



# LECTURE 3: ALGORITHM ANALYSIS

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# QUESTIONS BEFORE WE GET STARTED?



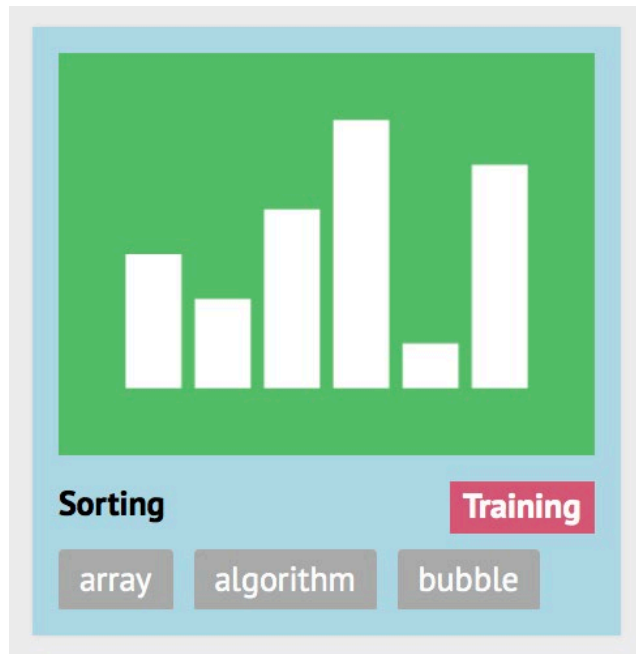
# LEARNING OUTCOMES

By the end of this session, you should be able to:

- **Determine the computational complexity** of an algorithm under the standard sequential computation model
- **Use Big-Oh Notation** to describe algorithm performance



# DID YOU DO YOUR HOMEWORK?



Did you revise the sorting material on Visualgo?

- A. Of course!
- B. Of course ... not!
- C. well, kind of half way...
- D. Ummm.. Visualgo?

# PROBLEM: CUSTOMER LOYALTY REWARDS

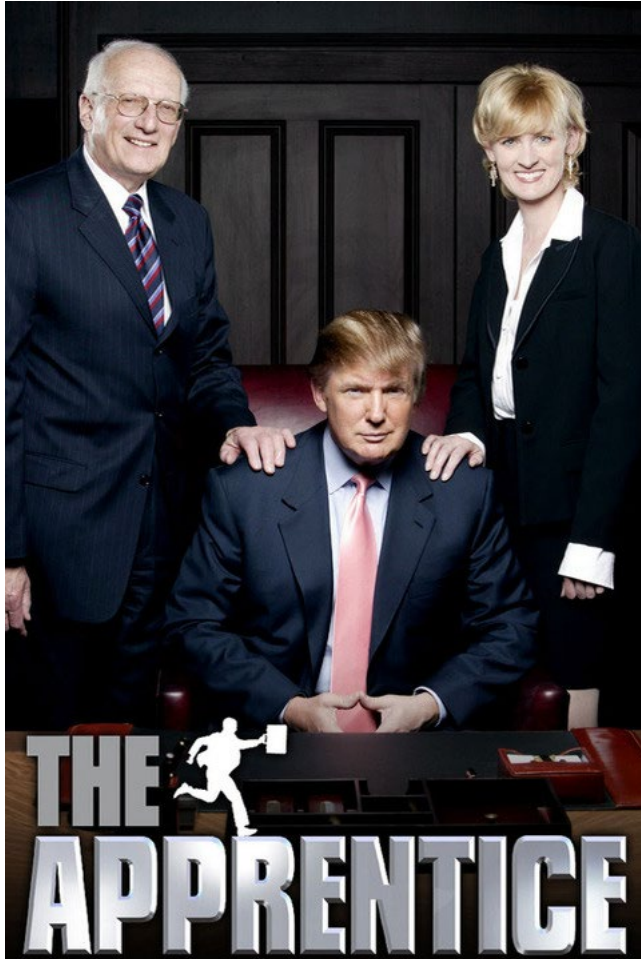


Get a list of customers ordered by their purchasing spend. Reward the  $n$  who spent the most.



BOSS HAS AN SOLUTION:

**BogoSort**



```
while items is not sorted  
    permute(items)
```

# NARUTO'S SOLUTION



```
mark first element as sorted
for each unsorted element X
    'extract' the element X
    for j = lastSortedIndex down to 0
        if current element j > X
            move sorted element to the right by 1
    break loop and insert X here
```



# NARUTO'S SOLUTION

What kind of sort is this?

- A. Bubble Sort
- B. Insertion Sort
- C. Merge Sort
- D. I obviously didn't do my homework

```
mark first element as sorted
for each unsorted element X
    'extract' the element X
    for j = lastSortedIndex down to 0
        if current element j > X
            move sorted element to the right by 1
    break loop and insert X here
```



# HOW DO WE GET AN IDEA OF WHICH IS BETTER?

without implementing and running the two algorithms on all kinds of data?

## Algorithm Analysis

# ALGORITHM ANALYSIS IN A NUTSHELL



*How long will my program take?*

*How much memory will it consume?*

# BUT OTHER CONCERNS MAY BE RELEVANT:

In general, we want to measure usage of some **resource**:

- communication bandwidth
- number of processing elements required
- human computation
- etc.

**but we will focus on time and memory (space)**



# A SIMPLE MODEL OF COMPUTATION

## Random-Access Machine (RAM)

- instructions are sequential (no concurrency)

Each operation takes some constant amount of time

What are instructions?

- arithmetic operations (+, -, /, \* etc.)
- control (branches, function call & returns)

Simple memory: no caching

Is exponentiation  $a^k$  considered one operation?

- A. Yes
- B. No
- C. It's up to us to define in our model.



# DO COMMENTS COUNT?

```
int a = 0;  
// the following is a loop  
for (int i=0; i<10; i++) {  
    a = a + 2;  
    A[i] = a;  
}
```

Do comments count?

- A. Yes
- B. No
- C. Only on Facebook.
- D. Only on Piazza



# RUNNING TIME

Number of primitive instructions / “steps” executed.

**Code**

```
int a = 0;  
for (int i=0; i<n; i++) {  
    a = a + 2;  
    A[i] = a;  
}
```

**Cost**

$c_1$

$c_2$

$c_3$

$c_4$

**Times**



# RUNNING TIME

Number of primitive instructions / “steps” executed.

**Code**

```
int a = 0;  
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**Cost**

$c_1$

$c_2$

$c_3$

$c_4$

**Times**

1



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$c_3$

$c_4$

**Times**

1

$n$





# RUNNING TIME

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## Times

1

$n$

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$c_1$

$c_2$

$c_3$

$c_4$

## Times

1

$n$

$n$

$n$



# RUNNING TIME

Number of primitive instructions / “steps” executed.

**Code**

```
int a = 0;  
for (int i=0; i<n; i++) {  
    a = a + 2;  
    A[i] = a;  
}
```

**Cost**

$c_1$

$c_2$

$c_3$

$c_4$

**Times**

1

$n$

$n$

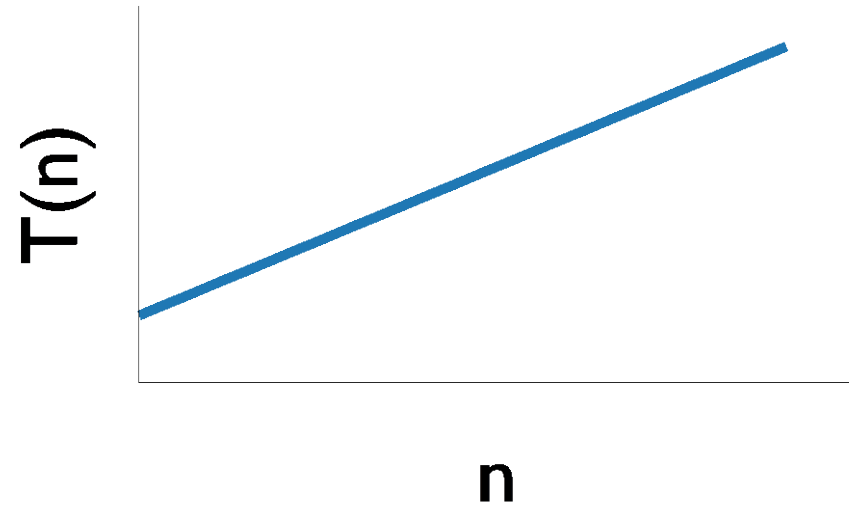
$n$

$$T(n) = c_1 + nc_2 + nc_3 + nc_4 = cn + c_1$$

# RUNNING TIME

Number of primitive instructions / “steps” executed.

$T(n) = cn + c_1$   
is a linear function





# WHICH TAKES LONGER?

