

## Tutorial - 1

Q1.

Ans → Asymptotic Notation are the mathematical notations used to describe the running time of an algorithm when the input tends towards a particular value or a limiting value.

There are mainly three asymptotic notations:

- Big-O notation
- Omega notation
- Theta notation

### Big-O-Notation

Big-O notation represents the upper bound of the running time of an algorithm. Thus, it gives the worst case complexity of an algorithm.

### Omega-Notation

Omega notation represents the lower bound of the running time of an algorithm. Thus it provides the best case complexity of an algorithm.

### Theta Notation

Theta Notation encloses the function from above and below. Since, it represents the upper and the lower bound of the running time of an algorithm, it is used for analyzing the average case complexity of an algorithm.

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Q2.

Sol: for (int  $i = 1$ ;  $i < n$ ;  $i = i * 2$ )  
 $T(1) = 1$

$$i = 1, 2, 4, 8, \dots, 2^{k-1}$$

$$T(n) = 1 + 2 + 4 + 8 + \dots + 2^{k-1}$$

$$T(n) = \frac{2(2^k - 1)}{(2 - 1)} = 2 \cdot 2^k - 2$$

~~Rec~~ ~~Rec~~

$$n = 2^k$$

$$\log n = k \log 2$$

$$k = \log_2 n$$

So, the time complexity is  $O(\log_2 n)$

Q3.  $T(n) = 3T(n-1)$  — (1)

$$T(1) = 1$$

Put  $n = n-1$  in eq (1)

$$\rightarrow T(n-1) = 3T(n-2) \text{ — (2)}$$

Put eq (2) in eq (1)

$$\rightarrow T(n) = 3(3T(n-2)) \text{ — (3)}$$

Put  $n = n-2$  — in eq (3)

$$T(n-2) = 3T(n-3) \text{ — (4)}$$

Put eq (4) in eq (3)

$$T(n) = 3 \cdot 3 \cdot (3T(n-3)) \text{ — (5)}$$

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

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$$T(n) = 3^n T(0)$$

Clearly, show that the complexity of this function is  $O(3^n)$ .

Q4.  $T(n) = 2T(n-1) - 1$  — (1)

Put  $n = n-1$  in eq (1).

$$T(n-1) = 2T(n-2) - 1$$
 — (2)

Put eq (2) in eq (1).

$$T(n) = 2(2T(n-2) - 1) - 1$$
 — (3)
$$= 4T(n-2) - 2 - 1$$

Put  $n = n-2$  in eq (1).

$$T(n-2) = 2T(n-3) - 1$$
 — (4)

Put eq (4) in (3)

$$T(n) = 2^3 T(n-3) - 2^2 - 2^1 - 2^0$$
 — (5)

$$T(n) = 2^{n-k} T(n-k) - 2^{n-k+1} - 2^{n-k+2} - \dots - 2^0$$

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

$$(n-k) \log 2 = 0$$

$$\boxed{k = n} \leftarrow \text{Put in eq (6)}$$

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

$$= 1 - 2(2^{n-1} - 1)$$

$$T(n) = 2^n + 3$$

$$O(2^n)$$

Ans

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Q5.

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

Q5.1 Q5.2 Q5.3 Q5.4

We can define the sum 's' according to relation  $S_i = S_{i-1} + i$ . 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' positive integers. If k is total number of iteration taken by the program, then while loop terminates if  $1 + 2 + 3 + \dots + k = \frac{k(k+1)}{2} > n$   
 So  $k = O(\sqrt{n})$   
 Time complexity is  $O(\sqrt{n})$ .

Q6.)

~~1 + 4 + 9 + \dots + (n-k)^2~~ ~~1 + 2 + 3 + \dots + (n-k)~~

~~1 + 4 + 9 + \dots + (n-k)^2~~

~~1 + 4 + 9 + \dots + (n-k)^2~~

$$T(n) = \frac{(n)(n+1)(n+2)}{6}$$

Ans  $\rightarrow O(n^3) \rightarrow$  Time Complexity

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Q7.)

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

→  $O(\log n)$

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

→  $O(\log n)$

for (int  $i = n/2$ ;  $i \leq n$ ;  $i++$ )

→  $O(n)$

Time Complexity  $O(n \log^2 n)$ .

Q8. function (int  $n$ )

{ if ( $n == 1$ ) return;

for ( $i=1$  to  $n$ )

→  $O(n)$  times

{ for ( $j=1$  to  $n$ ) {

print ("\*");

};

};

function ( $n-3$ ); →  $O(n)$  times.

};

Time Complexity  $O(n * n * n) = O(n^3)$ .

Q9.

void function (int  $n$ )

{

for ( $i=1$  to  $n$ )

{

for ( $j=1$ ;  $j \leq n$ ;  $j = j+1$ )

print ("\*")

};

}

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for inner loop  
 $O(n)$

for outer loop .  
 $n + n + \dots + n$  times .

$$= n^2$$

$O(n^2) \rightarrow$  time complexity .

i (10.) for the function,  $n^k$  and  $c^n$ , what is the asymptotic relationship between these function?

$k > 1$  &  $c > 1$  are constant .

It should be  $n^k$  is  $O(c^n)$