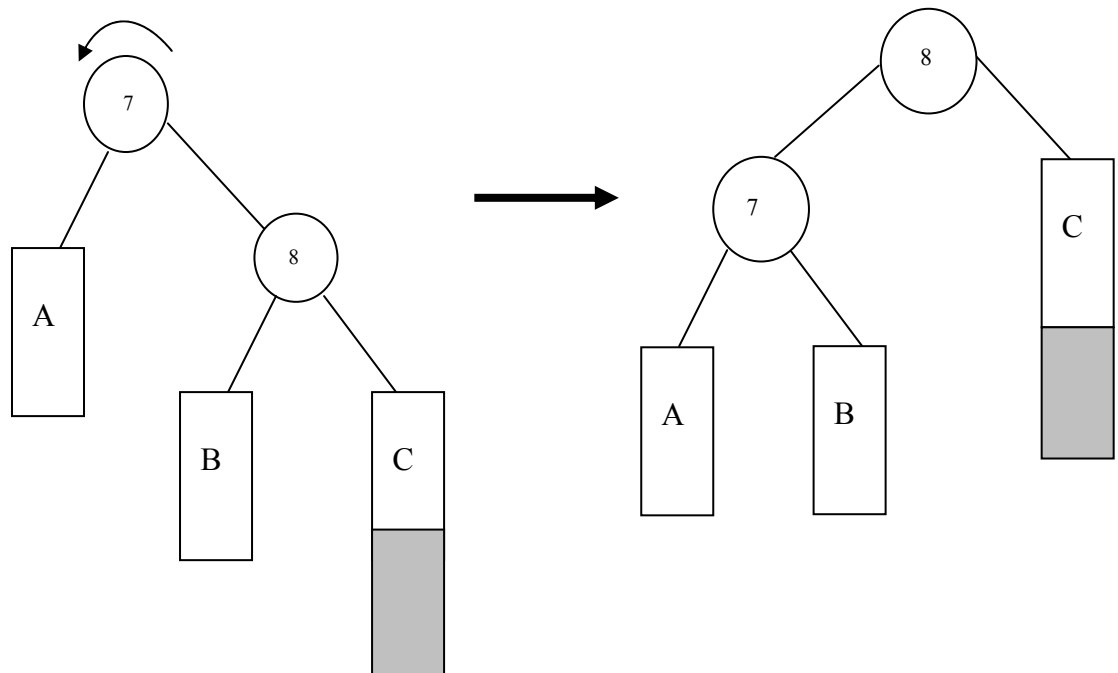


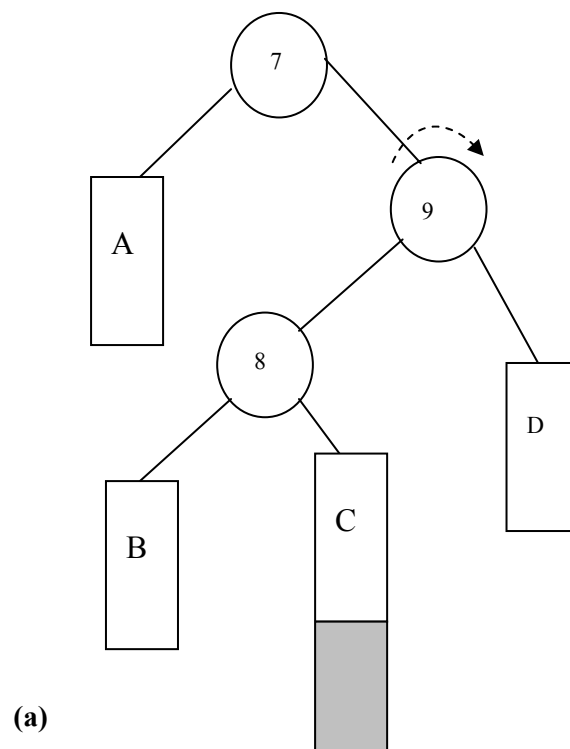
Figure 7.9: LL Rotation

The rectangles marked A, B and C are trees of equal height. The shaded rectangle stands for a new insertion in the tree C. Before the insertion, the tree was balanced, for the right child was taller then the left child by one.

The balance was broken when we inserted a node into the right child of 7, since the difference in height became 2.

To fix the balance we make 8 the new root, make c the right child move the old root (7) down to the left together with its left subtree A and finally move subtree B across and make it the new right child of 7.

Example: (Double left rotation when a new node is inserted into the AVL tree (RL rotation)) (refer to *Figure 7.10 (a),(b),(c)*).



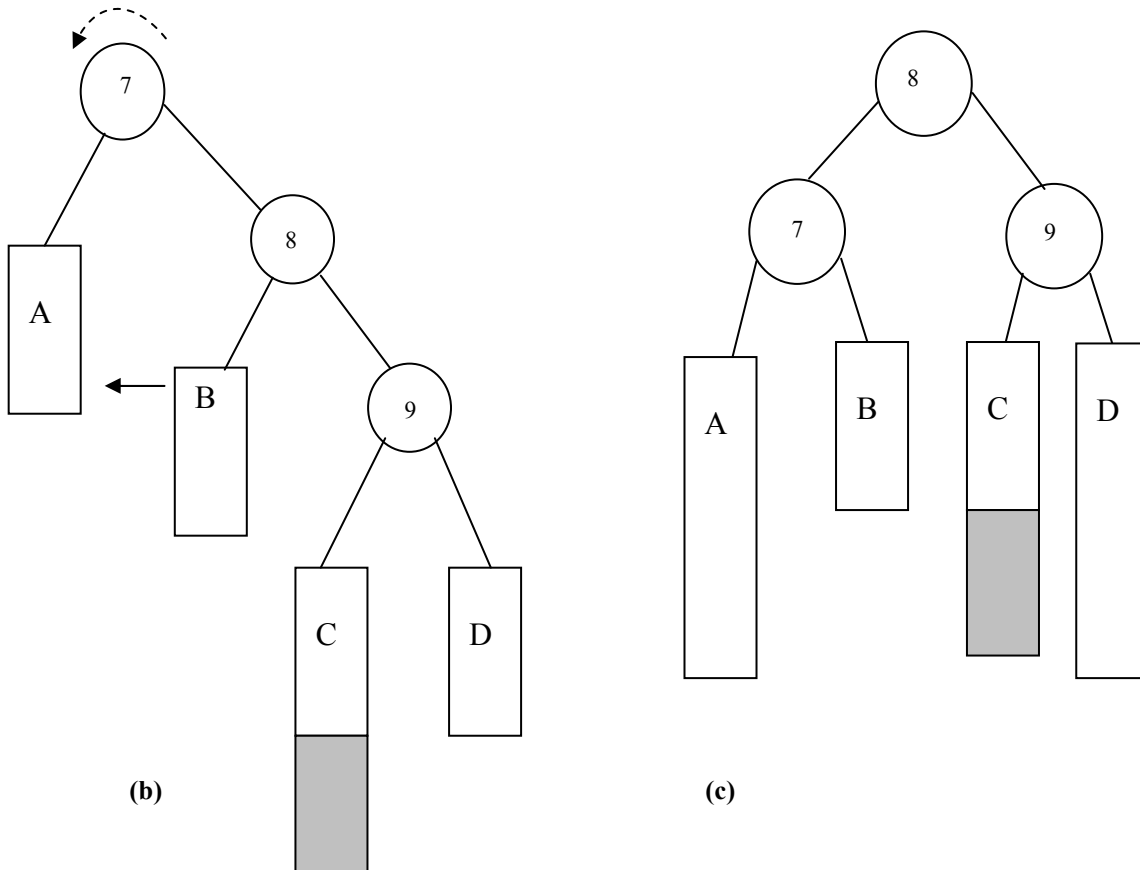


Figure 7.10: Double left rotation when a new node is inserted into the AVL tree

A node was inserted into the subtree C, making the tree off balance by 2 at the root. We first make a right rotation around the node 9, placing the C subtree into the left child of 9.

Then a left rotation around the root brings node 9 (together with its children) up a level and subtree A is pushed down a level (together with node 7). As a result we get correct AVL tree equal balance.

An AVL tree can be represented by the following structure:

```
struct avl {
    struct node *left;
    int info;
    int bf;
    struct node *right;
};
```

bf is the balance factor, *info* is the value in the node.

7.3.4 Applications of AVL Trees

AVL trees are applied in the following situations:

- There are few insertion and deletion operations
- Short search time is needed
- Input data is sorted or nearly sorted

AVL tree structures can be used in situations which require fast searching. But, the large cost of rebalancing may limit the usefulness.

Consider the following:

1. A classic problem in computer science is how to store information dynamically so as to allow for quick look up. This searching problem arises often in dictionaries, telephone directory, symbol tables for compilers and while storing business records etc. The records are stored in a balanced binary tree, based on the keys (alphabetical or numerical) order. The balanced nature of the tree limits its height to $O(\log n)$, where n is the number of inserted records.
2. AVL trees are very fast on searches and replacements. But, have a moderately high cost for addition and deletion. If application does a lot more searches and replacements than it does addition and deletions, the balanced (AVL) binary tree is a good choice for a data structure.
3. AVL tree also has applications in file systems.

Check Your Progress 2

- 1) Define the structure of an AVL tree.

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.....

7.4 B – TREES

B-trees are special m -ary balanced trees used in databases because their structure allows records to be inserted, deleted and retrieved with guaranteed worst case performance.

A B-Tree is a specialised multiway tree. In a B-Tree each node may contain a large number of keys. The number of subtrees of each node may also be large. A B-Tree is designed to branch out in this large number of directions and to contain a lot of keys in each node so that height of the tree is relatively small.

This means that only a small number of nodes must be read from disk to retrieve an item.

A B-Tree of order m is multiway search tree of order m such that

- All leaves are on the bottom level
- All internal nodes (except root node) have atleast $m/2$ (non empty) children
- The root node can have as few as 2 children if it is an internal node and can have no children if the root node is a leaf node
- Each leaf node must contain atleast $(m/2) - 1$ keys.

The following is the structure for a B-tree :

```
struct btree
{
    int count;           // number of keys stored in the current node
    item_type key[3];    // array to hold 3 keys
    long branch [4];     // array of fake pointers (records numbers)
};
```

Figure 7.11 depicts a B-tree of order 5.