```
void pushAll(int n) {  
    for (int i=0; i<= 100*n; i++) {  
        stack.push(i);  
    }  
}
```

```
void pushAdd(int n) {  
    for (int i=0; i<= n; i++) {  
        for (int j=0; j<= n; j++) {  
            stack.push(i+j);  
        }  
    }  
}
```

Which takes longer?

- A. pushAll
- B. pushAdd
- C. Runs the same.
- D. It depends!

# ANALYSIS OF PUSHALL

```
void pushAll(int n) {  
    for (int i=0; i<= 100*n; i++) {  
        stack.push(i);  
    }  
}
```

**Cost**

$c_1$

$c_2$

$c_3$

**Times**

1

$100n$

$100n$

$$T(n) = 100cn + c_1$$

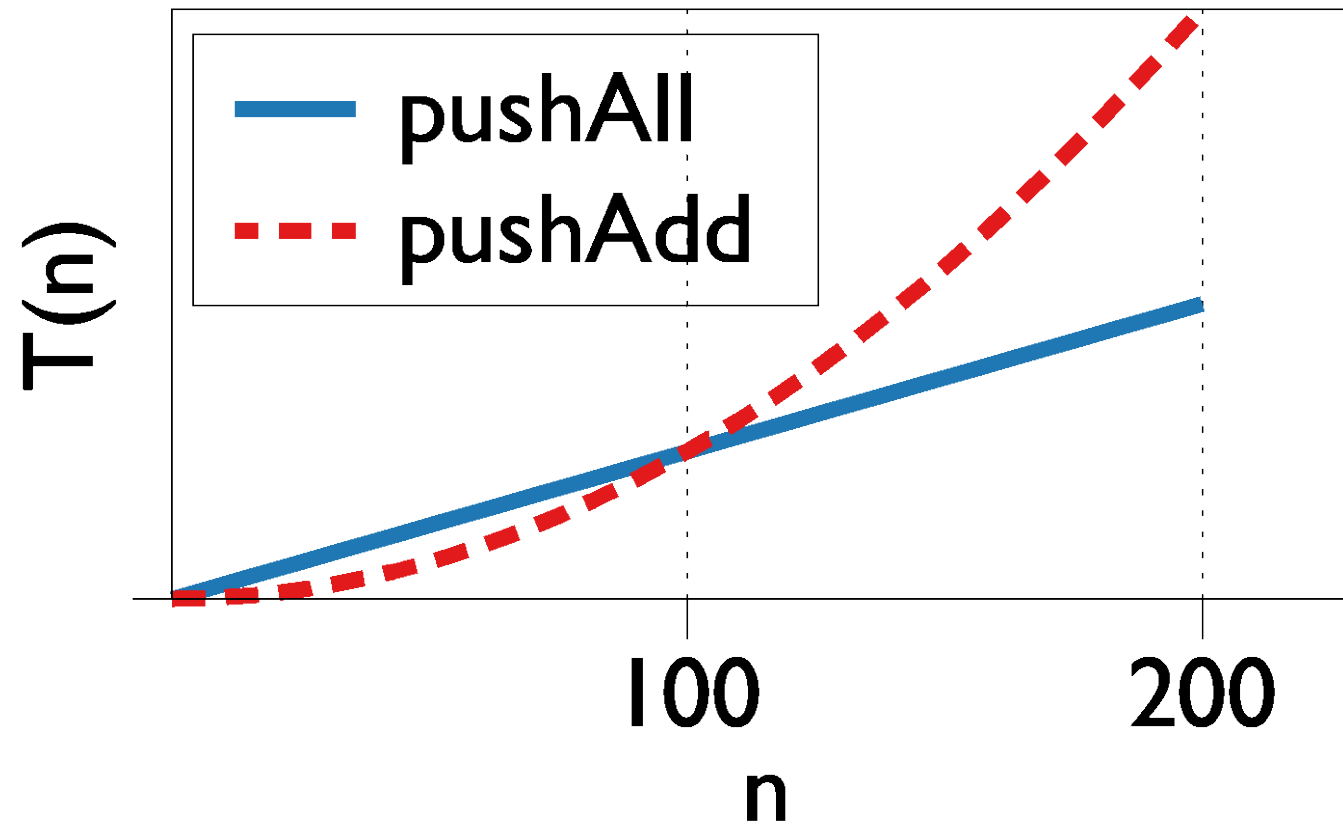
# ANALYSIS OF PUSHADD

	Cost	Times
<code>void pushAdd(int n) {</code>	$c_1$	1
<code>    for (int i=0; i&lt;= n; i++) {</code>	$c_2$	$n$
<code>        for (int j=0; j&lt;= n; j++) {</code>	$c_3$	$n^2$
<code>            stack.push(i+j);</code>	$c_4$	$n^2$
<code>        }</code>		
<code>    }</code>		
<code>}</code>		

$$T(n) = c_1 + nc_2 + n^2c_3 + n^2c_4 = cn^2 + c_2n + c_1$$

is a **quadratic** function

IF ALL THE CONSTANTS = 1







# WHICH TAKES LONGER?

```
void pushAll(int n) {  
    for (int i=0; i<= 100*n; i++) {  
        stack.push(i);  
    }  
}
```

```
void pushAdd(int n) {  
    for (int i=0; i<= n; i++) {  
        for (int j=0; j<= n; j++) {  
            stack.push(i+j);  
        }  
    }  
}
```

Which takes longer if  $n$  is large ( $n > 1000$ )?

- A. pushAll
- B. pushAdd
- C. Runs the same.
- D. It depends!



# WHICH TAKES **MORE MEMORY** ON THE STACK?

```
void pushAll(int n) {  
    for (int i=0; i<= 100*n; i++) {  
        stack.push(i);  
    }  
}
```

```
void pushAdd(int n) {  
    for (int i=0; i<= n; i++) {  
        for (int j=0; j<= n; j++) {  
            stack.push(i+j);  
        }  
    }  
}
```

Which takes more memory if  $n$  is large ( $n > 1000$ )?

- A. pushAll
- B. pushAdd
- C. takes the same amount of memory
- D. Naruto knows!

# ANALYSIS OF PUSHALL

```
void pushAll(int n) {  
    for (int i=0; i<= 100*n; i++) {  
        stack.push(i);  
    }  
}
```

**Cost**

$c$

**Times**

$100n$

$$S(n) = 100cn$$

# ANALYSIS OF PUSHADD

