

## ASSIGNMENT-1

1. Asymptotic notations are notations used to tell the complexity of an algorithm when the input is very big. Asymptotic notations are as follows:

i) Big O Notation: used to define the tight upper bound.

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

iff  $f(n) \leq c g(n) \forall n \geq n_0$  & some const  $c > 0$

Eg. for Merge sort:  $O(n \log n)$

ii) ~~The~~ Big  $\Omega$  Notation: used to define the tight lower bound.

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

iff  $f(n) \geq c g(n) \forall n \geq n_0$  & some const  $c > 0$

Eg. for Merge sort  $\Omega(n \log n)$ .

iii) Theta  $\Theta$  Notation: gives both tight lower and upper bound.

if  $f(n) = O(g(n))$  &  $f(n) = \Omega(g(n))$   
Then

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

iff  $c_2 g(n) \leq f(n) \leq c_1 g(n) \forall n \geq \max(n_1, n_2)$  &  $c_1, c_2 > 0$

iv) Small  $O$  Notation: used to give the upper bound.

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

if  $f(n) < c g(n) \forall n > n_0$  &  $\forall \text{const. } c > 0$

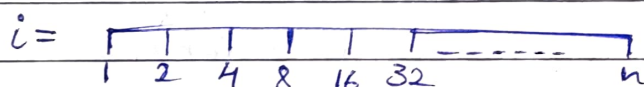
v) Small Omega  $\omega$  Notation: used to give lower bound.

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

if  $f(n) > c g(n) \forall n > n_0$  and  $\forall c > 0$ .

2. {for( $i=1$  to  $n$ )

{  $i = i * 2;$   
}



Let  $i$ 's final value be  $2^k$   
then

$$2^k = n.$$

taking  $\log$  on both sides.

$$k \log_2 2 = \log_2 n.$$

$$\Rightarrow k = \log_2 n.$$

$$\therefore T.C. = O(\log_2 n) //$$

$$3. T(n) = \begin{cases} 3T(n-1) & \text{if } n > 0, \\ 1 & \text{otherwise} \end{cases}$$

$$\therefore T(0) = 1. \text{ --- (a)}$$

$$T(n) = 3T(n-1) \text{ --- (1)}$$

$$\text{Let } n = n-1.$$

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

$$\text{then, } T(n) = 3 \times 3T(n-2) \text{ --- (2)}$$

$$\text{Let } n = n-2.$$

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

$$\text{then, } T(n) = 3 \times 3 \times 3T(n-3) \text{ --- (3)}$$

Generalizing (3)

$$T(n) = 3^k T(n-k) \text{ --- (4)}$$

$$1 = n-k \Rightarrow \boxed{n=k}$$

putting  $n=k$  in (4)

$$T(n) = 3^n T(k-k)$$

$$\Rightarrow T(n) = 3^n T(0) \quad \{T(0) = 1, \text{ from (a)}\}$$

$$T(n) = 3^n$$

$$\therefore f(n) = O(3^n)$$

$$4. T(n) = \begin{cases} 2T(n-1) - 1 & \text{if } n > 0, \\ 1 & \text{otherwise} \end{cases}$$

$$\therefore T(0) = 1. \text{ --- (1)}$$

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

$$\text{Let } n = 1$$

$$T(1) = 2T(1-1) - 1.$$

$$= 2T(0) - 1 \Rightarrow 2 - 1 \quad \{\text{from (1)}\}$$

$$T(1) = 1 \text{ --- (2)}$$

$$\text{Let } n = 2.$$

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

$$= 2T(1) - 1 \Rightarrow 2 - 1 \quad \{\text{from } \textcircled{2}\}$$

$$T(2) = 1$$

Let  $n=3$ .

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

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

$$= 2 - 1$$

$$T(3) = 1 \quad - \textcircled{3}$$

From the above examples, with increasing values of  $n$ , the time complexity stays 1.

$$\therefore T(n) = O(1)$$

5. `int i=1, s=1;`

`while (s <= n)`

`{`

`i++;`

`s = s + i;`

`printf("#");`

`}`

$i = 1 \ 2 \ 3 \ 4 \ 5 \ 6 \ 7 \ 8 \ \dots$

$s = 1 \ 3 \ 6 \ 10 \ 15 \ 21 \ 28 \ 36 \ \dots$

Let total ~~no. of~~ terms be  $k$ .

