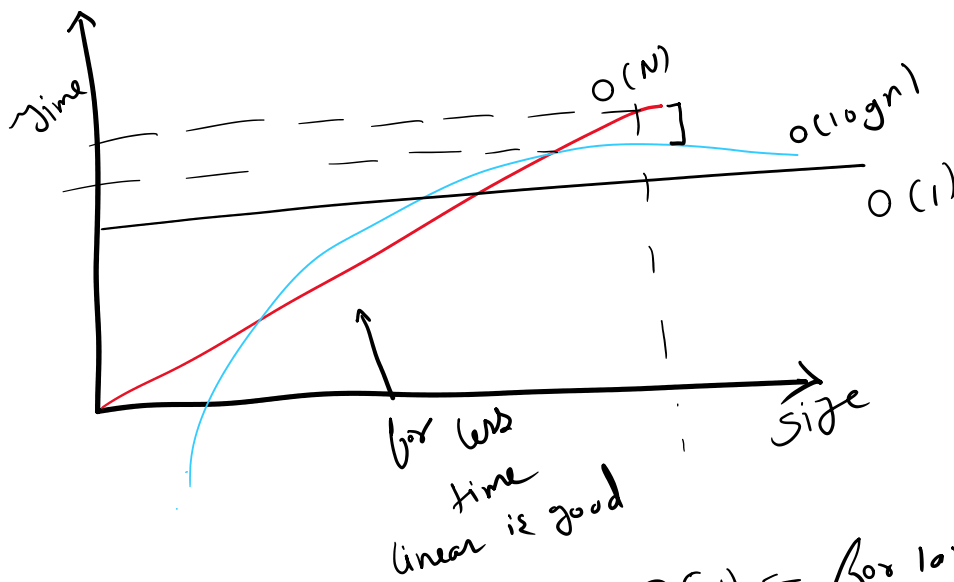
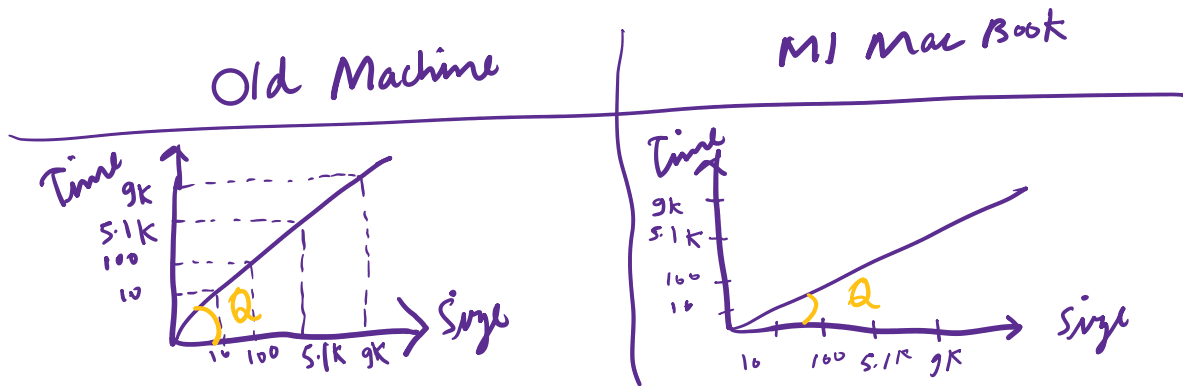


time complexity \neq Time Taken

* Function that gives us the relationship about how the time will grow as the input grows.



$$O(1) < O(\log n) < O(n) \leftarrow \text{for large}$$

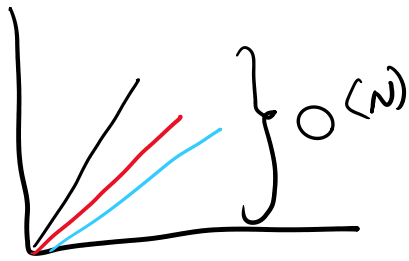
$$O(1) > O(\log n) > O(n) \leftarrow \text{for small}$$

what do we consider when thinking about complexity:-

- ① Always look for worst case complexity.
- ② Always look at complexity for large data.

data.