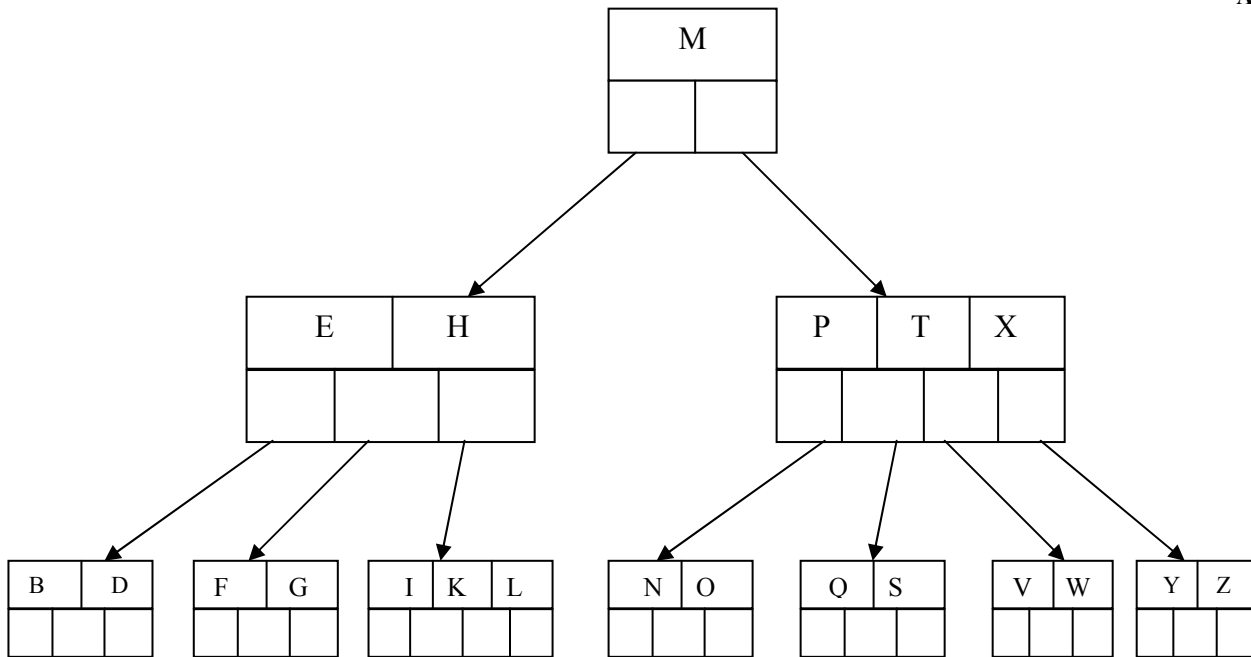


Figure 7.11: A B-tree of order 5

7.4.1 Operations on B-Trees

The following are various operations that can be performed on B-Trees:

- Search
- Create
- Insert

B-Tree strives to minimize disk access and the nodes are usually stored on disk

All the nodes are assumed to be stored in secondary storage rather than primary storage. All references to a given node are preceded by a read operation. Similarly, once a node is modified and it is no longer needed, it must be written out to secondary storage with write operation.

The following is the algorithm for searching a B-tree:

B-Tree Search (x, k)

```

i ← - 1
while i ≤ n [x] and k > keyi[x]
    do i ← i + 1
if i ≤ n [x] and k = keyi [x]
    then return (x, i)
if leaf [x]
    then return NIL
else Disk – Read (ci[x])
    return B – Tree Search (Ci[x], k)
  
```

The search operation is similar to binary tree. Instead of choosing between a left and right child as in binary tree, a B-tree search must make an n-way choice.

The correct child is chosen by performing a linear search of the values in the node. After finding the value greater than or equal to desired value, the child pointer to the immediate left to that value is followed.

The exact running time of search operation depends upon the height of the tree.

The following is the algorithm for the creation of a B-tree:

B-Tree Create (T)

```

x ← Allocate-Node ( )
Leaf [x] ← True
n [x] ← 0
Disk-write (x)
root [T] ← x

```

The above mentioned algorithm creates an empty B-tree by allocating a new root that has no keys and is a leaf node.

The following is the algorithm for insertion into a B-tree:

B-Tree Insert (T,K)

```

r ← root (T)
if n[r] = 2t - 1
    then S ← Allocate-Node ( )
        root[T] ← S
        leaf [S] ← FALSE
        n[S] ← 0
        C1 ← r
        B-Tree-Split-Child (s, l, r)
        B-Tree-Insert-Non full (s, k)
    else
        B-Tree-Insert-Non full (r, k)

```

To perform an insertion on B-tree, the appropriate node for the key must be located. Next, the key must be inserted into the node.

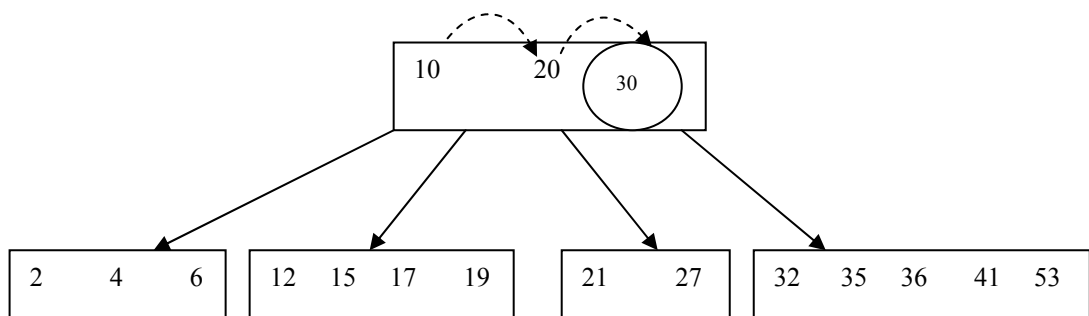
If the node is not full prior to the insertion, then no special action is required.

If node is full, then the node must be split to make room for the new key. Since splitting the node results in moving one key to the parent node, the parent node must not be full. Else, another split operation is required.

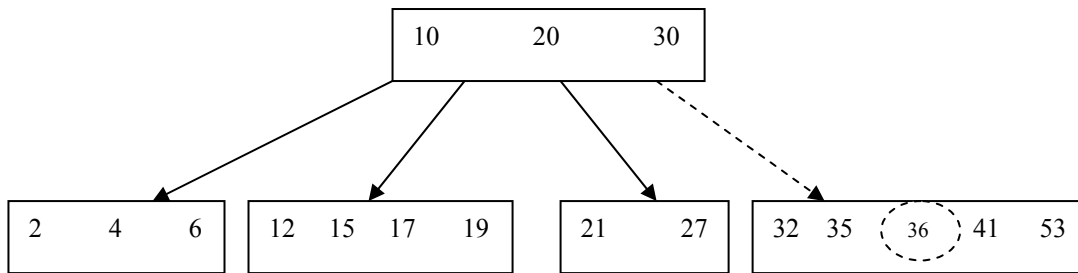
This process may repeat all the way up to the root and may require splitting the root node.

Example: Insertion of a key 33 into a B-Tree (w/split) (refer to *Figure 7.12*)

Step 1: Search first node for key nearest to 33. Key 30 was found.



Step 2: Node pointed by key 30, is searched for inserting 33. Node is split and 36 is shifted upwards.



Step 3: Key 33 is inserted between 32 and 35.

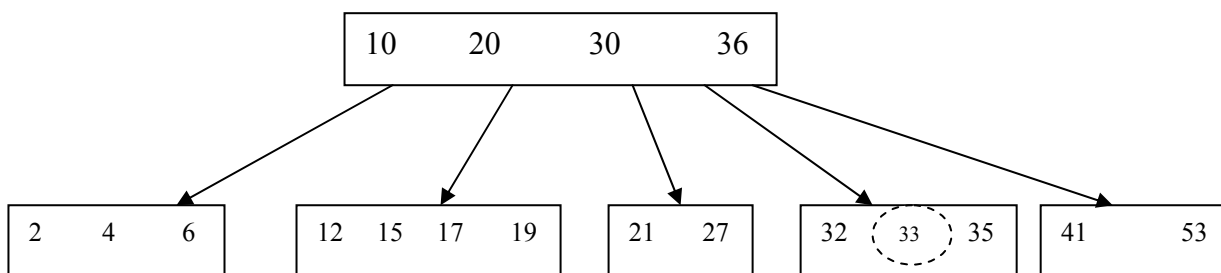
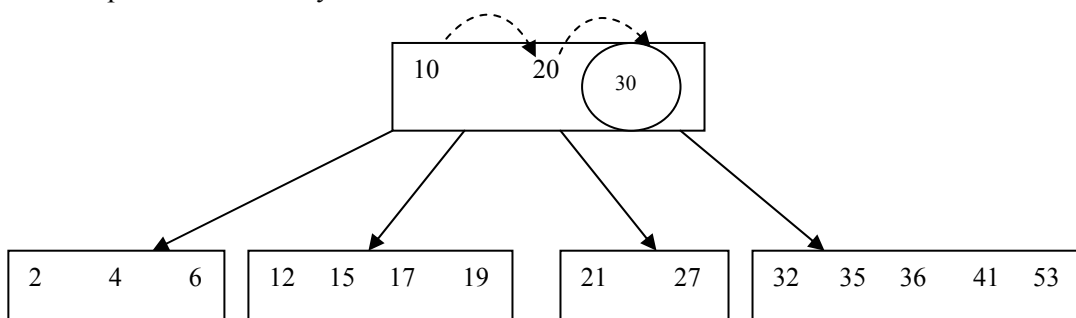


Figure 7.12 : A B-tree

Deletion of a key from B-tree is possible, but care must be taken to ensure that the properties of b-tree are maintained if the deletion reduces the number of keys in a node below the minimum degree of tree, this violation must be connected by combining several nodes and possibly reducing the height if the tree. If the key has children, the children must be rearranged.

Example (Searching of a B – Tree for key 21(refer to Figure 7.13))

Step 1: Search for key 21 in first node. 21 is between 20 and 30.



Step2 : Searching is conducted on the nodes connected by 30.

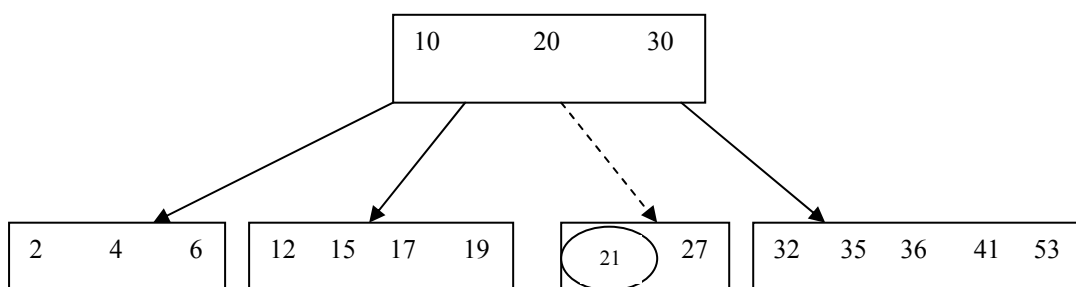


Figure 7.13 : A B-tree

7.4.2 Applications of B-trees

A database is a collection of data organised in a fashion that facilitates updation, retrieval and management of the data. Searching an unindexed database containing n keys will have a worst case running time of $O(n)$. If the same data is indexed with a b-tree, then the same search operation will run in $O(\log n)$ time. Indexing large amounts of data can significantly improve search performance.

