

- Algorithm: It is a combination of sequence of finite steps to solve a particular problem. ①

Properties of Algorithm:

- output should be generated after finite time.

- There should be at least one input.
- It's independent from programming language.

Difference between Algorithm and Program:

Algorithm

- Written at Design stage
- Need domain expert
- Written in any language
- H/W or OS independent
- Can be Analyze

Program

- Written at implementation stage
- Need programmer
- Written in programming language
- H/W or OS dependent
- Can be tested

Pseudo code: It's a description of an algorithm that is more structured than usual prose but less formal than a programming language.

- Pseudo code is our preferred notation for describing algorithms.

Measuring the running time of an algorithm:

App 1: Experimental study: Write a program that implements the algorithm.

App 2: Frequency count Method:

Prob 1:

```

main()
{
    x = y + z; → 1
}

```

$O(1)$

Prob 3:

```

main()
{
    x = y + z; → 1
    for (i = 1; i <= n; i++)
    {
        x = y + z; → n
    }
    for (i = 1; i <= n; i++)
    {
        for (j = 1; j <= n; j++)
        {
            x = y + z; → n^2
        }
    }
}

```

$$\frac{n^2 + n + 1}{O(n^2)}$$

Prob 2:

```

main()
{
    x = y + z; → 1
    for (i = 1; i <= n; i++)
    {
        x = y + z; → n
    }
}

```

$$\frac{n+1}{O(n)}$$

Prob 4:

```

main()
{
    while (n > 1)
    {
        n = n - 10 →  $\frac{n}{10}$ 
    }
}

```

$\Rightarrow O(n)$

Prob 5:

```

main()
{
    i = 0
    while (i <= n)
    {
        i = i + 5 →  $\frac{n}{5}$ 
    }
}

```

$\Rightarrow O(n)$

Prob 6:

```

main()
{
    while (n > 1)
    {
        n =  $\frac{n}{2}$ ;
    }
}

```

n	\vdots	$k = \log_2 n$
$\frac{n}{2}$	\vdots	$O(\log n)$
$\frac{n}{2^2}$	\vdots	$\frac{n}{2^k} = 1$

Prob 7:

```
main ()
{
  i = 1
  while (i <= n)
  {
    i = 2 * i;
  }
}
```

$$\begin{aligned}
 &i = 1 \\
 &2 * 1 = 2 \\
 &2^2 \\
 &\vdots \\
 &2^k = n \\
 &k = \log_2 n \\
 &O(\log_2 n)
 \end{aligned}$$

Prob 8:

(3)

```
main ()
{
  while (n > 2)
  {
    n = n1/2;
  }
}
```

$$\begin{aligned}
 &n \\
 &/ \\
 &n^{1/2} \\
 &/ \\
 &n^{1/4} \\
 &\vdots \\
 &n^{1/2^k}
 \end{aligned}$$

$$n^{\frac{1}{2^k}} = 2$$

$$\frac{1}{2^k} \log_2 n = \log_2 2$$

$$\log_2 n = 2^k$$

$$k = \log_2 \log_2 n$$

$O(\log_2 \log_2 n)$

Prob 9:

```
main ()
{
  while (n > 23)
  {
    n = n1/255;
  }
}
```

$$\begin{aligned}
 &n \\
 &/ \\
 &n^{1/255} \\
 &/ \\
 &n^{1/255^2} \\
 &\vdots \\
 &n^{1/255^k}
 \end{aligned}$$

$$n^{\frac{1}{255^k}} = 23$$

$$\frac{1}{255^k} \log_{23} n = \log_{23} 23$$

$$k = \log_{255} \log_{23} n$$

$O(\log_{255} \log_{23} n)$

Prob 10:

```
main()
{
  while (n > 15)
  {
    n = n1/5
  }
}
```

n
 $|$
 $n^{1/5}$
 $|$
 $n^{1/5^2}$
 $|$
 $n^{1/5^k}$

$$n^{1/5^k} = 15$$

$$\frac{1}{5^k} \log_{15} n = \log_{15} 15$$

$$\log_{15} n = 5^k$$

$$k = \log_5 \log_{15} n$$

$$O(\log_5 \log_{15} n)$$

Prob 11:

```
main()
{
  i = 2
  while (i < n)
  {
    i = i2
  }
}
```

2
 $|$
 2^2
 $|$
 $(2^2)^2$
 $|$
 $(2^2)^k = n$

$$2^k = \log_2 n$$

$$k = \log_2 \log_2 n$$

$$O(\log_2 \log_2 n)$$

Prob 12:

```
main()
{
  i = 3
  while (i < n)
  {
    i = i3
  }
}
```

3
 $|$
 3^2
 $|$
 $(3^2)^2$
 $|$
 $(3^2)^k = n$

$$3^k = \log_3 n$$

$$k = \log_2 \log_3 n$$

$$O(\log_2 \log_3 n)$$

→ Asymptotic Notations: The main idea of asymptotic analysis is to have a measure of efficiency of algorithms that doesn't depend on machine specific constants.

- Asymptotic notations are the mathematical notations used to describe the running time (Time complexity) of an algorithm.

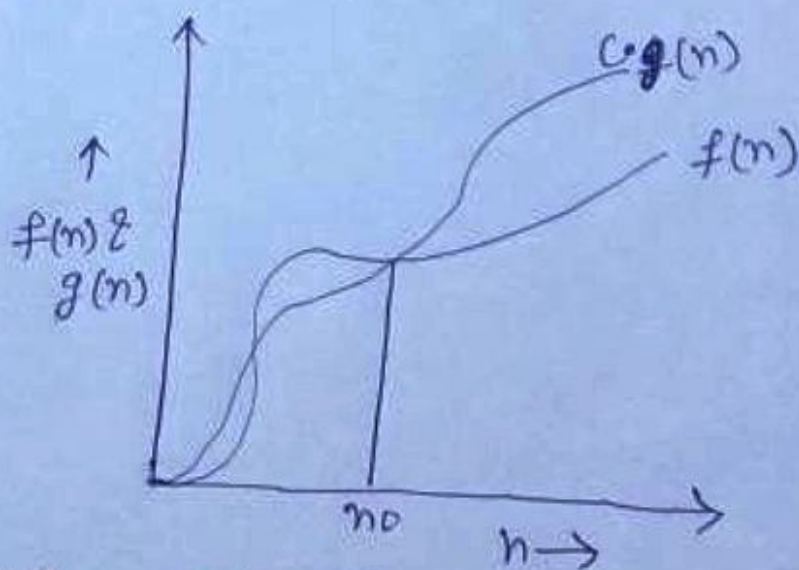
1. Big O Notation:
- Worst Case
 - Upper bound

Let $f(n)$ & $g(n)$ are two +ve functions.

$$f(n) = O(g(n))$$

iff

$f(n) \leq C \cdot g(n), \forall n, n > n_0$ where C is a constant and value of $C > 0$, n_0 is constant and value of $n_0 > 1$.



P-1 $g(n) = n^2$, $f(n) = n^2 + n + 10$

P-2 $f(n) = n + 10$, $g(n) = n - 10$

P-3 $f(n) = n^2$, $g(n) = n$

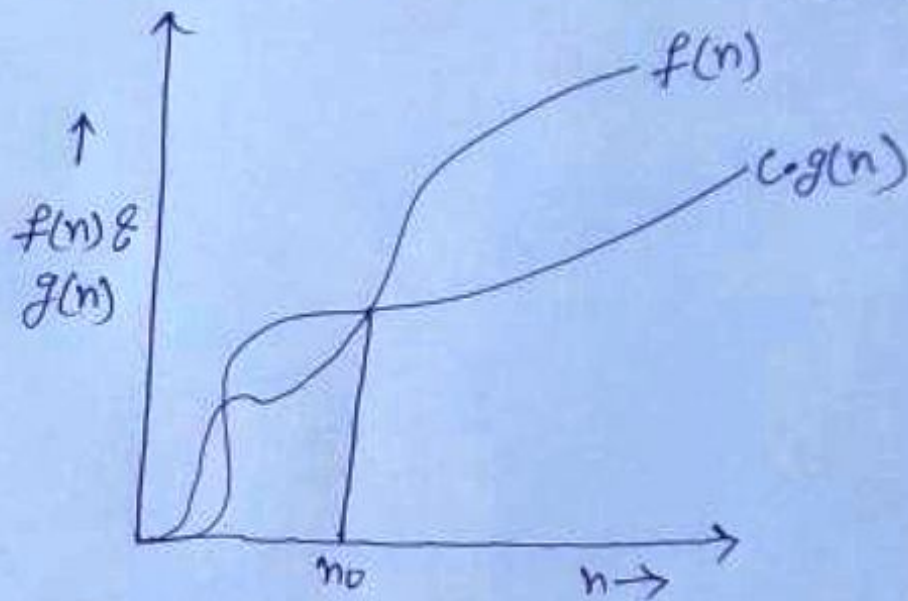
2. Big omega Notation (Ω)

Let $f(n)$ & $g(n)$ are two +ve functions.

$$f(n) = \Omega(g(n))$$

iff

$f(n) \geq c \cdot g(n), \forall n, n > n_0$, where c is a constant and value of $c > 0$, n_0 is constant and value of $n_0 > 1$.