```
void pushAdd(int n) {  
    for (int i=0; i<= n; i++) {  
        for (int j=0; j<= n; j++) {  
            stack.push(i+j);  
        }  
    }  
}
```

**Cost**

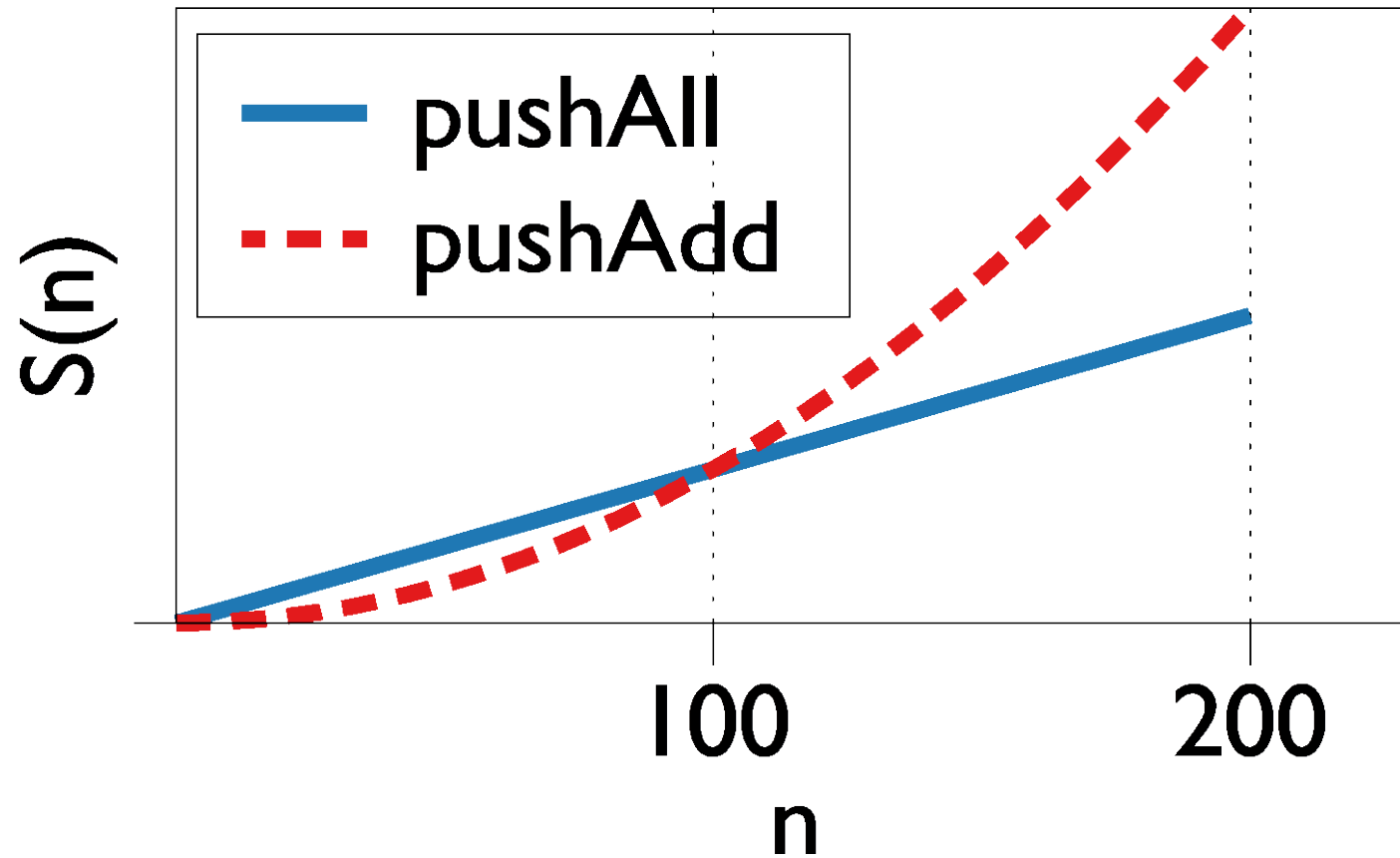
**Times**

$c$

$n^2$

$$S(n) = cn^2$$

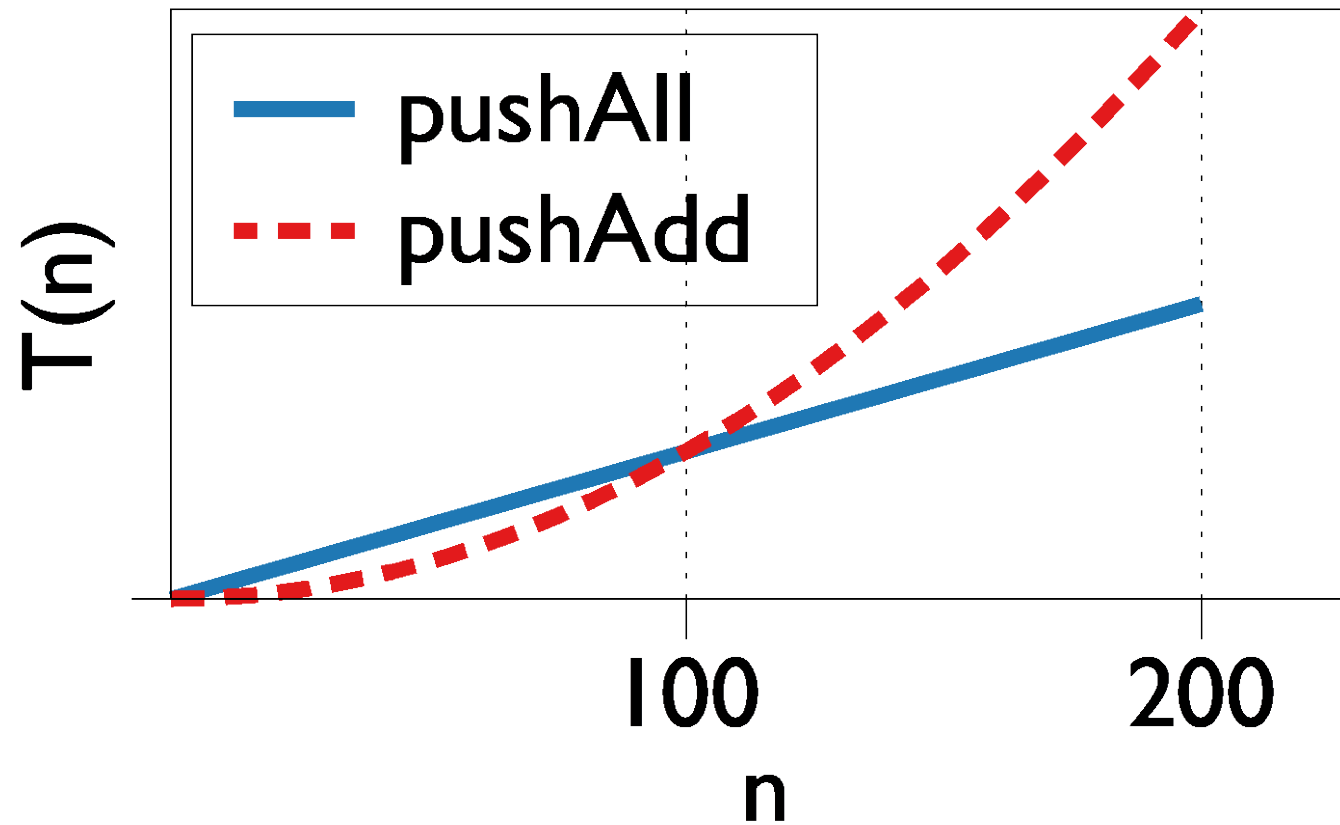
## SIMILAR GRAPH FOR MEMORY IF $C = 1$



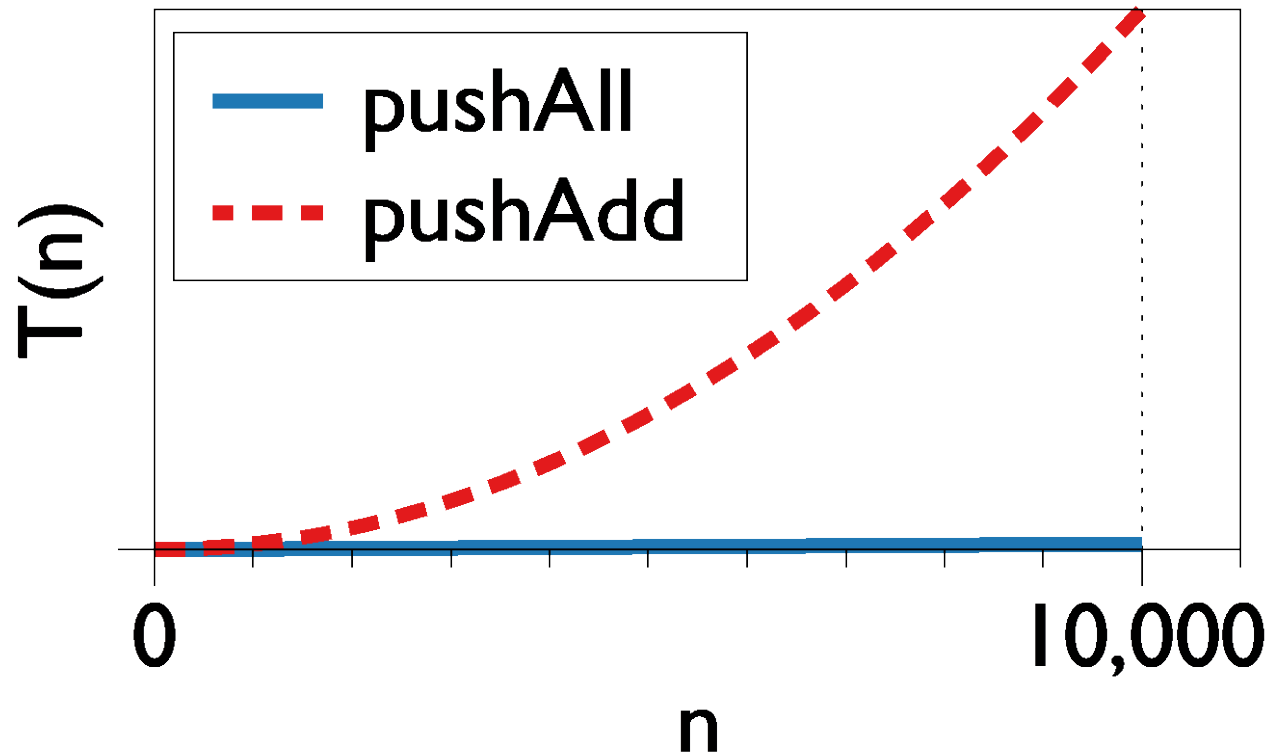
**QUESTIONS?**



# ORDER OF GROWTH



# ORDER OF GROWTH



$$T(n) = n^2 + c_2n$$
  
as  $n \rightarrow \infty$ , the term  $c_2n$  becomes insignificant compared to  $n^2$ .

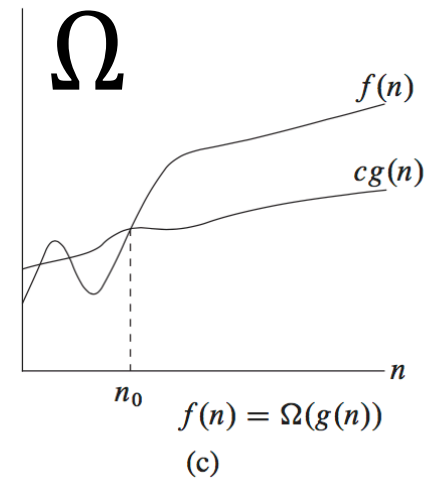
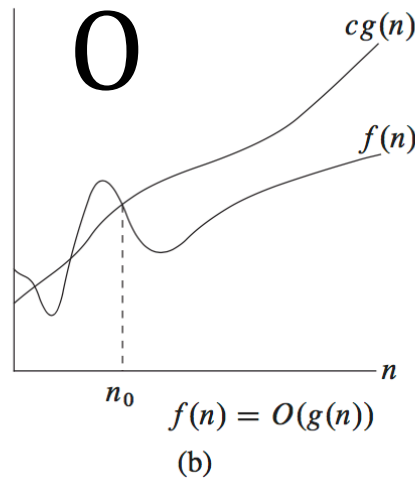
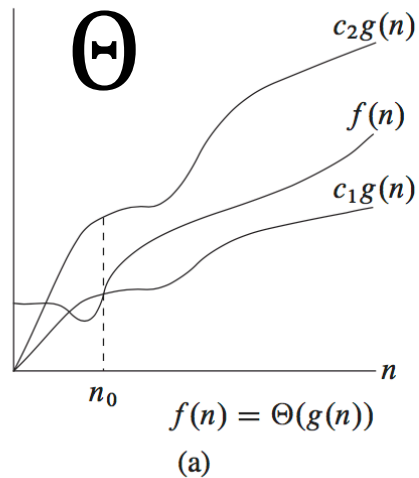
We can say the  $T(n)$  is “asymptotically equivalent” to  $n^2$ .



# ASYMPTOTIC EFFICIENCY & ORDER OF GROWTH

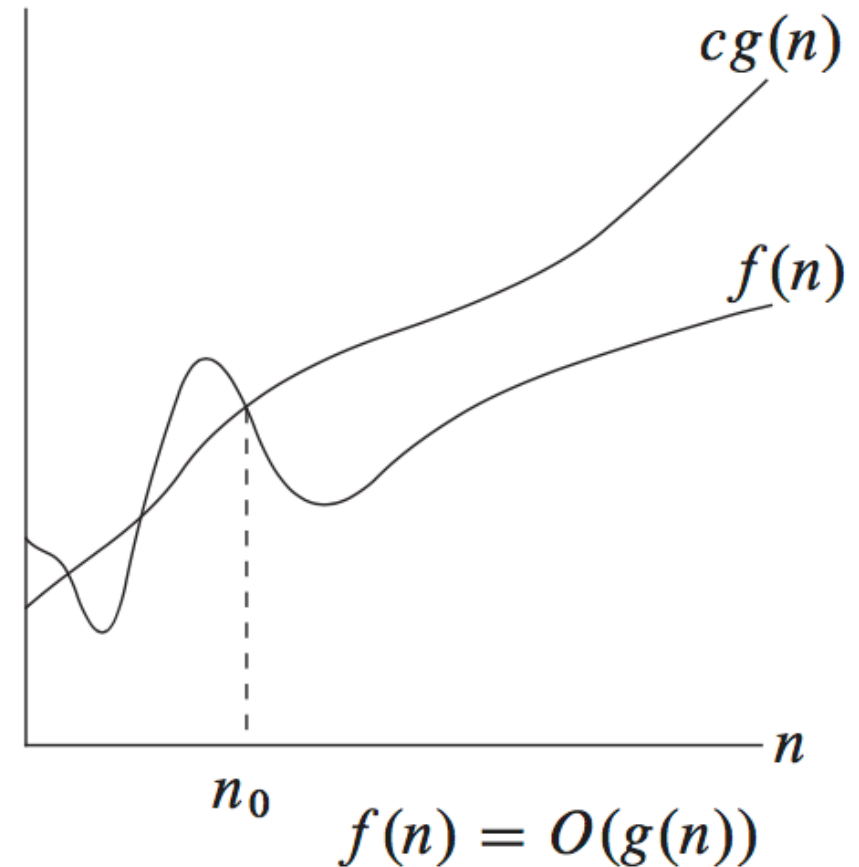
**Further simplify:** drop constants and lower order terms

How the running time of the algorithm increases with input size **in the limit** (*this is what asymptotic means*).



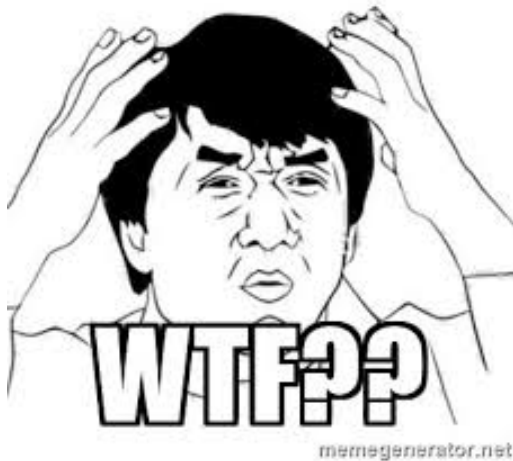
# BIG-OH $O(g(n))$

$O(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0\}$