Check Your Progress 3

- 1) Create a B – Tree of order 5 for the following:
CNGAHEKQMSWLTZDPRXYS

.....
.....

- 2) Define a multiway tree of order m .

.....
.....

7.5 SUMMARY

In this unit, we discussed Binary Search Trees, AVL trees and B-trees.

The striking feature of Binary Search Trees is that all the elements of the left subtree of the root will be less than those of the right subtree. The same rule is applicable for all the subtrees in a BST. An AVL tree is a Height balanced tree. The heights of left and right subtrees of root of an AVL tree differ by 1. The same rule is applicable for all the subtrees of the AVL tree. A B-tree is a m -ary binary tree. There can be multiple elements in each node of a B-tree. B-trees are used extensively to insert, delete and retrieve records from the databases.

7.6 SOLUTIONS/ANSWERS

Check Your Progress 1

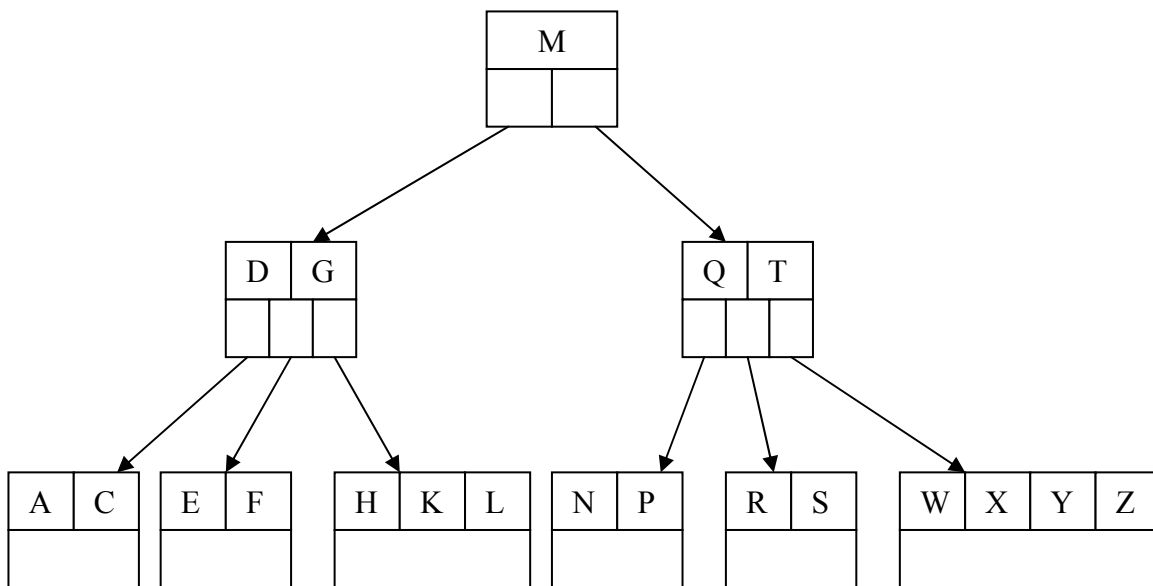
- 1) preorder, postorder and inorder
- 2) The major feature of a Binary Search Tree is that all the elements whose values is less than the root reside in the nodes of left subtree of the root and all the elements whose values are larger than the root reside in the nodes of right subtree of the root. The same rule is applicable to all the left and right subtrees of a BST.

Check Your Progress 2

- 1) The following is the structure of an AVL tree:

```
struct avl {  
    struct node *left;  
    int info;  
    int bf;  
    struct node *right;  
};
```

1)



- 2) A multiway tree of order n is an ordered tree where each node has at most m children. For each node, if k is the actual no. of children in the node, then $k-1$ is the number of keys in the node. If the keys and subtrees are arranged in the fashion of a search tree, then this is multiway search tree of order m .

7.7 FURTHER READINGS

1. *Data Structures using C and C++* by Yedidyah Hangsam, Moshe J. Augenstein and Aaron M. Tanenbaum, PHI Publications.
2. *Fundamentals of Data Structures in C* by R.B. Patel, PHI Publications.

Reference Websites

[http:// www.cs.umbc.edu](http://www.cs.umbc.edu)
<http://www.fredosaurus.com>

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8.0 INTRODUCTION

In this unit, we will discuss a data structure called Graph. In fact, graph is a general tree with no parent-child relationship. Graphs have many applications in computer science and other fields of science. In general, graphs represent a relatively less restrictive relationship between the data items. We shall discuss about both undirected graphs and directed graphs. The unit also includes information on different algorithms which are based on graphs.

8.1 OBJECTIVES

After going through this unit, you should be able to

- know about graphs and related terminologies;
- know about directed and undirected graphs along with their representations;
- know different shortest path algorithms;
- construct minimum cost spanning trees;
- apply depth first search and breadth first search algorithms, and
- finding strongly connected components of a graph.

8.2 DEFINITIONS

A graph G may be defined as a finite set V of vertices and a set E of edges (pair of connected vertices). The notation used is as follows:

Graph $G = (V, E)$

Consider the graph of *Figure 8.1*.

The set of vertices for the graph is $V = \{1, 2, 3, 4, 5\}$.

The set of edges for the graph is $E = \{(1,2), (1,5), (1,3), (5,4), (4,3), (2,3)\}$.

The elements of E are always a pair of elements.

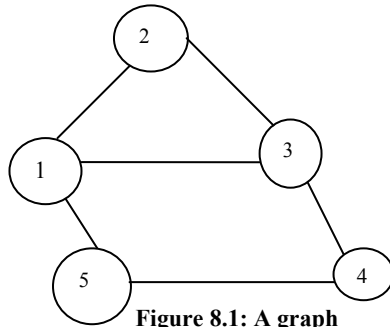


Figure 8.1: A graph

It may be noted that unlike nodes of a tree, graph has a very limited relationship between the nodes (vertices). There is no direct relationship between the vertices 1 and 4 although they are connected through 3.

Directed graph and Undirected graph: If every edge (a,b) in a graph is marked by a direction from a to b , then we call it a Directed graph (digraph). On the other hand, if directions are not marked on the edges, then the graph is called an Undirected graph.

In a Directed graph, the edges $(1,5)$ and $(5,1)$ represent two different edges whereas in an Undirected graph, $(1,5)$ and $(5,1)$ represent the same edge. Graphs are used in various types of modeling. For example, graphs can be used to represent connecting roads between cities.

Graph terminologies :

Adjacent vertices: Two vertices a and b are said to be adjacent if there is an edge connecting a and b . For example, in *Figure 8.1*, vertices 5 and 4 are adjacent.

Path: A path is defined as a sequence of distinct vertices, in which each vertex is adjacent to the next. For example, the path from 1 to 4 can be defined as a sequence of adjacent vertices $(1,5), (5,4)$.

A path, p , of length, k , through a graph is a sequence of connected vertices:

$$p = \langle v_0, v_1, \dots, v_k \rangle$$

Cycle : A graph contains *cycles* if there is a path of non-zero length through the graph, $p = \langle v_0, v_1, \dots, v_k \rangle$ such that $v_0 = v_k$.

Edge weight : It is the cost associated with edge.

Loop: It is an edge of the form (v,v) .

Path length : It is the number of edges on the path.

Simple path : It is the set of all distinct vertices on a path (except possibly first and last).

Spanning Trees: A *spanning tree* of a graph, G , is a set of $|V|-1$ edges that connect all vertices of the graph.

There are different representations of a graph. They are:

- Adjacency list representation
- Adjacency matrix representation

Adjacency list representation

An Adjacency list representation of a Graph $G = \{V, E\}$ consists of an array of adjacency lists denoted by *adj of V* list. For each vertex $u \in V$, $\text{adj}[u]$ consists of all vertices adjacent to u in the graph G .

Consider the graph of *Figure 8.2*.

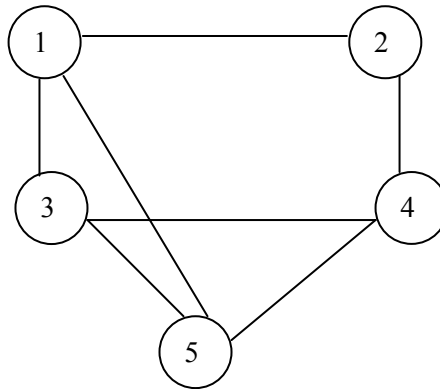


Figure 8.2: A Graph

The following is the adjacency list representation of graph of *Figure 8.2*:

$\text{adj}[1] = \{2, 3, 5\}$
 $\text{adj}[2] = \{1, 4\}$
 $\text{adj}[3] = \{1, 4, 5\}$
 $\text{adj}[4] = \{2, 3, 5\}$
 $\text{adj}[5] = \{1, 3, 4\}$

An adjacency matrix representation of a Graph $G=(V, E)$ is a matrix $A(a_{ij})$ such that

$$a_{ij} = \begin{cases} 1 & \text{if edge } (i, j) \text{ belongs to } E \\ 0 & \text{otherwise} \end{cases}$$

The adjacency matrix for the graph of *Figure 8.2* is given below:

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

Observe that the matrix is symmetric along the main diagonal. If we define the adjacency matrix as A and the transpose as A^T , then for an undirected graph G as above, $A = A^T$.

A connected graph is a graph in which path exists between every pair of vertices.

A strongly connected graph is a directed graph in which every pair of distinct vertices are connected with each other.

A weakly connected graph is a directed graph whose underlying graph is connected, but not strongly connected.

A complete graph is a graph in which there exists edge between every pair of vertices.

☞ Check Your Progress 1

- 1) A graph with no cycle is called _____ graph.
- 2) Adjacency matrix of an undirected graph is _____ on main diagonal.
- 3) Represent the following graphs(*Figure 8.3* and *Figure 8.4*) by adjacency matrix:

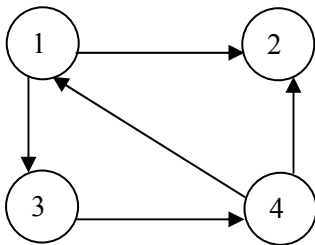


Figure 8.3: A Directed Graph

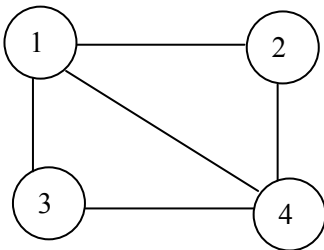


Figure 8.4: A Graph

8.3 SHORTEST PATH ALGORITHMS

A driver takes shortest possible route to reach destination. The problem that we will discuss here is similar to this kind of finding shortest route in a graph. The graphs are weighted directed graphs. The weight could be time, cost, losses other than distance designated by numerical values.

Single source shortest path problem : To find a shortest path from a single source to every vertex of the Graph.

Consider a Graph $G = (V, E)$. We wish to find out the shortest path from a single source vertex seV , to every vertex veV . The single source shortest path algorithm (Dijkstra's Algorithm) is based on assumption that no edges have negative weights.

