

⇒ Order of Growth

A function $f(n)$ is said to be growing faster than $g(n)$ if

$$\rightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty \quad \text{OR} \rightarrow \lim_{n \rightarrow \infty} \frac{g(n)}{f(n)} = 0 \quad \left[\begin{array}{l} n \geq 0 \\ f(n), \\ g(n) \geq 0 \end{array} \right]$$

$$\rightarrow \begin{array}{l} f(n) = 2n + 5 \\ g(n) = n^2 + n + 6 \end{array}$$

$$\lim_{n \rightarrow \infty} \frac{2n + 5}{n^2 + n + 6} \rightarrow \text{Dividing } N^x \text{ \& } D^x \text{ by } n^2$$

$$\lim_{n \rightarrow \infty} \frac{2/n + 5/n^2}{1 + 1/n + 6/n^2} = \underline{\underline{0}} \quad \therefore f(n) \text{ is growing faster than } g(n)$$

→ Direct way -

- ① Ignore the lower order terms
- ② Ignore leading constant

$$\rightarrow \text{Constant} < \log \log n < n^{1/3} < n^{1/2} < n < n^2 < n^3 < n^4 < n^4 < 2^n < n^n$$

$$\text{①)} \begin{array}{l} f(n) = 2n^2 + n + 6 \\ g(n) = 100n + 8 \end{array} \rightarrow \text{Ignore lower order terms \& leading coeff}$$

$f(n)$'s order is n^2
 $g(n)$'s order growth is n
 $\therefore g(n)$ is better.

Q2) $f(n) = C_1 \log n + C_2 \rightarrow \log n$
 $g(n) = C_3 n + C_4 \log \log n + C_5 \rightarrow n$

as n is higher order than $\log \log n$

$\therefore f(n)$ is better than $g(n)$

{ Order $\propto \frac{1}{\text{better}}$ }

Q3) $f(n) = C_1 n^2 + C_2 n + C_3 \rightarrow n^2$
 $g(n) = C_4 n \log n + C_5 n + C_6 \rightarrow n \log n$

$\therefore g(n)$ is better

\Rightarrow Best, Average & Worst Case

Asymptotic Notation

\rightarrow `int getSum (int arr[], int n)`
`{`

`int sum = 0`

`for (int i=0 ; i<n; i++)`

`sum = sum + arr[i];`

`return sum;`

`}`

\rightarrow $C_1 n + C_2 \rightarrow n$ is the order
 Linear Constant

```

Q2) int getSum (int arr[], int n)
{
    if (n % 2 != 0)
        return 0;
    for (int i = 0; i < n; i++)
        sum = sum + arr[i];
    return sum;
}

```

→ For Best Case → Constant

Average Case → Linear, under the assumption that even and odd are equally likely.

Worst Case → Linear, when n is even.

→ We make the decision of worst case, only when we have different cases.

⇒ Notations

→ Big O : Represent exact bound or upper bound.

→ Theta : Represents exact bound.

→ Omega : Represent exact or lower bound.

⇒ Big O Notation (Upper bound on Order of Growth)

We say $f(n) = O(g(n))$ iff there exists constants C and n_0 such that $f(n) \leq Cg(n)$ for all $n \geq n_0$.

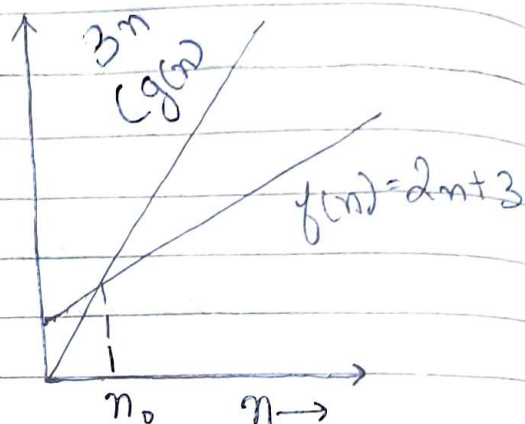
Ex $\rightarrow f(n) = 2n + 3$ can be written as $O(n)$