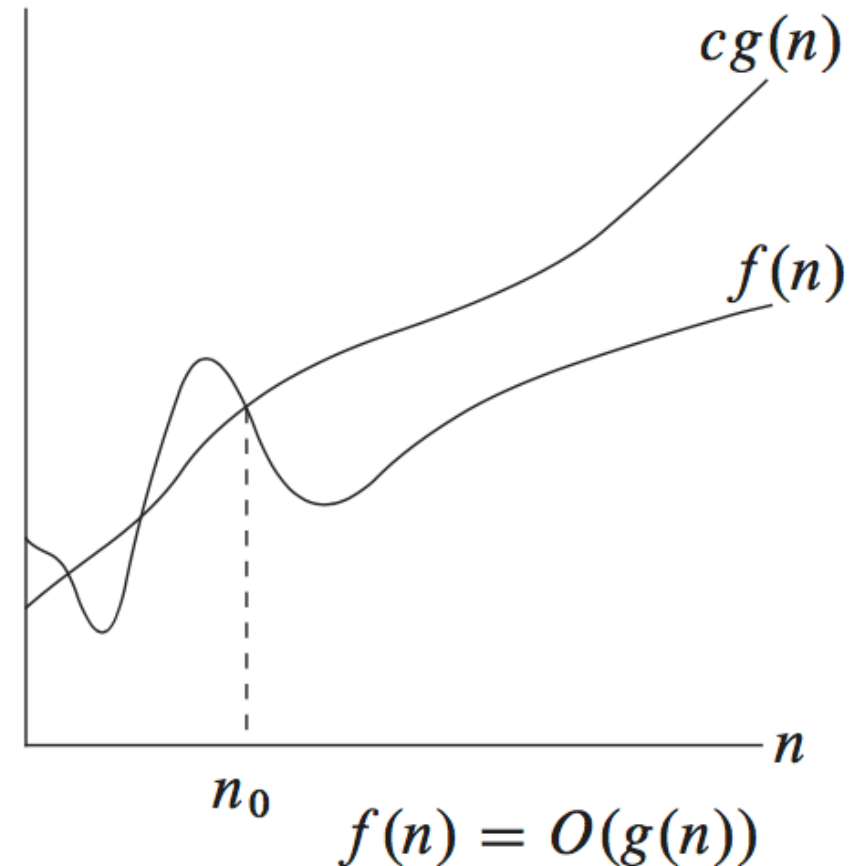


# BIG-OH $O(g(n))$

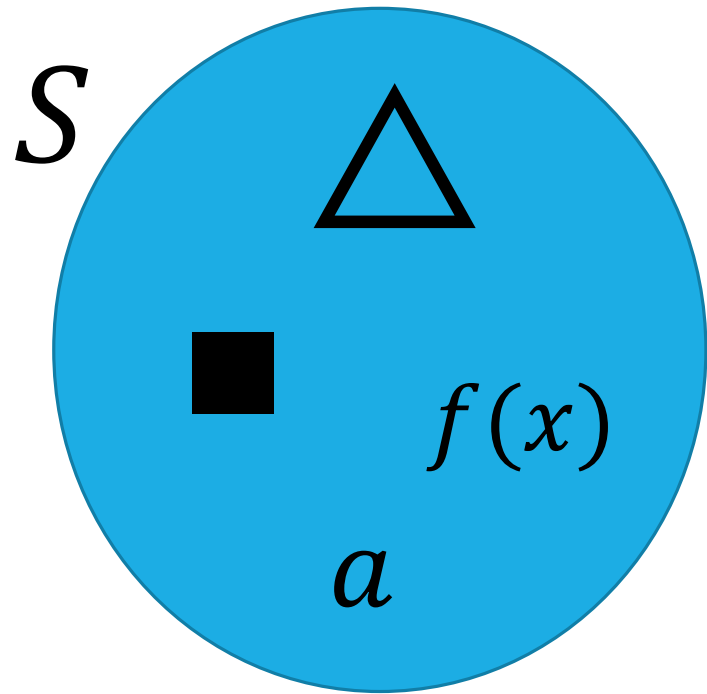
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**what does  
this mean?!**



## BRIEFLY: SET THEORY



$$S = \{a, \blacksquare, \triangle, f(x)\}$$

We say an element is a member of a set using the notation:

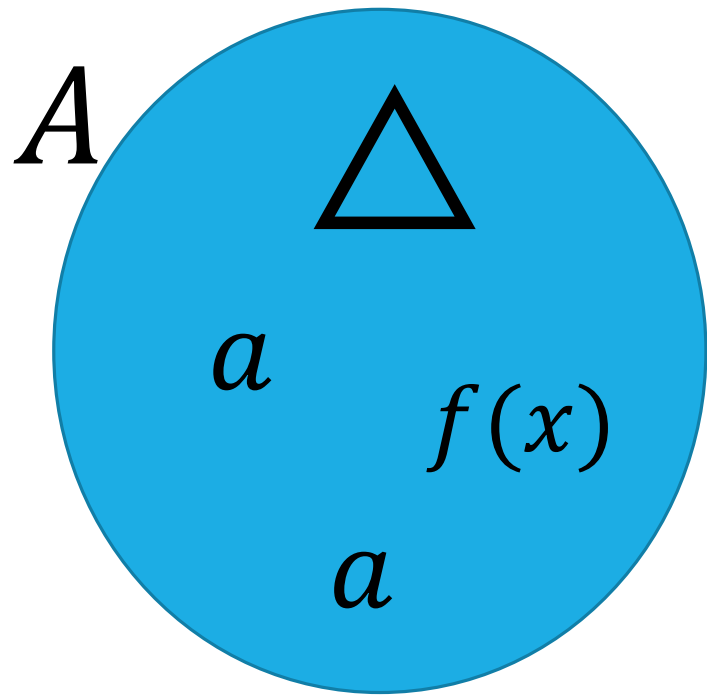
$$a \in S$$

and if  $b$  is **not** a member:

$$b \notin S$$



# BRIEFLY: SET THEORY



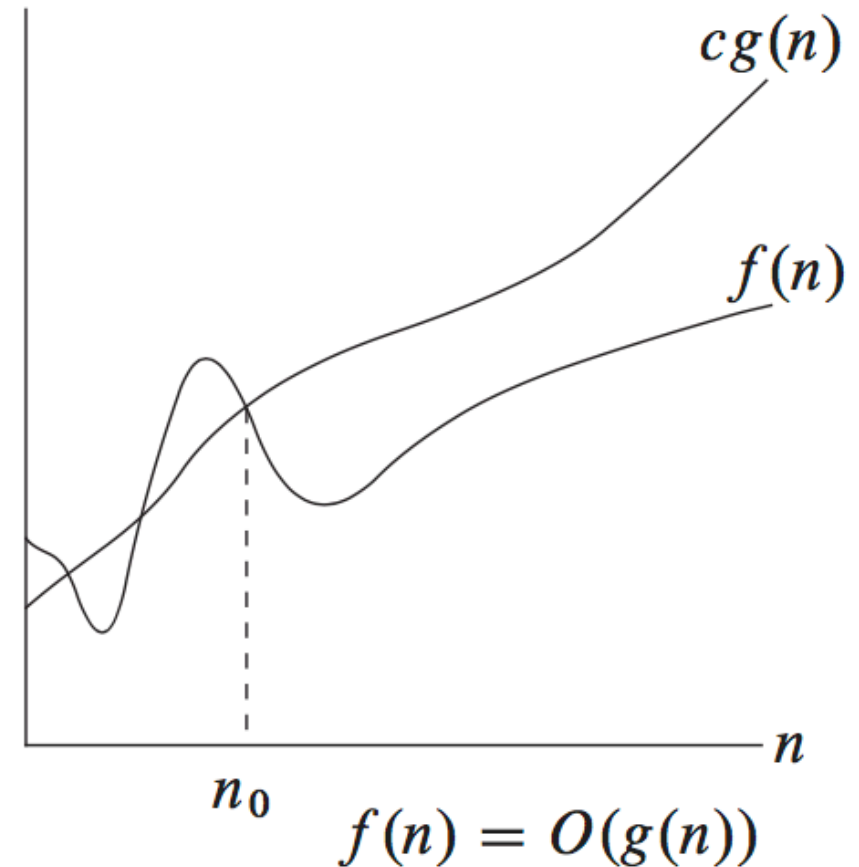
$$A = \{a, a, \Delta, f(x)\}$$

Is the object on the left a valid set?

- A. Yes
- B. No
- C. I don't know
- D. The triangle is pretty
- E. I'm so so confused...

# BIG-OH $O(g(n))$

$O(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0\}$

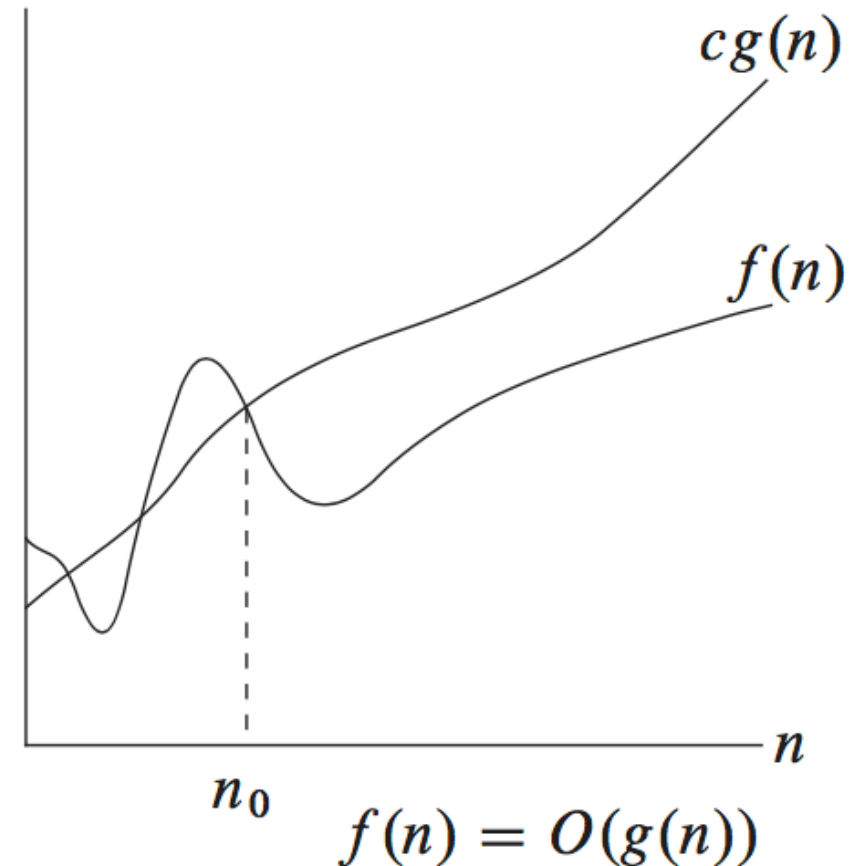


# BIG-OH $O(g(n))$

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Parsing the statement:

$O(g(n))$  is a SET that contains  
\_\_\_\_\_ that are  
smaller/larger than  $cg(n)$  for large  
 $n$

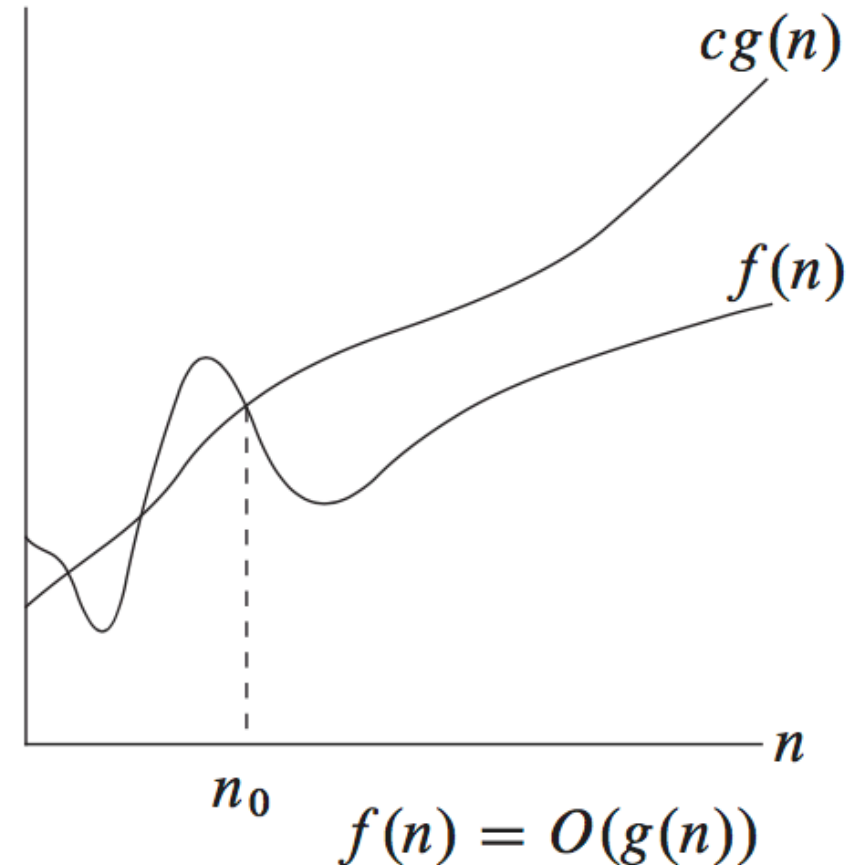


# BIG-OH $O(g(n))$

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Parsing the statement:

$O(g(n))$  is a SET that contains NON-NEGATIVE FUNCTIONS that are smaller/larger than  $cg(n)$  for large  $n$



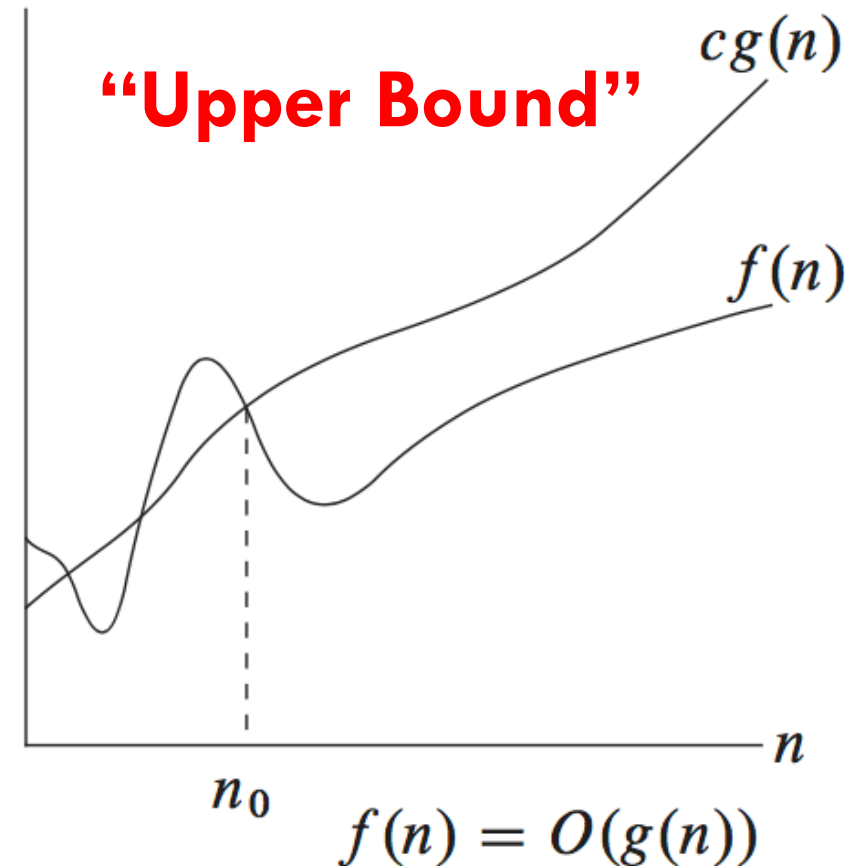


# BIG-OH $O(g(n))$

$O(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0\}$

Parsing the statement:

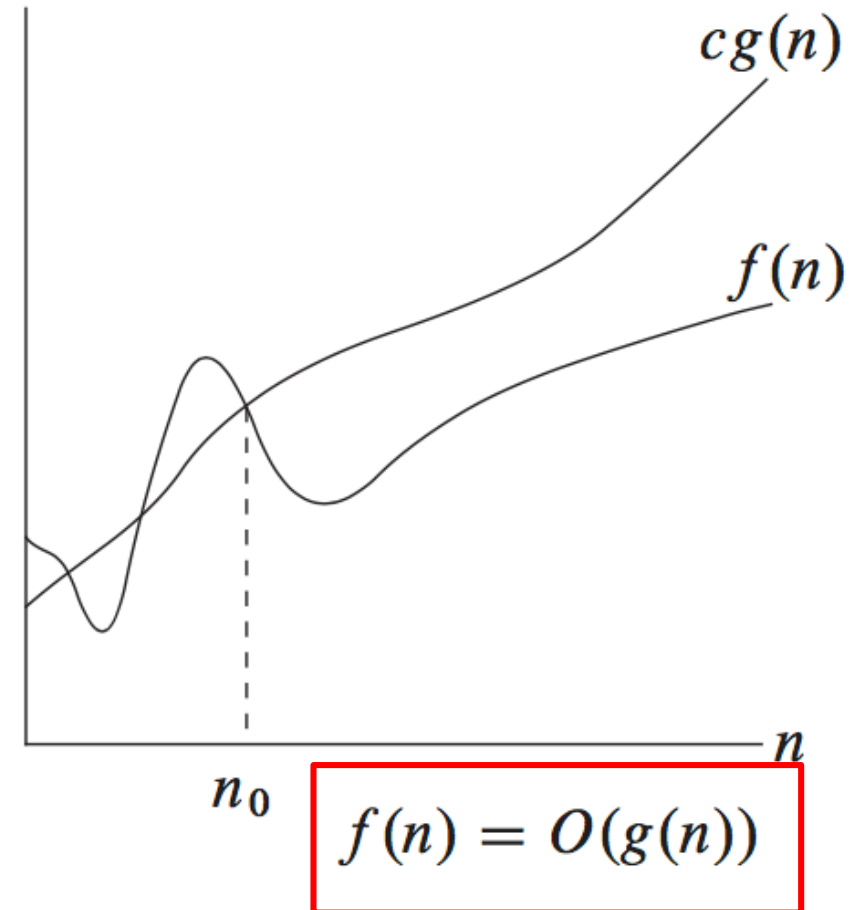
$O(g(n))$  is a SET that contains NON-NEGATIVE FUNCTIONS that are smaller/larger than  $cg(n)$  for large  $n$



# BUT WAIT...

If  $O(g(n))$  is a set (a collection of things), how can we say that

$$f(n) = O(g(n))?$$

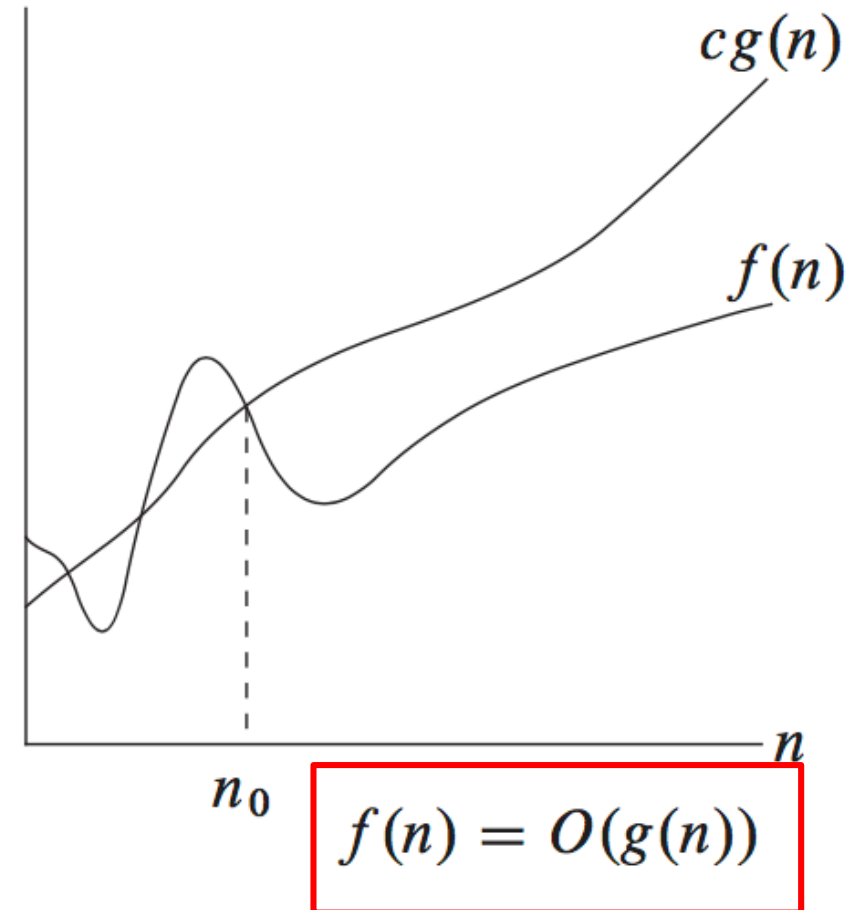


# BUT WAIT...

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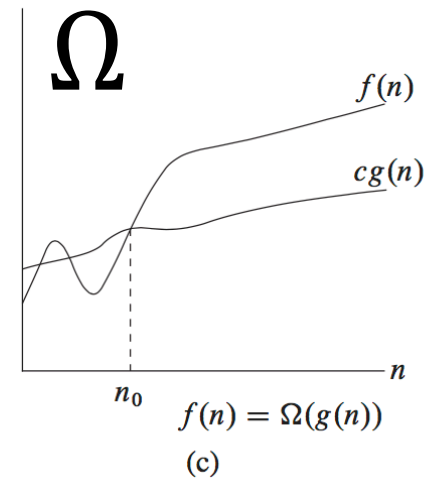
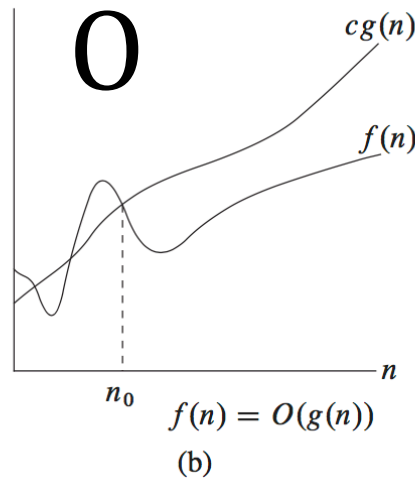
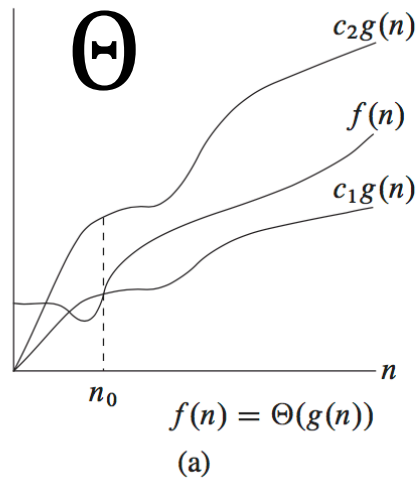
“Abuse of notation”: we write  
 $f(n) = O(g(n))$ , but mean  
 $f(n) \in O(g(n))$



# ASYMPTOTIC EFFICIENCY & ORDER OF GROWTH

**Further simplify:** drop constants and lower order terms

How the running time of the algorithm increases with input size **in the limit** (*this is what asymptotic means*).

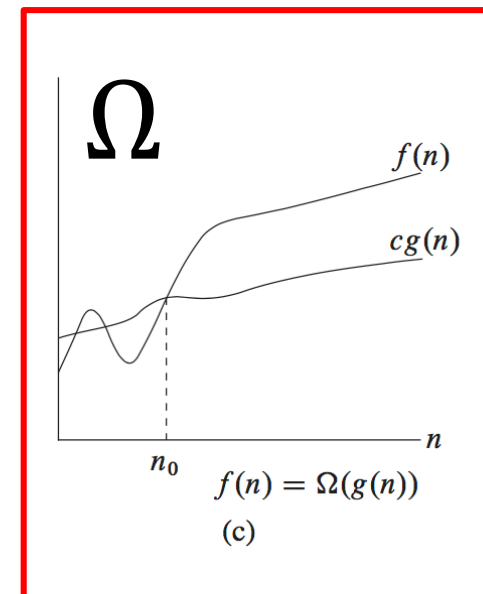
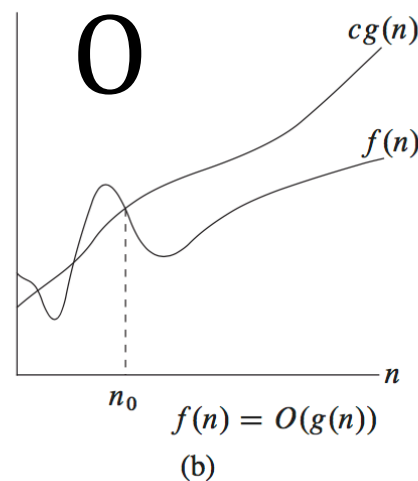
