

# Asymptotic Analysis

## Big-O: Definition

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of Algorithms I



# Big-Oh: English Definition

Let  $T(n)$  = function on  $n = 1, 2, 3, \dots$

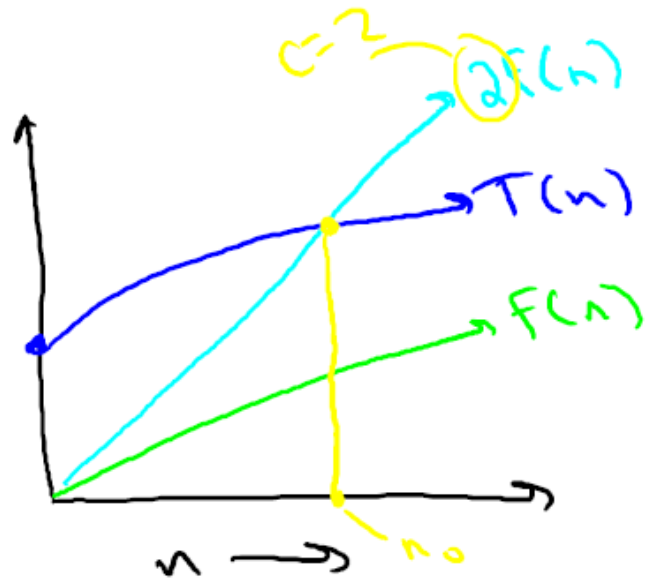
[usually, the worst-case running time of an algorithm]

Q : When is  $T(n) = O(f(n))$  ?

A : if eventually (for all sufficiently large  $n$ ),  $T(n)$  is bounded above by a constant multiple of  $f(n)$



# Big-Oh: Formal Definition



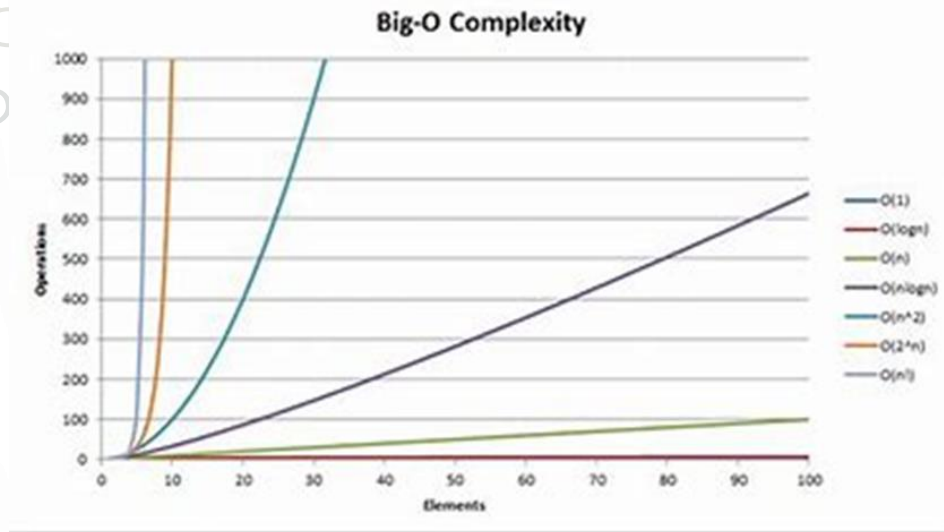
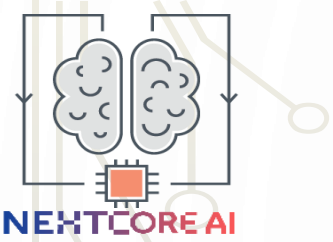
Picture  $T(n) = O(f(n))$

Formal Definition :  $T(n) = O(f(n))$  if and only if there exist constants  $c, n_0 > 0$  such that

$$T(n) \leq c \cdot f(n)$$

For all  $n \geq n_0$

Warning :  $c, n_0$  cannot depend on  $n$



# Asymptotic Analysis

## Big-Oh: Basic Examples

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## Example #1

Claim : if  $T(n) = a_k n^k + \dots + a_1 n + a_0$  then

$$T(n) = O(n^k)$$

Proof : Choose  $n_0 = 1$  and  $c = |a_k| + |a_{k-1}| + \dots + |a_1| + |a_0|$

Need to show that  $\forall n \geq 1, T(n) \leq c \cdot n^k$

We have, for every  $n \geq 1$ ,

$$\begin{aligned} T(n) &\leq |a_k|n^k + \dots + |a_1|n + |a_0| \\ &\leq |a_k|n^k + \dots + |a_1|n^k + |a_0|n^k \\ &= c \cdot n^k \end{aligned}$$



