

# Asymptotic Notations

Asymptotic Notations give us an idea about how good a given algorithm is compared to some other algorithm.  
let us see the mathematical definition of "Order of" now

primarily there are three types of widely used asymptotic notations.

- (1) Big Oh notation ( $O$ )
- (2) Big Omega notation ( $\Omega$ )
- (3) Big theta notation ( $\Theta$ )  $\rightarrow$  widely used One!

## Big oh notation

Big oh notation is used to describe asymptotic upper bound.  
Mathematically, if  $f(x)$  describes running time of an algorithm;  $f(x)$  is  $O(g(x))$  iff there exist positive constants  $c$  and  $n_0$  such that

$$0 \leq f(x) \leq cg(x) \text{ for all } n \geq n_0$$

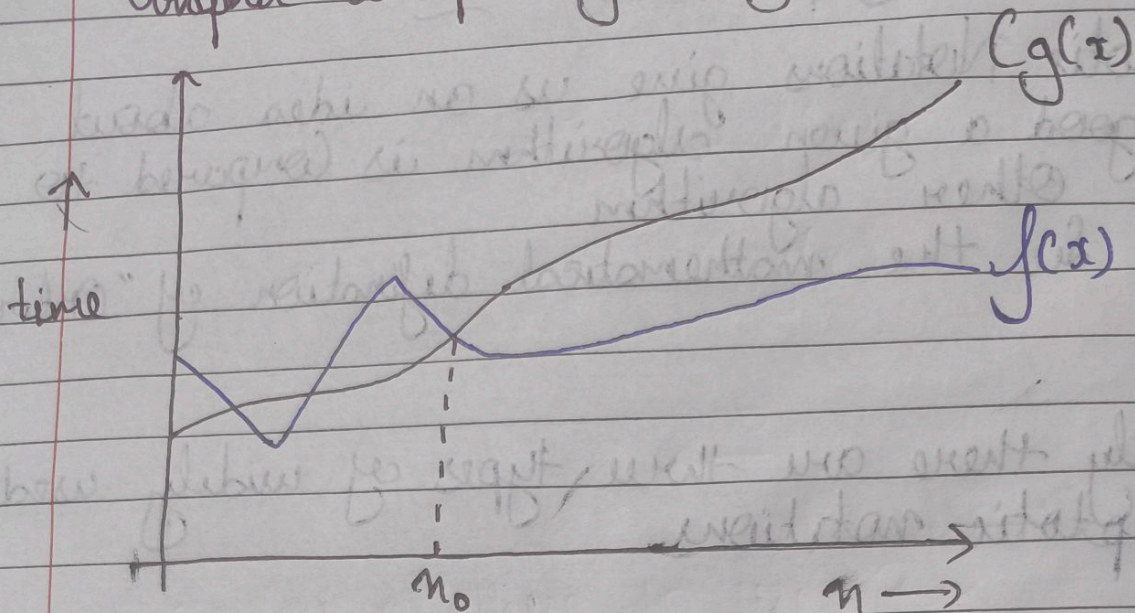
$\Downarrow$

used to give upper bound on a func<sup>n</sup>

if a function is  $O(n)$ , it is automatically  $O(n^2)$  as well!



## Graphic Example for Big Oh ( $O$ )



## Big Omega notation

Just like  $O$  notation provides an asymptotic upper bound,  $\Omega$  notation provides asymptotic lower bound. Let  $f(n)$  define running time of an algorithm;

$f(n)$  is said to be  $\Omega g(n)$  if there exists positive constants  $c$  and  $n_0$  such that

