

Tutorial Sheet-1

Sol 1: Asymptotic Notation: These notations are used to tell the complexity of an algorithm when the input is very large. It describes the algorithm efficiency and performance in a meaningful way.

The asymptotic notation of an algorithm is classified into 5 types :-

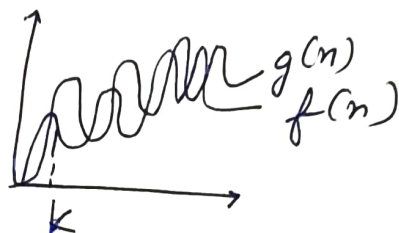
(i) Big-oh notation (O):

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

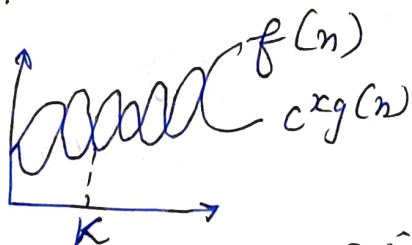
iff

$$f(n) \leq c \cdot g(n)$$

$\forall n \geq n_0$,
some constant $c > 0$



(ii) Big Omega notation (Ω): (A symmetric lower bound).

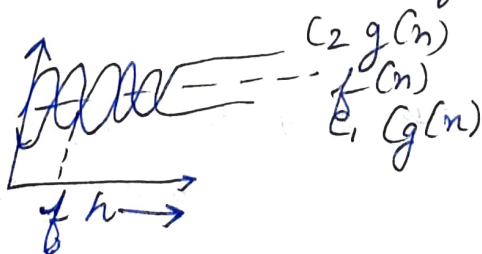


$$f(n) = \Omega(g(n)) \text{ iff}$$

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

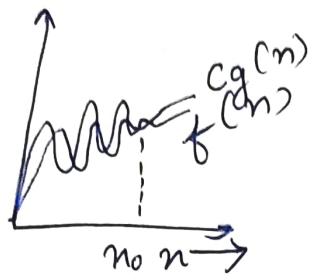
$\forall n \geq n_0$ for some
const $c > 0$.

(iii) Big Theta notation (Θ): (A symmetric tight bound)



$f(n) = \Theta(g(n))$ iff $C_1 g(n) \leq f(n) \leq C_2 g(n)$
 $\forall n \geq \max(n_1, n_2)$.

(iv) Small-oh(o):



$$f(n) < Cg(n)$$

$$\forall n > n_0 \ \& \ \forall C > 0$$

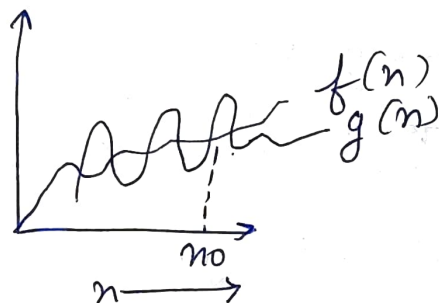
$$n = O(n^2)$$

$$n < \frac{1}{2} n^2$$

$$0 \leq n^2$$

$$n < 0.001 n^2 n_0$$

(v) Small-omega(ω):



lower bound

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

$$f(n) > C \cdot g(n)$$

$$\forall n > n_0 \ \& \ \forall C > 0$$

$$n^2 = \omega(n)$$

Sol 2: for ($i=1$ to n)

$i = i * 2;$

 }

Time complexity for a loop means
 no. of times a loop has run.

→ for the above loop, the loop will run for the following values of i :-

i	1	2	4	8	16	32	-----	2^k
values	2^0	2^1	2^2	2^3	2^4	2^5	---	n

$i = 1, 2, 4, 8, 16, 32, \dots, 2^k$ these means k times

$$\therefore \log_2 2 = \log_2 n$$

$$k = \log n \quad (\log_2 2 = 1)$$

$$\therefore T.C. = O(\log n)$$

Sol 3: $T(n) = \begin{cases} 3T(n-1), & n > 0 \\ 1. & \end{cases}$

By forward substitution,

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

$$T(0) = 1$$

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

$$= 3T(0)$$

$$= 3$$

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

$$= 3T(1)$$

$$= 3 \times 3 = 3^2$$

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

$$= 3T(2)$$

$$= 3 \cdot 3^2 = 3^3$$

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

Sol 4: $T(n) : \begin{cases} 2T(n-1) - 1, n > 0 \\ 1. \end{cases}$

By forward substitution,

$$T(0) = 1.$$

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

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

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

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

$$= 2^2 - 2^1 - 1$$

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

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

$$= 2(2^2 - 2^1 - 1) - 1$$

$$= 2^3 - 2^2 - 2^1 - 1$$

⋮

$$2^n - 2^{n-1} - 2^{n-2} - 2^{n-3} - \dots$$

$$- 2^2 - 2^1 - 2^0$$

$$\Rightarrow 2^n - (2^n - 1)$$

$$= 2^n - 2^n + 1 - 1$$

$$\therefore \boxed{TC = 1}$$

Sol 5: $int i = 1, s = 1;$

while $(s \leq n)$

{ $i++;$

$s = s + i;$

printf("#"); }

$$S_i^0 = S_i^0 - 1 + 1$$

The value of 'i' increases by one for each value contained in 'S' at the i^{th} iteration as the sum of the first 'i' +ve integers. If k is the total no. of iterations taken by any program then while loop terminates if : $1+2+3+ \dots + k$
 $= [R(k+1)/2] > n$.

$$\text{i.e. } k = O(\sqrt{n})$$

$$\therefore T.C. = O(\sqrt{n})$$

Sol 6:- void function (int n)

{ int i, count=0;

for (i=1; i<=n; i++) $O(n)$

count++;

}

Time complexity: $O(n)$

Sol 7:- void function (int n)

{ int i, j, k, count=0;

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

for (j=1; j<=n; j++)

for (k=1; k<=n; k=k*2) $O(\log n)$

for (k=1; k<=n; k=k*2) $O(\log n)$

count++;

}

$$T.C. = \log n * \log n = O(n \log^2 n)$$

$$T.C. = O(n \log^2 n)$$

Sol 8:- function (int n)

{ if (n == 1)
return;

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

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

{ printf(" * ");

}

function(n-3);

}

Time complexity = $O(n^2)$

Sol 9:- void function (int n)

{ for (i = 1 to n) {

for (j = 1 to j ≤ n; j = j + 1) $O(n)$

{ printf(" * ");

}

}

}

To Co = $O(n) * O(n) = O(n^2)$

To Co = $O(n^2)$

Sol 10:- n^k is $O(c^n)$ and $n^k = O(c^n)$