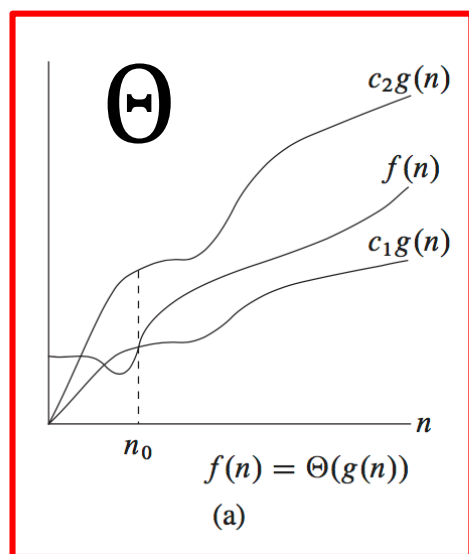


# BIG OMEGA $\Omega$ AND BIG THETA $\Theta$



What about  $\Omega$  and  $\Theta$  ?

Let's take a look...

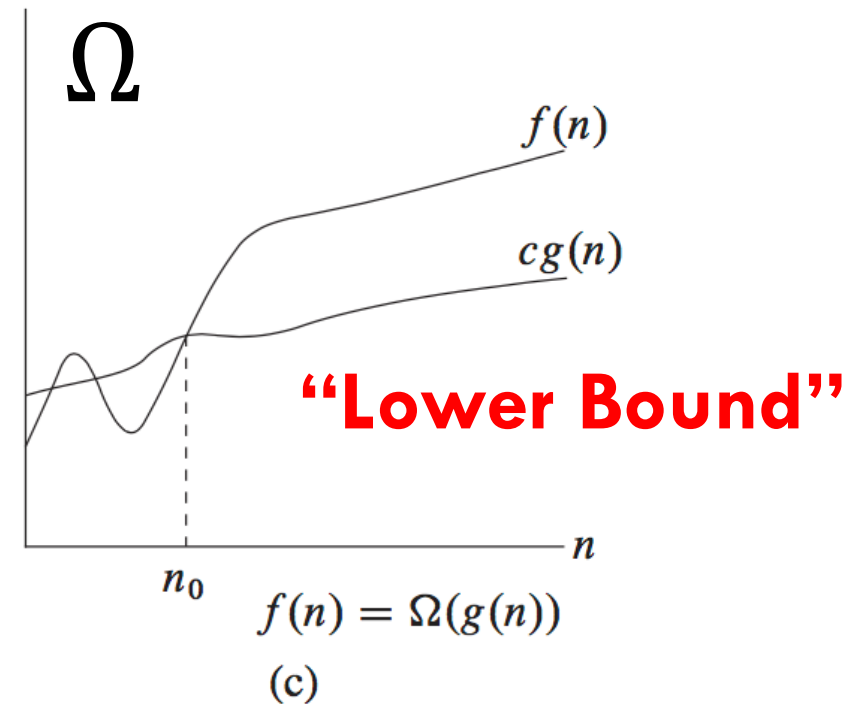


# BIG OMEGA $\Omega$



$\Omega(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that}$   
 $0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0\}$

$O(g(n))$  is a set that contains  
functions that are larger than  $cg(n)$   
for large  $n$  and some constant  $c$

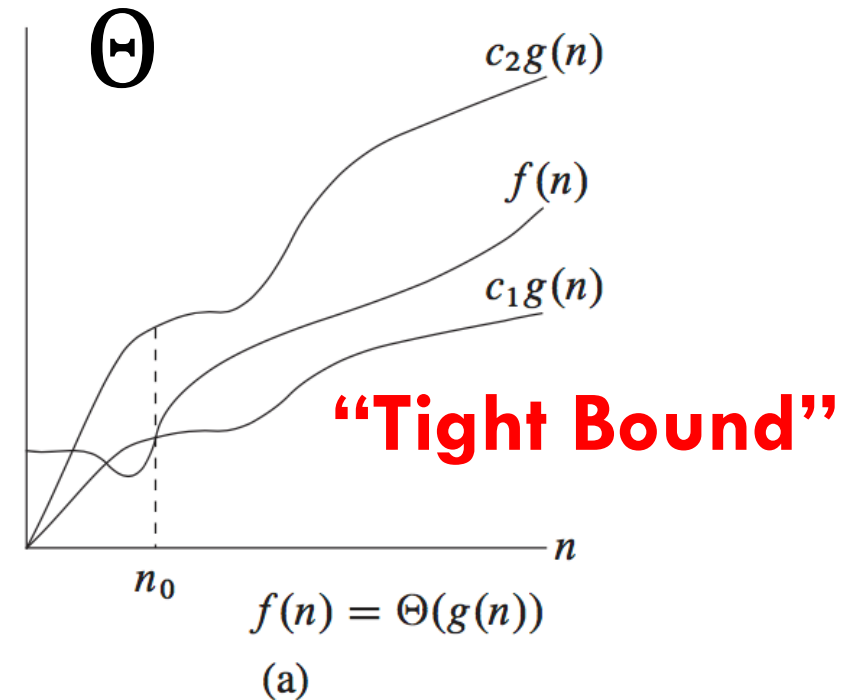


# BIG THETA $\Theta$



$\Theta(g(n)) = \{f(n): \text{there exist positive constants } c_1, c_2 \text{ and } n_0 \text{ such that } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0\}$

$O(g(n))$  is a set that contains functions that are larger than  $c_1 g(n)$  and smaller than  $c_2 g(n)$  for large  $n$  and some constants  $c_1$  and  $c_2$

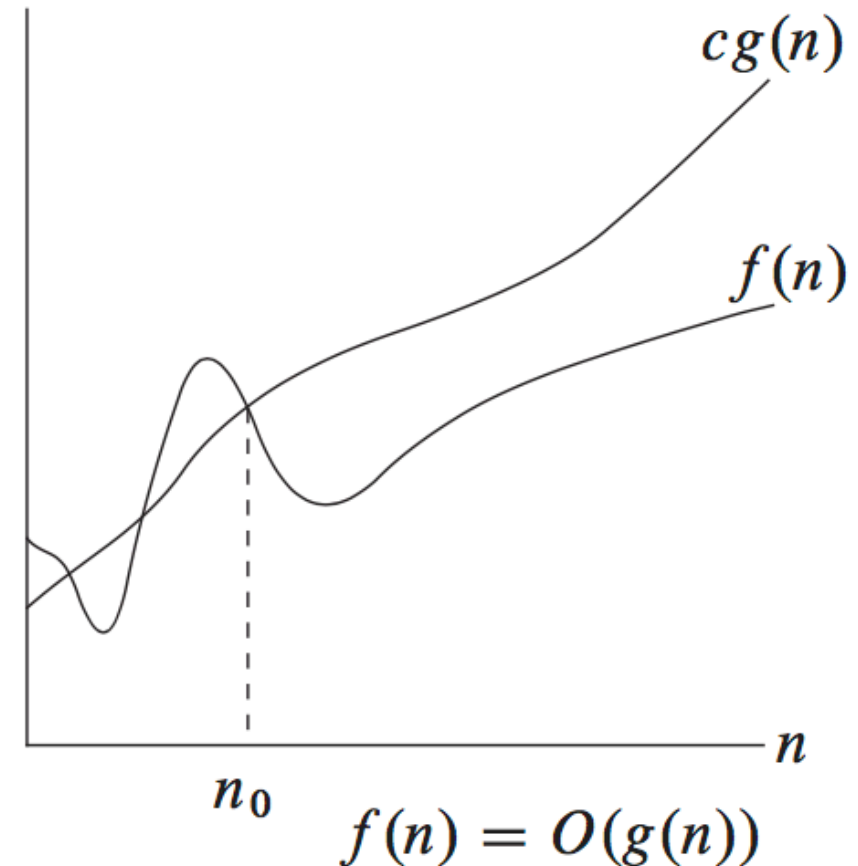




# WE FOCUS ON BIG-OH $O(g(n))$

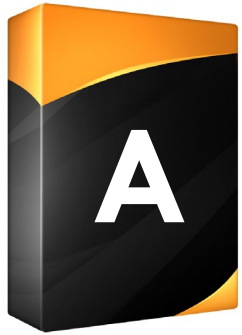
Why do we focus on  $O(g(n))$  ?

- A. we want to know the best case.
- B. we want to know the worst case.
- C. We are pessimistic people.
