

(Q1)

Asymptotic notations :- Asymptotic means tending to infinity or is used for large numbers of size. These are used to tell the complexities of an algorithm when input is very large.

Types \leftarrow Time complexity
Space complexity

Space complexity - any extra space that we take apart from the input. We do not consider the variables as extra space as it is constant if it comes in relevance to ' n '. Then only it is extra space.

(1)

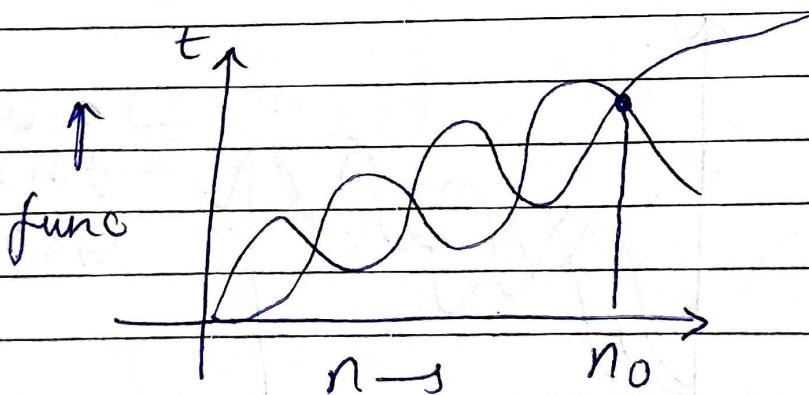
Big O(O) :-

$$\text{let } f(n) = O(g(n))$$

$$\text{if } f(n) \leq c(g(n))$$

$$n \geq n_0, c > 0$$

also $g(n)$ is tight upper bound of $f(n)$.



② Big Omega (Ω)

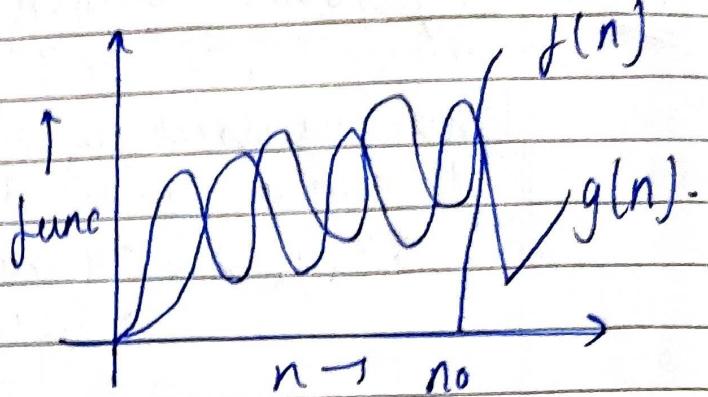
$$f(n) = \Omega(g(n))$$

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if and only if

$$f(n) \geq g(n)$$

for all $n \geq n_0$ and
some $c > 0$.



③ Theta (Θ) :-

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

{exact bound}

$$\text{if } c_1 g(n) \leq f(n) \leq c_2 g(n)$$

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

and constants $c_1 > 0$

$$c_2 > 0$$

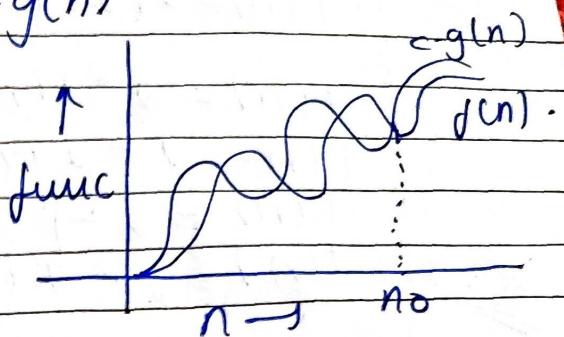


(4) small oh(δ) :-

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

if and only if $f(n) < c \cdot g(n)$
 $\forall (n > n_0)$

$g(n)$ is upper bound
 of $f(n)$
 for all $c > 0$.