The procedure followed to find shortest path are based on a concept called relaxation. This method repeatedly decreases the upper bound of actual shortest path of each vertex from the source till it equals the shortest-path weight. Please note that shortest path between two vertices contains other shortest path within it.

8.3.1 Dijkstra's Algorithm

Dijkstra's algorithm (named after its discover, Dutch computer scientist E.W. Dijkstra) solves the problem of finding the shortest path from a point in a graph (the *source*) to a destination with non-negative weight edge.

It turns out that one can find the shortest paths from a given source to *all* vertices (points) in a graph in the same time. Hence, this problem is sometimes called the *single-source shortest paths* problem. Dijkstra's algorithm is a greedy algorithm, which finds shortest path between all pairs of vertices in the graph. Before describing the algorithms formally, let us study the method through an example.

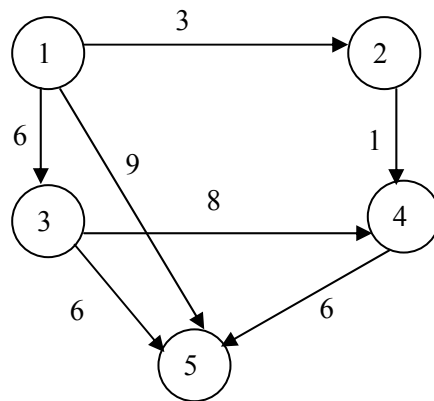


Figure 8.5: A Directed Graph with no negative edge(s)

Dijkstra's algorithm keeps two sets of vertices:

S is the set of vertices whose shortest paths from the source have already been determined

Q = V-S is the set of remaining vertices .

The other data structures needed are:

d array of best estimates of shortest path to each vertex from the source

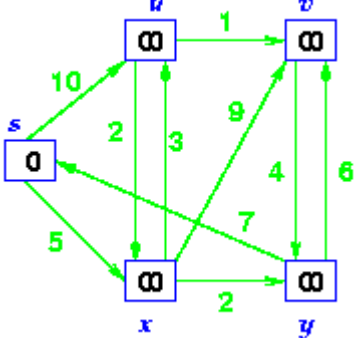
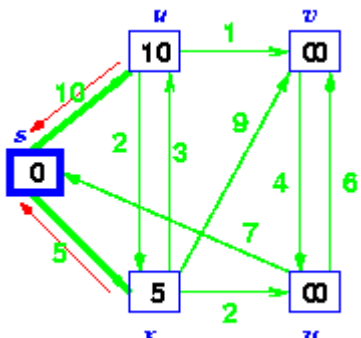
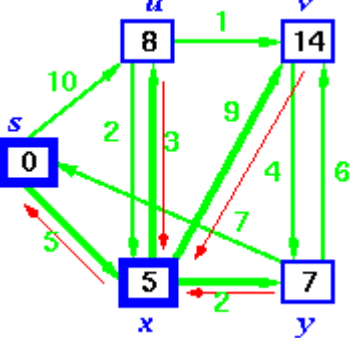
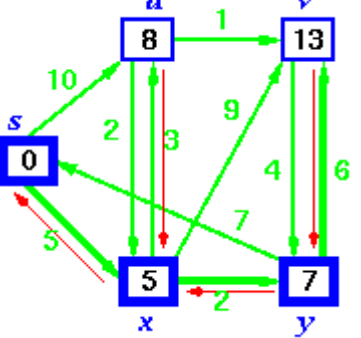
pi an array of predecessors for each vertex. *predecessor* is an array of vertices to which shortest path has already been determined.

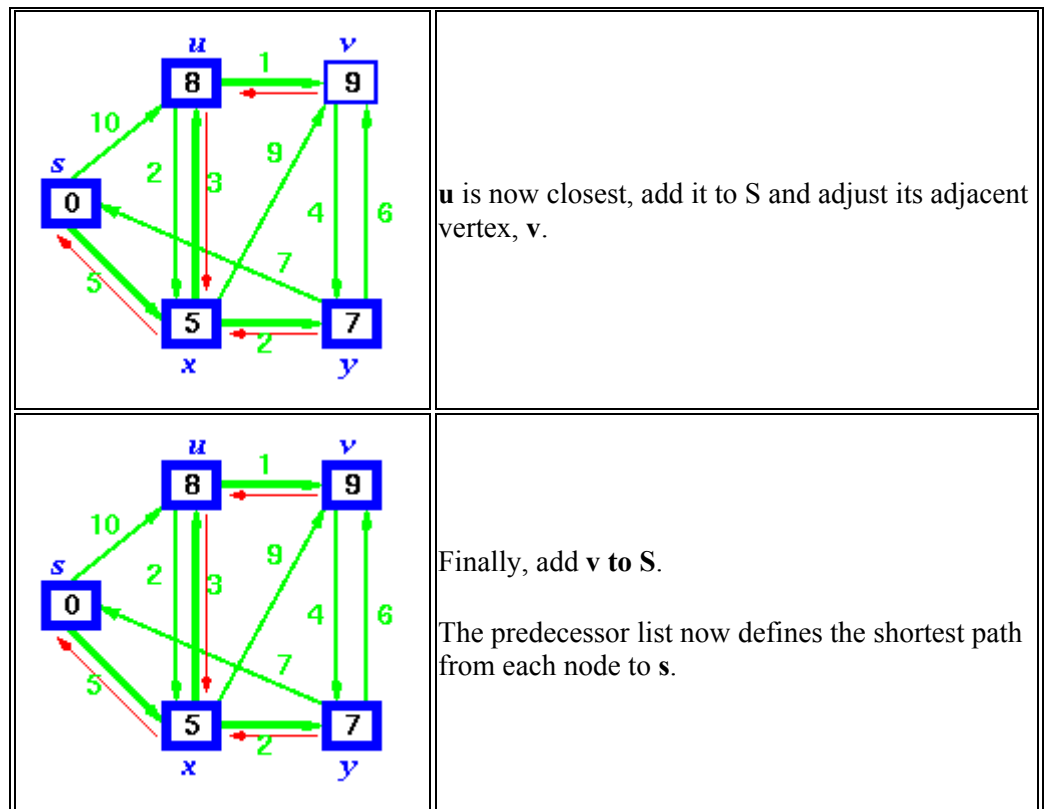
The basic operation of Dijkstra's algorithm is edge relaxation. If there is an edge from u to v , then the shortest known path from s to u can be extended to a path from s to v by adding edge (u,v) at the end. This path will have length $d[u]+w(u,v)$. If this is less than $d[v]$, we can replace the current value of $d[v]$ with the new value.

The predecessor list is an array of indices, one for each vertex of a graph. Each vertex entry contains the index of its predecessor in a path through the graph.

Operation of Algorithm

The following sequence of diagrams illustrate the operation of Dijkstra's Algorithm. The bold vertices indicate the vertex to which shortest path has been determined.

	<p>Initialize the graph, all the vertices have infinite costs except the source vertex which has zero cost</p>
	<p>From all the adjacent vertices, choose the closest vertex to the source s.</p> <p>As we initialized $d[s]$ to 0, it's s. (shown in bold circle)</p> <p>Add it to S</p> <p>Relax all vertices adjacent to s, i.e u and x</p> <p>Update vertices u and x by 10 and 5 as the distance from s.</p>
	<p>Choose the nearest vertex, x.</p> <p>Relax all vertices adjacent to x</p> <p>Update predecessors for u, v and y. Predecessor of x = s Predecessor of v = x, s Predecessor of y = x, s</p> <p>add x to S</p>
	<p>Now y is the closest vertex. Add it to S.</p> <p>Relax v and adjust its predecessor.</p>



Dijkstra's algorithm

** Initialise d and pi**

for each vertex v in $V(g)$

$g.d[v] := \text{infinity}$

$g.pi[v] := \text{nil}$

$g.d[s] := 0;$

** Set S to empty **

$S := \{ s \}$

$Q := V(g)$

** While (V-S) is not null**

while not Empty(Q)

1. Sort the vertices in **V-S** according to the current best estimate of their distance from the source
 $u := \text{Extract-Min}(Q);$
2. Add vertex **u**, the closest vertex in **V-S**, to **S**,
 $\text{AddNode}(S, u);$
3. Relax all the vertices still in **V-S** connected to **u**
 $\text{relax}(\text{Node } u, \text{Node } v, \text{double } w[u][v])$
if $d[v] > d[u] + w[u][v]$ then
 $d[v] := d[u] + w[u][v]$
 $pi[v] := u$

In summary, this algorithm starts by assigning a weight of infinity to all vertices, and then selecting a source and assigning a weight of zero to it. Vertices are added to the set for which shortest paths are known. When a vertex is selected, the weights of its adjacent vertices are relaxed. Once all vertices are relaxed, their predecessor's vertices

are updated (π_i). The cycle of selection, weight relaxation and predecessor update is repeated until the shortest path to all vertices has been found.

Complexity of Algorithm

The simplest implementation of the Dijkstra's algorithm stores vertices of set Q in an ordinary linked list or array, and operation $\text{Extract-Min}(Q)$ is simply a linear search through all vertices in Q . In this case, the running time is $\Theta(n^2)$.

8.3.2 Graphs with Negative Edge costs

We have seen that the above Dijkstra's single source shortest-path algorithm works for graphs with non-negative edges (like road networks). The following two scenarios can emerge out of negative cost edges in a graph:

- Negative edge with non-negative weight cycle reachable from the source.
- Negative edge with non-negative weight cycle reachable from source.

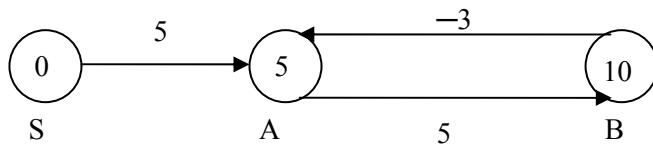


Figure 8.6 : A Graph with negative edge and non-negative weight cycle

The net weight of the cycle is 2(non-negative)(refer to Figure 8.6).

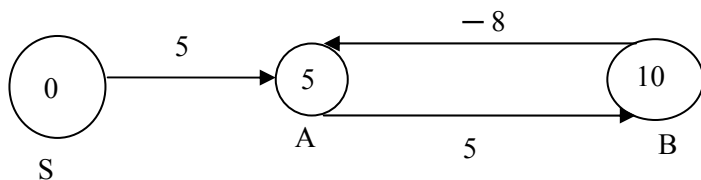


Figure 8.7: A graph with negative edge and negative weight cycle

The net weight of the cycle is -3 (negative) (refer to Figure 8.7). The shortest path from A to B is not well defined as the shortest path to this vertex are infinite, i.e., by traveling each cycle we can decrease the cost of the shortest path by 3, like (S, A, B) is path (S, A, B, A, B) is a path with less cost and so on.

Dijkstra's Algorithm works only for directed graphs with non-negative weights (cost).

