

ASSIGNMENT - 01

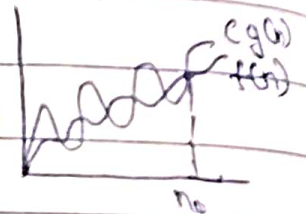
Q1 Asymptotic notations are used to tell the complexity of the algorithm when the input is very large.

(i) Big O Notation (O) \rightarrow

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

It describes that function $g(n)$ is tight upper bound of func $f(n)$. iff $f(n) \leq c * g(n)$

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



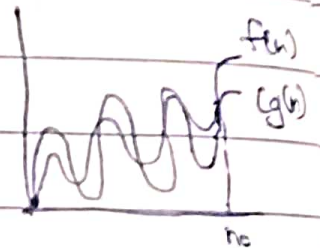
(ii) Big Omega (Ω) Notation \rightarrow

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

It describes that function $g(n)$ is tight lower bound of func $f(n)$.

$$\text{iff } f(n) \geq c * g(n)$$

$\forall n \geq n_0$ and for some constant $c > 0$



(iii) Theta Notation (Θ) \rightarrow

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

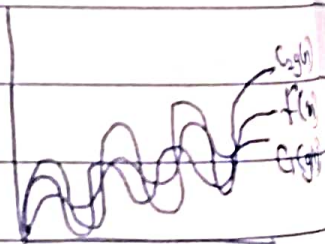
It describes that theta gives both tight upper bound and tight lower bound.

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

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

$\forall n > \max(n_1, n_2)$ and some constant ~~c_1, c_2~~

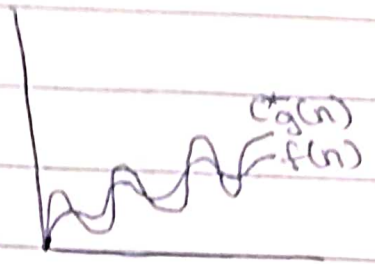
$c_1 > 0$ & $c_2 > 0$



(iv) Small (o) Notation \rightarrow

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

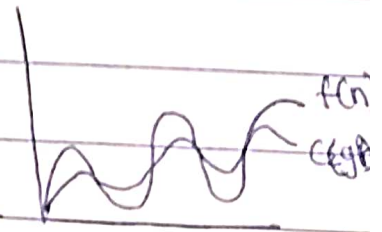
It describes that $g(n)$ is upper bound of $f(n)$. iff $f(n) < c * g(n)$
 $\forall n > n_0$ and $c > 0$



(v) Small Omega (ω) Notation \rightarrow

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

It describes that $g(n)$ is lower bound of func $f(n)$. iff $f(n) > c * g(n)$
 $\forall n > n_0$ and $c > 0$



2/

for ($i = 1$ to n)

{ $i = i * 2$;
 }

$$i = 1 \quad 2 \quad 4 \quad n/2 \quad n$$

\therefore Acc. to G.P

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

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

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

Logging on both sides

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

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

$$k = \log_2(n) + 1$$

$$k = \log_2 n$$

\therefore Time complexity $= O(\log_2 n)$

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

$\therefore T(0) = 1$

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

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

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

$\therefore T(n) = 3^n$

\therefore Time complexity = $O(3^n)$ Ans

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

$\therefore T(0) = 1$

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

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

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

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

\therefore Time complexity = $O(1)$ Ans

5) $\text{int } i = 1, s = 1;$

$\text{while } (s \leq n) \{$

$i++;$

$s = s + i;$

$\text{printf}("%d \# ");$

$\}$

$i = 1, 2, 3, 4, \dots$

$s = 1, 3, 6, 10, \dots, n$

\therefore sum of first n natural no.

$s = \frac{i(i+1)}{2}$

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$$\text{Also } S \leq n$$

$$\therefore \frac{i(i+1)}{2} \leq n$$

$$i^2 + i \leq 2n$$

$$i^2 < n$$

$$\therefore i < \sqrt{n}$$

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

void function (int n) {

61 int i=1, count=0

for (i=1; i*i <= n; i++)

{ count++;

}

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$$i=1$$

$$i*i \leq n$$

$$i=2$$

$$2*2 \leq n$$

$$i=3$$

$$3*3 \leq n$$

$$\therefore i \leq \sqrt{n}$$

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

71

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++;

}

| i | j | k | times |
|----------------------------|-------|-------|----------------|
| $n/2$ | $1-n$ | $1-n$ | $(\log_2 n)^2$ |
| $n/2+1$ | $1-n$ | $1-n$ | $(\log_2 n)^2$ |
| ⋮ | | | |
| n | $1-n$ | $1-n$ | $(\log_2 n)^2$ |
| <hr/> | | | |
| $\frac{n (\log_2 n)^2}{2}$ | | | |

$$\therefore \boxed{T.C = O(n (\log_2 n)^2)}$$

8.) function (int n) {
 if (n==1) return;
 for (i=1 to n) { $\longrightarrow O(n)$
 for (j=1 to n) { $\longrightarrow O(n^2)$
 printf("%x");
 }
 }
 function(n-3); $\longrightarrow T(n-3)$
 }

$$\therefore T(n) = T(n-3) + O(n^2) + O(n)$$

$$\therefore \boxed{T(n) = T(n-3) + O(n^2)} \quad \underline{\text{Ans}}$$

$$\therefore \boxed{T.C = O(n^2)} \quad \forall n \geq 2$$

9.) void function (int n) {
 for (i=1 to n) {
 for (j=1; j <= n; j=j+1) {
 printf("%x");
 }
 }
 }

| i | f _i | times |
|-------------------|----------------|-------|
| 1 | 1-n | n |
| 2 | 1-n | n/2 |
| 3 | 1-n | n/3 |
| ⋮ | | |
| n | 1-n | 1 |
| <hr/> | | |
| $n \times \log n$ | | |

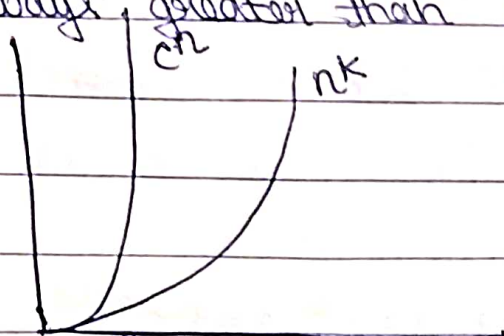
$$\therefore \boxed{\text{Time complexity} = O(n \log n)}$$

10/

 n^k and c^n $k \geq 1$ and $c > 1$

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

This is because exponential time complexity is always greater than polynomial time complexity.



~~Since~~, then ^{given} conditions does not give any value for c and n .