

Algorithm :

Algorithm is a combination of finite sequence to solve any problem.

It is combin.

Properties of Algorithm :

- 1.) It should produce atleast one o/p.
- 2.) It should take either 0 or more.
- 3.) It should produce o/p after finite step.
- 4.) It is independent to the programming language.
- 5.) Algorithm should be deterministic.  
↳ unambiguous.

Steps required to construct an algorithm :

- Problem definition
- Design algorithm
  - ↳ Divide & conquer
  - ↳ Backtracking
  - ↳ Dynamic programming
  - ↳ Greedy techniques.
  - ↳ Branch & Bound.
- Flowchart

4) `main()` Time complexity -  $O(n)$

```

{
    i = 1;
    while (i < n)
    {
        x = y + z;
        i++;
    }
}

```

5) `main()` Time complexity -  $O(\log_2 n)$

```

{
    i = 1;
    while (i < n)
    {
        x = y + z;
        i = 2 * i;
    }
}

```

$2^0 \rightarrow 1$   
 $2^1 \rightarrow 1$   
 $2^2 \rightarrow 1$   
 $2^3 \rightarrow 1$   
 $\vdots$   
 $2^k \rightarrow 1$

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

6) `main()`

```

{
    while (n > 1)
    {
        n = sqrt(n);
    }
}

```

$n = \frac{n}{2}$   
 $n = \frac{n}{2}$   
 $n = \left(\frac{n}{2}\right)^K$

$k = \frac{n}{2}$   
 $2k = n$   
 $\log_2 k = \log_2 \frac{n}{2}$   
 $n = \log_2 k$   
 $n = \left(\frac{k}{2}\right)^{\frac{1}{2}}$   
 $\log_2 n = \frac{1}{2} \log_2 \left(\frac{k}{2}\right)$   
 $2 \log_2 n = \log_2 \frac{k}{2}$

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

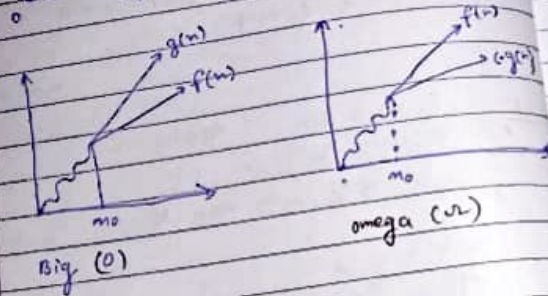
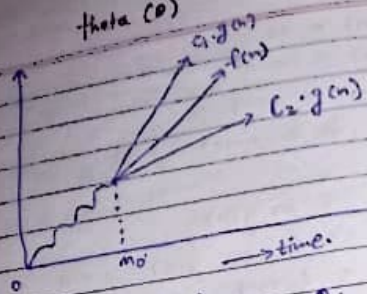
$\log_2 n^{\frac{1}{2^k}} = \log_2 1$   
 $\frac{1}{2^k} \log_2 n^k = 1$   
 $\log_2 n^k = 2^k$   
 $\log_2 (\log_2 n^k) = \log_2 2^k$

$k = \log \log_2 n$

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



$\theta(n)$



Ex: 1  $f(n) = n$   
 $g(n) = n + 10$   
 $f(n) \leq c_1 \cdot g(n)$   
 $n \leq c_1 (n + 10)$ ,  $c_1 = 1, n_0 = 1$

$f(n) \geq c_2 \cdot g(n)$   
 $n \geq c_2 (n + 10)$ ,  $c_2 = 1/2, n_0 = 10$

True ✓

Ques: 1

$f(n) = n$   
 $g(n) = n^2$   
 $f(n) = O(g(n))$  ?  
 $f(n) \leq c \cdot g(n)$   
 $f(n) \geq c_2 \cdot g(n)$

$n \leq c_1 \cdot n^2$  ✓  
 $n \geq c_2 \cdot n^2$  ✗  
 $\therefore f(n) \neq O(g(n))$

Ques: 2

$f(n) = n^2$   
 $g(n) = n^2$   
 $f(n) = O(g(n))$  ?  
 $f(n) \leq c \cdot g(n)$   
 $n^2 \leq c \cdot n^2$

Tightest upper bound  
 True (TUB)

$\left. \begin{matrix} n^3 \\ n^4 \\ \vdots \\ n^n \end{matrix} \right\} \rightarrow \text{upper bound}$

$n^2 \geq c \cdot n^2 \rightarrow \text{Tightest lower bound}$   
 increasing TC

$O(1) < \log \log n < \sqrt{\log n} < \log n < n < n \log n$   
 $< n^2 < n^3 \dots \text{polynomial} < 2^n \dots$   
 exponential

$\log \log n < \log n < \sqrt{\log n}$





$$(i) \quad 2^n = O(m^m)$$

$$2^n > m^{\log n}$$

$$m < \log_n^{\log n}$$

$$(ii) \quad m < m! < m^n$$

$$m > \log n$$

$$< \log_n^{\log n}$$

$$< \log n$$

•  $m > \log n$  taking log both side.