P-1 $f(n) = n, g(n) = n+10$

P-2 $f(n) = n^2 + n + 10, g(n) = n^2$

P-3 $f(n) = n, g(n) = n^2$

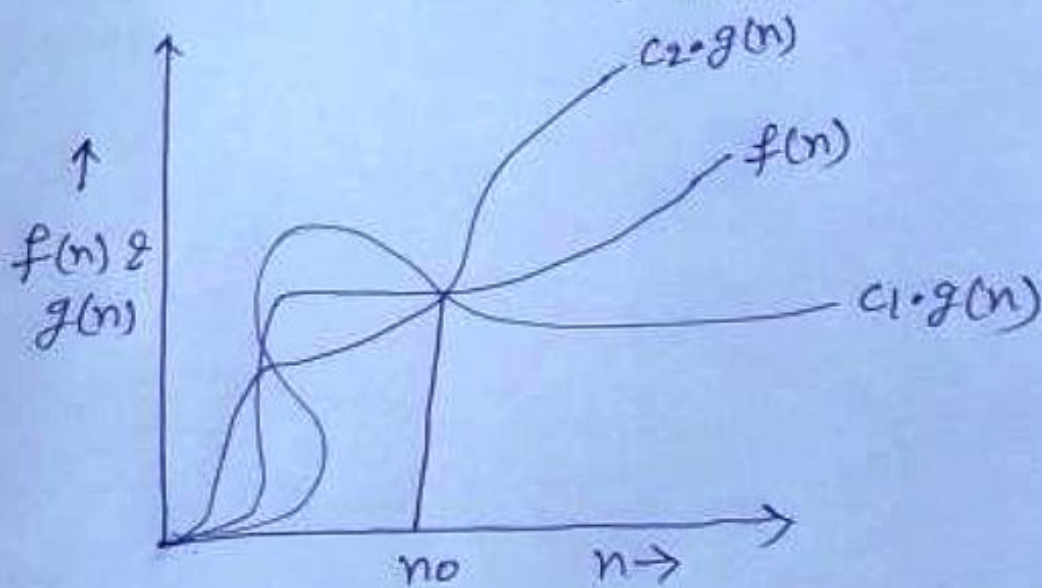
3. Theta Notation (Θ)

Let $f(n)$ & $g(n)$ are two +ve functions.

$$f(n) = \Theta(g(n))$$

iff

\exists $C_1 \cdot g(n) \leq f(n) \leq C_2 \cdot g(n), \forall n, n > n_0$,
where C_1, C_2 are constants and value of $C_1, C_2 > 0$,
 n_0 is a constant and value of $n_0 > 1$.



P-1 $f(n) = n, g(n) = n + 10$

P-2 $f(n) = n, g(n) = n$

P-3 $f(n) = n^2, g(n) = n^2 + n + 10$

P-4 $f(n) = n^2, g(n) = n$

→ Recursion:- A function is calling that itself to solve a particular problem is called a recursion. ①

- Recursion is nothing but solving bigger problem in terms of smaller problem.
- To execute the recursive program we used stack data structure.
- Every recursion program should have termination condition.

→ Recurrence relation of factorial:

$$\text{fact}(n) = \begin{cases} 1, & \text{if } n \leq 1 \\ n * \text{fact}(n-1), & \text{otherwise} \end{cases}$$

→ Recurrence relation of fibonacci series:

$$\text{fib}(n) = \begin{cases} n, & \text{if } n=0 \text{ || } n=1 \\ \text{fib}(n-1) + \text{fib}(n-2), & \text{otherwise} \end{cases}$$

→ Recurrence relation of GCD:

$$\text{GCD}(m, n) = \begin{cases} \infty, & \text{if } m=0 \text{ \& } n=0 \\ m, & n=0 \\ n, & m=0 \\ \text{GCD}(n \% m, m), & \text{otherwise} \end{cases}$$

→ Recurrence relation of multiplication of two numbers:

$$\text{mul}(m, n) = \begin{cases} 0, & m=0 \text{ || } n=0 \\ m + \text{mul}(m, n-1), & \text{otherwise} \end{cases}$$

→ Recurrence Relation Solving :-

- ① Iterative / Back Substitution Method
- ② Recursive Tree method
- ③ Master Theorem

Ex-1 $T(n) = \begin{cases} 1, & \text{if } n=1 \\ T(n-1)+n, & \text{if } n>1 \end{cases}$

$$T(n) = T(n-1) + n$$

$$= T(n-2) + (n-1) + n$$

$$= T(n-3) + (n-2) + (n-1) + n$$

⋮

$$T(n-k) + (n-k+1) + \dots + (n-1) + n$$

Put
 $n-k=1$

$$T(1) + 2 + \dots + (n-1) + n$$

$$1 + 2 + 3 + \dots + (n-1) + n$$

$$\frac{n(n+1)}{2}$$

$$\boxed{O(n^2)}$$

Ex-2 $T(n) = \begin{cases} 1, & \text{if } n=1 \\ T(n-1)+n, & \text{if } n>1 \end{cases}$

Ex-3 $T(n) = \begin{cases} 1, & \text{if } n=1 \\ T(n-1) + \log n, & \text{if } n>1 \end{cases}$

Ex-4

$$T(n) = \begin{cases} 0, & \text{if } n=0 \\ T(n-2) + n^2, & \text{if } n > 0 \end{cases}$$

Ex-5

$$T(n) = \begin{cases} 0, & \text{if } n=0 \\ T(n-2) + \log n, & \text{if } n > 0 \end{cases}$$

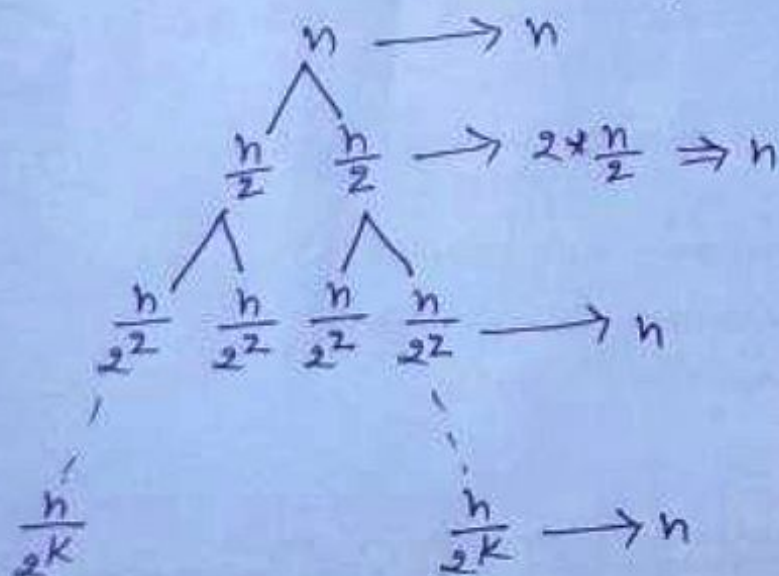
Ex-6

$$T(n) = \begin{cases} 1, & \text{if } n=1 \\ T\left(\frac{n}{2}\right) + n, & \text{if } n > 1 \end{cases}$$

→ Recursive Tree Method:

Ex-1

$$T(n) = n + T\left(\frac{n}{2}\right) + T\left(\frac{n}{2}\right)$$



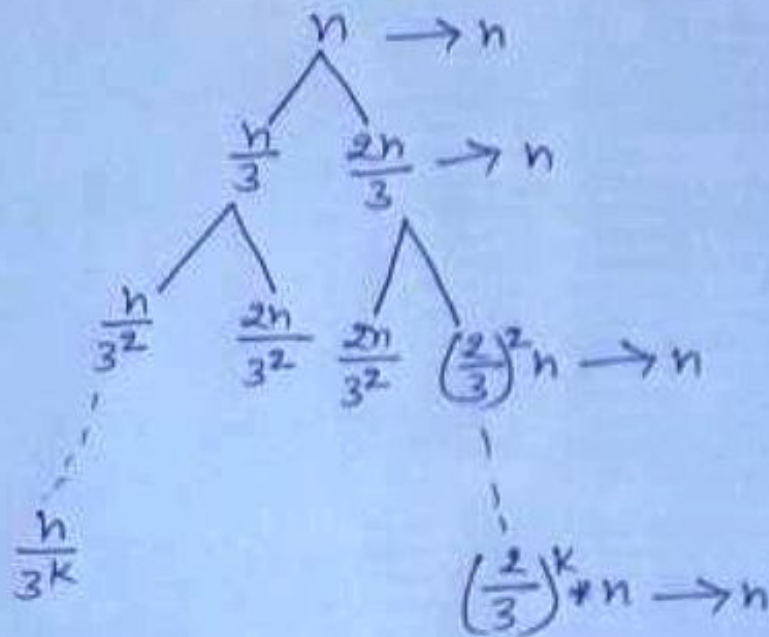
$$\frac{n}{2^K} = 1, \quad K = \log_2 n, \quad T(n) = O(n \log n)$$

$$T(n) = \Theta(n \log n)$$

EX-2

$$T(n) = n + T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right)$$

④



$$\frac{n}{3^k} = 1, k = \log_3 n$$

$$\left(\frac{2}{3}\right)^k n = 1, k = \log_{3/2} n$$

$$T(n) = \mathcal{O}(n \log_3 n)$$

$$, T(n) = \mathcal{O}(n \log_{3/2} n)$$

EX-3

$$T(n) = n + T\left(\frac{n}{100}\right) + T\left(\frac{99n}{100}\right)$$

EX-4

$$T(n) = n + T\left(\frac{n}{5}\right) + T\left(\frac{4n}{5}\right)$$

→ Master Theorem:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$a > 1, b > 1$, a & b are constants, $f(n) \rightarrow +ve$.

Find $n^{\log_b a}$

Case-I If $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$, then $T(n) = \theta(n^{\log_b a})$

Case-II If $f(n) = \theta(n^{\log_b a})$, then $T(n) = \theta(n^{\log_b a} \cdot \log n)$

Case-III If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$, and if $a f\left(\frac{n}{b}\right) \leq C f(n)$ for some constant $C < 1$ and all sufficiently large n , then

$$T(n) = \theta(f(n))$$

P-1 $T(n) = 8T\left(\frac{n}{2}\right) + n^2$

P-2 $T(n) = 8T\left(\frac{n}{2}\right) + n^4$

P-3 $T(n) = 2T\left(\frac{n}{2}\right) + n$

Note: If $n^{\log_b a}$ is logarithmic time smaller than $f(n)$ then if $f(n) = \theta(n^{\log_b a} \cdot (\log n)^K)$, where K is constant, $K > 0$

$$T(n) = \theta(n^{\log_b a} \cdot (\log n)^{K+1})$$

P-1 $T(n) = 2T\left(\frac{n}{2}\right) + n \log n$

Note : If recurrence relation contain root operator:

Ex-1 $T(n) = T(\sqrt{n}) + c$

Assume $n = 2^k$, $k = \log_2 n$

$$T(2^k) = T(2^{k/2}) + c$$

$$T(2^k) = S(k)$$

$$S(k) = S\left(\frac{k}{2}\right) + c$$

$$a=1, b=2, f(n)=c$$

Case III hold

$$S(k) = \theta(\log k)$$

$$T(2^k) = \theta(\log k)$$

$$\boxed{T(n) = \theta(\log \log n)}$$

Ex-2 $T(n) = 2T(\sqrt{n}) + \log n$

Assume $n = 2^k$

$$T(2^k) = 2T(2^{k/2}) + \log 2^k$$

$$T(2^k) = S(k)$$

$$S(k) = 2S\left(\frac{k}{2}\right) + \log 2^k$$

Case III hold

$$S(k) = \theta(k \log k)$$

$$T(2^k) = \theta(k \log k)$$

$$\boxed{T(n) = \theta(\log n \log \log n)}$$

HeapSort (A)