8.3.3 Acyclic Graphs

A path in a directed graph is said to form a cycle if there exists a path (A,B,C,...,P) such that $A = P$. A graph is called acyclic if there is no cycle in the graph.

8.3.4 All Pairs Shortest Paths Algorithm

In the last section, we discussed about shortest path algorithm which starts with a single source and finds shortest path to all vertices in the graph. In this section, we shall discuss the problem of finding shortest path between all pairs of vertices in a graph. This problem is helpful in finding distance between all pairs of cities in a road atlas. All pairs shortest paths problem is mother of all shortest paths problems.

In this algorithm, we will represent the graph by adjacency matrix.

The weight of an edge C_{ij} in an adjacency matrix representation of a directed graph is represented as follows

$$C_{ij} = \begin{cases} 0 & \text{if } i = j \\ \text{weight of the directed edge from } i \text{ to } j \text{ i.e. } (i,j) & \text{if } i \neq j \text{ and } (i,j) \text{ belongs to } E \\ \infty & \text{if } i \neq j \text{ and } (i,j) \text{ does not belong to } E \end{cases}$$

Given a directed graph $G = (V, E)$, where each edge (v, w) has a non-negative cost $C(v, w)$, for all pairs of vertices (v, w) to find the lowest cost path from v to w .

The All pairs shortest paths problem can be considered as a generalisation of single-source-shortest-path problem, by using Dijkstra's algorithm by varying the source node among all the nodes in the graph. If negative edge(s) is allowed, then we can't use Dijkstra's algorithm.

In this section we shall use a recursive solution to all pair shortest paths problem known as Floyd-Warshall algorithm, which runs in $O(n^3)$ time.

This algorithm is based on the following principle. For graph G let $V = \{1, 2, 3, \dots, n\}$. Let us consider a sub set of the vertices $\{1, 2, 3, \dots, k\}$. For any pair of vertices that belong to V , consider all paths from i to j whose intermediate vertices are from $\{1, 2, 3, \dots, k\}$. This algorithm will exploit the relationship between path p and shortest path from i to j whose intermediate vertices are from $\{1, 2, 3, \dots, k-1\}$ with the following two possibilities:

1. If k is not an intermediate vertex in the path p , then all the intermediate vertices of the path p are in $\{1, 2, 3, \dots, k-1\}$. Thus, shortest path from i to j with intermediate vertices in $\{1, 2, 3, \dots, k-1\}$ is also the shortest path from i to j with vertices in $\{1, 2, 3, \dots, k\}$.
2. If k is an intermediate vertex of the path p , we break down the path p into path p_1 from vertex i to k and path p_2 from vertex k to j . So, path p_1 is the shortest path from i to k with intermediate vertices in $\{1, 2, 3, \dots, k-1\}$.

During iteration process we find the shortest path from i to j using only vertices $\{1, 2, 3, \dots, k-1\}$ and in the next step, we find the cost of using the k^{th} vertex as an intermediate step. If this results in lower cost, then we store it.

After n iterations (all possible iterations), we find the lowest cost path from i to j using all vertices (if necessary).

Note the following:

Initialize the matrix

$C[i][j] = \infty$ if (i, j) does not belong to E for graph $G = (V, E)$

Initially, $D[i][j] = C[i][j]$

We also define a path matrix P where $P[i][j]$ holds intermediate vertex k on the least cost path from i to j that leads to the shortest path from i to j .

Algorithm (All Pairs Shortest Paths)

N = number of rows of the graph

$D[i][j] = C[i][j]$

For k from 1 to n

 Do for $i = 1$ to n

 Do for $j = 1$ to n

$D[i][j] = \text{minimum}(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)})$

 Enddo

 Enddo

Enddo

where $d_{ij}^{(k-1)}$ = minimum path from i to j using $k-1$ intermediate vertices

where $d_{ik}^{(k-1)}$ = minimum path from i to k using $k-1$ intermediate vertices

where $d_{kj}^{(k-1)}$ = minimum path from k to j using $k-1$ intermediate vertices

Program 8.1 gives the program segment for the All pairs shortest paths algorithm.

AllPairsShortestPaths(int N, Matrix C, Matrix P, Matrix D)

```
{
    int i, j, k

    if i = j then C[i][j] = 0
    for ( i = 0; i < N; i++)
    {
        for (j = 0; j < N; j++)
        {
            D[i][j] = C[i][j];
            P[i][j] = -1;
        }
        D[i][j] = 0;
    }

    for (k=0; k<N; k++)
    {
        for (i=0; i<N; i++)
        {
            for (j=0; J<N; J++)
            {
                if (D[i][k] + D[k][j] < D[i][j])
                {
                    D[i][j] = D[i][k] + D[k][j];
                    P[i][j] = k;
                }
            }
        }
    }
}

/***** End *****/
```

Program 8.1 : Program segment for All pairs shortest paths algorithm

From the above algorithm, it is evident that it has $O(N^3)$ time complexity.

Shortest path algorithms had numerous applications in the areas of Operations Research, Computer Science, Electrical Engineering and other related areas.

☞ Check Your Progress 2

- 1) _____ is the basis of Dijkstra's algorithm
 - 2) What is the complexity of All pairs shortest paths algorithm?
-

8.4 MINIMUM COST SPANNING TREES

A *spanning tree* of a graph is just a subgraph that contains all the vertices and is a tree (with no cycle). A graph may have many spanning trees.

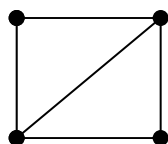


Figure 8.8: A Graph

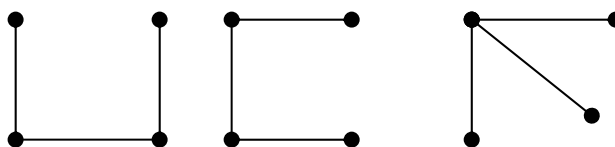


Figure 8.9 : Spanning trees of the Graph of Figure 8.8

Consider the graph of *Figure 8.8*. Its spanning trees are shown in *Figure 8.9*. Now, if the graph is a weighted graph (length associated with each edge). The weight of the tree is just the sum of weights of its edges. Obviously, different spanning trees have different weights or lengths. Our objective is to find the minimum length (weight) spanning tree.

Suppose, we have a group of islands that we wish to link with bridges so that it is possible to travel from one island to any other in the group. The set of bridges which will enable one to travel from any island to any other at minimum capital cost to the government is the minimum cost spanning tree.

8.4.1 Kruskal's Algorithm

Kruskal's algorithm uses the concept of *forest* of trees. Initially the forest consists of n single node trees (and no edges). At each step, we add one (the cheapest one) edge so that it links two trees together. If it forms a cycle, it would simply mean that it links two nodes that were already connected. So, we reject it.

The steps in Kruskal's Algorithm are as follows:

1. The forest is constructed from the graph G - with each node as a separate tree in the forest.
2. The edges are placed in a priority queue.
3. Do until we have added $n-1$ edges to the graph,
 1. Extract the cheapest edge from the queue.
 2. If it forms a cycle, then a link already exists between the concerned nodes. Hence reject it.
 3. Else add it to the forest. Adding it to the forest will join two trees together.

The forest of trees is a partition of the original set of nodes. Initially all the trees have exactly one node in them. As the algorithm progresses, we form a union of two of the trees (sub-sets), until eventually the partition has only one sub-set containing all the nodes.

Let us see the sequence of operations to find the Minimum Cost Spanning Tree(MST) in a graph using Kruskal's algorithm. Consider the graph of *Figure 8.10.*, *Figure 8.11* shows the construction of MST of graph of *Figure 8.10.*

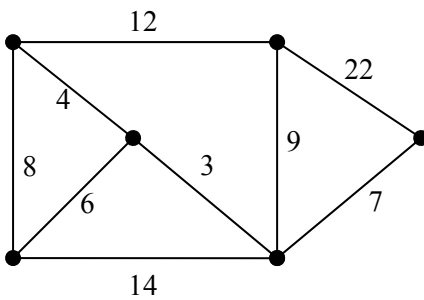
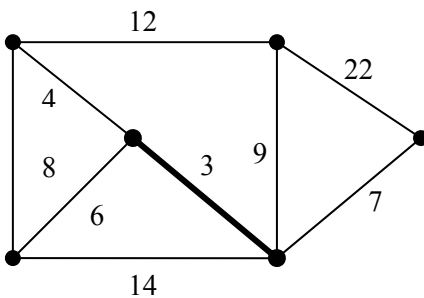


Figure 8.10 : A Graph



Step 1

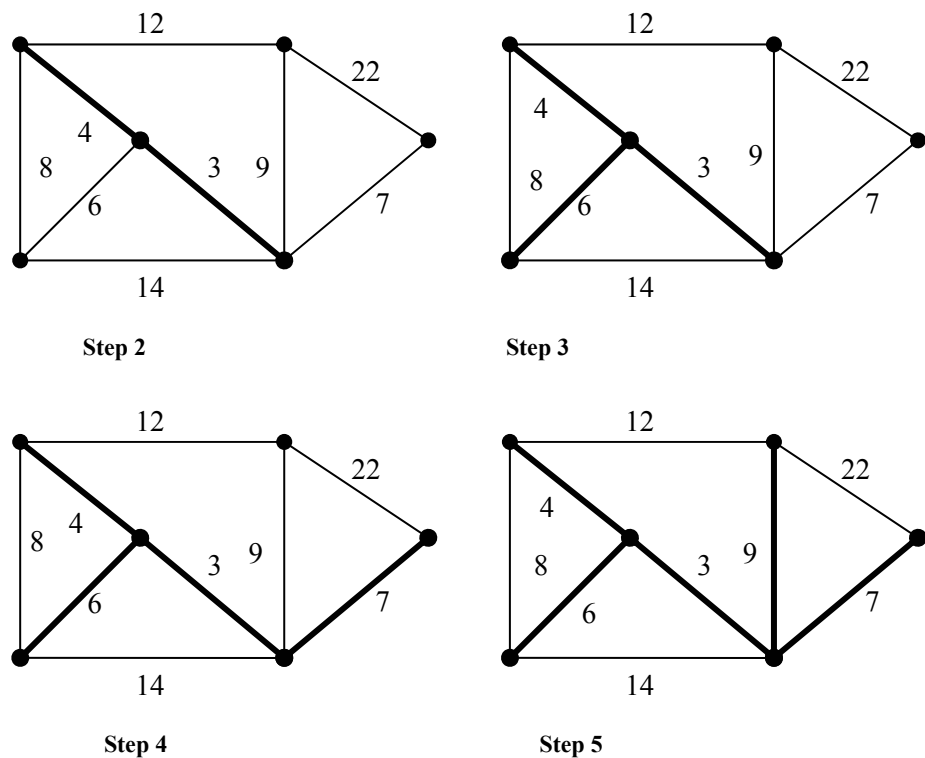


Figure 8.11 : Construction of Minimum Cost Spanning Tree for the Graph of Figure 8.10 by application of Kruskal's algorithm

The following are various steps in the construction of MST for the graph of *Figure 8.10* using Kruskal's algorithm.