$$\log m > \log \log n$$

set  $m = 2^{100}$

$$2^{100} > \log 2^{100}$$

$$2^{100} > 100$$

•  $\log_2 n \leq \log n$

$$\{m < m \log_2 m\}$$

$$(iii) \quad 1000 n \log n = O\left(\frac{n \log n}{1000}\right)$$

$$1000 n \log n < C \frac{n \log n}{1000} \quad \text{false. } T$$

$$(iv) \quad \sqrt{\log n} = O(\log \log n)$$

$$\sqrt{\log n} < C(\log \log n) \rightarrow f$$

$$(v) \quad \text{If } (0 < x < y)$$

then  $m^x = O(m^y)$

$$m^x \leq C m^y$$

$$m^2 \leq C m^4 \quad \begin{matrix} x=2 \\ y=4 \end{matrix}$$

$\hookrightarrow T$

$$(vi) \quad 2^n \neq O(m^k) \text{ where } k \text{ is constant.}$$

$$2^n < C m^k$$

$\hookrightarrow T$

$$(vii) \quad m^2 \cdot 3 \log_2 n = O(m^5)$$

$$f(n) \leq C_1 \cdot g(n) \quad (n \geq n_0)$$

$$f(n) \geq C_2 \cdot g(n) \quad (n \geq 0)$$

$$m^2 \cdot 3 \log_2 n \leq C_1 m^5$$

$$2 \log_2 n + 3 \log_2 n \leq C_1 5 \log_2 n$$

$$5 \log_2 n \leq C_1 5 \log_2 n \rightarrow T$$

$$m^2 \cdot 3 \log_2 n \geq C_2 m^5$$

$$5 \log_2 n \geq C_2 5 \log_2 n$$

$\hookrightarrow T$

(viii)  $4 \log_2 n = O(n^2) \rightarrow F$

(ix)  $\frac{4^n}{2^n} = O(2^n) \rightarrow T$   
 $\frac{2^{2n-1}}{(2^n)} = O(2^n)$   $O_v, O_v, O_v$

(x) if  $f(n) = O(g(n))$  then  $f(n) = O(2^{g(n)}) \rightarrow F$

$f(n) = 2^n, g(n) = n$

$2^n \leq O(2^n)$

$2^n \leq O(2^n), n=4$

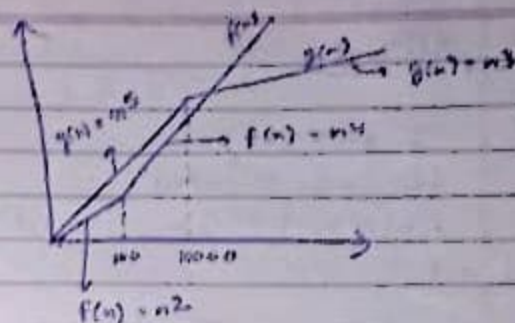
$2^0 \leq 2^4$

$O_x, \sqrt{x}, \sqrt{x}, O_x, O_x$

(xi)  $f(n) = \begin{cases} n^2 & 0 < n < 100 \\ n^4 & n \geq 100 \end{cases}$

$g(n) = \begin{cases} n^5 & 0 < n < 10000 \\ n^3 & n \geq 10000 \end{cases}$

find out relation b/w  $f(n)$  &  $g(n)$



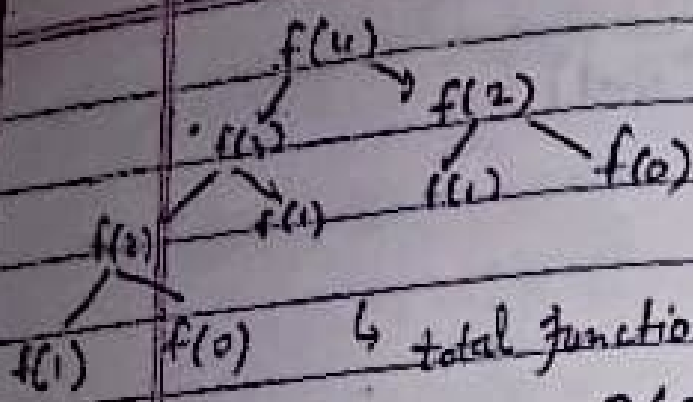
Properties of Asymptotic Notation:

(1) Reflexive:  $f(n) = O(f(n))$   
 $f(n) = O(f(n))$   
 $f(n) = O(f(n))$   
 $f(n) \neq O(f(n))$   
 $f(n) \neq w f(n)$

(2) Symmetric: if  $f(n) = O(g(n))$   
 then  $g(n) \neq O(f(n))$   
 only theta will pass.

