

# Asymptotic Analysis

- way to compare speed of two possible solutions to the same problem.
- Basic idea is to think of the rate of growth of the running time of an algorithm.
  - How much slower does the algorithm gets if we double the input size?
  - we try to ~~not~~ focus on big inputs.
- for asymptotic analysis, we will first define what is the rate of growth of a function.  
(Big Oh and related defs)
- In other words, asymptotic analysis is a way to describe the efficiency of algorithms by examining how their performance grows as the input becomes very large. (Trying to get a big picture, ignoring small details)

Key concept of asymptotic analysis will be the definition of growth rates, of functions  $f$  and  $g$ , and how these growth rates compare.

### Big O

Def: Let  $f, g : \mathbb{R}^{\geq 0} \rightarrow \mathbb{R}^{\geq 0}$ , we say that  $f$  grows no faster than  $g$  if there exist constant  $c > 0$  and  $n_0 \geq 0$  s.t

$$\forall n \geq n_0 : f(n) \leq c \cdot g(n)$$

In this case, we write

$$f(n) \text{ is } O(g(n)) \text{ or } f(n) = O(g(n))$$

The intuition behind  $f(n) = O(g(n))$  is that, for all large enough  $n$ , we have  $f(n) \leq c \cdot g(n)$

Example:  $f(n) = 3n^2 + 2$  is  $O(n^2)$

$$f(n) \leq c \cdot n^2$$

Def Big O says

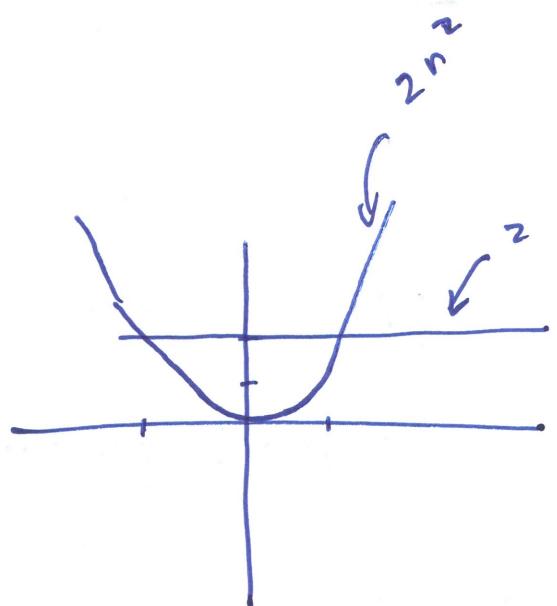
If  $f(n) = O(g(n))$ , Then

$$[\exists c > 0, n_0 \geq 0 : \forall n \geq n_0 : f(n) \leq c \cdot g(n)]$$

WTS  $3n^2 + 2 \leq c \cdot n^2$

$$\underline{\forall n \geq 1 : 2 \leq 2n^2}$$

$$\forall n \geq 1 : 3n^2 + 2 \leq 3n^2 + 2n^2$$



$$\forall n \geq 1 : \underbrace{3n^2 + 2}_{f(n)} \leq \underbrace{5n^2}_{g(n)}$$

Pick  $c = 5, n_0 = 1$ , then

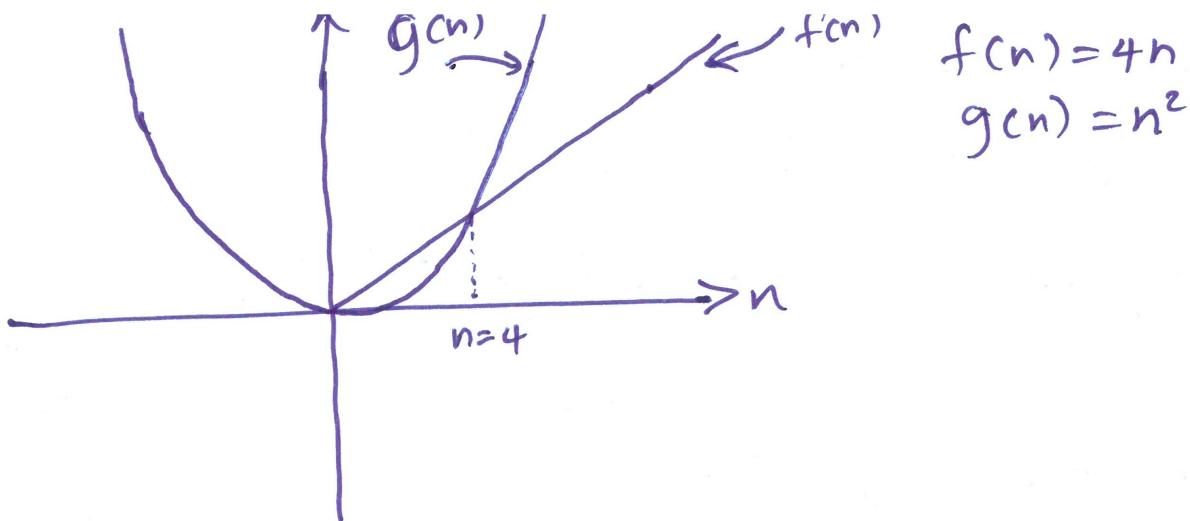
$$\underline{\forall n \geq 1 : f(n) \leq c \cdot g(n)}$$