Step 1 : The lowest cost edge is selected from the graph which is not in MST (initially MST is empty). The lowest cost edge is 3 which is added to the MST (shown in bold edges)

Step 2: The next lowest cost edge which is not in MST is added (edge with cost 4).

Step 3 : The next lowest cost edge which is not in MST is added (edge with cost 6).

Step 4 : The next lowest cost edge which is not in MST is added (edge with cost 7).

Step 5 : The next lowest cost edge which is not in MST is 8 but will form a cycle. So, it is discarded . The next lowest cost edge 9 is added. Now the MST contains all the vertices of the graph. This results in the MST of the original graph.

8.4.2 Prim's Algorithm

Prim's algorithm uses the concept of sets. Instead of processing the graph by sorted order of edges, this algorithm processes the edges in the graph randomly by building up disjoint sets.

It uses two disjoint sets A and \bar{A} . Prim's algorithm works by iterating through the nodes and then finding the shortest edge from the set A to that of set \bar{A} (i.e. outside A), followed by the addition of the node to the new graph. When all the nodes are processed, we have a minimum cost spanning tree.

Rather than building a sub-graph by adding one edge at a time, Prim's algorithm builds a tree one vertex at a time.

The steps in Prim's algorithm are as follows:

Let G be the graph with n vertices for which minimum cost spanning tree is to be generated.

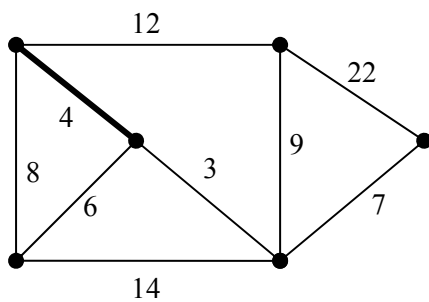
Let T be the minimum spanning tree.

Let T be a single vertex x .

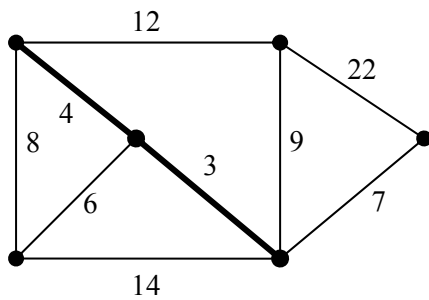
while (T has fewer than n vertices)

```
{
    find the smallest edge connecting  $T$  to  $G-T$ 
    add it to  $T$ 
}
```

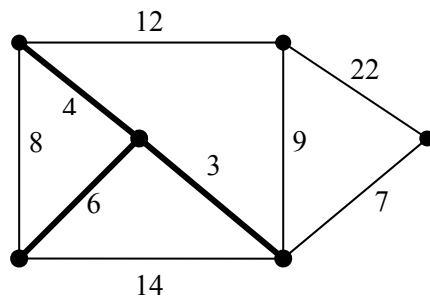
Consider the graph of Figure 8.10. Figure 8.12 shows the various steps involved in the construction of Minimum Cost Spanning Tree of graph of Figure 8.10.



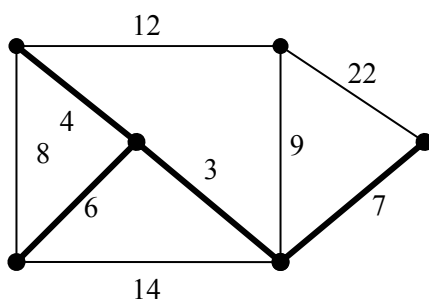
Step 1



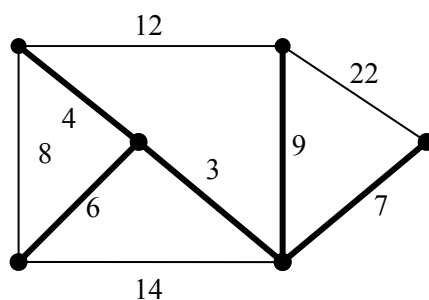
Step 2



Step 3



Step 4



Step 5

Figure 8.12 : Construction of Minimum Cost Spanning Tree for the Graph of Figure 8.10 by application of Prim's algorithm

The following are various steps in the construction of MST for the graph of Figure 8.10 using Prim's algorithm.

Step 1 : We start with a single vertex (node). Now the set A contains this single node and set A contains rest of the nodes. Add the edge with the lowest cost from A to A . The edge with cost 4 is added.

Step 2: Lowest cost path from shaded portion of the graph to the rest of the graph (edge with cost 3) is selected and added to MST.

Step 3: Lowest cost path from shaded portion of the graph to the rest of the graph (edge with cost 6) is selected and added to MST.

Step 4: Lowest cost path from shaded portion of the graph to the rest of the graph (edge with cost 73) is selected and added to MST.

Step 5: The next lowest cost edge to the set not in MST is 8 but forms a cycle. So, it is discarded. The next lowest cost edge 9 is added. Now the MST contains all the vertices of the graph. This results in the MST of the original graph.

Comparison of Kruskal's algorithm and Prim's algorithm

	Kruskal's algorithm	Prim's algorithm
Principle	Based on generic minimum cost spanning tree algorithms	A special case of generic minimum cost spanning tree algorithm. Operates like Dijkstra's algorithm for finding shortest path in a graph.
Operation	Operates on a single set of edges in the graph	Operates on two disjoint sets of edges in the graph
Running time	$O(E \log E)$ where E is the number of edges in the graph	$O(E \log V)$, which is asymptotically same as Kruskal's algorithm

For the above comparison, it may be observed that for dense graphs having more number of edges for a given number of vertices, Prim's algorithm is more efficient.

8.4.3 Applications

The minimum cost spanning tree has wide applications in different fields. It represents many complicated real world problems like:

1. Minimum distance for travelling all cities at most one (travelling salesman problem).
2. In electronic circuit design, to connect n pins by using $n-1$ wires, using least wire.
3. Spanning tree also finds their application in obtaining independent set of circuit equations for an electrical network.

8.5 BREADTH FIRST SEARCH (BFS)

When BFS is applied, the vertices of the graph are divided into two categories. The vertices, which are visited as part of the search and those vertices, which are not visited as part of the search. The strategy adopted in breadth first search is to start search at a vertex(source). Once you started at source, the number of vertices that are visited as part of the search is 1 and all the remaining vertices need to be visited. Then, search the vertices which are adjacent to the visited vertex from left to order. In this way, all the vertices of the graph are searched.

Consider the digraph of *Figure 8.13*. Suppose that the search started from S. Now, the vertices (from left to right) adjacent to S which are not visited as part of the search are B, C, A. Hence, B,C and A are visited after S as part of the BFS. Then, F is the unvisited vertex adjacent to B. Hence, the visit to B, C and A is followed by F. The unvisited vertex adjacent of C is D. So, the visit to F is followed by D. There are no

unvisited vertices adjacent to A. Finally, the unvisited vertex E adjacent to D is visited.

Hence, the sequence of vertices visited as part of BFS is S, B, C, A, F, D and E.

8.6 DEPTH FIRST SEARCH (DFS)

The strategy adopted in depth first search is to search deeper whenever possible. This algorithm repeatedly searches deeper by visiting unvisited vertices and whenever an unvisited vertex is not found, it backtracks to previous vertex to find out whether there are still unvisited vertices.

As seen, the search defined above is inherently recursive. We can find a very simple recursive procedure to visit the vertices in a depth first search. The DFS is more or less similar to pre-order tree traversal. The process can be described as below:

Start from any vertex (source) in the graph and mark it visited. Find vertex that is adjacent to the source and not previously visited using adjacency matrix and mark it visited. Repeat this process for all vertices that is not visited, if a vertex is found visited in this process, then return to the previous step and start the same procedure from there.

If returning back to source is not possible, then DFS from the originally selected source is complete and start DFS using any unvisited vertex.

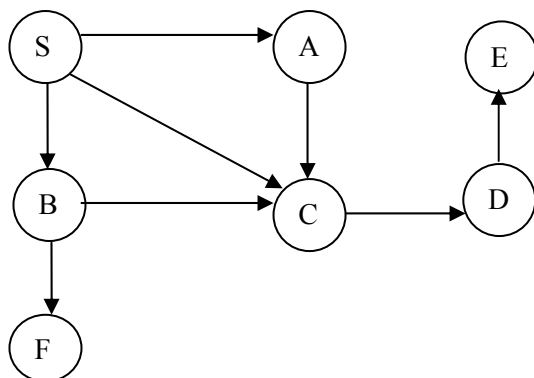


Figure 8.13 : A Digraph

Consider the digraph of *Figure 8.13*. Start with S and mark it visited. Then visit the next vertex A, then C and then D and at last E. Now there are no adjacent vertices of E to be visited next. So, now, backtrack to previous vertex D as it also has no unvisited vertex. Now backtrack to C, then A, at last to S. Now S has an unvisited vertex B. Start DFS with B as a root node and then visit F. Now all the nodes of the graph are visited.

Figure 8.14 shows a DFS tree with a sequence of visits. The first number indicates the time at which the vertex is visited first and the second number indicates the time at which the vertex is visited during back tracking.

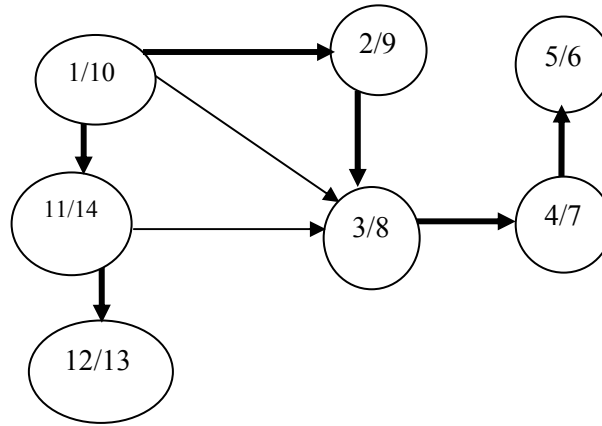


Figure 8.14 : DFS tree of digraph of Figure 8.13

The DFS forest is shown with shaded arrow in Figure 8.14.

Algorithm for DFS

Step 1: Select a vertex in the graph and make it the source vertex and mark it visited.

Step 2: Find a vertex that is adjacent to the source vertex and start a new search if it is not already visited.

Step 3: Repeat step 2 using a new source vertex. When all adjacent vertices are visited, return to previous source vertex and continue search from there.

If n is the number of vertices in the graph and the graph is represented by an adjacency matrix, then the total time taken to perform DFS is $O(n^2)$. If G is represented by an adjacency list and the number of edges of G are e , then the time taken to perform DFS is $O(e)$.

8.7 FINDING STRONGLY CONNECTED COMPONENTS

A beautiful application of DFS is finding a strongly connected component of a graph.