$$\rightarrow f(n) \leq g(n) \text{ for all } n > n_0$$

$$(2n + 3) \leq cn \text{ for all } n > n_0$$

(Highest growing term is 2)
 $c = 3$

\hookrightarrow Highest growing term + 1



$$(2n + 3) \leq 3n$$

$$3 \leq n$$

$$\underline{\underline{n_0 = 3}}$$

Only Highest order

$$\rightarrow \left\{ \frac{n}{4}, 2n + 3, \frac{n}{100} + \log n, n + 1000, \frac{n}{1000} \right\} \in [O(n)]$$

$$\rightarrow \{ n^2 + n, 2n^2, n^2 + 1000n, n^2 + 2 \log n \} \in [O(n^2)]$$

$$\rightarrow \{ 1000, 1, 2, 3, 12 \} \in [O(1)]$$

\Rightarrow

Omega Notation (Lower bound)

$f(n) = \Omega(g(n))$ iff there exist positive constants c and n_0 such that $0 \leq g(n) \leq cf(n)$ for all $n > n_0$.

Ex $\rightarrow f(n) = 2n + 3 \rightarrow \Omega(n)$

$$2) \overset{x}{2}n^2 + \overset{x}{3}n + \overset{x}{6} \rightarrow \underline{\Omega(n^2)}$$

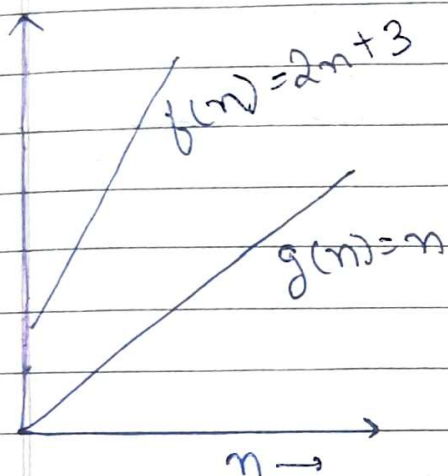
$$\rightarrow f(n) = 2n + 3 \rightarrow \Omega(n)$$

↪ Highest growing term = 2

$$C = 1 \sim [\text{Highest growing term} - 1]$$

$$Cg(n) = n$$

$$n \leq 2n + 3 \quad [n_0 = 0] \text{ -ve not allowed}$$



$$\rightarrow \left\{ \frac{n}{4}, \frac{n}{2}, 2n, 3n, 2n+3, n^2, 2n^2, \dots \right\} \in [\Omega(n)]$$

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

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

→ Omega notation is useful when we have lower bound on time complexity.

⇒

Theta Notation (Exact order of growth)

$f(n) = \Theta(g(n))$ iff there exist positive constants C_1, C_2 and n_0 such that

$$0 \leq C_1 g(n) \leq f(n) \leq C_2 g(n)$$

for all $n \geq n_0$

Ex: $f(n) = 2n + 3 \rightarrow \underline{\underline{\Theta(n)}}$

$$C_1 \times n \leq 2n + 3 \leq C_2 g(n)$$

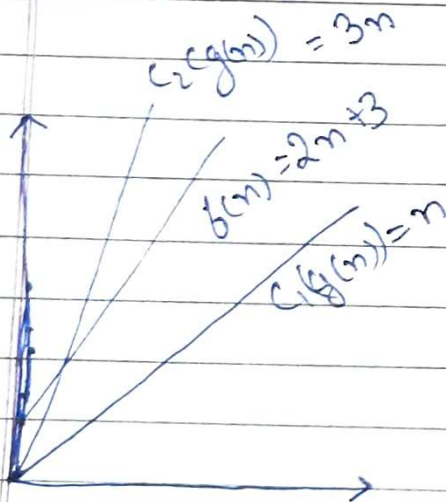
$C_1 = 1 \rightarrow$ Any value smaller than Highest growing term
 $C_2 = 3 \rightarrow$ " " " " " " " " " "

$$1 \times n \leq 2n + 3 \leq 3 \times n$$

$$n \geq 0$$

$$n \geq 3$$

$$[n_0 = 3] \checkmark$$



→ If $f(n) = \Theta(g(n))$
 then $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$
 and $g(n) = O(f(n))$ and $g(n) = \Omega(f(n))$