- D. I'm so very confused...





# SOFTWARE A V.S. SOFTWARE B



Guarantees in the **best case**,  
it will crash only once a week



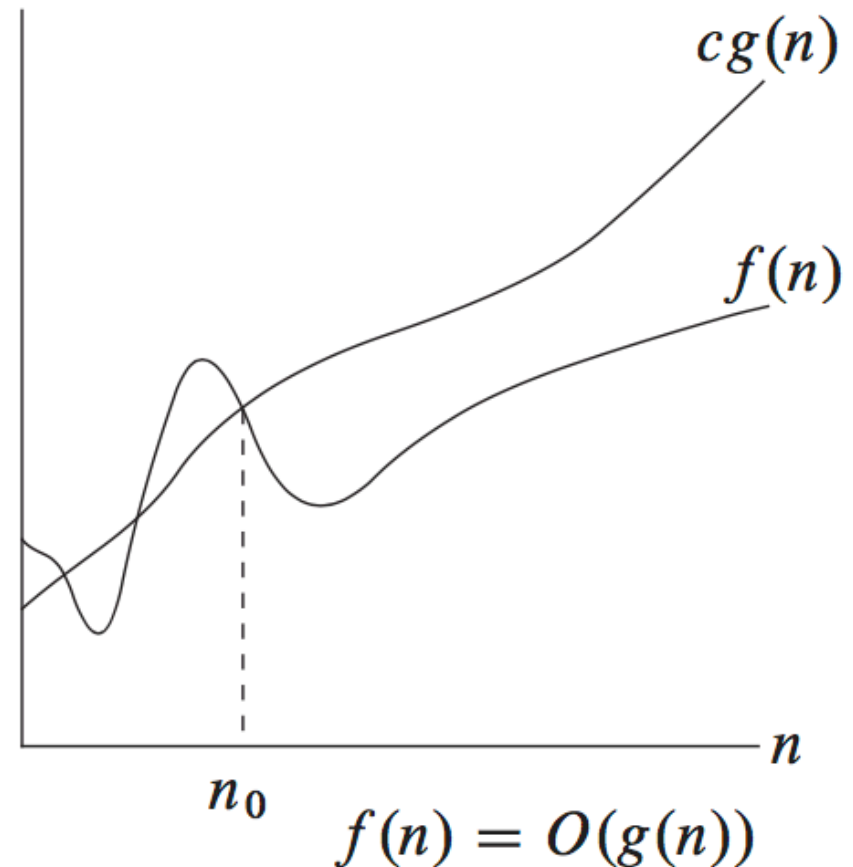
Guarantees in the **worst case**,  
it will crash only once a week

**Which do  
you buy?**

# WHAT DOES BIG-OH MEAN INTUITIVELY

When we say an algorithm is  $O(n^2)$ :

- the **worst case running time** is  $O(n^2)$ .
- guarantee that the running time will not exceed  $cn^2$  for large  $n$
- the running time is upper bounded by a  $cn^2$  for large  $n$





# DROPPING THE LOWER ORDER TERMS?

Given  $f(n) = an^2 + bn$  where  $a, b > 0$ .

Show that  $f(n) = O(n^2)$ , i.e., that we can drop the lower order terms and ignore the coefficient for the function.



# SOME PRACTICE WITH BIG-OH

Is  $2^{n+1} = O(2^n)$ ?

- A. Yes
- B. No
- C. It depends.
- D. My dog ate my math homework.



# SOME PRACTICE WITH BIG-OH

What is the worst case running time for the statement  $x = 1$ ?

- A.  $O(n)$
- B.  $O(n^2)$
- C.  $O(1)$
- D. Should have stayed at home...
- E. ... and watched Netflix



# SOME PRACTICE WITH BIG-OH: IF/ELSE

