

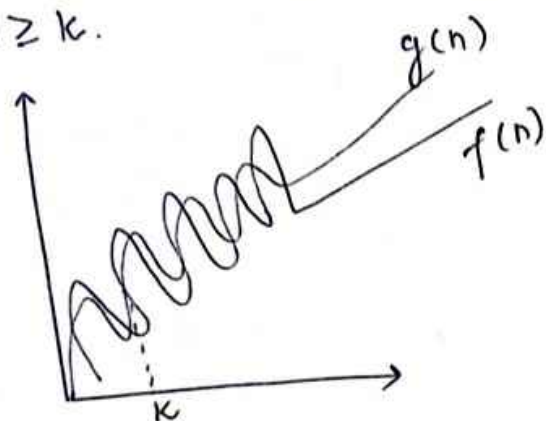
# Tutorial sheet-1

## Sol 1:- Asymptotic Notation:-

→ These notations are used to tell the complexity of an algorithm when the input is very large.  
→ It describes the algorithm efficiency and performance in a meaningful way. It describes the behaviour of time or space complexity for large instance characteristics.

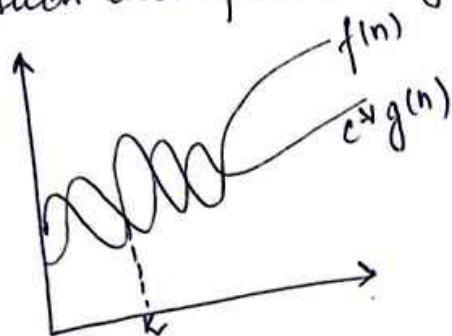
• The asymptotic notation of an algorithm is classified into 5 types:-

(i) Big Oh notation ( $O$ ) :- (Asymptotic upper Bound) The function  $f(n) = O(g(n))$ , if and only if there exist a +ve constant  $C$  and  $k$  such that  $f(n) \leq C * g(n)$  for all  $n \geq k$ .



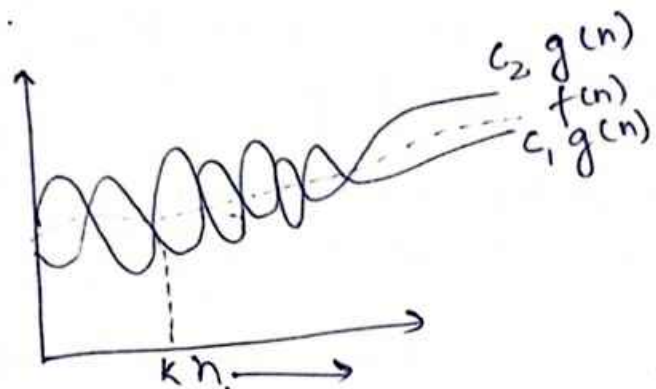
$$\begin{aligned} f(n) &= O(g(n)) \\ \text{iff} \\ f(n) &\leq C \cdot g(n) \\ \forall n \geq n_0, \\ \text{Some constant } C > 0. \end{aligned}$$

(ii) Big Omega notation ( $\Omega$ ) :- (Asymptotic lower bound) The function  $f(n) = \Omega(g(n))$ , iff there exists a +ve constant  $C$  and  $k$  such that  $f(n) \geq C * g(n)$  for all  $n, n \geq k$ .



$$\begin{aligned} f(n) &= \Omega(g(n)) \text{ iff} \\ f(n) &\geq C \cdot g(n) \\ \forall n \geq n_0 \text{ \& some} \\ \text{const } C > 0. \end{aligned}$$

(iii) Big theta notation ( $\Theta$ ): (Asymptotic tight bound) The function  $f(n) = \Theta(g(n))$ , iff there exists a +ve constant  $c_1, c_2$  &  $k$  such that  $c_1 \cdot g(n) < f(n) < c_2 \cdot g(n)$  for all  $n, n \geq k$ .



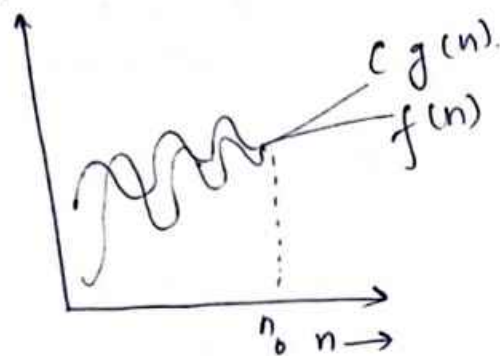
$$f(n) = \Theta(g(n))$$

iff

$$c_1 g(n) \leq f(n) \leq c_2 g(n)$$

$$\forall n \geq \max(n_1, n_2)$$

(iv) Small-oh ( $o$ ):  $o$  gives us upper bound.  
 $f(n) = o(g(n))$ .