$$0 \leq c g(n) \leq f(n) \text{ for all } n \geq n_0$$

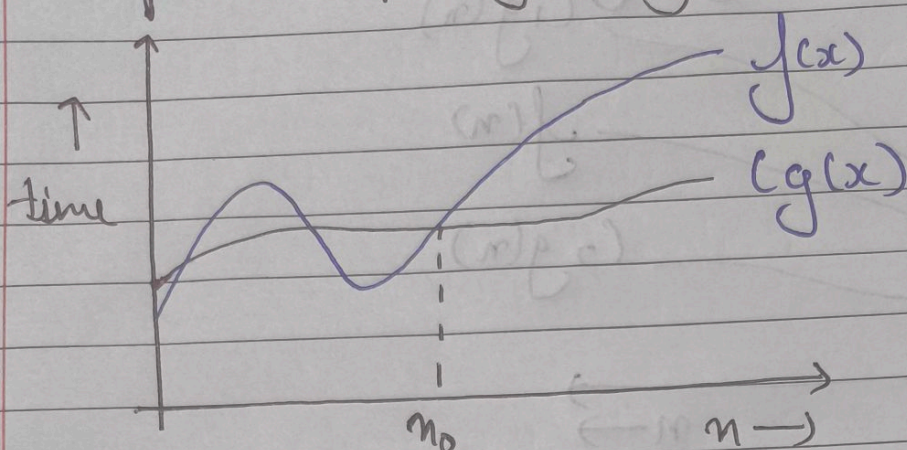
$\Downarrow$

used to give lower bound on a function

if a function is  $O(n^2)$  it is automatically  $\Omega(n)$  as well



## Graphic Example for Big Omega ( $\Omega$ )



## Big theta notation

let  $f(x)$  define running time

of an algorithm

$f(x)$  is said to be  $\Theta(g(n))$  iff  $f(n)$  is  $O(g(n))$   
and  $f(n)$  is  $\Omega(g(n))$

Mathematically,

$$0 \leq f(n) \leq c_1 g(n) \quad \forall n \geq n_0$$

Sufficiently large

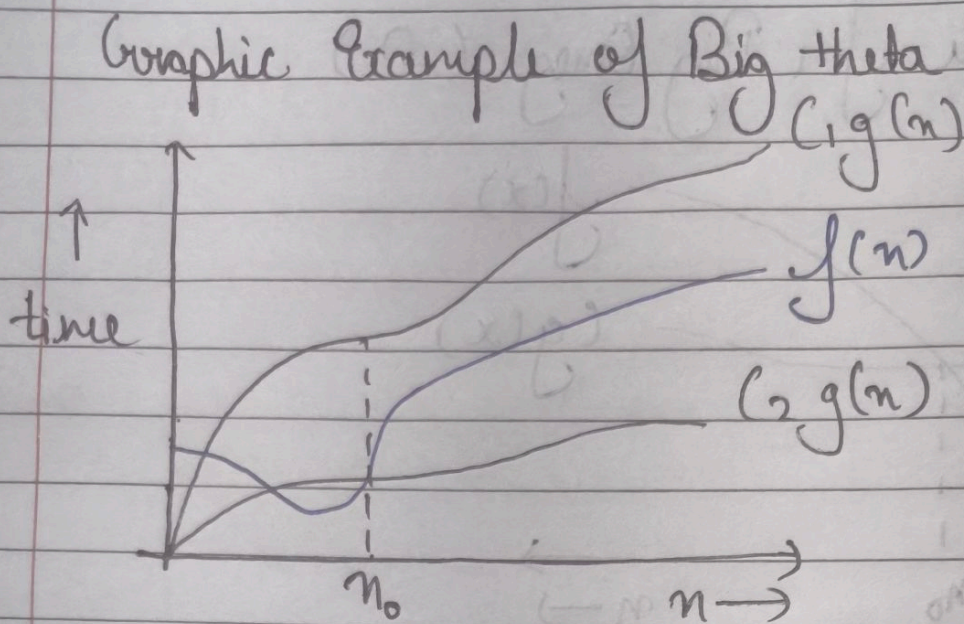
$$0 \leq c_2 g(n) \leq f(n) \quad \forall n \geq n_0$$

Merging both the eq<sup>n</sup>, we get

$$0 \leq c_2 g(n) \leq f(n) \leq c_1 g(n) \quad \forall n \geq n_0$$

The eq<sup>n</sup> simply means there exists positive constants  $c_1$  and  $c_2$  such that  $f(n)$  is sandwich between  $c_2 g(n)$  and  $c_1 g(n)$





Which One of these to use?

Since Big theta gives a better picture of runtime for a given algorithm, most of the interviewers expect you to provide an answer in terms of Big theta when they say "Order of".

Increasing Order of Common runtimes

$$1 < \log n < n < n \log n < n^2 < n^3 < 2^n < n^n$$

Better



worse

Common runtime

from better to worse

