

CS 2009  
Design and Analysis of Algorithms

Lecture 2 and 3:  
Growth Of Function

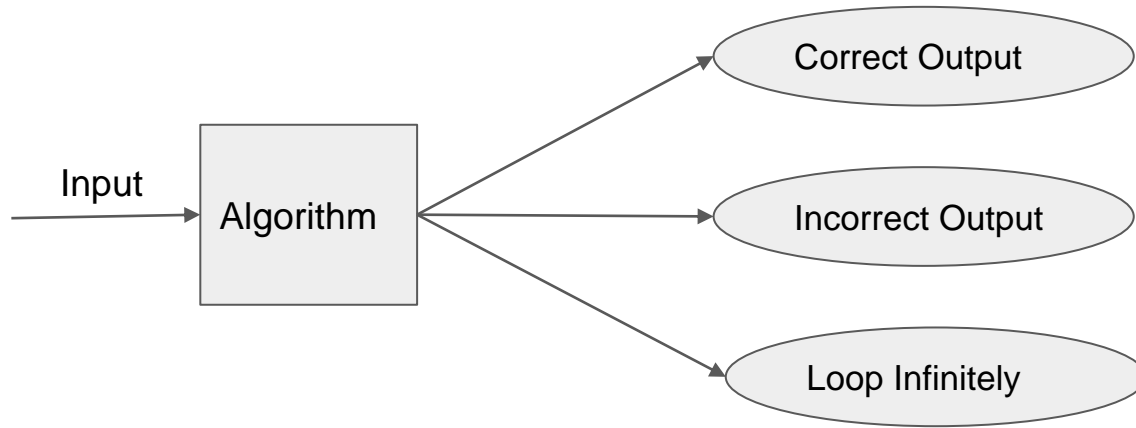
*September 8-9, 2021*

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# Correct Algorithm

An algorithm is said to be **correct** if, for every input instance, it halts with the correct output.



# MULTIPLICATION PROBLEM

**How efficient is this algorithm?**

(How many single-digit operations are required?)

**Algorithm description (informal\*):**

compute partial products (using multiplication & “carries” for digit overflows), and add all (properly shifted) partial products together

$$\begin{array}{r} 2143 \\ \times 9112 \\ \hline 4286 \\ 21430 \\ 214300 \\ 19187000 \\ \hline 19427016 \end{array}$$

# MULTIPLICATION PROBLEM

**How efficient is this algorithm?**

(How many single-digit operations are required?)

**n partial products:  $\sim 2n^2$  ops** (at most n multiplications & n additions per partial product)

**adding n partial products:  $\sim 2n^2$  ops**  
(a bunch of additions & “carries”)

**$\sim 4n^2$  operations in the worst case**

$$\begin{array}{r} 2143 \\ \times 9112 \\ \hline 4286 \\ 21430 \\ 214300 \\ 19187000 \\ \hline 19427016 \end{array}$$

# MULTIPLICATION PROBLEM

*n digits*

$$\begin{array}{r} 12345678998765432101 \\ \times 98765432112345678901 \\ \hline \end{array}$$

**How efficient is this  
algorithm?**

(How many single-digit operations  
are required?)

# Which Running Time Is Better?

Computer A (**Faster**): Run algorithm of  **$2n^2$**  complexity. Run 10 billions instruction per second.

Computer B (**Slower**): Run Algorithm  **$50 n \log n$**  complexity. Run 10 millions instruction per second.

Input length  **$n = 10$**  millions

$$\frac{2 \cdot (10^7)^2 \text{ Instructions}}{10^{10} \text{ Instructions/second}} = 20,000 \text{ seconds } (> 5.5 \text{ hours})$$

$$\frac{50 \cdot 10^7 \log 10^7 \text{ Instructions}}{10^7 \text{ Instructions/second}} = 1163 \text{ seconds } (< 20 \text{ minutes})$$

# Growth Rate Ranking of Function?

## Comparison of running times

For each function  $f(n)$  and time  $t$  in the following table, determine the largest size  $n$  of a problem that can be solved in time  $t$ , assuming that the algorithm to solve the problem takes  $f(n)$  microseconds.

	1 second	1 minute	1 hour	1 day	1 month	1 year	1 century
$\lg n$							
$\sqrt{n}$							
$n$							
$n \lg n$							
$n^2$							
$n^3$							
$2^n$							
$n!$							

# Efficiency of Algorithm

INTRODUCING...