Ex:  $f(n) = 4n$

(claim):  $f(n) = O(n^2)$   $\left( f(n) = O(n) \right)$

We must show a  $c > 0, n_0 \geq 0$  s.t

$$\forall n \geq n_0 : f(n) \leq c \cdot g(n)$$



Pick  $c=1, n_0 = 4$

$$\forall n \geq 4 : 4n \leq 1 \cdot n^2$$

Q: Is  $4n = O(4n^2)$  yes

$$\text{is } 3n^2 + 2 = O(\frac{1}{2}n^2)$$

Pick  $c=100$

$$\forall n \geq 50 \quad 3n^2 + 2 \leq 100 \cdot \frac{1}{2} \cdot n^2$$

Ex: ~~Is~~ Is  $n^3 = O(n^2)$  (no)

can we find  $c > 0, n_0 \geq 0$  s.t  $\left. \begin{array}{l} \\ \forall n \geq n_0 : n^3 \leq c \cdot n^2 \end{array} \right\}$  we cannot

$n^3$  is not  $O(n^2)$   $\leftarrow$  How to prove this?

If  $n^3 = O(n^2)$ , then

$\exists c > 0, n_0 \geq 0 : \forall n \geq n_0 : f(n) \leq c \cdot g(n)$

since this is incorrect, the negation of this ~~is~~ claim should be true.

$$\neg [\exists c > 0, n_0 \geq 0 : \forall n \geq n_0 : f(n) \leq c \cdot g(n)]$$

$$\neg [\exists x \in S : P(x)] \equiv [\forall x \in S : \neg P(x)]^* \quad \left\{ \begin{array}{l} \text{Predicate} \\ \text{logic} \\ \text{theorems.} \end{array} \right.$$

$$\neg [\forall x \in S : P(x)] \equiv [\exists x \in S : \neg P(x)]^*$$

$$\rightarrow [\forall c > 0, n_0 \geq 0 : \neg [\forall n \geq n_0 : f(n) \leq c \cdot g(n)]]$$

$$[\forall c > 0, n_0 \geq 0 : [\exists n \geq n_0 : \neg (f(n) \leq c \cdot g(n))]]$$

$$[\forall c > 0, n_0 \geq 0 : [\exists n \geq n_0 : f(n) > c \cdot g(n)]]$$

$\forall c > 0, n_0 \geq 0$ : I need to show the existence of  $n$  s.t.

$$n^3 > c \cdot n^2$$

$$f(n) > c \cdot g(n)$$

What if we pick  $n = c+1$

$$(c+1)^3 > c \cdot (c+1)^2$$

$$n = \max(n_0, c+1)$$

When  $n = \max(n_0, c+1)$ ,  $\forall c > 0, n_0 \geq 0, f(n) > c \cdot g(n)$

$\therefore f(n)$  is not  $O(n^2)$

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Note that

$$f(n) = 4n^3 + 4n + 5$$

$$f(n) = \cancel{O(n^3)}$$

$$f(n) = O(n^4)$$

$$f(n) = O(n^5)$$

However, the tightest  $O(\cdot)$  function for  $f(n)$  is  $O(n^3)$

Note: When we write the Big O function for  $f(n)$ , we ~~must~~ want to use the tightest bound, as it is more useful.

(common distinct functions.