Heap Sort

Build-Max-Heap(A)

for $i = \text{length}(A)$ to 2

Swap($A[i]$ with $A[i]$)

heap-size[A] = heap-size[A] - 1

Max-Heapify(A, i)

- Comparison based sorting
- Unstable sorting
- In place sorting

Build-Max-Heap(A)

heap-size[A] = length[A]

for $i = \text{length}[A]/2$ to 1

Max-Heapify(A, i)

Time Complexity

BC = $O(n)$

AC = $O(n \log n)$

WC = $O(n \log n)$

Max-Heapify(A, i)

L = left(i)

R = right(i)

if $L \leq A.\text{heap-size}$ and $A[L] > A[i]$

largest = L

else

largest = i

if $R \leq A.\text{heap-size}$ and $A[R] > A[\text{largest}]$

largest = R

if largest $\neq i$

Swap $A[i]$ with $A[\text{largest}]$

Max-Heapify(A, largest)

End

Shell Sort

```
for (gap =  $\frac{n}{2}$ ; gap > 1; gap /= 2)
{
    for (j = gap; j < n; j++)
    {
        for (i = j - gap; i > 0; i -= gap)
        {
            if (a[i+gap] > a[i])
            {
                break;
            }
            else
            {
                swap(a[i+gap], a[i]);
            }
        }
    }
}
```

Array:

0	1	2	3	4	5	6	7	8
23	29	15	19	31	7	9	5	2

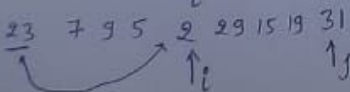
Pass-1 $gap_1 = \lfloor \frac{N}{2} \rfloor = \lfloor \frac{9}{2} \rfloor \Rightarrow 4$

23 7 15 19 31 29 9 5 2
 ↑ ↑
 i j

23 7 9 19 31 29 15 5 2
 ↑ ↑
 i j

23 7 9 5 31 29 15 19 2
 ↑ ↑
 i j

23 7 9 5 2 29 15 19 31
 ↑ ↑
 i j



2 7 9 5 23 29 15 19 31
 i j

$$gap_2 = \frac{gap_1}{2} = \frac{4}{2} = 2$$

pass-2

2 7 9 5 23 29 15 19 31
 i i j j

2 5 9 7 23 29 15 19 31
 i i j i j j

2 5 9 7 15 29 23 19 31
 i i j j

2 5 9 7 15 19 23 29 31
 i i j j

pass-3

2 5 9 7 15 19 23 29 31
 i j i j i j

$$gap_3 = \frac{gap_2}{2}$$

2 5 7 9 15 19 23 29 31 $\Rightarrow \frac{2}{2} = 1$
 i j i j i j i j i j

2	5	7	9	15	19	23	29	31
---	---	---	---	----	----	----	----	----

works as
like Insertion
sort

B-Trees

- B-trees are balanced search trees designed to work well on disks or other direct access secondary storage devices.
- B-Trees are better at minimizing disks I/O operations.
- Database systems use B-Trees and its variants for indexing.

→ A B-tree T is a rooted tree (whose root is $T.root$) having following properties:

1. Every node x has the following attributes:
 - a. $x.n$, the number of keys currently stored in node x .
 - b. The $x.n$ keys themselves, $x.key_1, x.key_2, \dots, x.key_n$, stored in non-decreasing, so that $x.key_1 \leq x.key_2 \leq \dots \leq x.key_n$.
 - c. $x.leaf$, a Boolean value that is true if x is a leaf and false if x is an internal node.
2. Each internal node x also contains $x.n+1$ pointers $x.c_1, x.c_2, \dots, x.c_{n+1}$ to its children. Leaf nodes have no children, and so their c_i attributes are undefined.
3. The key $x.key_i$ separates the ranges of keys stored in each subtree, if k_i is any key stored in the subtree with root $x.c_i$ then,

$$k_i \leq x.key_1 \leq x.key_2 \leq \dots \leq x.key_n \leq k_{n+1}$$

4. All leaves have the same depth, which is the tree's height h .

5. Nodes have lower and upper bounds, on the number of keys, they can contain.

These bounds are fixed integers $t \geq 2$ is called minimum degree of B-tree:

a. Every node other than the root, must have at least $(t-1)$ keys.

Every internal node other than the root thus has at least t children. If the tree is non-empty, the root must have at least one key.

b. Every node must contain at most $(2t-1)$ keys.

Therefore an internal node may have at most $2t$ children. A node is full if it has $(2t-1)$ keys.