Definition: For graph $G = (V, E)$, where V is the set of vertices and E is the set of edges, we define a strongly connected components as follows:

U is a sub set of V such that u, v belongs to U such that, there is a path from u to v and v to u . That is, all pairs of vertices are reachable from each other.

In this section we will use another concept called transpose of a graph. Given a directed graph G a transpose of G is defined as G^T . G^T is defined as a graph with the same number of vertices and edges with only the direction of the edges being reversed. G^T is obtained by transposing the adjacency matrix of the directed graph G .

The algorithm for finding these strongly connected components uses the transpose of G , G^T .

$G = (V, E)$, $G^T = (V, E^T)$, where $E^T = \{ (u, v) : (v, u) \text{ belongs to } E \}$

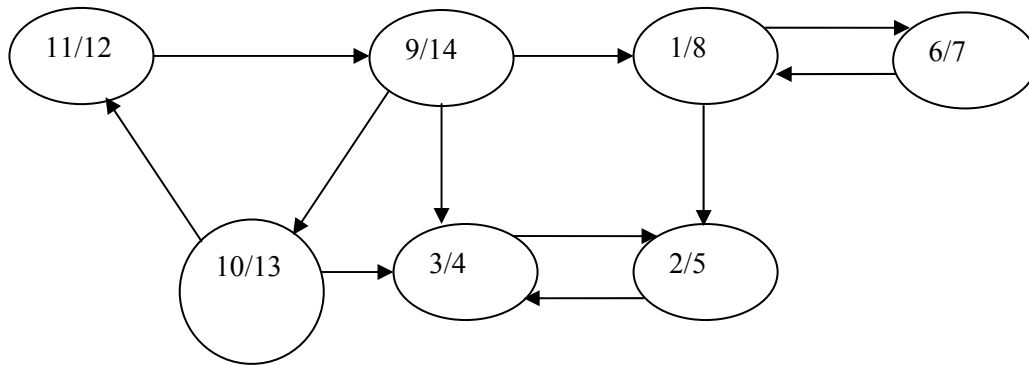


Figure 8.15: A Digraph

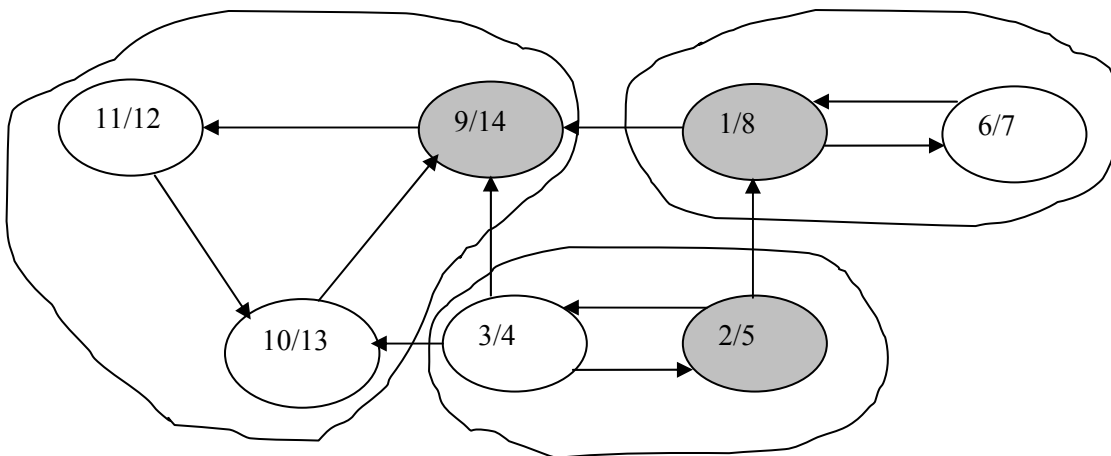


Figure 8.16: Transpose and strongly connected components of digraph of Figure 8.15

Figure 8.15 shows a directed graph with sequence in DFS (first number of the vertex shows the discovery time and second number shows the finishing time of the vertex during DFS. Figure 8.16 shows the transpose of the graph in Figure 8.15 whose edges are reversed. The strongly connected components are shown in zig-zag circle in Figure 8.16.

To find strongly connected component we start with a vertex with the highest finishing time and start DFS in the graph G^T and then in decreasing order of finishing time. DFS with vertex with finishing time 14 as root finds a strongly connected component. Similarly, vertices with finishing times 8 and then 5, when selected as source vertices also lead to strongly connected components.

Algorithm for finding strongly connected components of a Graph:

Strongly Connected Components (G)

where $d[u]$ = discovery time of the vertex u during DFS, $f[u]$ = finishing time of a vertex u during DFS, G^T = Transpose of the adjacency matrix

Step 1: Use DFS(G) to compute $f[u] \forall u \in V$

Step 2: Compute G^T

Step 3: Execute DFS in G^T

Step 4: Output the vertices of each tree in the depth-first forest of Step 3 as a separate strongly connected component.

☞ Check Your Progress 3

- 1) Which graph traversal uses a queue to hold vertices that are to be processed next ?

- 2) Which graph traversal is recursive by nature?

- 3) For a dense graph, Prim’s algorithm is faster than Kruskal’s algorithm
True/False
- 4) Which graph traversal technique is used to find strongly connected component of a graph?

8.8 SUMMARY

Graphs are data structures that consist of a set of vertices and a set of edges that connect the vertices. A graph where the edges are directed is called directed graph. Otherwise, it is called an undirected graph. Graphs are represented by adjacency lists and adjacency matrices. Graphs can be used to represent a road network where the edges are weighted as the distance between the cities. Finding the minimum distance between single source and all other vertices is called single source shortest path problem. Dijkstra’s algorithm is used to find shortest path from a single source to every other vertex in a directed graph. Finding shortest path between every pair of vertices is called all pairs shortest paths problem.

A spanning tree of a graph is a tree consisting of only those edges of the graph that connects all vertices of the graph with minimum cost. Kruskal’s and Prim’s algorithms find minimum cost spanning tree in a graph. Visiting all nodes in a graph systematically in some manner is called traversal. Two most common methods are depth-first and breadth-first searches.

8.9 SOLUTIONS/ANSWERS

Check Your Progress 1

- 1) an acyclic
- 2) symmetric
- 3) The adjacency matrix of the directed graph and undirected graph are as follows:

$$\begin{bmatrix} 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix}$$

(Refer to Figure 8.3)

$$\begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{bmatrix}$$

(Refer to Figure 8.3)

Check Your Progress 2

- 1) Node relaxation
- 2) $O(N^3)$

Check Your Progress 3

- 1) BFS
- 2) DFS
- 3) True
- 4) DFS

8.10 FURTHER READINGS

1. *Fundamentals of Data Structures in C++* by E.Horowitz, Sahni and D.Mehta; Galgotia Publications.
2. *Data Structures and Program Design in C* by Kruse, C.L.Tonodo and B.Leung; Pearson Education.
3. *Data Structures and Algorithms* by Alfred V.Aho; Addison Wesley.

Reference Websites

<http://www.onesmartclick.com/engineering/data-structure.html>
<http://msdn.microsoft.com/vcsharp/programming/datastructures/>
http://en.wikipedia.org/wiki/Graph_theory

UNIT 9 SEARCHING

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9.0 INTRODUCTION

Searching is the process of looking for something: Finding one piece of data that has been stored within a whole group of data. It is often the most time-consuming part of many computer programs. There are a variety of methods, or algorithms, used to search for a data item, depending on how much data there is to look through, what kind of data it is, what type of structure the data is stored in, and even where the data is stored - inside computer memory or on some external medium.

Till now, we have studied a variety of data structures, their types, their use and so on. In this unit, we will concentrate on some techniques to *search* a particular data or piece of information from a large amount of data. There are basically two types of searching techniques, *Linear or Sequential Search and Binary Search*.

Searching is very common task in day-to-day life, where we are involved some or other time, in searching either for some needful at home or office or market, or searching a word in dictionary. In this unit, we see that if the things are organised in some manner, then search becomes efficient and fast.

All the above facts apply to our computer programs also. Suppose we have a telephone directory stored in the memory in an array which contains Name and Numbers. Now, what happens if we have to find a number? The answer is search that number in the array according to name (given). If the names were organised in some order, searching would have been fast.

So, basically a search algorithm is an algorithm which accepts an argument 'a' and tries to find the corresponding data where the match of 'a' occurs in a file or in a table.

9.1 OBJECTIVES

After going through this unit, you should be able to:

- know the basic concepts of searching;
- know the process of performing the Linear Search;
- know the process of performing the Binary Search and
- know the applications of searching.

9.2 LINEAR SEARCH

Linear search is not the most efficient way to search for an item in a collection of items. However, it is very simple to implement. Moreover, if the array elements are arranged in random order, it is the only reasonable way to search. In addition, efficiency becomes important only in large arrays; if the array is small, there aren't many elements to search and the amount of time it takes is not even noticed by the user. Thus, for many situations, linear search is a perfectly valid approach.

Before studying Linear Search, let us define some terms related to search.

A **file** is a collection of records and a record is in turn a collection of fields. A field, which is used to differentiate among various records, is known as a '**key**'.

For example, the telephone directory that we discussed in previous section can be considered as a file, where each record contains two fields: name of the person and phone number of the person.

Now, it depends on the application whose field will be the 'key'. It can be the name of person (usual case) and it can also be phone number. We will locate any particular record by matching the input argument 'a' with the key value.

The simplest of all the searching techniques is *Linear or Sequential Search*. As the name suggests, all the records in a file are searched sequentially, one by one, for the matching of key value, until a match occurs.

The Linear Search is applicable to a table which it should be organised in an array. Let us assume that a file contains 'n' records and a record has 'a' fields but only one key. The values of key are organised in an array say 'm'. As the file has 'n' records, the size of array will be 'n' and value at position R(i) will be the key of record at position i. Also, let us assume that 'el' is the value for which search has to be made or it is the search argument.

Now, let us write a simple algorithm for Linear Search.

Algorithm

Here, m represents the unordered array of elements
 n represents number of elements in the array and
 el represents the value to be searched in the list

Sep 1: [Initialize]

k=0

flag=1

Step 2: Repeat step 3 for k=0,1,2,...n-1

Step 3: if (m[k]=el)

then

flag=0

print "Search is successful" and element is found at location (k+1)

stop

endif

Step 4: if (flag=1) then

print "Search is unsuccessful"

endif

Step 5: stop

Program 9.1 gives the program for Linear Search.

```
/*Program for Linear Search*/
/*Header Files*/
#include<stdio.h>
#include<conio.h>
/*Global Variables*/
int search;
int flag;
/*Function Declarations*/
int input (int *, int, int);
void linear_search (int *, int, int);
void display (int *, int);
/*Functions */
void linear_search(int m[ ], int n, int el)
{
    int k;
    flag = 1;
    for(k=0; k<n; k++)
    {
        if(m[k]==el
        {
            printf("\n Search is Successful\n");
            printf("\n Element : %i Found at location : %i", element, k+1);
            flag = 0;
        }
    }
    if(flag==1)
        printf("\n Search is unsuccessful");
}
void display(int m[ ], int n)
{
    int i;
    for(i=0; i< 20; i++)
    {
        printf("%d", m[i];
    }
}
int input(int m[ ], int n, int el)
{
    int i;
    n = 20;
    el = 30;
    printf("Number of elements in the list : %d", n);
    for(i=0; i<20; i++)
    {
        m[i]=rand( )%100;
    }
    printf("\n Element to be searched :%d", el);
    search = el;
    return n;
}
/* Main Function*/
```

```

void main( )
{
    int n, el, m[200];
    number = input(m, n, el);
    el = search;
    printf("\n Entered list as follows: \n");
    display(m, n);
    linear_search(m, n, el);
    printf("\n In the following list\n");
    display(m, n);
}

```

Program 9.1: Linear Search

Program 9.1 examines each of the key values in the array ‘m’, one by one and stops when a match occurs or the total array is searched.