Name	Example fcn)	generic fcn)	tightest O(.)
constant	$f(n) = 10$	$f(n) = c, c \in \mathbb{R}^{>0}$	$O(1)$
log	$f(n) = 2 \log_{10}(n)$		$O(\log n)$
linear	$f(n) = 10 \cdot n$		$O(n)$
$n \log n$	$f(n) = 8n \log_{12}(n)$		$O(n \log n)$
quadratic	$f(n) = 12n^2$		$O(n^2)$
cubic	$f(n) = 20n^3$		$O(n^3)$
deg-k polynomial		$f(n) = a_1 n^k + a_2 n^{k-1} + \dots + a_{n-1}$	$O(n^k)$
exponential	$f(n) = 3 \cdot 2^n$	$2^n$	$O(2^n)$
	$f(n) = 5 \cdot 3^n$	$3^n$ ⋮	$O(3^n)$

functions later on this list  $\neq O(\text{functions earlier on this list})$

$$2^n \neq O(n^{100})$$

$$3^n \neq O(n)$$

- However,

functions earlier =  $O(\text{functions later on the list})$

$$n = O(2^n)$$

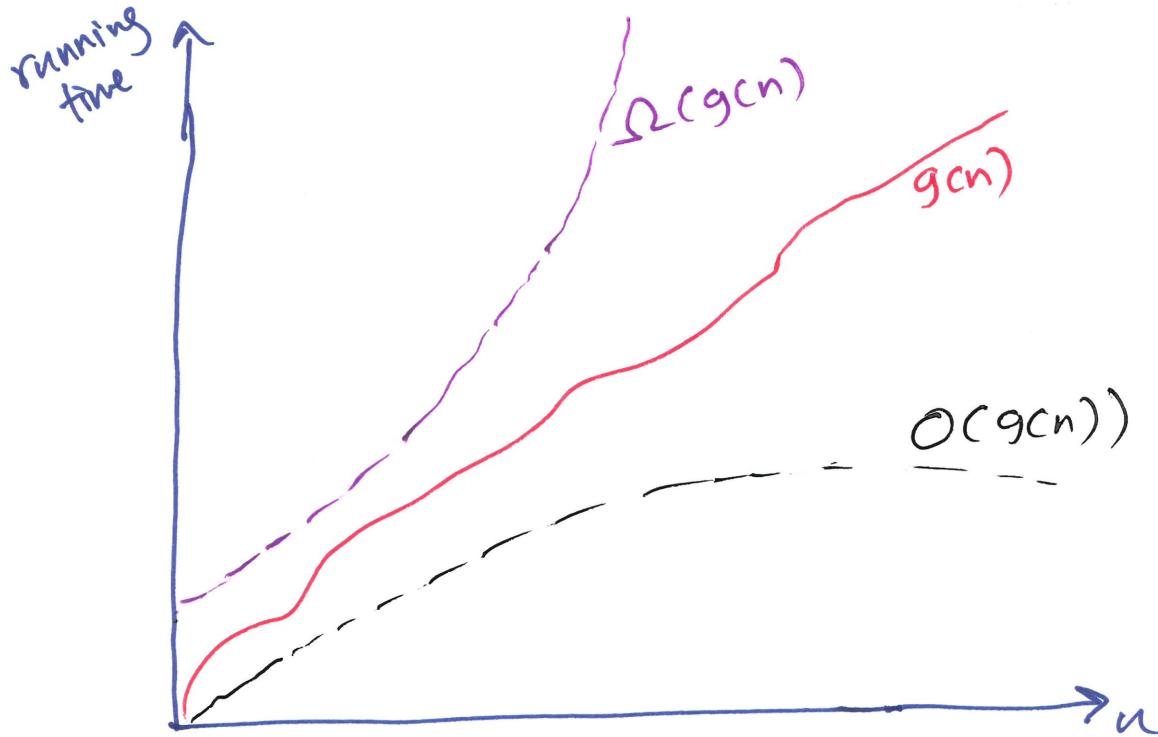
$$n^{100} = O(2^n)$$

$$n^{100} = O(3^n)$$

## Other asymptotic notations

Big-Omega ( $\Omega$ ) -  $f$  grows no slower than  $g$ .

$f$  is  $\Omega(g)$  if  $\exists d > 0, n_0 \geq 0$  s.t  $\forall n \geq n_0: f(n) \geq d \cdot g(n)$



Big-Theta ( $\Theta$ )

$f$  is  $\Theta(g)$  if  $f = O(g)$  and  $f = \Omega(g)$

In computer science this is what we are interested in.

## Properties of Big O

Lemma 6.2 Asymptotic equivalence of max and sum

We have,

$$f(n) = O(g(n) + h(n)) \iff f(n) = O(\max(g(n), h(n)))$$