B-Trees - Creation

Array = 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15

Degree $t = 2$

Soln: Min Keys = $(t-1) = 1$
Max Keys = $(2t-1) = 3$

Step-1 Insert-1

1		
---	--	--

Step-2 Insert-2

1	2	
---	---	--

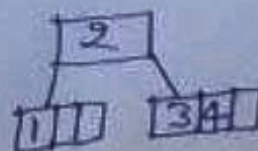
Step-3 Insert-3

1	2	3
---	---	---

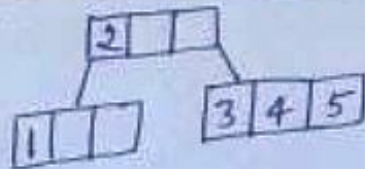
Step-4 Insert-4

1	2	3
---	---	---

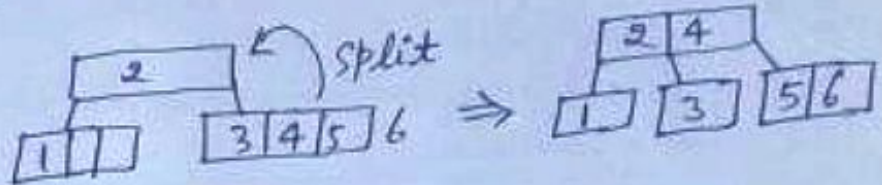
 4 \Rightarrow Split



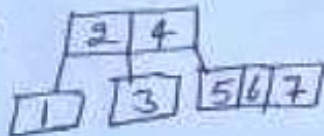
Step-5 Insert-5



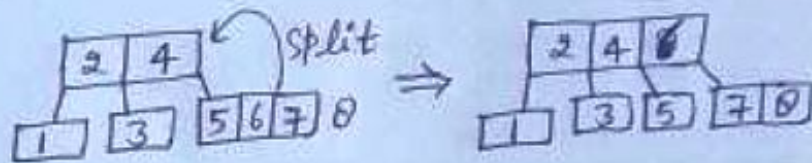
Step-6 Insert-6



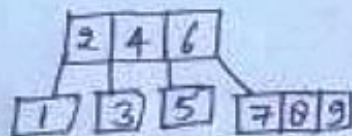
Step-7 Insert-7



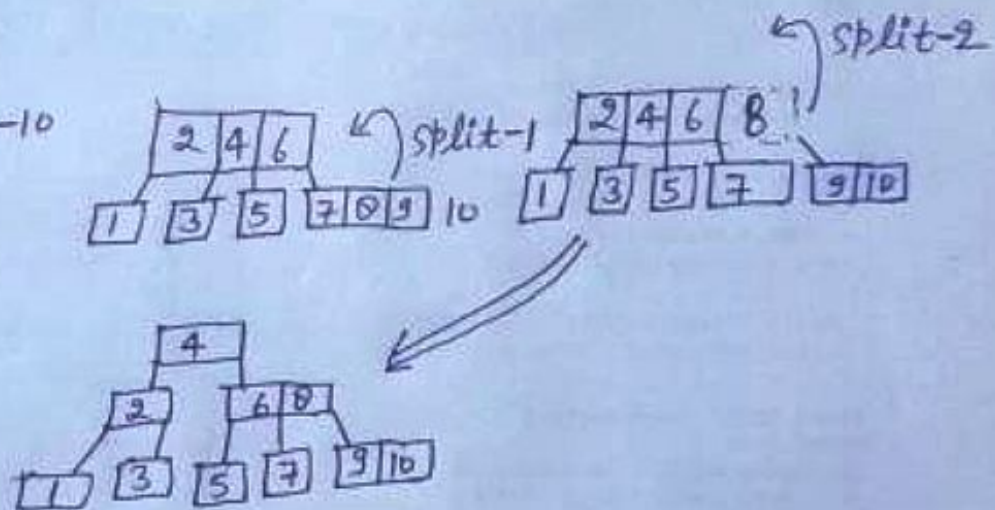
Step-8 Insert-8



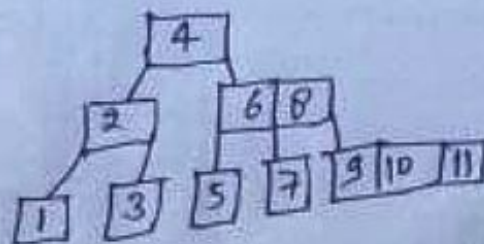
Step-9 Insert-9



Step-10 Insert-10

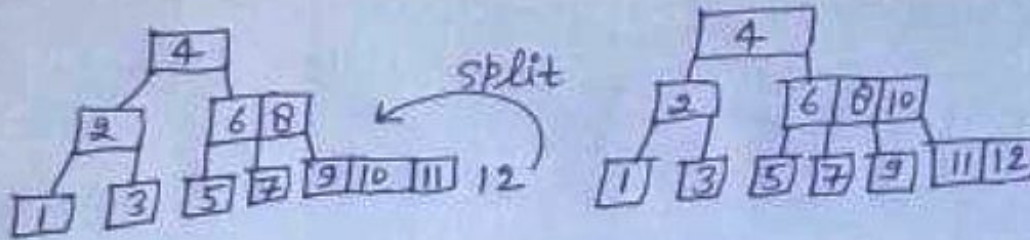


Step-11 Insert-11



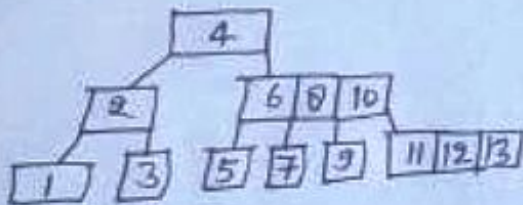
Step-12

Insert -12



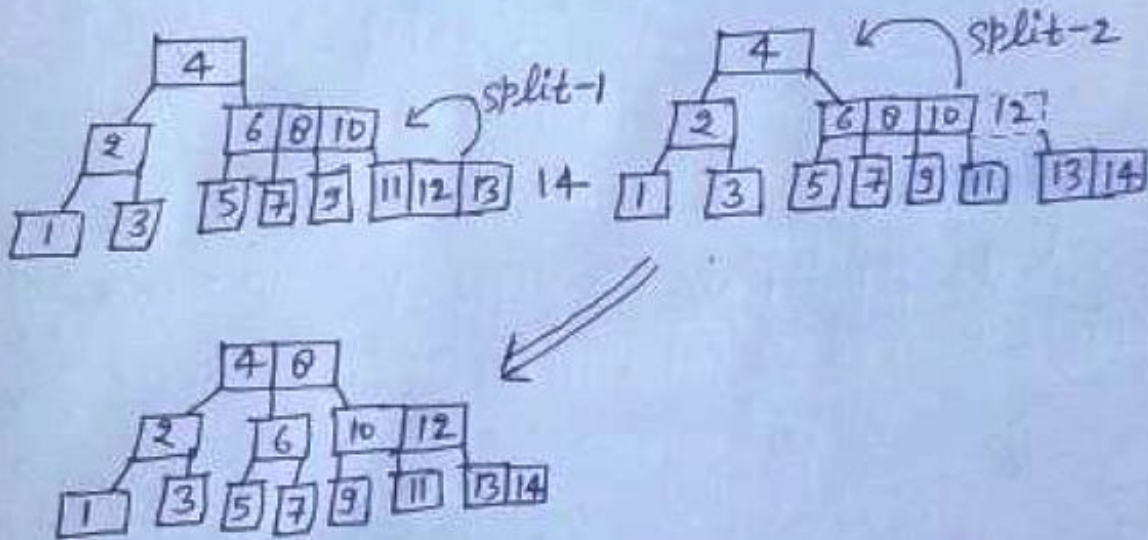
Step-13

Insert -13



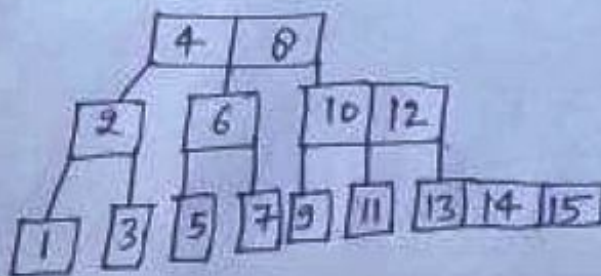
Step-14

Insert -14



Step-15

Insert -15



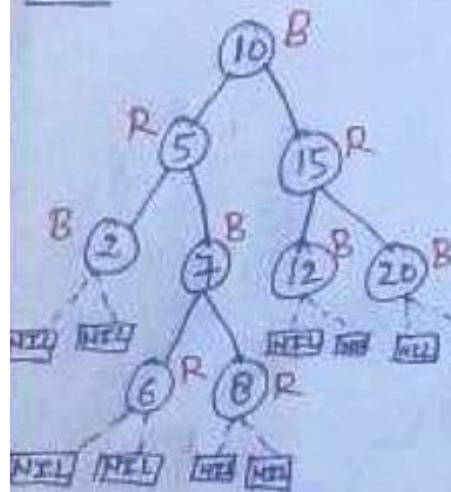
Red-Black Trees

A red-black tree is a binary search tree with one extra bit of storage per node: its color, which can be either RED or BLACK. By constraining the node colors on any simple path from the root to a leaf, red-black trees ensures that no such path is more than twice as long as any other, so that the tree is approximately balanced.

A red-black tree is a binary tree that satisfies the following red-black properties: \rightarrow Red-Black tree is a self-balancing BST.

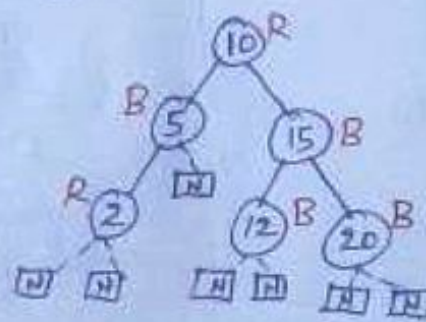
1. Every node is either red or black.
2. The root is black.
3. Every leaf (NIL) is black.
4. If a node is red, then both its children are black.
5. For each node, all simple paths from the node to descendant leaves contain the same number of black nodes.

EX-1



R-B Tree (✓)

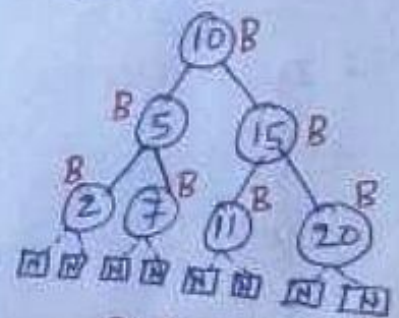
EX-2



R-B Tree (X)

(Violate Property 2)

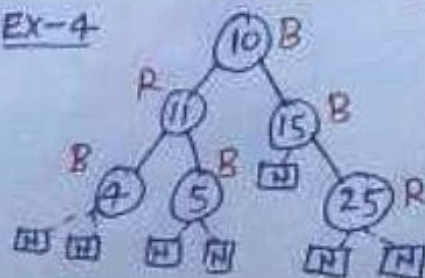
EX-3



R-B Tree (✓)

Even not a single red color node in a tree.

EX-4



R-B Tree (X)

Even though it follows all the properties of R-B tree but it's not a BST.

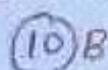
Insertion in Red-Black Tree

Algorithms:

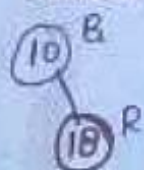
- ① If tree is empty, ~~new~~ create new node as root node with color Black.
- ② If tree is not empty, create new node as leaf node with color Red.
- ③ If Parent of new node is 'black' then exit.
- ④ If Parent of new node is 'red' then check the color of parent's sibling or (uncle):
 - Ⓐ If color is black or null then do suitable rotation & recolor.
 - Ⓑ If color is red then recolor both parent and sibling & also check if parent's parent of new node is not root node then recolor it & recheck.