(5) small omega (Ω)

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

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

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

$\forall n > n_0$ for all $c > 0$

Q2 :-

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 1$$

Q3-1

$$T(n) = 3T(n-1) \quad \text{if } n > 0, \text{ otherwise } 1.$$

$$T(1) = 0$$

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

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

put the value of $T(n-1)$ from ② to ①.

$$\begin{aligned} T(n) &= \cancel{3} \times [3T(n-2)] \\ &= 9T(n-2) \end{aligned}$$

$$\begin{aligned} T(n-2) &= 3T(n-3) \\ &= 9[3T(n-3)] \\ &= 3^3 [T(n-3)] \end{aligned}$$

Let us

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

$$\text{put } n-k = 0$$

$$\boxed{n=k}$$

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

$$= 3^n (1)$$

$$= 3^n \boxed{(n \geq n)}$$

Q4+

$$T(n) = [T(n-1) - 1] \text{ if } n > 0 \\ \text{otherwise 1.}$$

Sol-

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

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

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

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

$$T(n) = 2^2 [2T(n-3) - 1] - 2 - 1$$

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

\vdots k times

$$\begin{aligned} T(n) &= 2^k T(n-k) - 2^{k-1} - 2^{k-2} - \dots - 2^1 - 2^0 \\ &= 2^n - 2^{n-1} - 2^{n-2} - 2^{n-3} - \dots - 2^2 - 2^1 - 2^0 \\ &= 2^n - (1 + 2 + 2^2 + \dots + 2^{n-2} + 2^{n-1}) \\ &= 2^n - (1 - (2^n - 1)) \\ &= 2^n - 2^n + 1 \\ &= 1 \end{aligned}$$

$O(1)$

$$\boxed{s = \frac{a(r^n - 1)}{r - 1} \quad r \neq 1}$$



Q3)

i)

1

1

2

$$1+2=3$$

3

$$1+2+3=6$$

4

$$1+2+3+4=10$$

:

:

m times $1+2+3+\dots+m$ Assume $\boxed{J > m}$

$$\therefore S = (1+2+3+\dots+n)$$

$$= \frac{m(m+1)}{2}$$

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

$$m^2 > n$$

$$m = \sqrt{n}$$

$$\boxed{O(\sqrt{n})}$$

Q6)-1

i)

 1^2 2^2 3^2 4^2

?

:

 k^2

$$i * i > n$$

$$i * i = k^2$$

$$k^2 > n$$

$$k > \sqrt{n}$$

$$\boxed{O(\sqrt{n})}$$

Q7-

void function (int n)

{

 int i, j, k, count = 0;

 for (i = n/2; i <= n; i++) // log n

{

 for (j = 1; j <= n; j = j + 2) // log n x log n

{

 for (k = 1; k <= n; k = k + 2) // log n

{

 count++;

}

}

}

$\boxed{\Theta(\log^3 n)}$

Q8-

function (int n)

 if (n == 1) → 1

 return;

 for (i = 1 to n) → n+1

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

{ point ("*");

}

} function(n-3)

{

return n;

DATA STRUCTURE

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

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

$$\left\{ \begin{array}{l} T(n) = aT(n-b) + f(n) \\ a > 0 \quad f(n) = O(n^k) \end{array} \right.$$

where $k \geq 0$

$$a = 1$$

$$b = 3$$

$$f(n) = n^2 = O(n^2)$$

$$k = 2$$

case 2: if $a = 1$
 $O(n^{k+1})$
 $O(n^2+1)$

$$\boxed{O(n^3)}$$

⑨

i J

$$1 \quad 1 = 1 + 1$$

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

$$3 \quad 6 = 1 + 2 + 3 + 1$$

$$4 \quad 10 = 1 + 2 + 3 + 4 + 1$$

⋮

⋮

m terms

$$\text{arithmetic sum } j \geq n$$

$$J = 1 + 2 + 3 + \dots + m \quad J = m(m+1)$$

$$m^2 \geq n$$

$$m \geq \sqrt{n}$$

$$\boxed{O(n^{3/2})}$$



(10)

If k is a constant $\delta c > 1$

Then,

c^n grows faster than n^k as $n \rightarrow \infty$
because;

rate of growth of c^n is exponential
while rate of growth of n^k is polynomial

So we can say $O = \boxed{O(n^k)}$.

(9)