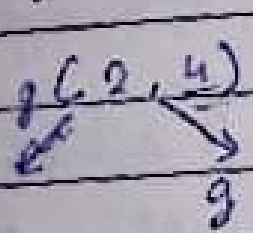
(3) Transitive:  $f(n) = O(g(n))$   
 $g(n) = O(h(n))$   
 then  $f(n) = O(h(n))$

(4) Multiplicative: if  $f(n) = O(g(n))$   
 then  $f(n) \cdot f(n) = O(h(n) \cdot g(n))$



total function call = 24  
 $O(2^n)$

Ques 3: Write a recurrence relation and recursive program to find gcd of two (Ave no)



2 (2) → 2 sol.  
 gcd = 2 (2)  
 2

- gcd(100, 200) =
- gcd(23, 25) =
- gcd(0, 0) = 0
- gcd(0, 10) = 0

gcd(int a, int b)  
 {



## Divide and Conquer:

Recursion: function calling itself is called recursion.

```

if n=0 || n=1
    return n
else
    return n * f(n-1)
    } → factorial
    
```

• solving big problems using smaller problem.

• To execute recursive program we use stack data structure.

• Every recursive program must have termination condition else it will go  $\infty$ .

• In recursive programs from one function call to another parameter value will only change but not the no. of parameters.

Ques 1 write recursive program and recursive sol<sup>n</sup> to find product of two (five) numbers a & b.

```

if n=0
    return 0
else n=1
    } X
    
```

```

else
    return n * f(n-1)
    } X
    mul(2, 5)
    
```

```

if a=0 || b=0
    return 0
else
    return a + f(a, b-1)
    } → recursive program
    ⇒ O(B)
    
```

$$mul(A, B) = \begin{cases} 0 & \text{if } A=0 \text{ || } B=0 \\ A + mul(a, b-1) & \text{otherwise} \end{cases}$$

Ques 2 write a recursive program and recursive func<sup>n</sup> to find n<sup>th</sup> FIBONACCI no.  
 0 1 1 2 3 5 . . .

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

$$fib(a, b) \Rightarrow \text{return}$$

```

if (n==0 || n==1)
    return n
    
```

```

else
    return fib(n-1) + fib(n-2)
    
```



$$1 + 2^2 + 4^2 + \dots + n^2$$

$$O(n^3)$$

$$T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2T(\frac{n}{2}) + n & \text{otherwise} \end{cases}$$

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

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

$$= 2^2 \left[ \frac{n}{2^2} \right] + n + n$$

Master Class :-

Master Theorem :-

Case 1:  $f(n) = O(n^{\log_b a - \epsilon})$   
where  $\epsilon$  is a constant.

$$T(n) = O(n^{\log_b a})$$

Case 2:  $f(n) = \Omega(n^{\log_b a + \epsilon})$   
 $T(n) = O(f(n))$

$$\text{Case 3: } f(n) = O(n^{\log_b a})$$

$$T(n) = O(n^{\log_b a} \log n)$$

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

# of subproblems  $\rightarrow$  time of conquer & combine  
 $\hookrightarrow$  size of problem

where  $a, b$  are two positive constants.

$$a \geq 1$$

$$b > 1$$

$f(n)$  is a positive function.

$f(n) = 8T(n/2) + n^2 \Rightarrow$  it lies under case

$T(n) = 2T(n/2) + n^2 \Rightarrow$  it lies under case

$$T(n) = T(n/2) + C$$

$$n^{\log_2 1} = \log_2 n = 1 \text{ (Case 1)}$$

$(C=1) \Rightarrow$  same multiply by  $\log n$

$$\therefore TC = O(\log n)$$

Ques:  $T(n) = 0.5T(n/2) + n^2$

$\downarrow x$

$$T(n) = 4T(n/2) - n \log n$$

$\hookrightarrow f(n) = O(n \log n)$

Ques.)

$$T(n) = 2T(n/2) + n \log n$$

$$a = 2, b = 2, f(n) = n \log n$$

Because difference isn't polynomial.

(By di)

Extended:

If log is coming

$$f(n) = O(\log^a n \cdot (\log n)^{k+1})$$

$$\rightarrow f(n) = O(n \log^a n \cdot \log n^k \text{ where } k \geq 0)$$

- If recurrence relation contain root operator.

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

(1) assume  $m = 2^k$

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

S

$$S(k) = S(k/2) + c$$

$$O(\log k)$$

$$m = 2^k$$

$$\log m = k \log 2$$

$$k = \log_2 m$$

$$\{k = \log_2 n\}$$

Ques.)

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

$$= 2T(2^{k/2}) + \log_2 k$$

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

$$= k \log k$$

$$= \log n \cdot \log \log n$$

24 March 2023

Linear Search: Array may be sorted  
x may not be.

