

20-01-22 | See Abdul Bari videos on youtube for computing complexity.

TIME COMPLEXITY: <sup>relationship of</sup>  $f(n)$  that tells us how the time is going to grow as the input grows. <sub>size grows</sub>  
mathematical  $f(n)$

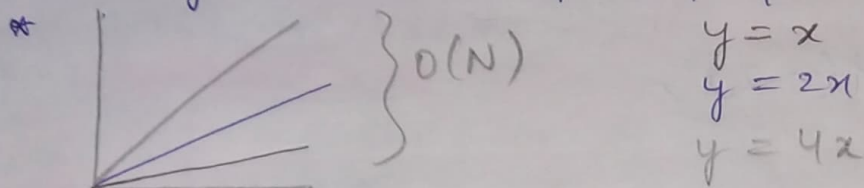
→ Time complexity  $\neq$  Time taken

→ Time taken varies from machine to machine, so we see relationship b/w time grows when input grows

→ Things consider when thinking about complexity:

\* Always look for worst case complexity.

\* Always look at complexity for large /  $\infty$  data.



Even though the value ( $x, 2x, 4x$ ) is different, they are all growing linearly.

We don't care about actual time.

Do we need constants?

\* We only care about relationship of how time grows, when input grows.

So, No we don't need constants. Above are some points, why we ignore all constants.

\* Always ignore less dominating terms.

eg. Let's say, you've complexity

$$O(N^3 + \log(N))$$

put 2). We'll look complexity for large /  $\infty$  data.

$\therefore$  we data amount.  $\rightarrow$  1 million/sec

$$N = 1 \text{ million}$$

$$\Rightarrow O((1M)^3 + \log(1M))$$

$$\Rightarrow (1M)^3 \text{ sec} + \boxed{6 \text{ sec}}$$

very small compared to  $(1M)^3$ . Hence, ignore it.

Example

$$O(3N^3 + 4N^2 + 5N + 6)$$

① ignore constants

$$N^3 + N^2 + N$$

② ignore less dominating terms

$$N^3 \rightarrow O(N^3)$$

$$O(3N^3 + 4N^2 + 5N + 6) \Rightarrow O(N^3)$$

→ we write const.  
as  $O(1)$  because  
constants doesn't  
count and necessary  
so for all we  
write it as  
 $O(1)$

Big-oh Notation <sup>o</sup> ~~imp~~ denoted by  $O$   
in simple language →

let say, we have  $O(N^3) \rightarrow$  upper bound

Here,  $N^3$  refers to size and Big-oh saying, that the  
complexity will not exceed  $N^3$ . ( $N^3$  is the upper bound)  
meaning →

mathematically →

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

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

It is actually some finite  
value.

we have to look for worst case & complexity for large  $\infty$   
data

\* Our algo will not exceed the complexity.

It can be solved in less complexity but in any case  
it will never exceed.

It can be better but never exceed.



~~not imp~~  
Big-omega notation  $\frac{0}{0}$  opposite of Big-oh.  
in words  $\rightarrow$   $\frac{0}{0}$  denoted by  $\Omega$

eg -  $\Omega(N^3) \rightarrow$  lower bound

it means that, it'll take atleast  $N^3$  time complexity.  
It can take above  $N^3$  but not below  $N^3$ .

lower bound  $\rightarrow$  minimum  $N^3$  time complexity required.

mathematically  $\rightarrow$

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} > 0$$

$\Rightarrow$  we always look at the worst case, so we do actually care about Big-oh-notation.

~~not imp~~  
Big-theta Notation  $\frac{0}{0}$  combining big-oh-notation and big-omega notation.  
 $\rightarrow \Theta$

$$0 < \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} < \infty$$

~~not imp~~  
little-oh-Notation  $\frac{0}{0}$  denoted by  $o$ .  
in words  $\rightarrow$  This is also giving upper bound but, this is not strict upper bound.

$\rightarrow$  loose upper bound

$\rightarrow$  mathematically  $\rightarrow$

If  $f$  is strictly slower than  $g$ , then you can say numerator is slower than denominator, it should give us  $0$ /zero.

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

$$f = n^2, g = n^3$$

$$\lim_{n \rightarrow \infty} \frac{n^2}{n^3} \Rightarrow \lim_{n \rightarrow \infty} \frac{1}{n} = 0$$

~~explanation~~  
Big-oh

little-oh

$$f = O(g)$$

$$f = o(g)$$

growth of  $f$  is no faster than  $g$

$f$  is smaller than  $g$

$$\therefore f \leq g$$

$$\therefore f < g$$

$\rightarrow$  strictly lower than  $g$ .

not imp

little omega  $\frac{\omega}{\omega}$  loosely lower bound.

$\frac{\omega}{\omega}$  denoted by  $\omega$

mathematically

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

Big $\Omega$	little $\omega$
$f = \Omega(g)$ means $f \geq g$ lower bound	$f = \omega(g)$ means $f > g$ strictly greater difference.

SPACE COMPLEXITY  $\frac{\omega}{\omega}$

$\frac{\omega}{\omega}$  input space + auxiliary space  
 $\frac{\omega}{\omega}$  Total space taken by algo w.r.t.  
input size.

→ Auxiliary Space — extra space or temporary space used by algorithm.