3



$$\begin{aligned} y &= x \\ y &= 2x \\ y &= 4x \end{aligned}$$

* Even tho values of actual time is different they are all growing linearly.

* we don't care about actual time.
↳ This is why, we ignore all constants.

4 $O(N^3 + \log N)$

* from Point No. ②

$$1 \text{ mil} = (1 \text{ mil})^3 + \log(1 \text{ mil})$$
$$= (1 \text{ mil}^3 + 6 \text{ sec})$$

very small
hence ignore

Always ignore less dominating term.

Big-OH Notation :

* Word definition :-

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

* Maths :-

$$f(N) = O(g(N))$$

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

$$O(n^3) = O\left(\frac{6n^3 + 3n + 5}{f(n)}\right)$$

$$\lim_{n \rightarrow \infty} \frac{6n^3 + 3n + 5}{n^3}$$

$$= \lim_{n \rightarrow \infty} 6 + \frac{3}{n^2} + \frac{5}{n^3} = 6 + \frac{3}{\infty} + \frac{5}{\infty}$$

$$= 6 + 0 + 0$$

$$= 6 < \infty$$

*finite
value*

Big Omega \rightarrow Opposite of Big O

words $\Omega(n^3)$ (lower bound)

maths :-

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

Q1. What if an algo has lower bound & upper bound as (n^2) .

$$= O(n^2) \text{ \& } \Omega(n^2)$$

Theta notation (Combining Both)

$\Theta(N^2) \Rightarrow$ words \rightarrow Both upper bound & lower bound

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

Little O notation \rightarrow

* This is also giving upper bound.
words: lower upper bound

Big O	little O
$f = O(g)$ $f \leq g$	$f = o(g)$ $f < g$ strictly lower

Maths \rightarrow

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

Ex. $f = N^2$ $g = N^3$

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

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

* little Omega \rightarrow

<u>Big Ω</u> $f = \Omega(g)$ $f \geq g$		<u>little ω</u> $f = \omega(g)$ $f > g$
-----------------------------------------------------------------	--	-----------------------------------------------------------------

Maths \rightarrow

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

Space Complexity or Auxiliary Space?

Auxiliary Space is the extra space or temporary space used by an algorithm.

Space Complexity of an algorithm is total space taken by the algorithm with respect to the input size. Space complexity includes both Auxiliary space and space used by input.

For example, if we want to compare standard sorting algorithms on the basis of space, then Auxiliary Space would be a better criteria than Space Complexity. Merge Sort uses $O(n)$ auxiliary space, Insertion sort and Heap Sort use $O(1)$ auxiliary space. Space complexity of all these sorting algorithms is $O(n)$ though.