## Example #2

Claim : for every  $k \geq 1$ ,  $n^k$  is not  $O(n^{k-1})$

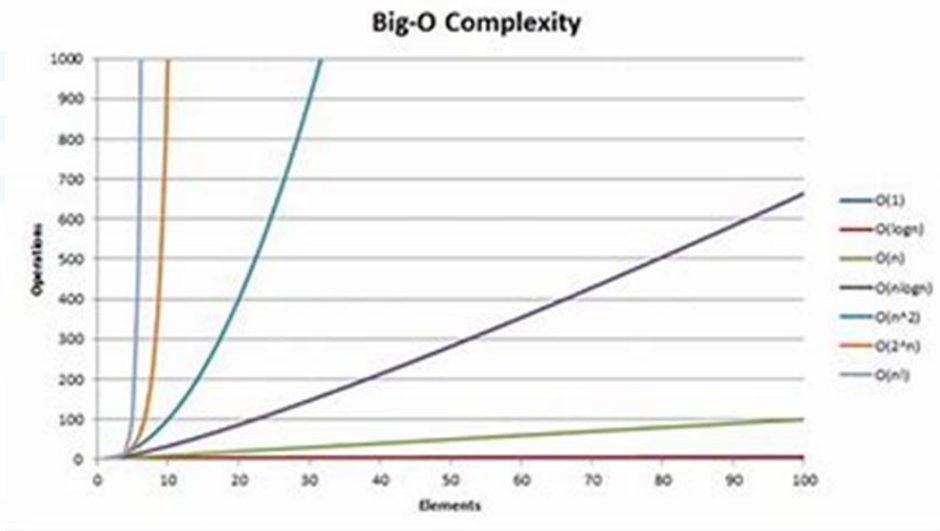
Proof : by contradiction. Suppose  $n^k = O(n^{k-1})$   
Then there exist constants  $c, n_0$  such that

$$n^k \leq c \cdot n^{k-1} \quad \forall n \geq n_0$$

But then [cancelling  $n^{k-1}$  from both sides]:

$$n \leq c \quad \forall n \geq n_0$$

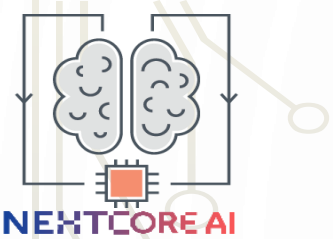
Which is clearly False [contradiction]



# Asymptotic Analysis

## Big Oh: Relatives (Omega & Theta)

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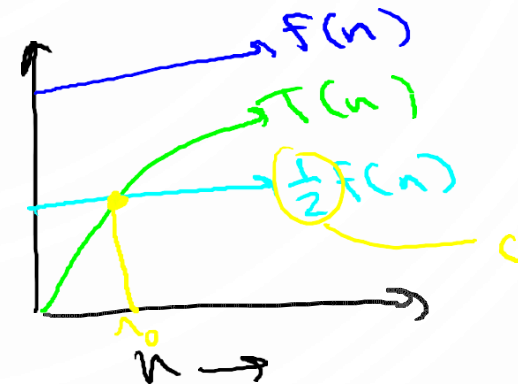
# OMEGA NOTATION

Definition :  $T(n) = \Omega(f(n))$

If and only if there exist constants  $c, n_0$  such that

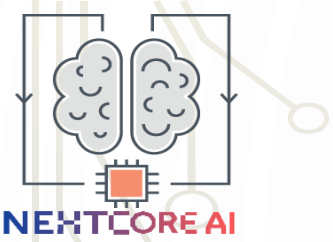
$$T(n) \geq c \cdot f(n) \quad \forall n \geq n_0.$$

Picture



$$T(n) = \Omega(f(n))$$





# Theta Notation

Definition :  $T(n) = \theta(f(n))$  if and only if  
 $T(n) = O(f(n))$  and  $T(n) = \Omega(f(n))$

Equivalent : there exist constants  $c_1, c_2, n_0$  such that  
$$c_1 f(n) \leq T(n) \leq c_2 f(n)$$
$$\forall n \geq n_0$$



Let  $T(n) = \frac{1}{2}n^2 + 3n$ . Which of the following statements are true? (Check all that apply.)

$T(n) = O(n)$ .

