

Data Types:-

int, floating point, char, string, (1B)
(small int, long int)
boolean (1 bit).

Asymptotic Notations:-

To choose the best algorithm, we need to check efficiency of each algorithm.

The efficiency can be measured by computing time complexity of each algorithm.

Asymptotic notation is a shorthand way to represent the time complexity.

Using asymptotic notations we can give time complexity as "fastest possible", "slowest possible" or "average time".

Various notations such as Ω , Θ and O are used & called as asymptotic notions.

- large values of n
1. Big-Oh Notation : O - upper bound
 2. Big-omega notation : Ω - lower bound
 3. Theta Notation : Θ - Avg. / Tight bound
 4. Little-oh notation : o
 5. Little-omega notation : ω

small values of n : $1 < \log n < \sqrt{n} < n < n \log n < n^2 < n^3 < \dots < 2^n < 3^n$

Big-Oh Notation (O) \rightarrow worst case running time of algo.

A function $f(n)$ is said to be in $O(g(n))$, if $f(n)$ is denoted as $f(n) \in O(g(n))$, if $f(n)$ is bounded above by some constant multiple of $g(n)$ for all large n .

i.e. if there exist some positive constant c and some non-negative integer n_0 such that

$$f(n) \leq c(g(n)) \text{ for all } n \geq n_0.$$

- * $O(g(n))$: class of ftns/ $f(n)$ that grow no faster than $g(n)$
- * $O \rightarrow$ puts asymptotic UB on $f(n)$.

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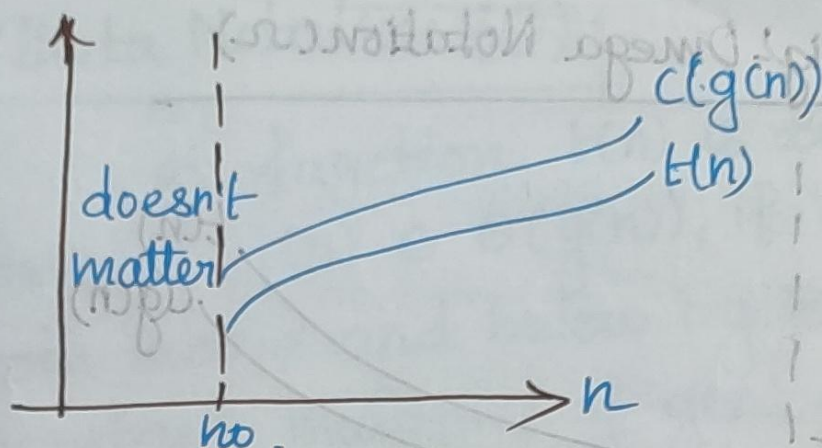


Fig:- Big-oh(O)