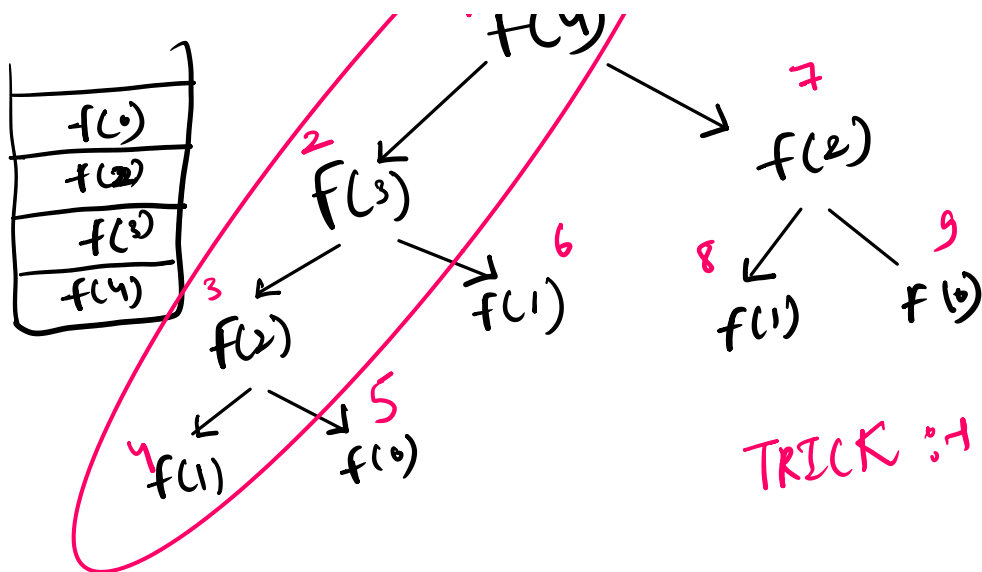
Recursive Algorithm \rightarrow

[]

f(y)

TC $\neq O(N)$

7



TRICK :-

Only call that are interlinked with each other will be in the stack at some time

2 types of recursion :-

① Linear

$$f(n) = f(n-1) + f(n-2)$$

② Divide & Conquer

$$f(n) = f\left(\frac{n}{2}\right) + O(1)$$

* Divide & Conquer Recurrence :-

Form :-

$$T(x) = a_1 T(b_1 x + \epsilon_1(n)) + a_2 T(b_2 x + \epsilon_2(n)) + \dots + a_k T(b_k x + \epsilon_k(n))$$

$$+ \dots + a_k (b_k^x T(x))$$

$$+ g(x)$$

for $x \geq x_0$
 \uparrow
 same unit

$$\sqrt{T(N)} = T\left(\frac{N}{2}\right) + C$$

$$a_1 = 1$$

$$g(x) = C$$

$$b_1 = 1/2$$

$$c_1(x) = 0$$

$$T(N) = 9 T(N/3) + \frac{4}{3} T\left(\frac{5}{6}N\right) + 4N^3$$

$\swarrow \quad \nwarrow$ $\swarrow \quad \nwarrow$ \swarrow
 $a_1 \quad b_1$ $a_2 \quad b_2$ $g(N)$

$$T(N) = 2 T\left(\frac{N}{2}\right) + \frac{(N-1)}{g(N)}$$

$\swarrow \quad \nwarrow$
 $a_1 \quad b_2$

when you get an answer from this + what you are doing with that answer take how much time.

* How to actually solve to get complexity :)

① Plug & chug

$$f(N) = f(N/2) + C$$

$$f(N) = f(N/2) + C$$

② Master's Theorem

③ Akra Bazzi (1996)

Akra Bazzi :-

$$T(x) = O\left(x^p + x^p \int_1^x \frac{g(u)}{u^{p+1}} du\right)$$

What is p ?

$$a_1 b_1^p + a_2 b_2^p + \dots = 1$$

$$\sum_{i=1}^K a_i b_i^p = 1$$

Ex: $T(N) = 2T\left(\frac{N}{2}\right) + (N-1)$

$$a_1 = 2$$

$$g(x) = n-1$$

$$a_1 = 2$$

$$b_1 = 1/2$$

0

$$2 \times \left(\frac{1}{2}\right)^p = 1$$

$$\boxed{p=1}$$

Put p in formula :-

$$T(n) = O\left(x' + x' \int_1^x \frac{u-1}{u^2} du\right)$$

$$= O\left(x + x \int_1^x \left(\frac{1}{u} - \frac{1}{u^2}\right) du\right)$$

$$= O\left(x + x \left[\int_1^x \frac{du}{u} - \int_1^x \frac{du}{u^2} \right]\right)$$

$$= O\left(x + x \left[\log u + \frac{1}{u} \right]\right)$$

$$= O\left(x + x \left(\log x + \frac{1}{x} - 1 \right)\right)$$

$$= O(x + x \log x + 1 - x)$$

$$= O(x \log x + 1)$$

// Time Complexity

$$= O(n \log n) \quad // \text{Time complexity}$$