20	62	70	19	5	101
----	----	----	----	---	-----

Best case TC =  $O(1)$

Worst case TC =  $O(n)$

Avg case TC =  $O(n)$

Binary Search:

Graph

Tree (DPG)

1.  $G(V, E)$

2. No root element present in the graph.

3. Graph may be connected or disconnected.

4. Graph may contains cycle.

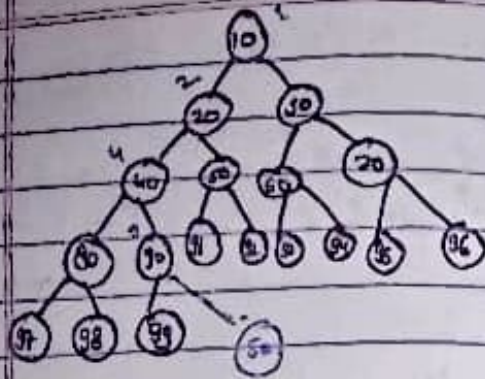
$T(V, E)$

Root element is present.

connected.

Tree can not cycle.





array :-  $\{ 10, 20, 30, 40, 50, 60, 70, 80, 90, 91, 92, 93, 94, 95, 96, 97, 98, 99, 5 \}$

insert = 5

insert last  $\rightarrow$  to maintain property of complete binary tree.

compare 5 with  $\rightarrow (90, 40, 20, 10)$  root.

$$\text{parent} = \left\lfloor \frac{19}{2} \right\rfloor_{\text{floor}} = \left\lfloor \frac{19}{2} \right\rfloor = 9$$

$$\left\lfloor \frac{9}{2} \right\rfloor = 4$$

$$\left\lfloor \frac{4}{2} \right\rfloor = 2$$

$$\left\lfloor \frac{2}{2} \right\rfloor = 1$$

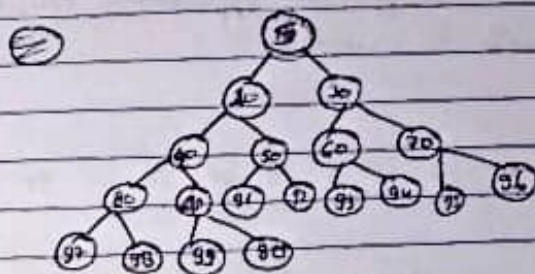
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Best case =  $O(1)$

Worst case =  $O(\log_2 n)$  = Average case.

min  
deletion in heap:

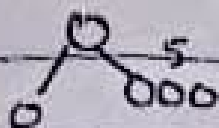


deletions always takes place from top.

deletion of 1 element in min heap or max heap =  $O(\log n)$

$$\begin{array}{l|l} \text{Best case} = O(1) & \Rightarrow O(n) \\ \text{Worst} = O(\log_2 n) & \Rightarrow n \log n \end{array}$$

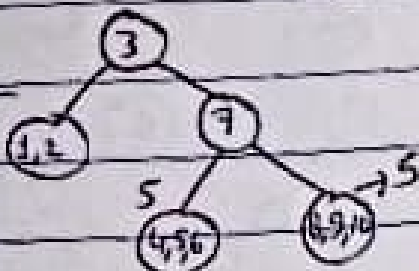
$n=5$



$= 42 \text{ BST}$

$n=6$

Root is 3 and right is 7



$$= 2 \times 5 \times 5$$

$$= 50$$

$$\text{Total no of BST} = \frac{2^n C_n}{n+1} = \frac{2^{n+1}}{n+1}$$

$$\# \text{ of BST} = \sum_{i=1}^n (\text{BST}(i-1) \times \text{BST}(n-i))$$

$$n=3 = \frac{6!}{3!3!} = \frac{6!}{3!3!} = \frac{6 \times 5 \times 4 \times 3 \times 2 \times 1}{3 \times 2 \times 1 \times 3 \times 2 \times 1}$$

$$= \frac{20}{3+1} = 5$$



Quick sort (a, i, j)

```

{
    if (i == j)
        return (a[i]);
    else
    {
        p = partition(a, i, j);
        QuickSort(a, i, p-1);
        QuickSort(a, p+1, j);
        return (a);
    }
}

```

Recurrence Relation:

$$\begin{cases} a[i] & ; m=1 \\ m + T(i-i) + T(j-i) & \text{otherwise} \end{cases}$$

100 90 80 70 60

60 90 80 70 100

$$\begin{aligned}
 &\Rightarrow m + T(0) + T(n-1) \\
 &= T(n-1) + m \\
 &= O(n^2) \Rightarrow \text{worst case}
 \end{aligned}$$

If the array is sorted quick sort not preferable.

Ques 1 In quick sort the sorting of n numbers the  $(n/4)^{\text{th}}$  smallest element is selected as pivot element using order of n time. then what is + will be the time complexity of quick sort algorithm.

Ans:  $m + m + T(n/4) + T(3n/4)$   
 $= m \log_{4/3} m$

Ques 2 In quick sort the sorting of n numbers the  $(n/5)^{\text{th}}$  element is selected as pivot using order of  $O(n^2)$  time. what will be the worst case time complexity of quick sort.