## ASYMPTOTIC ANALYSIS

### Some guiding principles:

- we want some measure of runtime that's independent of hardware, programming language, memory layout, etc.
  - We want to reason about high-level algorithmic approaches rather than lower-level details
- we care about how the running time/number of operations *scales* with the size of the input (i.e. the runtime's *rate of growth*),
- Not concerned with small values of  $n$ , Concerned with VERY LARGE values of  $n$ .
- Asymptotic –refers to study of function  $f$  as  $n$  approaches infinity



# ASYMPTOTIC ANALYSIS (High Level Idea)

*We'll express the asymptotic runtime of an algorithm using*

## BIG-O NOTATION

- We would say Multiplication **“runs in time  $O(n^2)$ ”**
  - Informally, this means that the runtime “scales like”  $n^2$

*THE POINT OF ASYMPTOTIC NOTATION*

**suppress constant factors and lower-order terms**

*too system dependent*

*irrelevant for large inputs*

# ASYMPTOTIC ANALYSIS (High Level Idea)

## BIG-O NOTATION

*THE POINT OF ASYMPTOTIC NOTATION*

**suppress constant factors and lower-order terms**

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**Example  $f(n) = 2n^2 + 4n + 1$**

**$f(n) = O(n^2)$** : 2 is constant,  $n^2$  is the dominant term, and the term  $4n + 1$  becomes insignificant as  $n$  grows larger.

# ASYMPTOTIC ANALYSIS (High Level Idea)

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**suppress constant factors and lower-order terms**

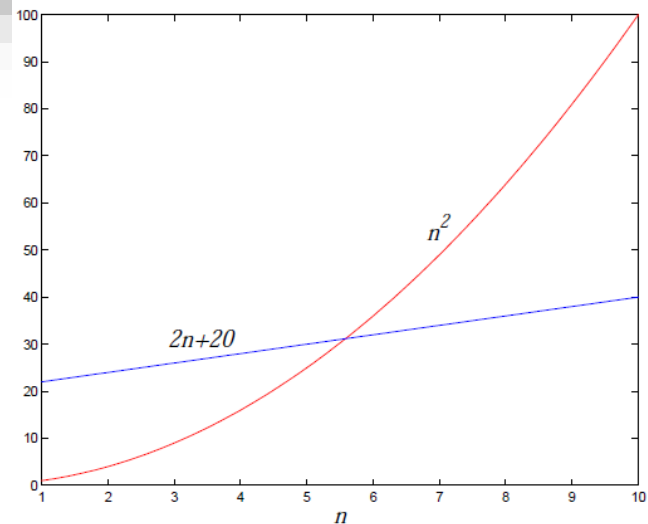
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$$f_1(n) = n^2$$

$$f_2(n) = 2n + 20$$

**Which is better?**



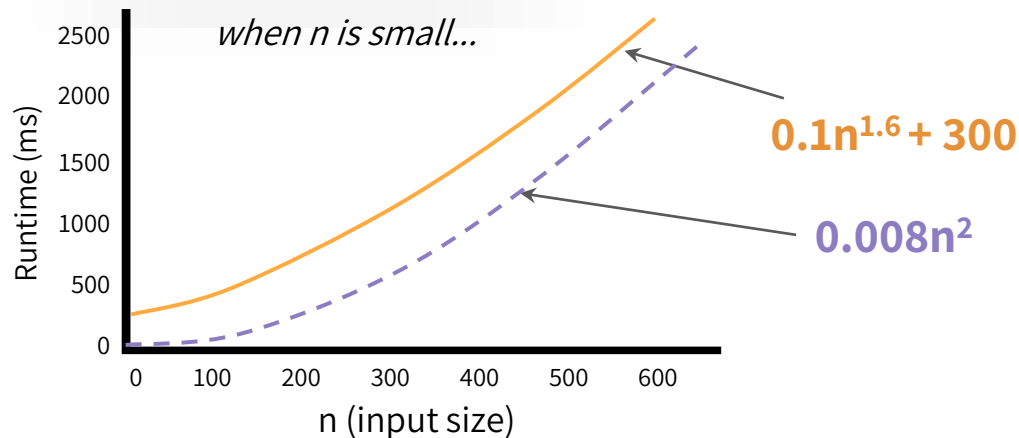
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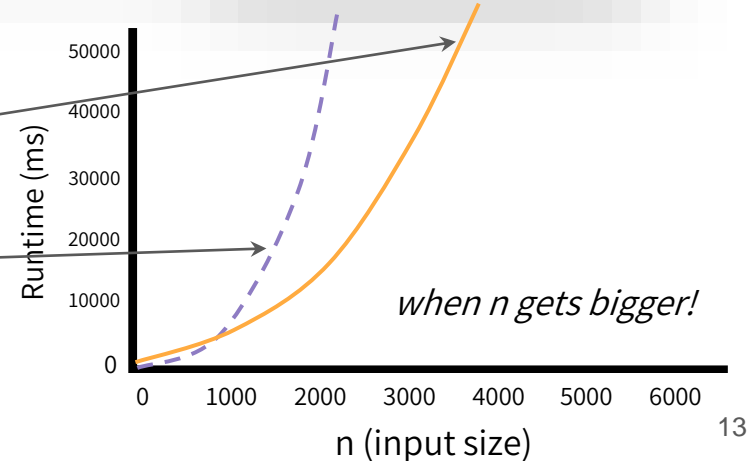
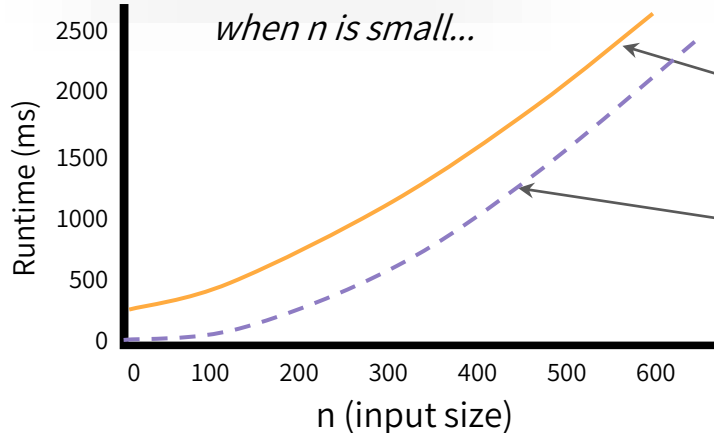
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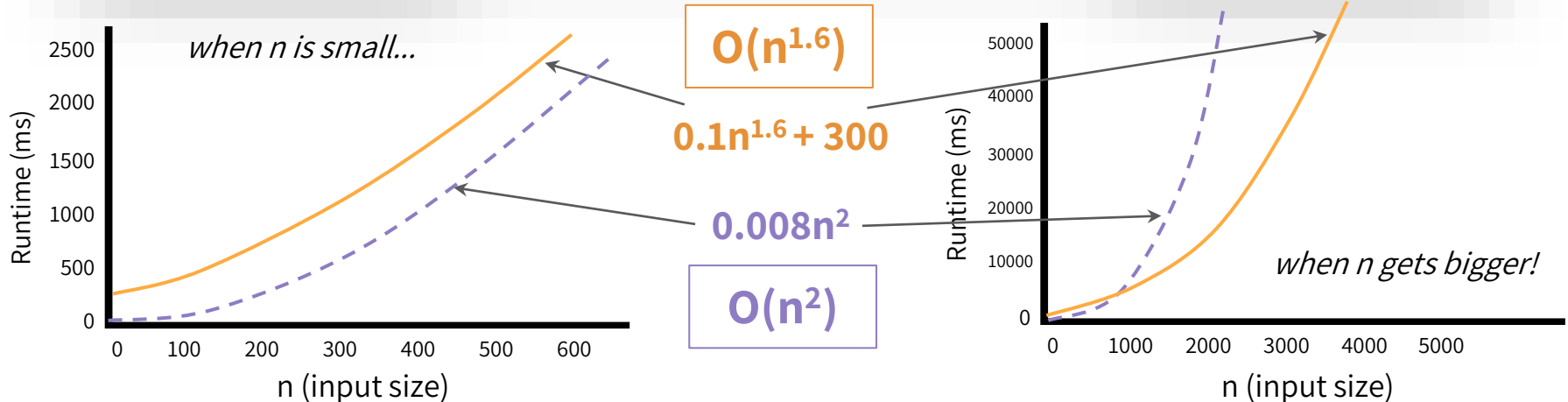
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# ASYMPTOTIC ANALYSIS (High Level Idea)

