

23-02-2024 | DAA

## Assignment - 1

Name: - Harshita Bhatt

Section: - ML

Roll No.: - 30

Semester: - IV

① Asymptotic notations is used to describe the running time of an algorithm - how much time an algorithm takes with a given input,  $n$ . The types of notations are:-

a) Big O notation ( $O$ )

Denotes the  $[g(n)]$  tight upper bound of  $f(n)$

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

$$\text{iff } f(n) = (g(n))$$

$$\forall n = n_0$$

for some constant  $c > 0$

Example: -

$$f(n) = 2n^2 + 3n + 1$$

is  $O(n^2)$

b) Big Omega notation ( $\Omega$ )

Denotes the  $[g(n)]$  tight lower bound of  $f(n)$ .

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

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

$$\forall n = n_0$$

for some constant  $c > 0$ .

Example: -

$$f(n) = n^2 \text{ is } \Omega(n)$$

c) Theta ( $\theta$ ) Notation

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

Theta gives both tight upper bound and tight lower bound.

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

$$f(n) = o(g(n)) \text{ and } \Omega(g(n))$$

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

$$\text{i.e. } c_1 g(n) \leq f(n) \leq c_2 g(n)$$

$$\forall n \geq \max(n_1, n_2) \text{ and}$$

for some constant  $c_1 > 0$  and  $c_2 > 0$

Example:-

$$f(n) = n^2 \text{ is } \theta(n^2)$$

d) Small O notation ( $o$ )

Denotes the  $[g(n)]$  upper bound of  $f(n)$

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

$$\text{i.e. } f(n) < g(n)$$

$$\forall n > n_0$$

$$(\text{for all}) \quad c > 0$$

Example:

$$f(n) = n \text{ is } o(n^2)$$



e) Small Omega ( $\omega$ ) notation

Denotes the strict lower bound of a functions growth rate.

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

$g(n)$  is lower bound of  $f(n)$

$$f(n) > g(n)$$

$$\text{iff } f(n) > g(n)$$

$$\forall n > n_0$$

$$\text{for all } \forall c > 0$$

Example:

$$\cancel{f(n) = \omega(g(n))}$$

$$f(n) = n^2 \text{ is } \omega(n)$$

② for ( $i = 1$  to  $n$ )

{

$$i = i * 2;$$

}

$$\Rightarrow \text{for } (i = 1; i \leq n; i = i * 2)$$

$$\underbrace{1, 2, 4, 8, 16, \dots}$$

$\downarrow$   
G.P.

$$n = a r^{k-1}$$

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

$$n = \frac{2^k}{2}$$

$$2n = 2^k$$

$$k = \log_2(2n)$$

$$k = \log_2(2) + \log_2(n)$$

$$K = 1 + \log_2 n$$

$$\text{Time complexity} = O(\log_2 n)$$

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

$$T(0) = 1$$

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

$$\text{for } T(1)$$

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

$$\text{for } T(2)$$

$$\begin{aligned} T(2) &= 3T(2-1) \\ &= 3T(1) \\ &= 3T(0) \\ &= 3 \times 3 \times 1 \end{aligned}$$

$$\text{for } T(3)$$

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

$$\text{for } T(4)$$

$$\begin{aligned} T(4) &= 3T(4-1) \\ &= 3T(3) \\ &= 3 \times 3 \times 3 \times 3 \times 1 \end{aligned}$$



for  $T(n)$

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

$$= 3 \times 3 \times 3 \times 3 \times \dots \times 3$$

$$= 3^n$$

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

$$(9) T(n) = \begin{cases} 2T(n-1) - 1 & \text{if } n > 0, \\ 1 & \text{otherwise} \end{cases}$$

$$T(0) = 1$$

for  $T = 1$

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

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

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

$$= 2 - 1$$

$$= 1$$

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

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

$$= 2 \times 1 - 1$$

$$= 2 - 1$$

$$= 1$$

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

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

$$= 2 \times 1 - 1$$

$$= 2 - 1$$

$$= 1$$

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

$$= 2n - (2n-2) - \dots - 4 - 2 - 1$$

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

⑤

```

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

```

after first iteration:-

$$s = s + 1$$

after second iteration:-

$$s = s + 1 + 2$$

it goes on for  $x$  iterations.

$$1 + 2 + \dots + x \leq n$$

$$(x * (x + 1)) / 2 \leq n$$

$$O(x^2) \leq n$$

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

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++;
}

```



⑥ void function (int n)

{

int i, count = 0;

for (i = 1; i \* i ≤ n; i++)

count++

}