$$\text{Let } t(n) = 2n + 3$$

$$2n + 3 \leq \dots ?$$

$$2n + 3 \leq 5n \quad n \geq 1$$

$$\text{here } c = 5$$

$$\& g(n) = n$$

$$\boxed{t(n) = O(n)}$$

Also: $2n + 3 \leq 5n^2 \quad n \geq 1$

Where $c = 5$ & $g(n) = n^2$

$$\boxed{\therefore t(n) = O(n^2)}$$

Big-Omega Notation (Ω) \rightarrow Best case running time

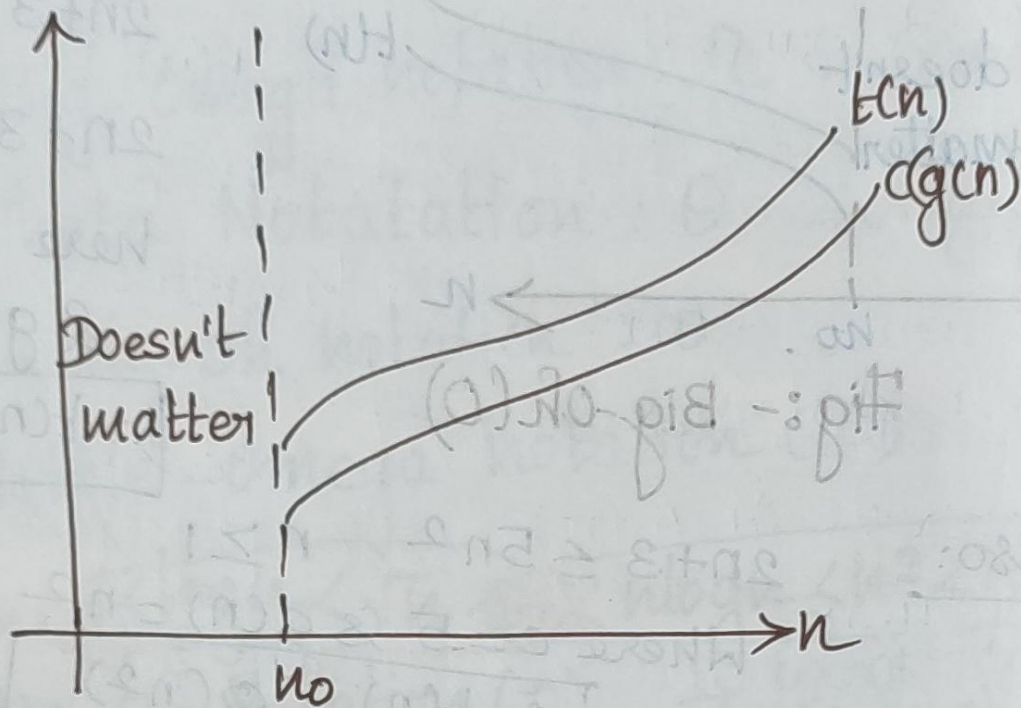
A function $t(n)$ is said to be in $\Omega(g(n))$, denoted as $t(n) \in \Omega(g(n))$, if $t(n)$ is bounded below by some positive constant multiple of $g(n)$ for all large n .
i.e., there exist some positive constant c & some non-negative integer n_0 .

Such that $\boxed{t(n) \geq c(g(n)) \text{ for all } n \geq n_0}$

* It represents the lower bound of the resources required to solve a problem.

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Fig: Big-Omega Notation (Ω)



$$\text{Let } f(n) = 2n + 3$$

$$2n + 3 \geq \text{---} ??$$

$$2n + 3 \geq 1n$$

$$\text{here } c = 1 \text{ and } g(n) = n$$

$$\boxed{f(n) = \Omega(n)}$$

$$2n + 3 \geq 1 \log n \quad n \geq 1$$

$$\text{here } c = 1 \text{ and } g(n) = \log n$$

$$\boxed{f(n) = \Omega(\log n)}$$

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Theta Notation (Θ):

A function $t(n)$ is said to be in $\Theta(g(n))$, denoted $t(n) \in \Theta(g(n))$, if $t(n)$ is bounded both above and below by some positive constant multiples of $g(n)$ for all large n . i.e., if there exist some positive constant c_1 and c_2 and some non-negative integer n_0 such that

$$c_2 g(n) \leq t(n) \leq c_1 g(n) \quad \forall n > n_0.$$

$$\boxed{\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))} //$$

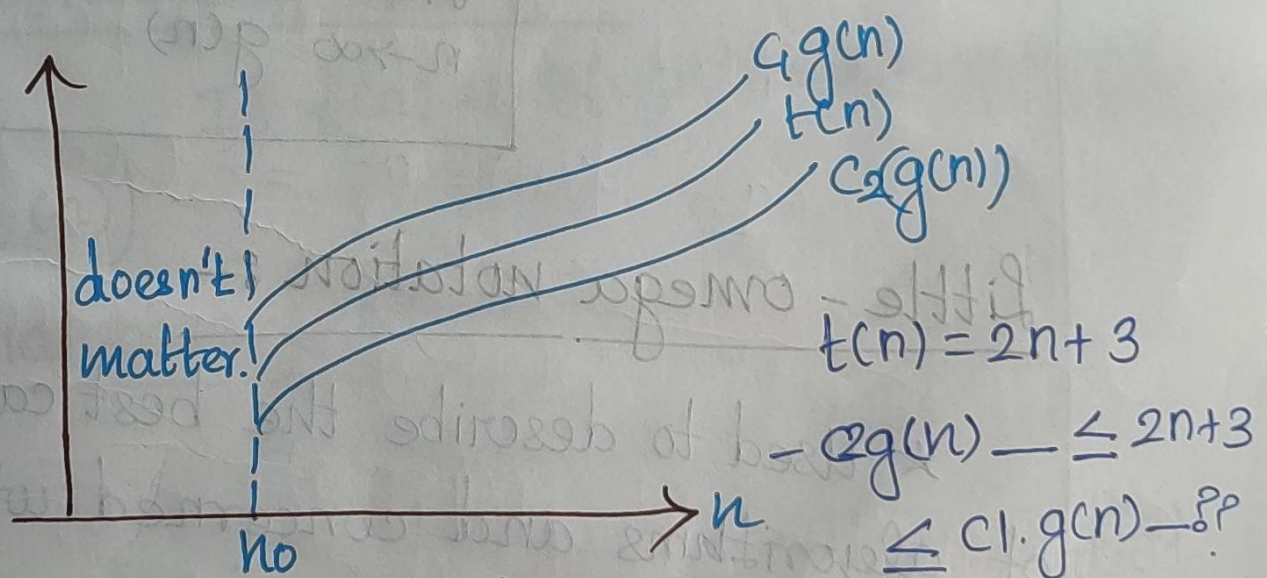


Fig: Θ -Notation.