Array: 10, 18, 7, 15, 16, 30, 25, 40, 60, 20, 70

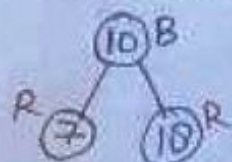
Step 1: Insert element 10



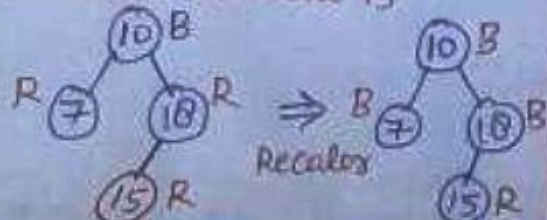
Step 2: Insert element 18



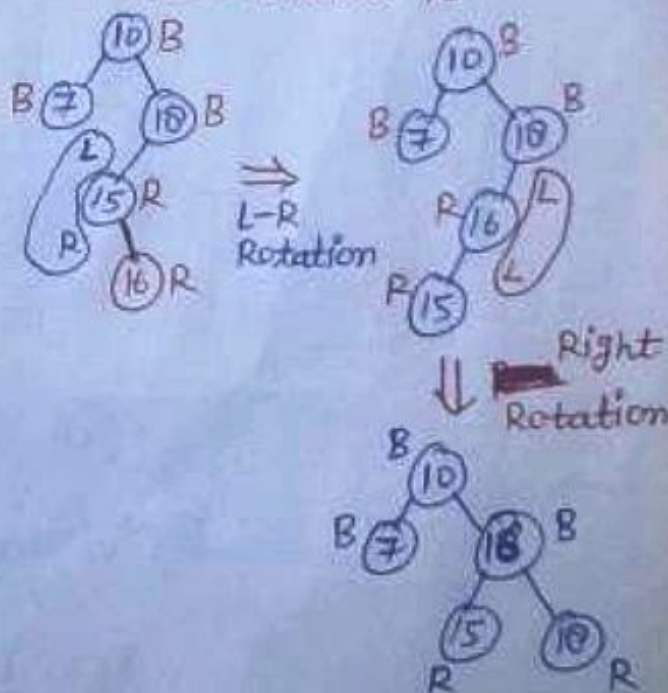
Step 3: Insert element 7



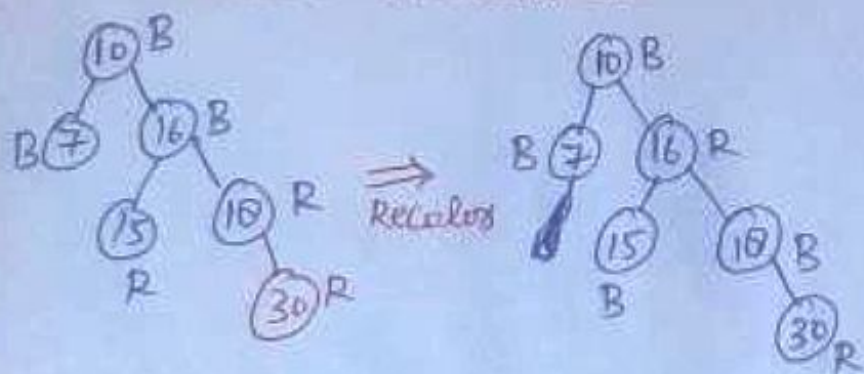
Step 4: Insert element 15



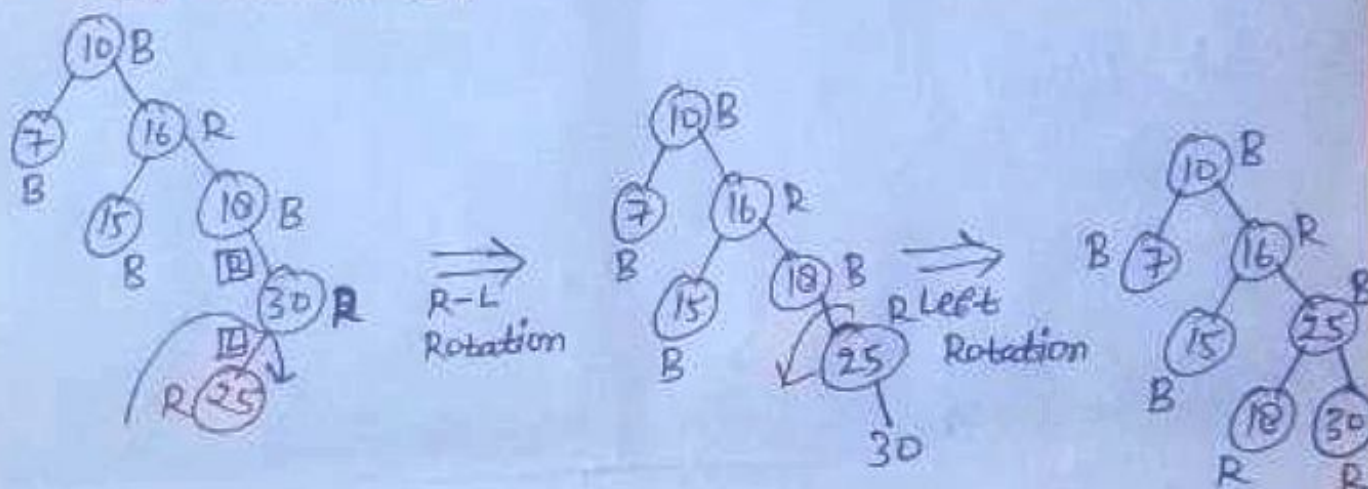
Step 5: Insert element 16



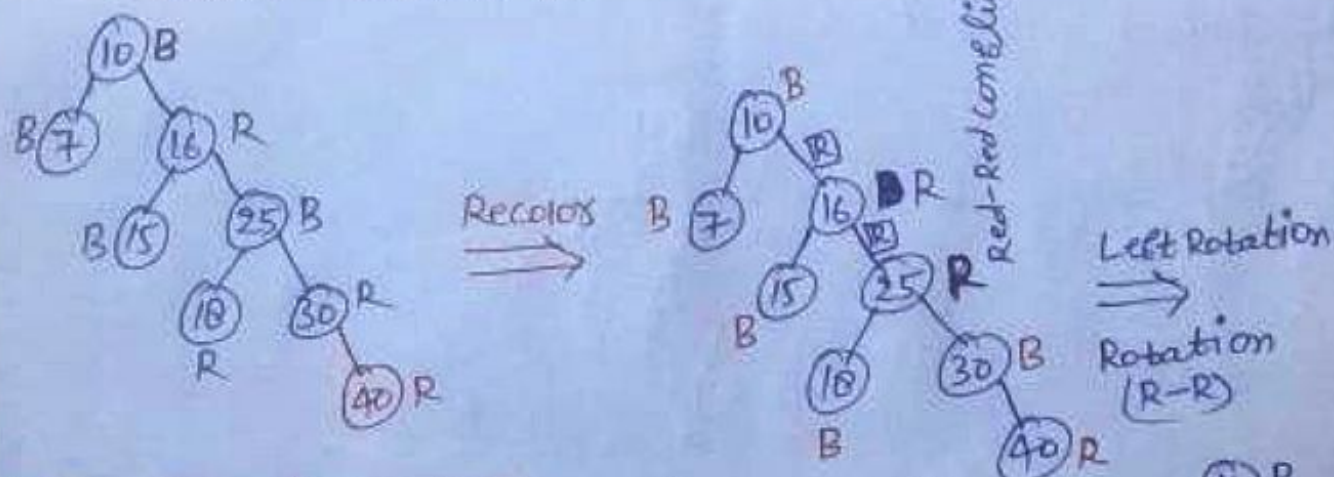
Step 6: Insert element 30



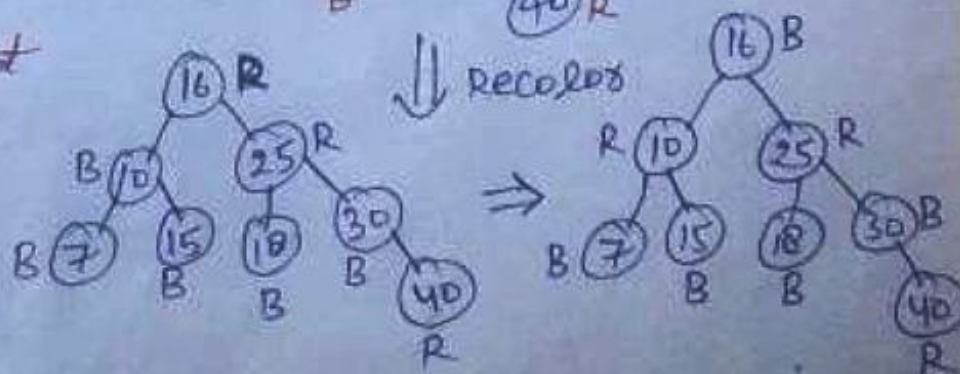
Step 7: Insert element 25



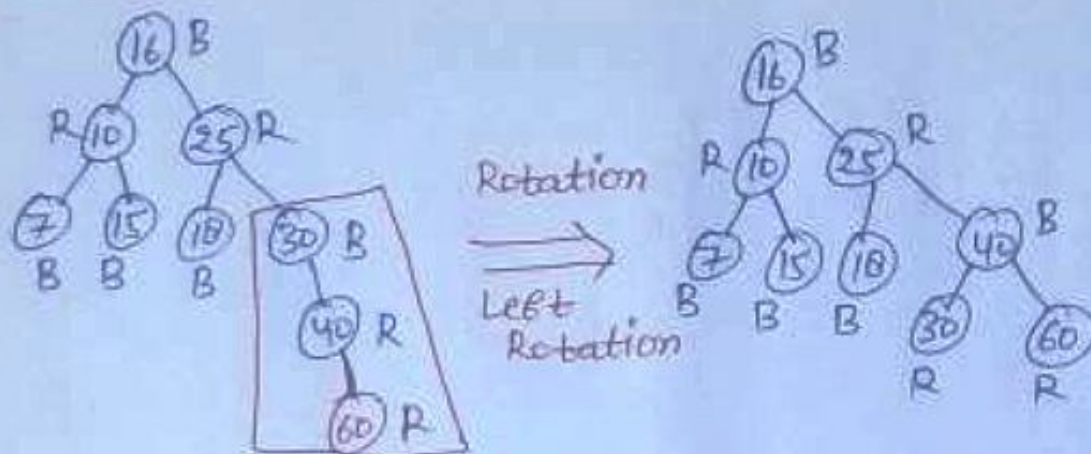
Step 8: Insert element 40



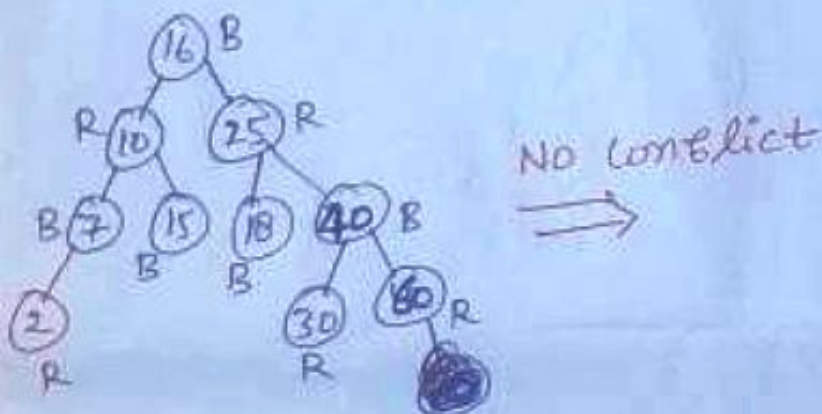
~~Step 9: Insert element~~



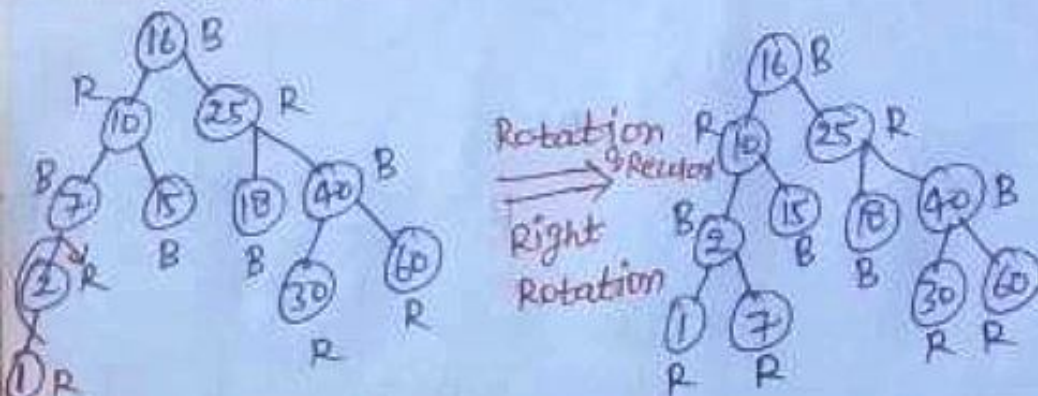
Step 9: Insert element 60



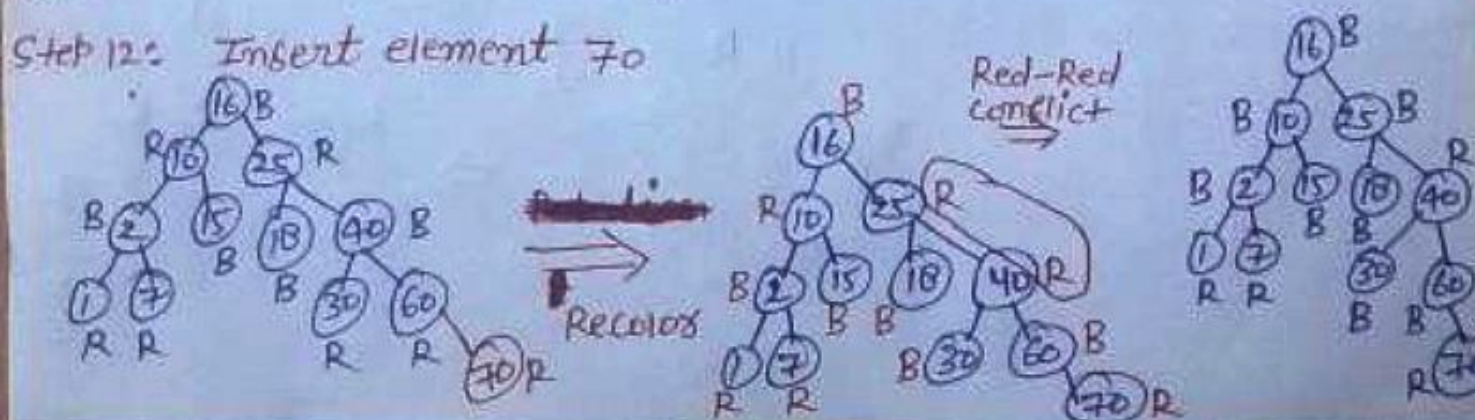
Step 10: Insert element 2



Step 11: Insert element 1



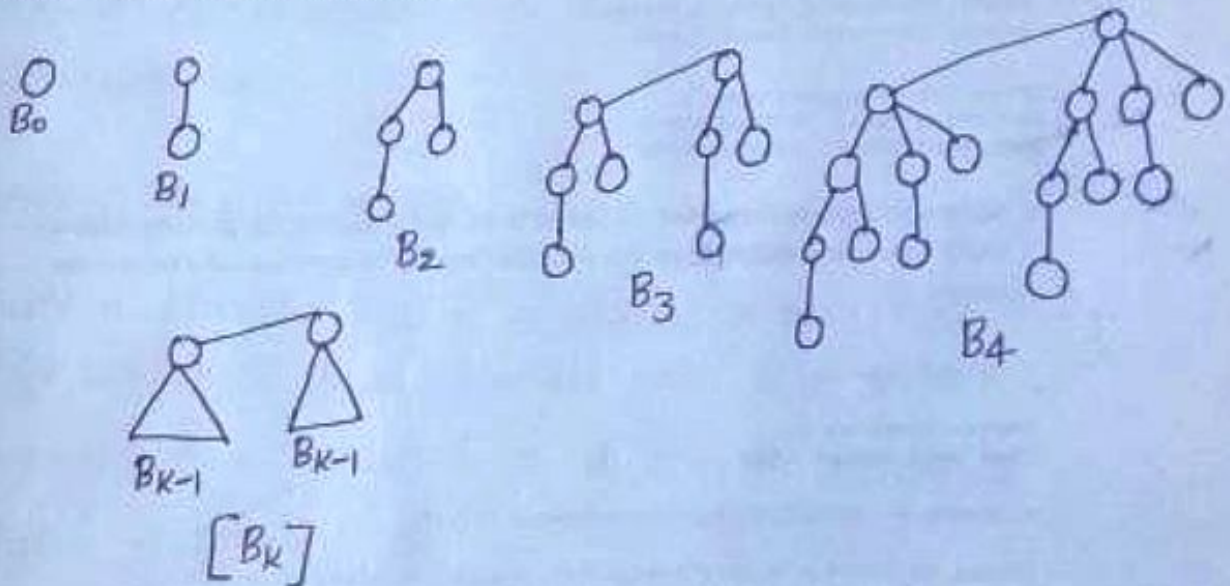
Step 12: Insert element 70



Binomial Heap

①

