

Ans 1 → Asymptotic Notation :-

Mathematical Notation used to describe the running time of an algorithm.

Types of Asymptotic Notation -

1. Big-O Notation : It represent upper bound of Algorithm.

$$f(n) = O(g(n)) \text{ if } f(n) \leq Cg(n)$$

2. Omega Notation : It represent lower bound of Algorithm

$$f(n) = \Omega(g(n)) \text{ if } f(n) \geq Cg(n)$$

3. Theta Notation : It represents upper & lower bound of Algorithm.

$$f(n) = \Theta(g(n)) \text{ if } C_1g(n) \leq f(n) \leq C_2g(n)$$

Ans 2 →  
for (i = 1 to n)  
{  
    i = i \* 2;  
}

i = 1, 2, 4, 8, 16, ---- n (It is forming GP)

$$a_n = ar^{n-1}$$

$$n = ar^{k-1}$$

$$\therefore k = \log n + 1$$

$$n = 1 \times (2)^{k-1}$$

$$\log n = \log 2^{k-1}$$

$$\log n = (k-1) \log 2$$

$$\text{Time Complexity} = O(\log n)$$

Ans.

Ans 3 →  $T(n) = 3T(n-1)$  if  $n > 0$ , otherwise 1

$$T(1) = 3T(0)$$

$$T(0) = 1$$

$$T(1) = 3 \times 1$$

$$T(2) = 3T(1) = 3 \times 3 \times 1$$

$$T(3) = 3T(2) = 3 \times 3 \times 3 \times 1$$

⋮

$$T(n) = 3 \times 3 \times 3 \dots \times n = 3^n$$

$$\text{Time Complexity} = O(3^n)$$

Ans.

Ans 4,  $T(n) = 2T(n-1) - 1$  if  $n > 0$ , otherwise 1

$$T(0) = 1$$

$$T(1) = 2T(0) - 1$$

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

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

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

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

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

⋮

$$T(1) = 1$$

$$\text{Time Complexity} = O(1)$$

Ans.

Ans 5, `int i = 1, s = 1`

`while (s ≤ n)`

`{`

`i++;`

`s = s + i;`

`printf("#");`

`}`

$$i = 1 \quad s = 1$$

$$i = 2 \quad s = 1 + 2$$

$$i = 3 \quad s = 1 + 2 + 3$$

$$i = 4 \quad s = 1 + 2 + 3 + 4$$

$$1 + 2 + 3 + 4 + \dots + k > n$$

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

$$k^2 > n$$

$$k > \sqrt{n}$$

$$TC = O(\sqrt{n})$$

Ans.

Ans 6, `void func(int n)`

`{`

`int i, count = 0;`

`for (int i = 1; i * i ≤ n; i++)`

`count++;`

`}`

loop ends when  $i * i > n$   
 $k * k > n$   
 $k^2 > n$   
 $k > \sqrt{n}$

Time Complexity =  $O(\sqrt{n})$

Ans.

Ans 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)  
     {  
       for (k = 1; k ≤ n; k = k + 2)  
       {  
         count++;  
       }  
     }  
   }  
 }

1<sup>st</sup> loop: i = n/2 to n; i++  
           =  $O(n/2) = O(n)$

2<sup>nd</sup> loop: j = 1 to n, j = j \* 2  
           =  $O(\log n)$

3<sup>rd</sup> loop: k = 1 to n, k = k \* 2  
           =  $O(\log n)$

Time Complexity =  $O(n * \log n * \log n) = O(n \log^2 n)$

Ans.

Ans 8, function (int n)  
 {  
   if (n == 1) return;      —  $O(1)$   
   for (i = 1 to n) {  
     for (j = 1 to n) {      —  $O(n^2)$   
       printf("\*");  
     }  
   }  
   function (n-3);      —  $T(n-3)$   
 }



$$T(n) = T(n-3) + n^2, \quad T(1) = 1$$

$$T(4) = T(4-3) + 4^2 = 1 + 4^2$$

$$T(7) = T(7-3) + 7^2 = 1 + 4^2 + 7^2$$

$$T(10) = T(10-3) + 10^2 = 1 + 4^2 + 7^2 + 10^2$$

So,

$$T(n) = 1^2 + 4^2 + 7^2 + 10^2 + \dots + n^2 = \frac{n(n+1)(2n+1)}{6} = n^3$$

$$\text{Time Complexity} = O(n^3)$$

Ans.

Ans 9 → void function (int n)  
 {  
     for (i = 1 to n) {  
         for (j = 1; j ≤ n; j = j + 1) {  
             printf ("\*");  
         }  
     }  
 }

for 1<sup>st</sup> loop :-  $O(n)$

for 2<sup>nd</sup> loop :-  $O(n \times n) = O(n^2)$

$$\text{Time Complexity} = O(n^2)$$

Ans.

Ans 10 →  $f_1(n) = n^k$   
 $f_2(n) = C^n$        $k \geq 1, C > 1$

Asymptotic Notation b/w  $f_1$  &  $f_2$  is Big O i.e

$$f_1(n) = O(f_2(n))$$

$$= O(C^n) \quad n^k \leq C \cdot C^n$$

C is some constant.

Ans.