$n > 1$

$$1n \leq 2n + 3 \leq 5n$$

here $c_1 = 5$ & $c_2 = 1$

$$\& g(n) = n //$$

$$\boxed{t(n) = \Theta(n)} //$$

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Little-oh notation :- (o)

★ used to describe the worst case analysis of algorithms and concerned with small values of n .

A fty. $f(n)$ is said to be in $o(g(n))$, denoted $f(n) \in o(g(n))$, if there exist some positive constant c and some non-negative integer such that

$$\begin{aligned} f(n) &\leq c g(n) \\ \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} &= 0. \end{aligned}$$

Little-omega notation :-

★ Used to describe the best case analysis of algorithms and concerned with small values of n .

The function $f(n) = \omega(g(n))$ iff

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty \quad \text{or} \quad \lim_{n \rightarrow \infty} \frac{g(n)}{f(n)} = 0$$

71. # Properties of O , Ω and Θ

① General property:- (Ω & Θ)

If $t(n)$ is $O(g(n))$ and then $a * t(n)$ is $O(g(n))$. Similar for Ω and Θ .

② Transitive property:- ($\forall \Omega, \Theta, O$ & ω)

If $f(n) \in O(g(n))$ and $g(n) \in O(h(n))$, then $f(n) \in O(h(n))$; that is O is transitive. Also Ω , Θ , O and ω are transitive.

③ Reflexive property:-

If $f(n)$ is $\Theta(g(n))$ then $g(n)$ is $\Theta(f(n))$.

④ Transpose property:-

If $f(n) = O(g(n))$ then $g(n)$ is $\Omega(f(n))$.

⑤ Symmetric property:-

If $f(n)$ is $\Theta(g(n))$ then $g(n)$ is $\Theta(f(n))$.

⑦② ① General properties:- (True $\forall \Omega$ & θ)

If $f(n)$ is $O(g(n))$ then $a \times f(n)$ is $O(g(n))$

eg:- $f(n) = 2n^2 + 5$ is $O(n^2)$

then $\Rightarrow 7 \cdot f(n) = 7(2n^2 + 5)$

$= 14n^2 + 35$ is $\Rightarrow O(n^2)$

$\therefore f(n) = \Omega(g(n)) \rightarrow a \times f(n) \rightarrow \Omega(g(n))$

② Reflexive property:-

If $f(n)$ is given then $f(n)$ is $O(f(n))$

eg:- $f(n) = n^2 \Rightarrow O(n^2)$ //

\uparrow

$f(n)$

③ Transitive property:- [$\forall O, \Omega$ & θ]

If $f(n)$ is $O(g(n))$ and $g(n)$ is $O(h(n))$

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

(73) eg. $f(n) = n \xrightarrow{UB} g(n) = n^2 \xrightarrow{UB} h(n) = n^3$
 n is $O(n^2)$ & n^2 is $O(n^3)$
 then n is $O(n^3)$

④ Symmetric property:-

If $f(n)$ is $\Theta(g(n))$ then $g(n)$ is $\Theta(f(n))$

eg:- $f(n) = n^2 \rightarrow g(n) = n^2$ *

$$f(n) = \Theta(n^2)$$

$$g(n) = \Theta(n^2)$$

When both the functions are same,
 then they are symmetric.

⑤ Transpose symmetric:- $[O \& \Omega]^{dy.}$

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

$$\Omega(f(n))$$

LB.

eg:- $f(n) = n$ $g(n) = n^2$

then n is $O(n^2)$ and n^2 is $\Omega(n)$.

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If one function forms an upper bound for other fty. then the other fty. will form a lower bound for the other function.

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

When same fty. is acting both as UB & LB

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

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

* If $f(n) = O(g(n))$
and $d(n) = O(e(n))$

then $f(n) + d(n) = O(\max(g(n), e(n)))$

eg:- $f(n) = n \Rightarrow O(n)$

$d(n) = n^2 \Rightarrow O(n^2)$

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

* If $f(n) = O(g(n))$
 $d(n) = O(e(n))$

(#5)

then $f(n) * d(n) = O(g(n) * e(n))$

$n \quad n^2 = n^3 //$

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