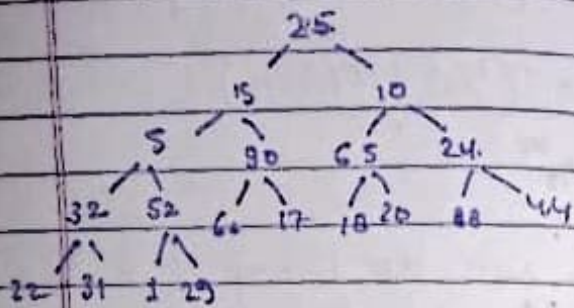
Ans:  $n^2 \log$

Pivot Partition  
 $n^2 + m + T(0) + T(n-1) \Rightarrow \text{Pivot smaller}$   
 OR  
 $n^2 + m + T(n-1) + T(0) \Rightarrow \text{largest}$   
 $n^2 + m + T(n-1)$   
 $T(n-1) + m^2$   
 $O(n^3)$

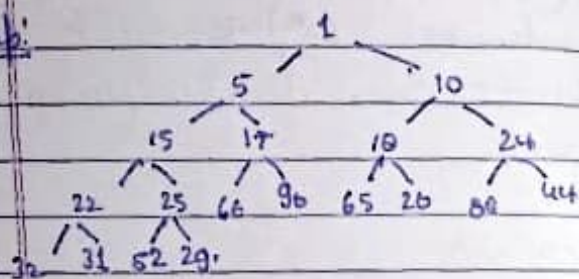
Sing: Create a min heap.

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25, 15, 10, 5, 90, 65, 24, 32, 52, 68,  
18, 20, 88, 44, 22, 31, 1, 29,



min heap:


$$\hookrightarrow 0(n)$$

Build heap (a, n)

3.

for (  $i = \lceil n/2 \rceil$  to 1 )

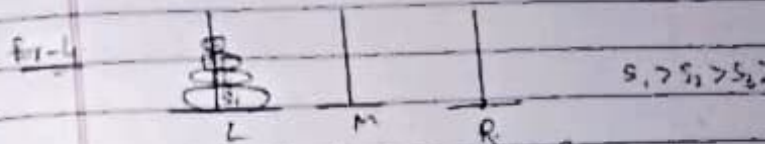
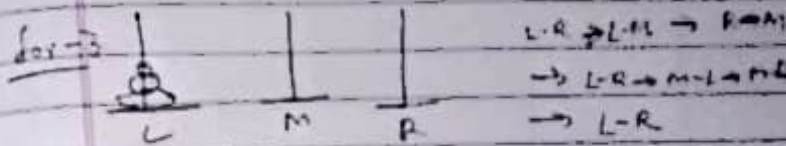
```
minHeapifydown(a, i);
```

3

- Construct min heap using Build heap function  $\rightarrow O(n)$

② delete one by one and add at end  
( $\log_2 n = O(\log n)$ )

Tower of Hanoi :  $2^n - 1$  (moves)

Step  $\Rightarrow$  15.

L-R

 $L = 100$ 

R-M

L-R

34. 3

$$\frac{L \quad \overline{R} \quad M}{\downarrow}$$
$$\begin{array}{cc} 1 & 3 \\ \text{L} & \text{M} \\ \text{---} & \text{---} \\ \text{R} & \text{R} \\ | & | \\ \text{---} & \text{---} \\ \text{---} & \text{---} \end{array}$$

10  $\rightarrow$  dist.



## Quick Sort :-

1. It is based on divide and conquer technique.
2. It is Inplace.
3. It is not stable algorithm.
4. In all practical applications quick sort will be used.

50 20 15 5 10 45 70 80 90

- If we change pivot place everytime it is called as randomized quick sort.
- If we choose the pivot place as same everytime it simply called as quick sort.
- It doesn't include merge step.

### Partition Algorithm :-

Partition (a, i, j)

{

x = a[i];

p = i;

```
for (q = i+1; q ≤ j; q++)
{
```

```
    if (a[q] ≤ x)
```

```
    {
```

```
        p = p+1;
```

```
        swap(a[p], a[q]);
```

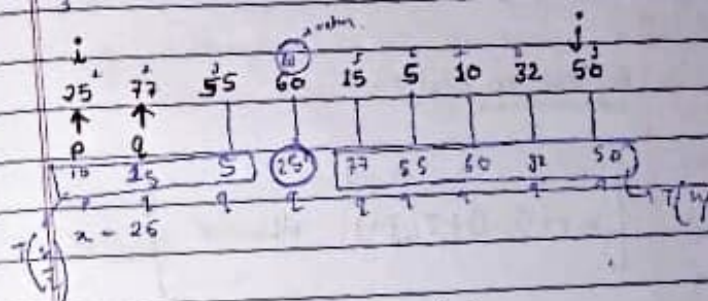
```
    }
```

```
}
```

```
swap(a[p], a[j]);
```

```
return (p);
```

```
}
```



Time complexity =  $O(n)$

~~Recurrence Relation~~

$$\begin{aligned} \text{Recurrence Relation: } & O(1) + O(n) + T(n/2) + T(n/2) \\ &= O(T(n/2)) + n \\ &= n \log n \end{aligned}$$

Quick sort algorithm :-

$$T(n) = \begin{cases} O(1) & n=1 \\ T(\frac{n}{2}) + c & \text{otherwise} \end{cases}$$

Merge Sort:

Merging two sorted <sup>Sub</sup>array.