Example:

A *telephone directory* with $n = 10$ records and Name field as key. Let us assume that the names are stored in array ‘m’ i.e. $m(0)$ to $m(9)$ and the search has to be made for name “Radha Sharma”, i.e. $\text{element} = \text{“Radha Sharma”}$.

Telephone Directory

<i>Name</i>	<i>Phone No.</i>
Nitin Kumar	25161234
Preeti Jain	22752345
Sandeep Singh	23405678
Sapna Chowdhary	22361111
Hitesh Somal	24782202
R.S.Singh	26254444
Radha Sharma	26150880
S.N.Singh	25513653
Arvind Chittora	26252794
Anil Rawat	26257149

The above algorithm will search for $\text{element} = \text{“Radha Sharma”}$ and will stop at 6th index of array and the required phone number is “26150880”, which is stored at position 7 i.e. $6+1$.

Efficiency of Linear Search

How many number of comparisons are there in this search in searching for a given element?

The number of comparisons depends upon where the record with the argument key appears in the array. If record is at the first place, number of comparisons is ‘1’, if record is at last position ‘n’ comparisons are made.

If it is equally likely for that the record can appear at any position in the array, then, a successful search will take $(n+1)/2$ comparisons and an unsuccessful search will take ‘n’ comparisons.

In any case, the order of the above algorithm is $O(n)$.

Check Your Progress 1

- 1) Linear search uses an exhaustive method of checking each element in the array against a key value. When a match is found, the search halts. Will sorting the array before using the linear search have any effect on its order of efficiency?
.....
- 2) In a best case situation, the element was found with the fewest number of comparisons. Where, in the list, would the key element be located?
.....

9.3 BINARY SEARCH

An unsorted array is searched by *linear search* that scans the array elements one by one until the desired element is found.

The reason for sorting an array is that we search the array “quickly”. Now, if the array is sorted, we can employ *binary search*, which brilliantly halves the size of the search space each time it examines one array element.

An array-based binary search selects the middle element in the array and compares its value to that of the key value. Because, the array is sorted, if the key value is less than the middle value then the key must be in the first half of the array. Likewise, if the value of the key item is greater than that of the middle value in the array, then it is known that the key lies in the second half of the array. In either case, we can, in effect, “throw out” one half of the search space or array with only one comparison.

Now, knowing that the key must be in one half of the array or the other, the binary search examines the mid value of the half in which the key must reside. The algorithm thus narrows the search area by half at each step until it has either found the key data or the search fails.

As the name suggests, binary means two, so it divides an array into *two* halves for searching. This search is applicable only to an **ordered table** (in either ascending or in descending order).

Let us write an algorithm for Binary Search and then we will discuss it. The array consists of elements stored in ascending order.

Algorithm

Step 1: Declare an array ‘k’ of size ‘n’ i.e. k(n) is an array which stores all the keys of a file containing ‘n’ records

Step 2: $i \leftarrow 0$

Step 3: $low \leftarrow 0$, $high \leftarrow n-1$

Step 4: while ($low \leq high$)do
 $mid = (low + high)/2$
 if ($key = k[mid]$) then
 write “record is at position”, $mid+1$ //as the array
 starts from the 0th position
 else
 if ($key < k[mid]$) then
 $high = mid - 1$


```

        else
            low = mid + 1
        endif
    endif
endwhile

```

Step 5: Write “Sorry, key value not found”

Step 6: Stop

Program 9.2 gives the program for Binary Search.

```

/*Header Files*/
#include<stdio.h>
#include<conio.h>
/*Functions*/
void binary_search(int array[ ], int value, int size)
{
    int found=0;
    int high=size-1, low=0, mid;
    mid = (high+low)/2;
    printf("\n\n Looking for %d\n", value);
    while((!found)&&(high>=low))
    {
        printf("Low %d Mid%d High%d\n", low, mid, high);
        if(value==array[mid] )
        {printf("Key value found at position %d",mid+1);
          found=1;
        }
        else
        {if (value<array[mid])
            high = mid-1;
          else
            low = mid+1;
          mid = (high+low)/2;
        }
    }
    if (found==1
    printf("Search successful");
    else
    printf("Key value not found");
}
/*Main Function*/
void main(void)
{
    int array[100], i;
    /*Inputting Values to Array*/
    for(i=0;i<100;i++)
    { printf("Enter the name:");
      scanf("%d", array[i]);
    }
    printf("Result of search %d\n", binary_searchy(array,33,100));
    printf("Result of search %d\n", binary_searchy(array, 75,100));
    printf("Result of search %d\n", binary_searchy(array,1,100));
}

```

Program 9.2 : Binary Search

Example:

Let us consider a file of 5 records, i.e., $n = 5$
And k is a sorted array of the keys of those 5 records.

k	
11	0
22	1
33	2
44	3
55	4

Let key = 55, low = 0, high = 4

Iteration 1: $mid = (0+4)/2 = 2$

$k(mid) = k(2) = 33$

Now $key > k(mid)$

So $low = mid + 1 = 3$

Iteration 2: low = 3, high = 4 (low \leq high)

$Mid = 3+4 / 2 = 3.5 \sim 3$ (integer value)

Here $key > k(mid)$

So $low = 3+1 = 4$

Iteration 3: low = 4, high = 4 (low \leq high)

$Mid = (4+4)/2 = 4$

Here $key = k(mid)$

So, the record is at $mid+1$ position, i.e., 5

Efficiency of Binary Search

Each comparison in the binary search reduces the number of possible candidates where the key value can be found by a factor of 2 as the array is divided in two halves in each iteration. Thus, the maximum number of key comparisons are approximately $\log n$. So, the order of binary search is **$O(\log n)$** .

Comparative Study of Linear and Binary Search

Binary search is lots faster than linear search. Here are some comparisons:

NUMBER OF ARRAY ELEMENTS EXAMINED

array size	linear search (avg. case)	binary search (worst case)
8	4	4
128	64	8
256	128	9
1000	500	11
100,000	50,000	18

A **binary search** on an array is $O(\log_2 n)$ because at each test, you can “throw out” one half of the search space or array whereas a **linear search** on an array is $O(n)$.

It is noteworthy that, for very small arrays a **linear search** can prove faster than a **binary search**. However, as the size of the array to be searched increases, the binary

search is the clear winner in terms of number of comparisons and therefore overall speed.

Still, the binary search has some drawbacks. First, it requires that the data to be searched be in sorted order. If there is even one element out of order in the data being searched, it can throw off the entire process. When presented with a set of unsorted data, the efficient programmer must decide whether to sort the data and apply a binary search or simply apply the less-efficient linear search. Is the cost of sorting the data is worth the increase in search speed gained with the binary search? If you are searching only once, then it is probably to better do a linear search in most cases.

Check Your Progress 2

- 1) State True or False
 - a. The order of linear search in worst case is $O(n^2)$ True/False
 - b. Linear search is more efficient than Binary search. True/False
 - c. For Binary search, the array has to be sorted in ascending order only. True/False
- 2) Write the Binary search algorithm where the array is sorted in descending order.

9.4 APPLICATIONS

The searching techniques are applicable to a number of places in today's world, may it be Internet, search engines, on line enquiry, text pattern matching, finding a record from database, etc.

The most important application of searching is to track a particular record from a large file, efficiently and faster.

Let us discuss some of the *applications of Searching* in the world of computers.

1. Spell Checker

This application is generally used in **Word Processors**. It is based on a program for checking spelling, which it checks and searches sequentially. That is, it uses the concept of *Linear Search*. The program looks up a word in a list of words from a dictionary. Any word that is found in the list is assumed to be spelled correctly. Any word that isn't found is assumed to be spelled wrong.

2. Search Engines

Search engines use software robots to survey the Web and build their databases. Web documents are retrieved and indexed using keywords. When you enter a query at a search engine website, your input is checked against the search engine's keyword indices. The best matches are then returned to you as hits. For checking, it uses any of the Search algorithms.

Search Engines use software programs known as robots, spiders or crawlers. A robot is a piece of software that automatically follows hyperlinks from one document to the next around the Web. When a robot discovers a new site, it sends information back to its main site to be indexed. Because Web documents are one of the least static forms of publishing (i.e., they change a lot), robots also update previously catalogued sites. How quickly and comprehensively they carry out these tasks vary from one search engine to the next.

3. String Pattern matching

Document processing is rapidly becoming one of the dominant functions of computers. Computers are used to edit, search and transport documents over the Internet, and to display documents on printers and computer screens. Web ‘surfing’ and Web searching are becoming significant and important computer applications, and many of the key computations in all of this document processing involves character strings and string pattern matching. For example, the Internet document formats HTML and XML are primarily text formats, with added tags for multimedia content. Making sense of the many terabytes of information on the Internet requires a considerable amount of text processing. This is accomplished using *trie* data structure, which is a tree-based structure that allows for faster searching in a collection of strings.

9.5 SUMMARY

Searching is the process of looking for something. Searching a list consisting of 100000 elements is not the same as searching a list consisting of 10 elements. We discussed two searching techniques in this unit namely Linear Search and Binary Search. Linear Search will directly search for the key value in the given list. Binary search will directly search for the key value in the given sorted list. So, the major difference is the way the given list is presented. Binary search is efficient in most of the cases. Though, it had the overhead that the list should be sorted before search can start, it is very well compensated through the time (which is very less when compared to linear search) it takes to search. There are a large number of applications of Searching out of whom a few were discussed in this unit.

9.6 SOLUTIONS / ANSWERS

Check Your Progress 1

- 1) No
- 2) It will be located at the beginning of the list

Check Your Progress 2

- 1) (a) F
(b) F
(c) F

9.7 FURTHER READINGS

Reference Books

1. *Fundamentals of Data Structures in C++* by E. Horowitz, Sahai and D. Mehta, Galgotia Publications.
2. *Data Structures using C and C ++* by Yedidyah Hangsam, Moshe J. Augenstein and Aaron M. Tanenbaum, PHI Publications.
3. *Fundamentals of Data Structures in C* by R.B. Patel, PHI Publications.

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