- To compare algorithm runtimes in this class, we compare their Big-O runtimes
  - Ex: a runtime of  $O(n^2)$  is considered “better” than a runtime of  $O(n^3)$
  - Ex: a runtime of  $O(n^{1.6})$  is considered “better” than a runtime of  $O(n^2)$
  - Ex: a runtime of  $O(1/n)$  is considered “better” than  $O(1)$  ?

# Which Running Time Is Better?

Is  $1000000n$  operations better than  $4n^2$ ?

Is  $0.000001n^3$  operations better than  $4n^2$ ?

Is  $3n^2$  operations better than  $4n^2$ ?

- **The answers for the first two depend on what value  $n$  is...**
  - $1000000n < 4n^2$  only when  $n$  exceeds a certain value (in this case, 250000)
- **These constant multipliers are too environment-dependent...**
  - An operation could be faster/slower depending on the machine, so  $3n^2$  ops on a slow machine might not be “better” than  $4n^2$  ops on a faster machine



# Growth of Function

Skip in Class

$n$	$\log_2 n$	$n \log_2 n$	$n^2$	$2^n$
1	0	0	1	1
2	1	2	4	8
4	2	8	16	16
8	3	24	64	256
16	4	64	256	65536
32	5	160	1024	4294967296

# Growth Rate Ranking of Function?

$$f(n) = n^n$$

$$f(n) = 2^n$$

$$f(n) = n^3$$

$$f(n) = n^2$$

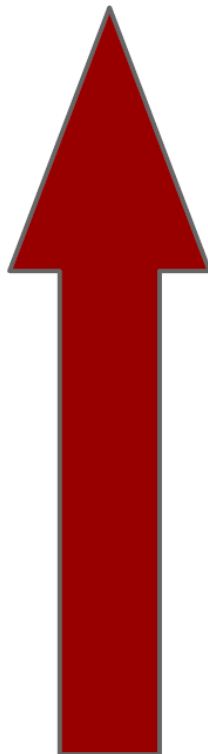
$$f(n) = n \log n$$

$$f(n) = n$$

$$f(n) = \sqrt{n}$$

$$f(n) = \log n$$

$$f(n) = 1$$



**grow fast**

**grow slowly**