Merge algorithm:-  
 $A$  - array  
 $q$  - mid of array  
 $r$  - last index  
 $p$  - starting index  
 $q$

$$m_1 = q - p + 1$$

$$m_2 = r - q$$

Let  $L[1 \dots m_1]$  and  $R[1 \dots m_2]$  be two arrays.

for  $i = 1$  to  $m_1$

$$L[i] = A[p + i - 1]$$

for  $j = 1$  to  $m_2$

$$R[j] = A[q + j]$$

$$i = 1, j = 1$$

for  $k = p$  to  $r$

if  $(L[i] \leq R[j])$

$$A[k] = L[i]$$

$$i = i + 1$$

else  $A[k] = R[j]$

$$j = j + 1$$

mergesort( $A, p, r$ )  
 $\{$

if  $(p < r)$

$$q = p + r / 2$$

mergesort( $A, p, q$ )

mergesort( $A, q+1, r$ )

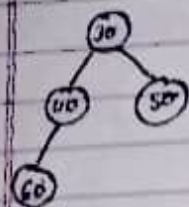
merge( $A, p, q, r$ )

$$TC = m \log n.$$

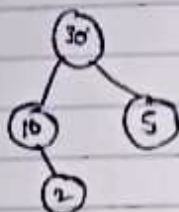


## Heap Sort:

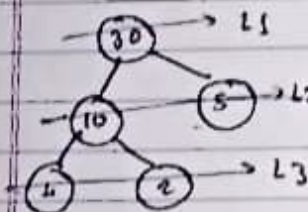
• Complete Binary tree ✓



→ min Heap.



⇒ Not complete BT.



→ max heap.

store heap in the arrays :-

30	10	5	1	2	
1	2	3	4	5	

Left child of  $i^{\text{th}}$  child =  $2i$

eg.  $2(2) = 4$  index = 1 (left child)

$$\text{Right child} = (2i+1)$$

total no of elements =  $n$

$$\text{then total leaf nodes} = \left\lceil \frac{n}{2} \right\rceil \cdot \left\lfloor \frac{n}{2} \right\rfloor + 3$$

ceil value

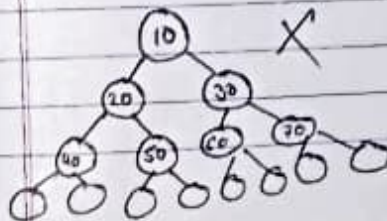
$$\text{internal nodes} = \left\lfloor \frac{n}{2} \right\rfloor \text{ floor value} = \left\lfloor \frac{n}{2} \right\rfloor$$

Min Heap tree: In the given complete binary tree at every node root is minimum or equal compare its children.

Max heap tree: In the given complete binary tree root is maximum or equal.

Insertion in min heap:

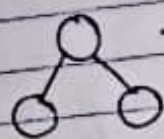
Insertion always takes place at  $L$  so that complete binary tree should be disturbed.



5. Graph may be divided. Tree is always directed.

• Every Tree is Graph but Graph isn't tree.

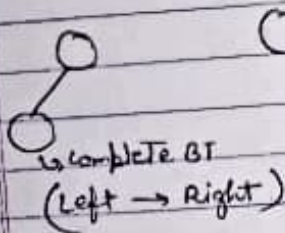
• Binary Tree has atmost 2 child.



→ strict BT or Full BT.

height = level - 1

func<sup>n</sup> calling — Pre order  
" execution — Post order



incomplete BT  
(Left → Right)



→ Not complete BT

Total no of nodes =  $2^n - 1$   
 $n = \text{level}$

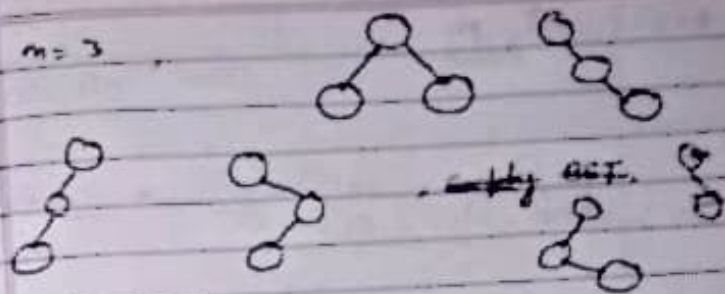
total nodes at  $n^{\text{th}}$  level =  $2^{n-1}$

Ques: Find the no. of Binary search tree where no. of nodes are 1, 2, 3, 4, 5, ..., 10  
Root is 3 & right is 7.