- Binomial Heap is a collection of Binomial Tree.
- The Binomial Tree B_k is an ordered tree defined recursively.
- The Binomial Tree B_0 consist of a single node.
- The Binomial Tree B_k consist of two Binomial Tree B_{k-1} that are linked together.



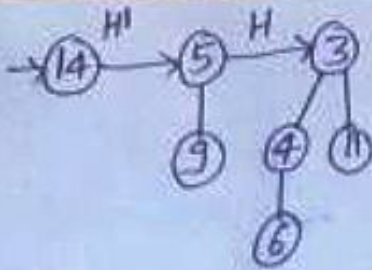
Properties of Binomial Tree (B_k)

1. There are 2^k nodes.
2. The height of the tree is k .
3. There are exactly kC_i nodes at depth i for $i=0, 1, 2, \dots, k$.
4. The root has degree k , which is greater than that of any other node.
5. If i the children of the root are numbered from left to right by $k-1, k-2, \dots, 0$ child i is the root of a subtree B_i .

Binomial Tree:- Binomial Tree B_k is an ordered tree defined recursively.

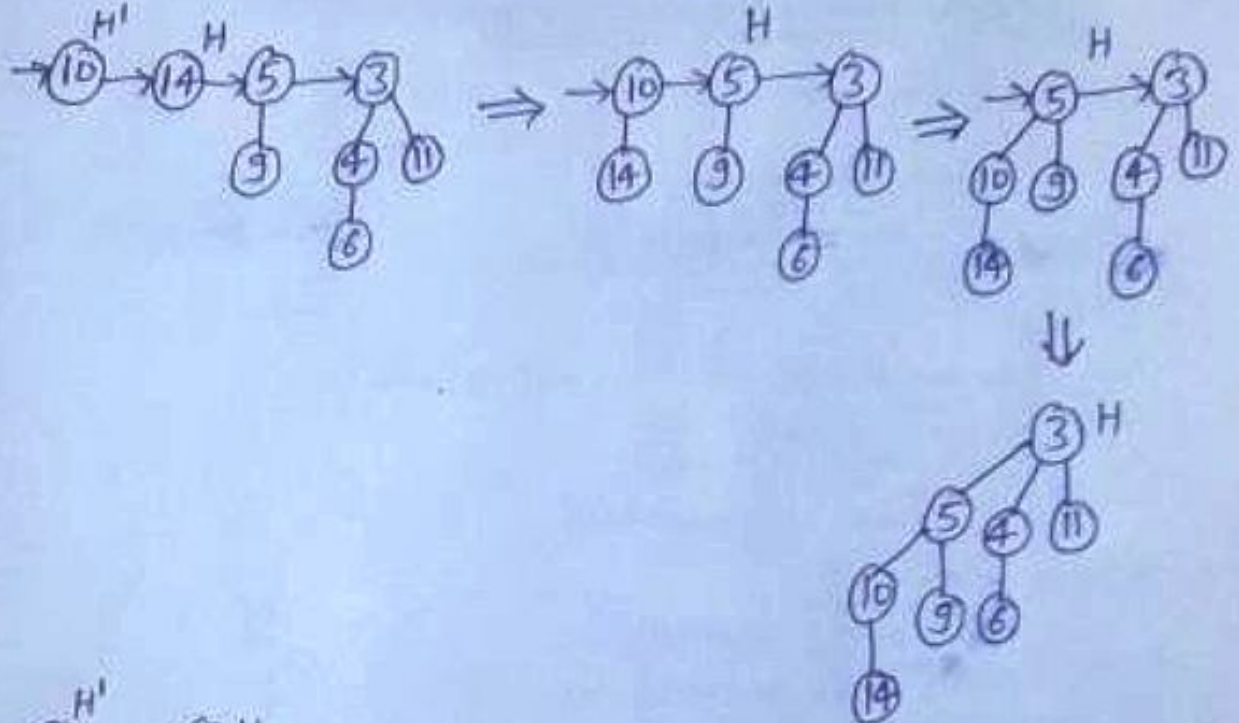
- The Binomial tree B_0 consist node.
- The Binomial Tree B_k consist two Binomial Tree B_{k-1} and B_{k-1} are linked together.

Step-7

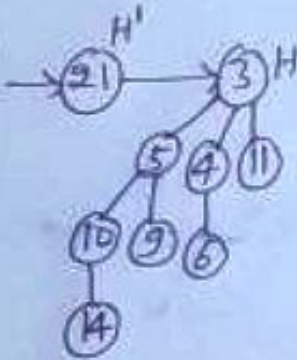


(3)