$$f(n) \leq c g(n)$$

$$\forall n > n_0 \text{ \& \& } \forall c > 0$$

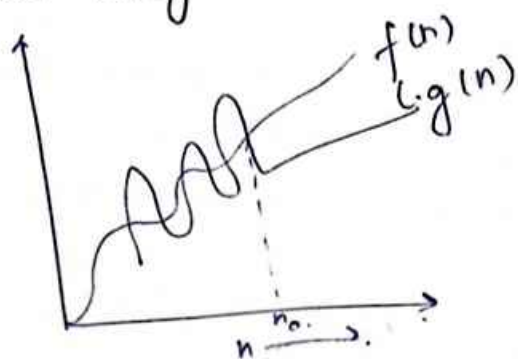
$$n = o(n^2)$$

$$n < \frac{1}{2} n^2$$

$$0.5 n^2$$

$$n < 0.001 n^2 n_0$$

(v) Small-omega ( $\omega$ ):



lower bound

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

$$f(n) > c \cdot g(n)$$

$$\forall n > n_0 \text{ \& \& } \forall c > 0$$

$$n^2 = \omega(n)$$

Sol 2:- for ( $i = 1$  to  $n$ )

{  $i = i * 2$ ;

}

Time complexity for a loop means no. of times loop has run.

→ For the above loop, the loop will run for the following values of  $i$  :-

$i$	1	2	4	8	16	32	...	$2^k$
value	$2^0$	$2^1$	$2^2$	$2^3$	$2^4$	$2^5$	...	$n$

$i = 1, 2, 4, 8, 16, 32, \dots, 2^k$  this means  $k$  times

$$i \leq 2^k = n$$

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

$$k = \log n \quad \left\{ \log_2 2 = 1 \right.$$

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

Sol 3:-  $T(n) = \begin{cases} 3T(n-1), & n > 0 \\ 1 & \end{cases}$

By forward substitution,

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

$$T(0) = 1$$

$$\begin{aligned} T(1) &= 3T(1-1) \\ &= 3T(0) \\ &= 3 \end{aligned}$$

$$\begin{aligned} T(2) &= 3T(2-1) \\ &= 3T(1) \\ &= 3 \times 3 = 3^2 \end{aligned}$$

$$\begin{aligned} T(3) &= 3T(3-1) \\ &= 3T(2) \\ &= 3 \times 3^2 = 3^3 \end{aligned}$$

For  $\vdots$

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

$$\boxed{\therefore T.C = O(3^n)}$$

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

By forward substitution,

$$T(0) = 1$$

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

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

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

$$\vdots \\ 2^n - 2^{n-1} - 2^{n-2} - 2^{n-3} \dots - 2^2 - 2^1 - 2^0$$

$$\Rightarrow 2^n - (2^n - 1) \\ = 2^n - 2^n + 1 = 1$$

$$\therefore \boxed{T.C = 1}$$

sol 5:-  

```

int = 1, s = 1;
while (s <= n)
{
    i++;
    s = s + i;
    printf("#");
}

```



$S_i = S_{i-1} + 1$   
 The value of 'i' increases by one for each iteration. The value contained in 's' at the  $i^{th}$  iteration is the sum of the first 'i' +ve integers. If k is the total no. of iterations taken by any program then while loop terminates if:

$$= [k(k+1)/2] > n$$

$$\text{So, } k = O(\sqrt{n})$$

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

Sol 6:- void function (int n)

{  
  int i, count = 0;

  for (i = 1; i <= n; i++)  $O(n)$   
    count++;

}

Time complexity :-  $O(n)$ .

Sol 7:- void function (int n)

{  
  int i, j, k, count = 0;

  for (i = n/2; i <= n; i++)

    for (j = 1; j <= n; j = j \* 2)  $O(\log n)$

      for (k = 1; k <= n; k = k \* 2)  $O(\log n)$

        count++;

}

$$J.C = \log n * \log n = O(\log^2 n)$$

$$\boxed{T.C = O(n \log^2 n)}$$

Sol 8:- function (int n)

```
{
    if (n == 1)
        return;
    for (i = 1 to n)           O(n) times
    {
        for (j = 1 to n)       O(n) times
        {
            printf("*");
        }
    }
    function(n-3);
}
```

Time complexity:-  $O(n^2)$  ans.

Sol 9:-

```
void function (int n)
{
    for (i = 1 to n)           O(n)
    {
        for (i = 1; j <= n; j = j + 1) O(n)
            printf("*");
    }
}
```

$$T.C = O(n) * O(n) = O(n^2)$$

$$\boxed{T.C = O(n^2)}$$

Sol 10:-  $n^k$  is  $O(c^n)$  ans.

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