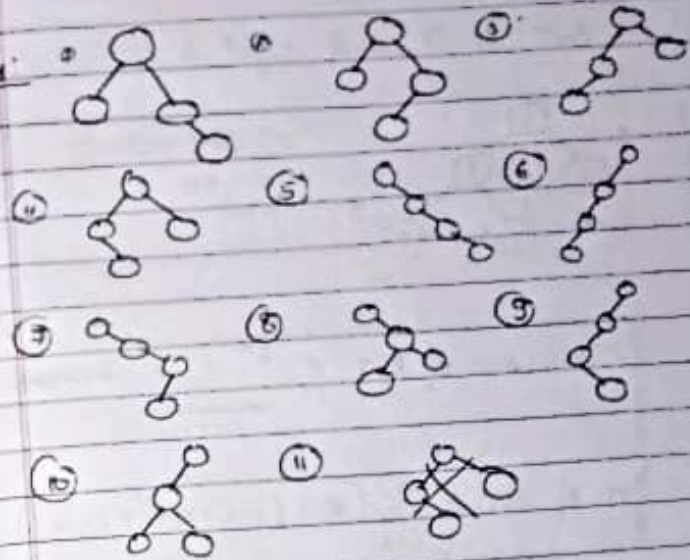
Total BST if  $n$  is no. of nodes = 25

If  $n = 0$  BST = 1 (Empty BST)

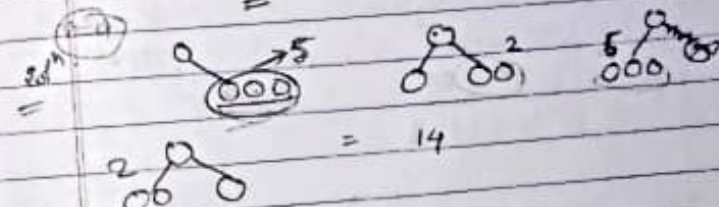
$n = 3$



no. 4



(14 BST)





(5)

Additive:

$$\text{if } f(n) = o(g(n))$$

$$\text{if } e(n) = o(h(n))$$

$$f(n) + e(n) = O(\max(g(n), h(n)))$$

17 March 2023

(1) Consider: the following,  $\mathbb{R}^n$  relation

$$f(n) = n^{1+\sin n}$$

$$g(n) = n$$

Find the relation among  $f(n)$  &  $g(n)$ .

$$n = (0, 90, 270, 360)$$

$$f(n) = n^{1+0} = n = 0$$

$$g(n) = n = 0$$

$$f(n) = n^{1+\sin 90} = n^2 = (90)^2$$

$$g(n) = n = 90$$

$$f(n) = n^{1+(-1)} = n^0 = 1$$

$$g(n) = n = 270$$

$$f(n) = n = 360$$

$$g(n) = n = 360$$

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n	f(n)	g(n)
n=0	0	0
90	90 <sup>2</sup>	90
180	180	180
270	1	270
360	360	360

Both the function are now comparable

(2) Consider: the following,  $\mathbb{R}^n$ 

$$f(n) = n^{2+\sin n}$$

$$g(n) = n^{\cos n}$$

Find the relation among  $f(n)$  &  $g(n)$ 

n	f(n)	g(n)
0	0	0
90	90 <sup>3</sup>	1
180	(180) <sup>2</sup>	1/180
270	1	1

$$f(n) \geq g(n)$$

$$f(n) = o(g(n))$$

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

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

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

~~$$T(n-k) = T(n-(k-1)) + \log(n-k)$$~~

$$\therefore = T(n-k) + \log(n-k+1) + \log(n-k+2) + \dots + \log n$$

$$n-k=1$$

$$= 1 + \log(2) + \log(3) + \dots + \log n$$

$$= 1 + \log(2 \cdot 3 \cdot \dots \cdot n)$$

$$= 1 + \log(n!)$$

$$n! \approx n^n$$

$$= 1 + \log n^n$$

$$= 1 + n \cdot \log n$$

$$= \underline{(n \cdot \log n)}$$

Ques:  $T(n) = \begin{cases} 1 & \text{if } n=1 \\ T(n-1) + \frac{1}{n} & \text{if } n>1 \end{cases}$

$$T(n) = T(n-1) + \frac{1}{n}$$

$$= T(n-2) + \frac{1}{n-2} + \frac{1}{n}$$

$$= T(n-3) + \frac{1}{n-3} + \frac{1}{n-2} + \frac{1}{n}$$

$$\therefore \frac{1}{(n-k)} + \frac{1}{n-k+1} + \frac{1}{n-k+2} + \dots + \frac{1}{n}$$