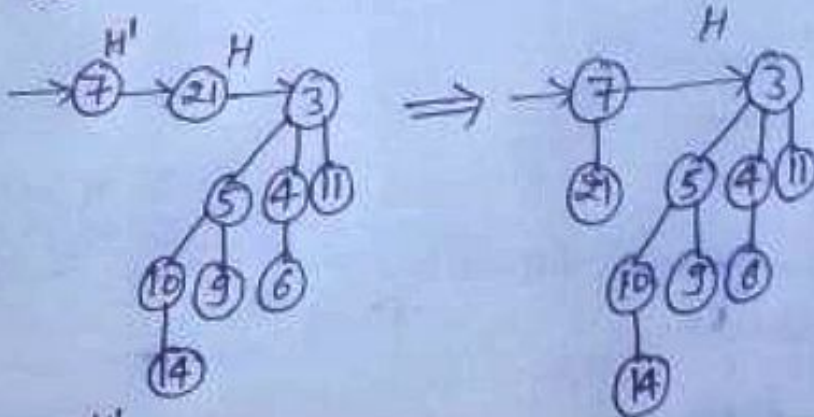
Step-8



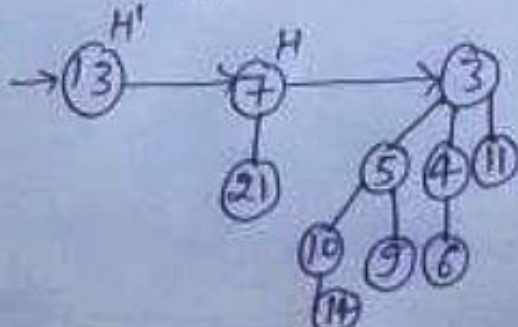
Step-9



Step-10

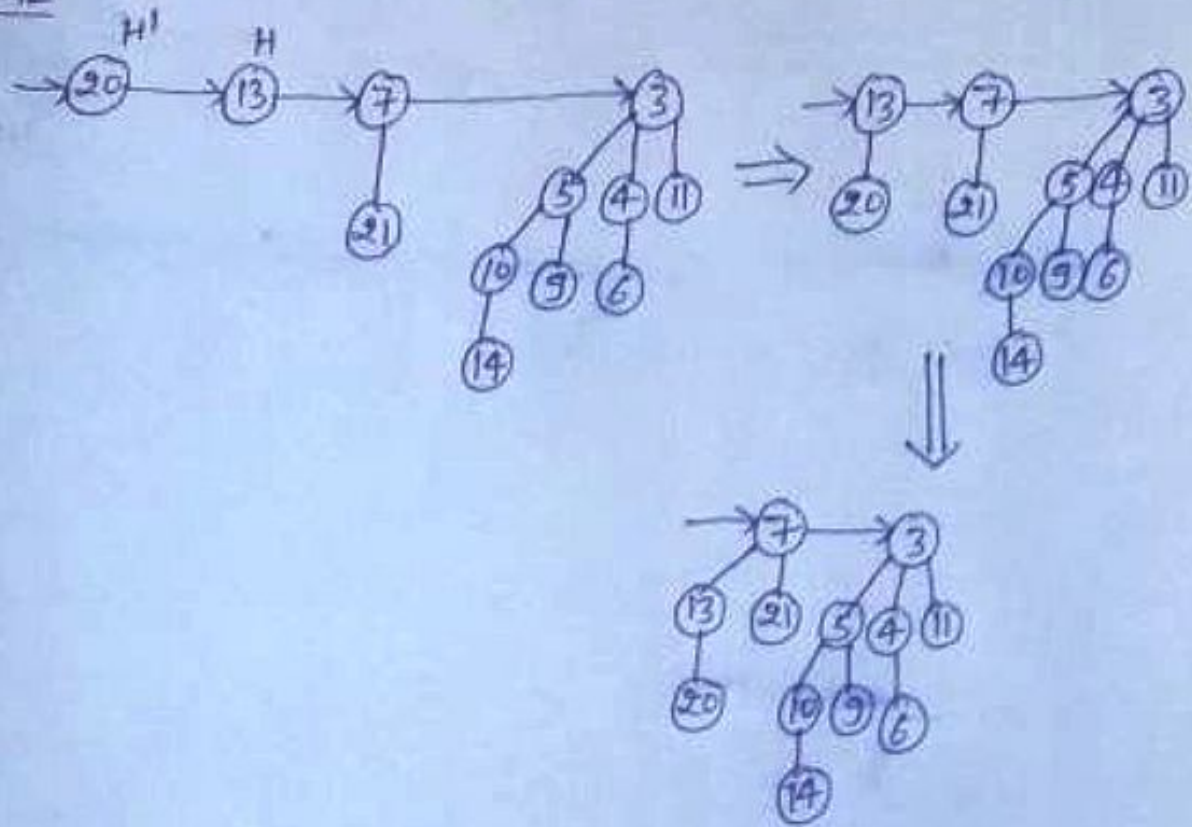


Step-11

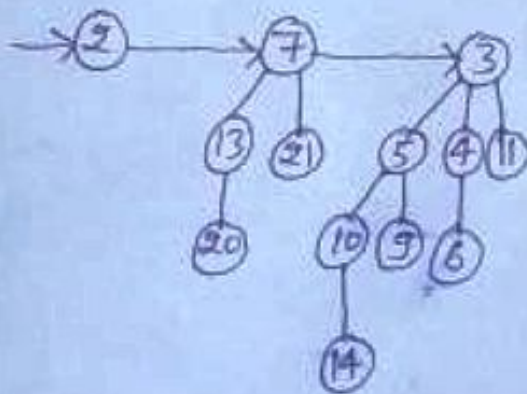


Step-12

4

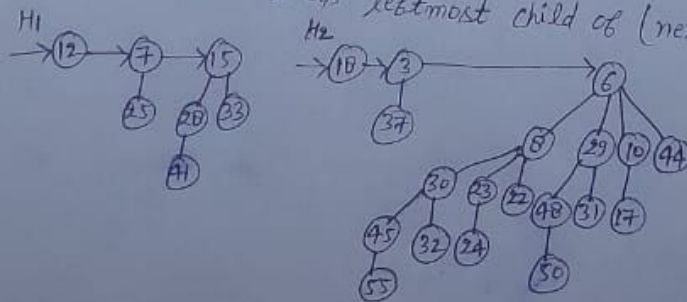


Step-13



Binomial min Heap UNION

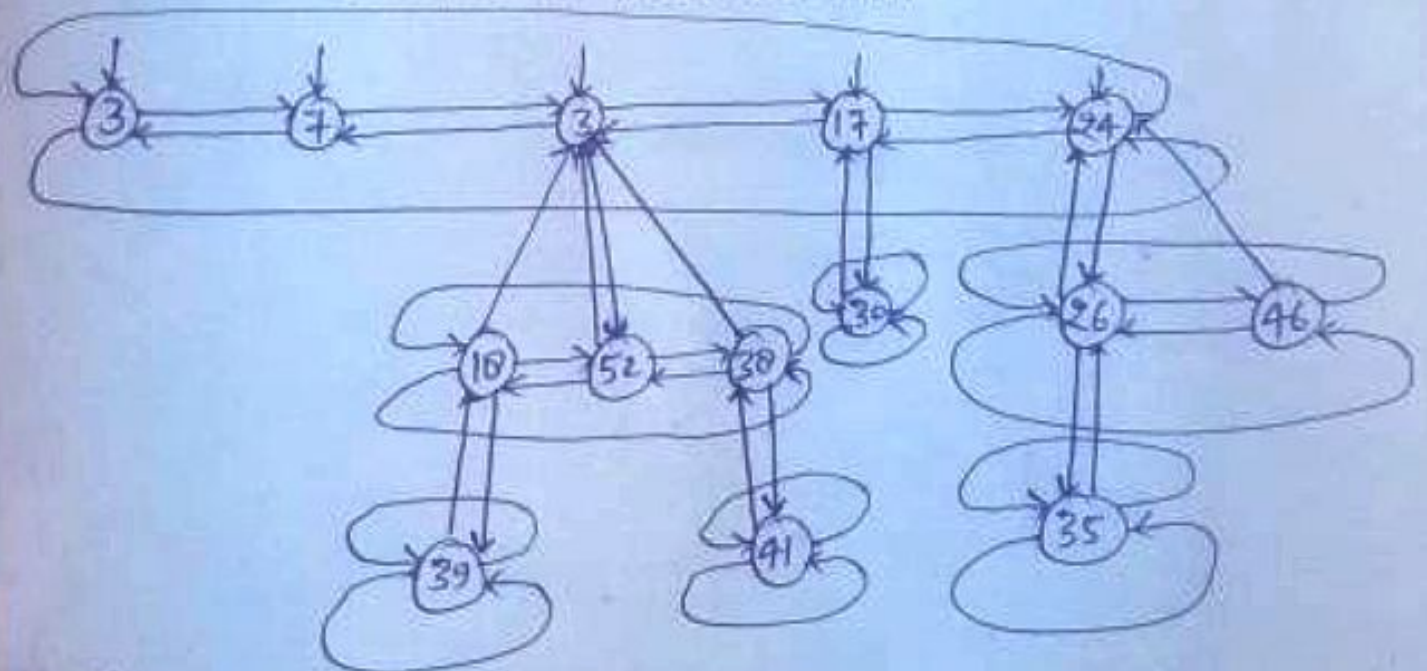
- ① Merge the root lists of binomial min heaps H_1 and H_2 into a single linked list that is stored in non decreasing order of their degree.
- ② Link the roots of equal degree until atmost one root remains of each degree.
 - (a) if ($\text{degree}[x] \neq \text{degree}[\text{next-}x]$ or $\text{degree}[x] = \text{degree}[\text{next-}x] = \text{degree}[\text{sibling}[\text{next-}x]]$) then move the pointers one position right.
 - (b) if ($\text{degree}[x] = \text{degree}[\text{next-}x] \neq \text{degree}[\text{sibling}[\text{next-}x]]$)
 - ① if ($\text{key}(x) < \text{key}(\text{next-}x)$) then make ($\text{next-}x$) as leftmost child of x .
 - ② if ($\text{key}(x) > \text{key}(\text{next-}x)$) then make x as leftmost child of ($\text{next-}x$).



Fibonacci Heaps

- Collection of trees with each tree following the heap ordering (min heap) property.
- Trees may be in any order in the root list. (Unlike Binomial Heaps)
- A pointer to the min element of the heap always maintained.
- Siblings are connected through a circular doubly linked list.
- Each child points to its parent.
- Each parent points to any one child.
- $\text{degree}(x)$: No. of children of root of a tree.
- $\text{Mark}(x)$:
 - 1 - Lost one of its child
 - 0 - Lost no child

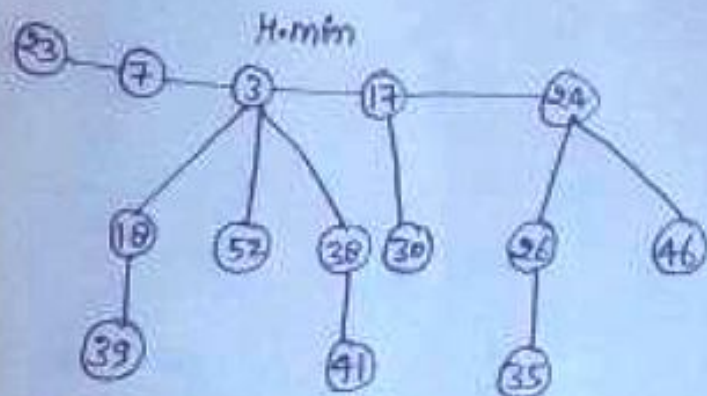
Structure of Fibonacci Heap



Notation:

- n : Number of nodes in heap.
- $\text{rank}(x)$: Number of children of node x .
- $\text{rank}(H)$: Max rank of any node in heap H .
- $\text{trees}(H)$: Number of trees in heap H .
- $\text{marks}(H)$: Number of marked nodes in heap H .