$S = 1 \ 3 \ 6 \ 10 \ 15 \ \dots \dots T_{k-1} \ T_k$

$S = 1 \ 3 \ 6 \ 10 \ 15 \ \dots \dots T_{k-2} \ T_{k-1} \ T_k$

Subtracting:

$0 = 1 \ 2 \ 3 \ 4 \ 5 \ \dots \dots k$

$$\Rightarrow T_n = \frac{n(n+1)}{2}$$



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

$$2k \leq n^2 + n$$

$$k \approx \sqrt{n}$$

$$T.C. = O(\sqrt{n})$$

6.

```
void function(int n)
{
```

```
    int i, count = 0;
```

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

```
        count ++;
```

```
}
```

loop runs till  $i^2 < n$ .

or till  $i$  reaches  $\sqrt{n}$ .

$$\therefore T.C. = O(\sqrt{n})$$

7.

```
void function(int n)
{
```

```
    int i, j, k, count = 0;
```

```
    for(i = n/2; i <= n; i++) —  $n/2$ 
```

```
    {
        for(j = 1; j <= n; j = j*2) —  $\log n$ 
```

```
        {
            for(k = 1; k <= n; k = k*2) —  $\log n$ 
                count ++;
```

```
        }
```

```
    }
```

for( $j=1; j \leq n; j=j*2$ )

$j$ 's final value =  $2^k$ .

$$2^k = n.$$

taking log on both sides,

$$k \log_2 2 = \log_2 n.$$

$$\Rightarrow k = \log_2 n$$

$\therefore$  for  $j$ 's loop T.C. =  $O(\log_2 n)$   
Similarly for  $k$ ,  
T.C. =  $O(\log_2 n)$

$$\therefore f(n) = O\left(\frac{n}{2} \times \log_2 n \times \log_2 n\right)$$

$$T.C. = O(n \log^2 n) //$$

8. ~~Find~~

```
function(int n)
{
    if (n==1)
        return;
    for(i=1 to n)           — n
        for(j=1 to n)       — n
            printf("%*");
}
```

$$T.C. = O(n \times n) \\ = O(n^2)$$

9.

```
void function(int n)
for(i=1 to n)                — n times
    for(j=1; j<=n; j=j+1) — n times
        printf("%*");
```

i = 1   2   3   4   5   ---   n  
j = 1

no. of times = n   n/2   n/3   n/4   n/5   ---   1  
inner loop runs

for inner loop iterations  
=  $n + \frac{n}{2} + \frac{n}{3} + \frac{n}{4} + \dots + 1$   
=  $\log n$ .

∴ T.C. =  $O(n \log n)$  //

10.  $f(n) = n^k$  ,  $k \geq 1$   
 $f(c) = c^n$  ,  $c > 1$

Let  $k=1$  &  $c=2$ .

$c^n$  grows faster than  $n^k$ . In other words  
 $c^n$  has greater growth rate than  $n^k$ .

$$\Rightarrow n^k = O(c^n)$$

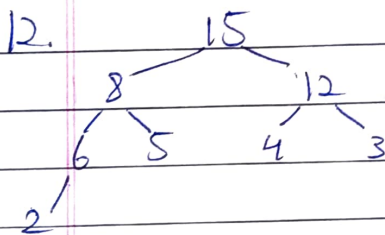
It can be said that  $c^n$  is the upper bound of  $n^k$ .

11. time taken to extract minimum element =  $O(1)$  since minimum element is always at the top of a minheap.

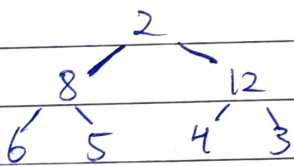
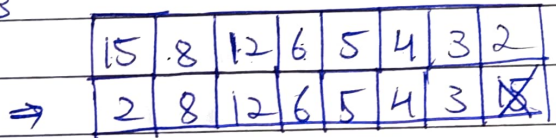
time taken to Heapify the remaining heap.

$$= O(\log n)$$

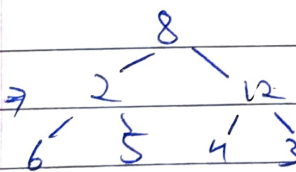
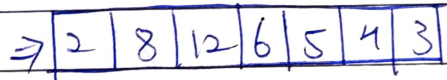
$$\therefore T.C. = O(\log n)$$



delete 15 and swap 15 with 2.



Heapify



Heapify