→ Theta is useful to represent time complexity

when we know exact bound. For example, Time Complexity to find sum, max and min in an array is $\Theta(n)$.

$$(3) \left\{ \frac{n^2}{4}, \frac{n^2}{2}, \dots, 2n^2, 2n^2 + 100n \dots \right\} \in [\Theta(n^2)] \Leftarrow$$

→ Analysis of some common loop :-

① for (int i=0; i < n; i=i+c)

{
// Some $\Theta(1)$ work
}

$$\begin{array}{ccccccc} n=10 & & & & & & \\ c=2 & & & & & & \\ 0 & 2 & 4 & 6 & 8 & & \\ \hline & & & & 5 & & \end{array}$$

$$\{10/2=5\} \Rightarrow \underline{n/c}$$

$$\Theta(n/c)$$

$$= [\Theta(n)] \Leftarrow$$

② for (int i=n; i > 0; i=i-c)

{
// Some $\Theta(1)$ work
}

$$\begin{array}{l} n=10 \\ c=2 \end{array}$$

$$i = 10, 8, 6, 4, 2$$

$$\begin{array}{l} n=11 \\ c=2 \end{array}$$

$$i = 11, 9, 7, 5, 3, 1$$

$$(\text{approx}) \lfloor n/c \rfloor \sim \underline{\underline{\Theta(n)}}$$

3) `for (int i=1; i < n; i = i * c)`

`{`
 // Some $O(1)$ work
`}`

$n=32, c=2$	$n=33, c=2$
1, 2, 4, 8, 16	1, 2, 4, 8, 16, 32

$\rightarrow 1, c, c^2, c^3, \dots, c^{k-1}$

$$[c^{k-1} < n] \Rightarrow k-1 < \log_c n$$

$$\therefore [O(\log n)] \leq \lceil \log_c n + 1 \rceil$$

4) `for (int i=n; i > 1; i = i/c)`

`{`
 // Some $O(1)$ work
`}`

$n=32, c=2$	$n=33, c=2$
32, 16, 8, 4, 2	33, 16, 8, 4, 2

- Similar to previous question
 $[O(\log n)] \leq$

5) `for (int i=2; i < n; i = pow(i, c))`

`{`
 // Some $O(1)$ work
`}`

$$c=2, n=32$$

$$2, 2^2, (2^2)^2 = 2, 4, 16$$

$$\rightarrow 2, 2^c, (2^c)^c \dots 2^{c^{K-1}}$$

$$\rightarrow 2, 2^c, 2^{c^2}, \dots 2^{c^{K-1}}$$

$$\rightarrow 2^{c^{K-1}} < n$$

$$\rightarrow c^{K-1} < \log_2 n$$

$$K-1 < \log_c \log_2 n$$

$$\lfloor K < \log_c \log_2 n + 1 \rfloor \ll \Rightarrow [\Theta(\log \log n)] \ll$$

⑥ void fun (int n)

```
{
  for (int i=0; i<n; i++) ] O(n)
    // some O(1) work
```

```
  for (int i=1; i<n; i=i*2) ] O(log n)
    // some O(1) work
```

```
  for (int i=1; i<100; i++) ] O(1)
    // some O(1) work
}
```

$$\therefore \underbrace{O(n)}_x + \underbrace{O(\log n)}_x + \underbrace{O(1)}_x = \underline{\underline{O(n)}} \ll$$

⑦ void fun (int n)

```
{
  for (int i=0; i<n; i++)
    for (int j=1; j<n; j=j*2)
      // some O(1) work
```

```
}
```

→ In nested loops we simply multiply the no. of times the loops run

$$\rightarrow O(n) \times O(\log n) \Rightarrow [O(n \log n)] //$$

⑧ Void fun (int n)

```

{
    for (int i = 0; i < n; i++)
        for (int j = 1; j < n; j = j * 2)
            // some O(1) work
}

```

$O(n \log n)$

```

for (int i = 0; i < n; i++)
    for (int j = 1; j < n; j++)
        // some O(1) work
}

```

$O(n^2)$

$$\Rightarrow \underset{\times}{O(n \log n)} + O(n^2) \Rightarrow [O(\underline{n^2})] //$$