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(3)  $f(n) = m$   $\Rightarrow f(n)$  is  $O(n)$   
 $g(n) = n-10$

$f(n) = O(n-10)$

$n \leq (n-10) \cdot C$

$n \leq m-10$   $C=1$

$(n-10) \geq 8 \Rightarrow$  for  $C=1$  not

for  $C=2$

(4)  $f(n) = n^2$   
 $g(n) = n$

$f(n) = O(g(n)) \rightarrow f(n) \neq O(g(n))$

$n^2 \leq C \cdot n$   $n=10$

$C=1$

### $\Rightarrow$ Omega Notation ( $\Omega$ )

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

$f(n) \geq C \cdot g(n)$

$C > 0$

$n_0 > 0 \quad \forall (n \geq n_0)$

EX (1)  $f(n) = n+10$

$g(n) = n$

$n+10 \geq \Omega g(n)$

$n+10 \geq C \cdot n$

is always true.

EX 2:  $f(n) = m$

$g(n) = n-10$

$f(n) \geq g(n)$

$m \geq C \cdot (n-10) \Rightarrow$  True.

EX 3:  $f(n) = \Omega g(n)$

$f(n) = m$  ,  $g(n) = n+10$

$m \geq C \cdot (n+10) \Rightarrow$  True.  $C=1/2$   
 $n_0=10$

EX 4:

$f(n) = n$  ,  $g(n) = n^2$

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

$n \geq C \cdot n^2 \Rightarrow$  X  $C=1$   $n=1$

$\frac{1}{n} \geq \frac{1}{n^2}$   
Theta ( $\Theta$ ) :

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

$f(n) \leq C_1 \cdot g(n)$  ( $n \geq n_0$ )

$f(n) \geq C_2 \cdot g(n)$  ( $n \geq n_0$ )

EX:

$f(n)$   $g(n)$

$m = \Theta n$

$m \leq C_1 n$

$m \geq C_2 n$

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

$f(n) \leq C_1 \cdot g(n)$

$f(n) \geq C_2 \cdot g(n)$

Ques:  $W(n) = i \leq n$   
 $1 \quad n = \frac{n}{2} ; \rightarrow O(1) \times$

Ques:  $A(n)$   
 $\{ \begin{aligned} & \text{if } (n \leq 2) \text{ return } n \\ & \text{else return } (\sqrt{n}) \end{aligned}$   
 $(2)^{\frac{1}{2^k}} = 1$   
 $\log_2 \log_2 n = k$

Ques:  $APP(n)$   
 $\{ \begin{aligned} & \text{for } (i=1; i \leq n; i++) \\ & \{ j = n; \\ & \text{while } (j > 1) \rightarrow O(n \log n) \\ & \{ j = j/2; \end{aligned}$

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

$$2^k n = 2^k$$

$$n = 1$$

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

return

Ques:  $A(n)$   
 $\{ \text{for } (i=1; i \leq n; i++) \rightarrow n$

$\{ \text{for } (j=1; j \leq i; j++) \rightarrow n$

$\{ \begin{aligned} & \text{for } (k=1; k \leq i^2; k++) \\ & \{ z = x + z; \end{aligned}$

i	j	k
1	1	1 $\rightarrow 1 \times 133$
2	4	133 $\rightarrow 2 \times 133$
	2	133
n	n	$n \times 133$

$$O(n) \quad 133(1+2+3+\dots+n)$$

$$133 \cdot \frac{n(n+1)}{2}$$

$$= \frac{133 n^2}{2} + \frac{133 n}{2} = \frac{133}{2} n^2$$

$$= O(n^2)$$

Ques:  $A(n)$   
 $\{ \text{for } (i=1; i \leq n; i++)$   
 $\{ \text{for } (j=1; j \leq i^2; j++)$   
 $\{ \text{for } (k=1; k \leq n; k=2k)$



## Analysis:

Analysis: two types of analysis.

- ① Apriory
- ② Apostiory.

### Apriory:

- 1) It is independent on programming language.
- 2) It is giving same answers in every system.
- 3) It will give approximate answers.

### Apostiory:

- ① It is dependent on programming language.
- 2) It is giving diff. answers in every system.
- 3) It will give exact answers.

1) main ()

{

a = a + b;

}

Time complexity =  $O(1)$   
counts no. of operations

2) main ()

{

x = y + z

for (i = 1; i <= n; i++)

{

x = y + z;

}

}

Time complexity =  $O(n^2)$   
=  $O(n)$

3) main ()

{

x = y + z;  $\rightarrow O(1)$

for (i = 1 to n)

{

x = y + z;  $\rightarrow O(n)$

}

for (i = 1 to n)  $\rightarrow O(n)$

{

for (j = 1 to n)  $\rightarrow O(n)$

{

}

}

Time complexity  
=  $O(1) + O(n) + O(n^2)$   
=  $O(n^2)$