

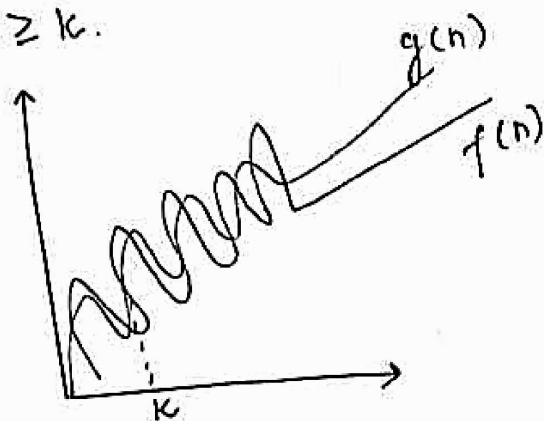
Tutorial sheet-1

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.

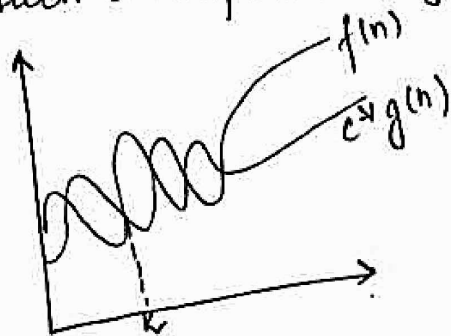
- The asymptotic notation of an algorithm is classified into 5 types:-

- (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 $n \geq k$.



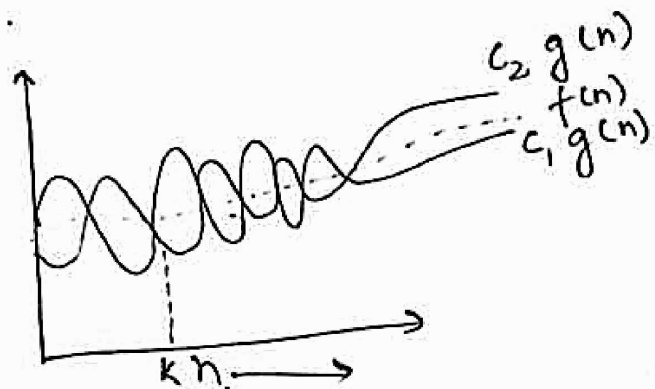
$$f(n) = O(g(n)) \text{ iff } f(n) \leq C \cdot g(n) \quad \forall n \geq n_0, \text{ some constant } C > 0.$$

- (ii) Big Omega notation (Ω) :- (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 $n, n \geq k$.



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

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



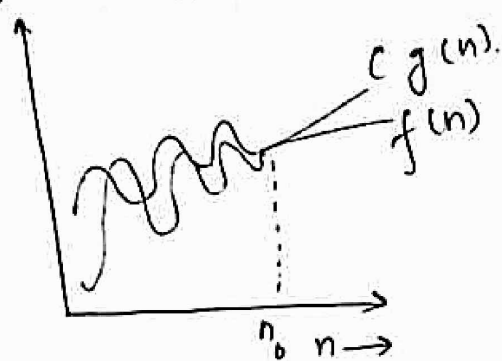
$$f(n) = \Theta(g(n))$$

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



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

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

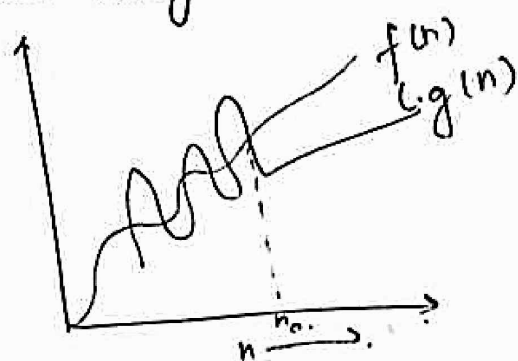
$$n = o(n^2)$$

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

$$0.5 n^2$$

$$n < 0.001 n^2 n_0$$

(v) Small-omega (ω):



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

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

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

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

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

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

By forward substitution,

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

$$T(0) = 1$$

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

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

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

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 : $1 + 2 + 3 + \dots + k$

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

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

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

Sol 6:- void function (int n)

{
 int i, count = 0;

 for (i = 1; ~~i~~ 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.

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