Insertion of key



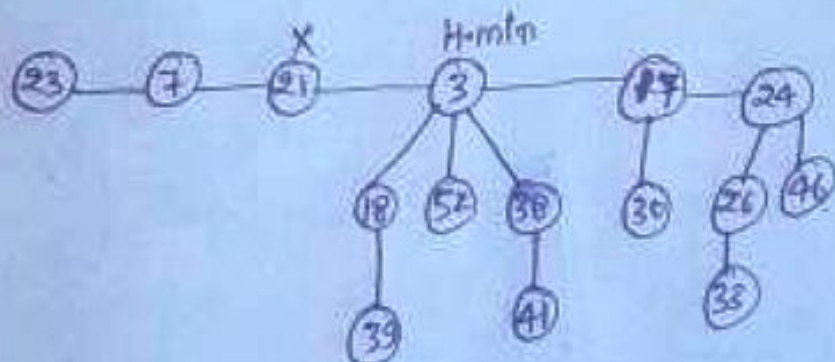
21^x

x.degree
x.p
x
x.maxK
x.child

0
NULL
x
FALSE
NULL

Fib-Heap-Insert(H, x)

1. $x.degree = 0$
2. $x.p = NIL$
3. $x.child = NIL$
4. $x.maxK = FALSE$
5. If $H.min == NIL$
6. create a root list for H containing just x .
7. $H.min = x$
8. else insert x into H 's root list
9. If $x.key < H.min.key$
10. $H.min = x$
11. $H.n = H.n + 1$

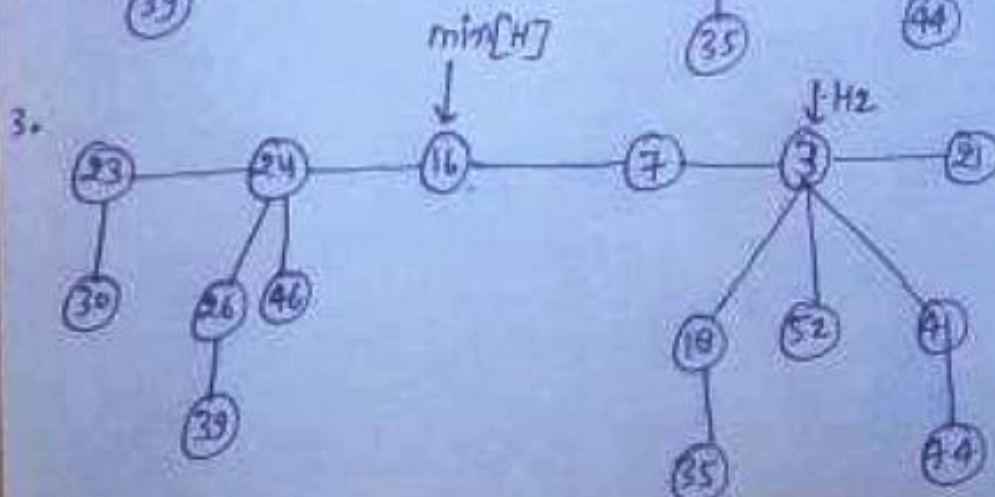
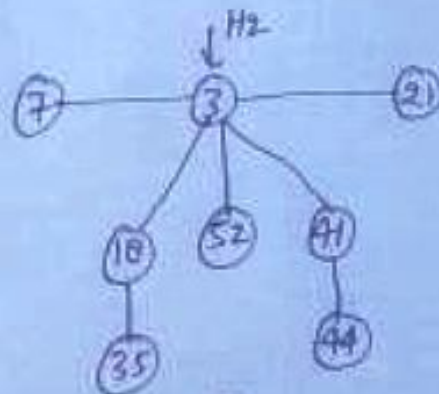
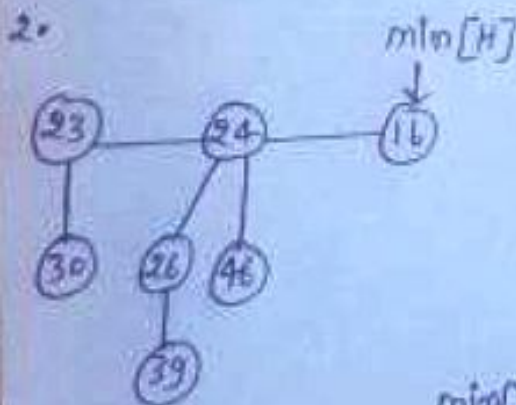
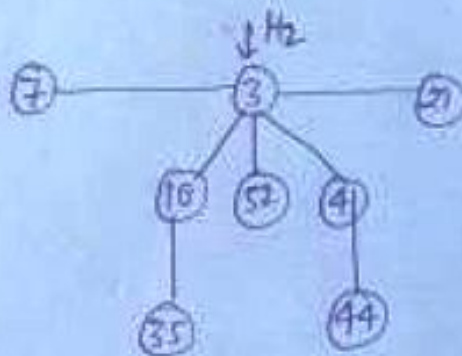
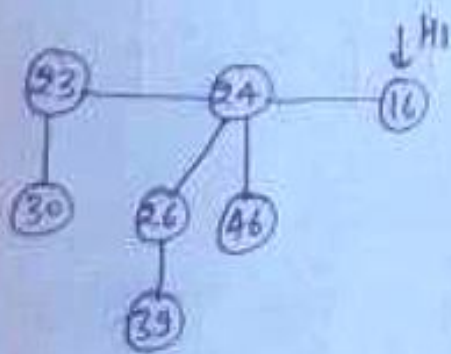


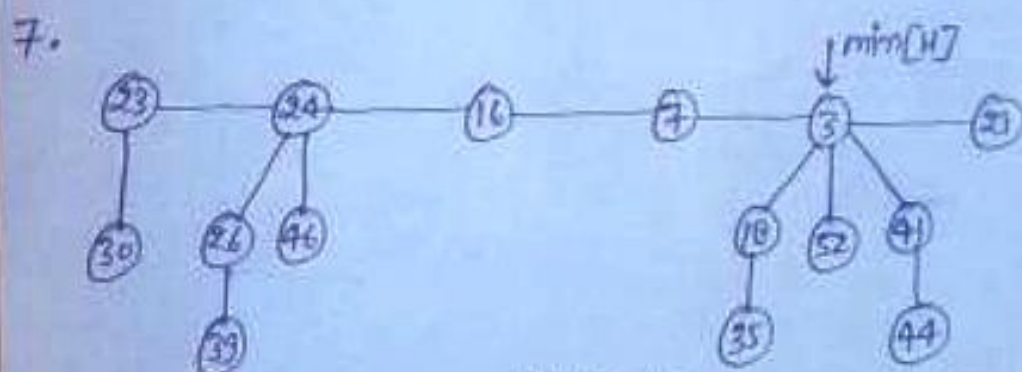
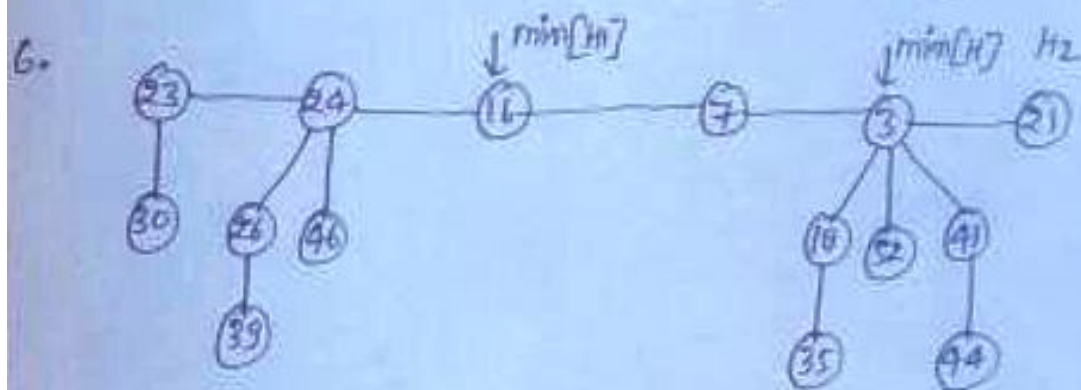
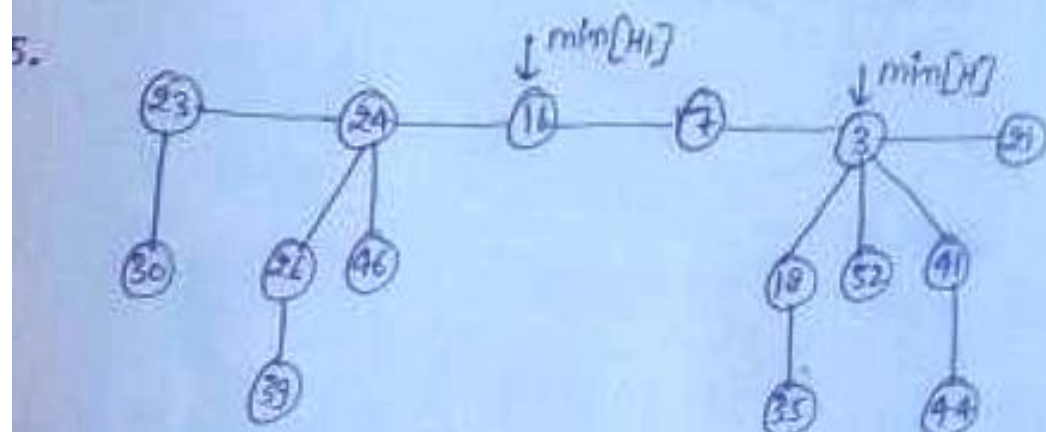
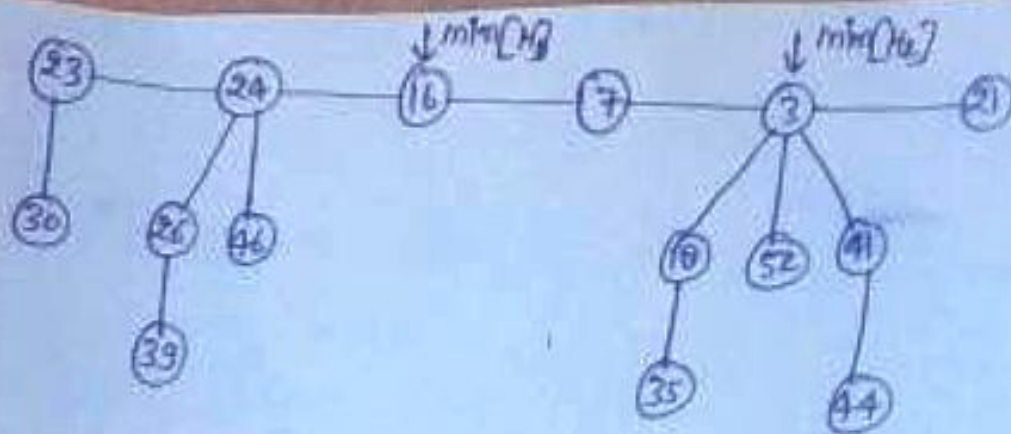
$$TC = O(1)$$

UNION of Fibonacci Heaps

Fib-Heap-UNION(H_1, H_2)

1. $H \leftarrow \text{Make-Fib-Heap}(H_1)$
2. $\text{min}[H] \leftarrow \text{min}[H_1]$
3. Concatenate the root list of H_2 with the root list of H
4. IF ($\text{min}[H_1] = \text{NIL}$) OR ($\text{min}[H_2] \neq \text{NIL}$ and $\text{min}[H_2] < \text{min}[H_1]$)
5. then $\text{min}[H] \leftarrow \text{min}[H_2]$
6. $n[H] \leftarrow n[H_1] + n[H_2]$
7. free the objects H_1 and H_2
8. return H





Applications

$$T_c = O(1)$$

1. Priority Queue Implementation.
2. Large amount of data representation.
3. Decrease key operation is used in minimum spanning tree algorithm.
4. Single source shortest path.