$T(n) = \Omega(n)$ .  $[n_0 = 1, c = \frac{1}{2}]$

$T(n) = \Theta(n^2)$ .  $[n_0 = 1, c_1 = 1/2, c_2 = 4]$

$T(n) = O(n^3)$ .  $[n_0 = 1, c = 4]$

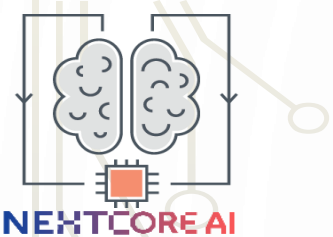


## Little-Oh Notation

Definition :  $T(n) = o(f(n))$  if and only if for all constants  $c > 0$ , there exists a constant  $n_0$  such that

$$T(n) \leq c \cdot f(n) \quad \forall n \geq n_0$$

Exercise :  $\forall k \geq 1, n^{k-1} = o(n^k)$



## Where Does Notation Come From?

“On the basis of the issues discussed here, I propose that members of SIGACT, and editors of computer science and mathematics journals, adopt the  $O$ ,  $\Omega$ , and  $\Theta$  notations as defined above, unless a better alternative can be found reasonably soon”.

*-D. E. Knuth, “Big Omicron and Big Omega and Big Theta”, SIGACT News, 1976. Reprinted in “Selected Papers on Analysis of Algorithms.”*



# Asymptotic Analysis

## Additional Examples

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## EXAMPLE #1

Claim :  $2^{n+10} = O(2^n)$

Proof : need to pick constants  $c, n_0$  such that

$$(*) \quad 2^{n+10} \leq c \cdot 2^n \quad n \geq n_0$$

Note :  $2^{n+10} = 2^{10} \times 2^n = (1024) \times 2^n$

So if we choose  $c = 1024, n_0 = 1$  then  $(*)$  holds.

Q.E.D



## EXAMPLE #2

Claim :  $2^{10n} \neq O(2^n)$

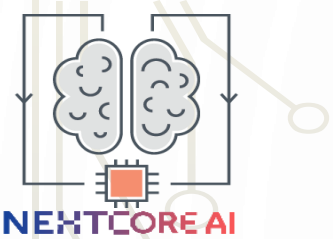
Proof : by contradiction. If  $2^{10n} = O(2^n)$  then there exist constants  $c, n_0 > 0$  such that

$$2^{10n} \leq c \cdot 2^n \quad n \geq n_0$$

But then [cancelling  $2^n$ ]

Which is certainly false.

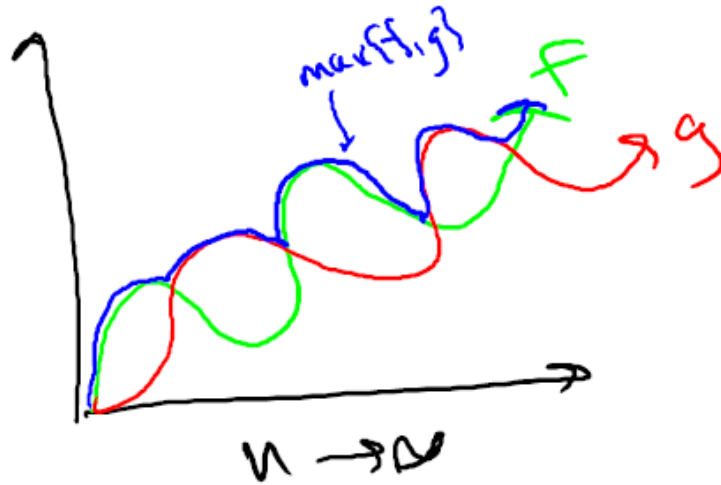
Q.E.D



## Example #3

Claim : for every pair of (positive) functions  $f(n)$ ,  $g(n)$ ,

$$\max\{f, g\} = \theta(f(n) + g(n))$$







## Example #3 (continued)

Proof:  $\max\{f, g\} = \theta(f(n) + g(n))$

For every  $n$ , we have

$$\max\{f(n), g(n)\} \leq f(n) + g(n)$$

And

$$2 * \max\{f(n), g(n)\} \geq f(n) + g(n)$$

Thus  $\frac{1}{2} * (f(n) + g(n)) \leq \max\{f(n), g(n)\} \leq f(n) + g(n) \quad \forall n \geq 1$   
 $\Rightarrow \max\{f, g\} = \theta(f(n) + g(n))$  [where  $n_0 = 1, c_1 = 1/2, c_2 = 1$ ]

**Q.E.D**