```
if (n > 100) {  
    for (int i=0; i<n; i++) {  
        for (int j=0; j<n; j++) {  
            result += 2*i + 3*j + i*j;  
        }  
    }  
    return result;  
} else {  
    for (int i=0; i<n; i++) {  
        result += 2*i + i*i;  
    }  
    return result;  
}
```

The algorithm on the left is:

- A.  $O(n)$
- B.  $O(n^2)$
- C.  $O(n^3)$
- D. I'm confused!



# SOME PRACTICE WITH BIG-OH: RECURSION

```
int addSum(int n) {
    if (n<=1) return n;
    if (isOdd(n)) return n;

    return addSum(n-2) + 10;
}
```

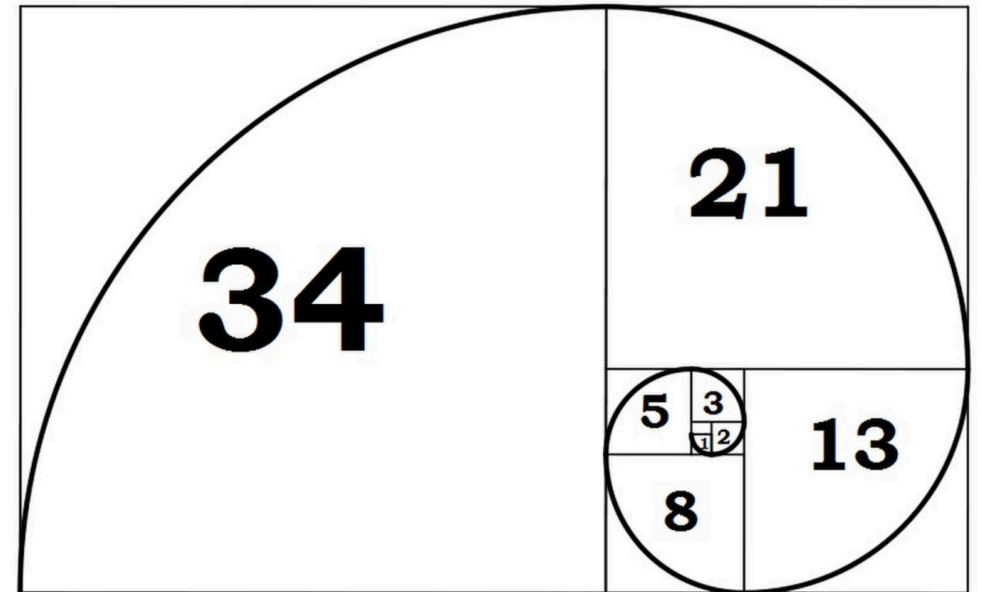
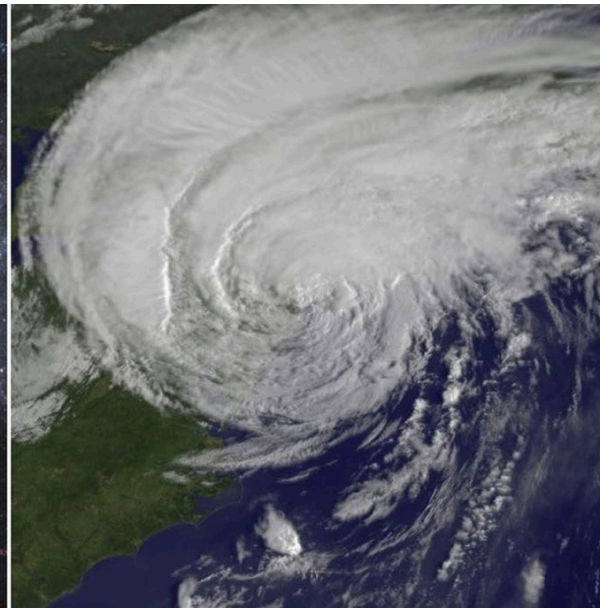
What is the worst case running time for addSum(n)?

- A.  $O(n)$
- B.  $O(n^3)$
- C.  $O(2^n)$
- D.



# THE SHAPES OF SPIRAL GALAXIES AND HURRICANES FOLLOW THIS SEQUENCE. WHAT IS IT?

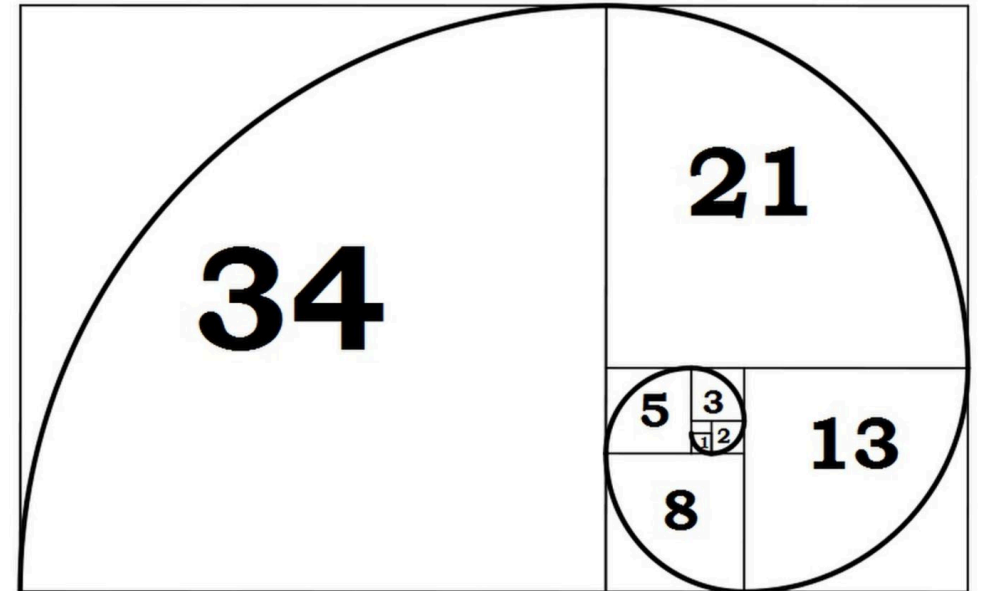
0, 1, 2, 3, 5, 8, 13, 21, 34, ...





# FIBONACCI SEQUENCE

0, 1, 2, 3, 5, 8, 13, 21, 34, ...





# SOME PRACTICE WITH BIG-OH: ITERATION

```
static int fibItr(int n)
{
    if (n == 0) return 0;
    if (n == 1) return 1;
    int prev_2 = 0;
    int prev_1 = 1;
    int result = 0;
    for (int i = 2; i <= n; i++)
    {
        result = prev_1 + prev_2;
        prev_2 = prev_1;
        prev_1 = result;
    }
    return result;
}
```

What is the worst case running time for fibItr(n)?

- A.  $O(n)$
- B.  $O(n^3)$
- C.  $O(2^n)$
- D.





# SOME PRACTICE WITH BIG-OH: LINKED LISTS

What is the worst case running time for inserting an object into a linked list?

- A.  $O(1)$
- B.  $O(n)$
- C.  $O(n^2)$
- D. how to remember? that was yesterday!



# SOME PRACTICE WITH BIG-OH: STACKS

What is the worst case running time for pushing an item onto a stack (implemented as an array)?

- A.  $O(1)$
- B.  $O(n)$
- C.  $O(n^2)$
- D. Longer than it takes to read this question.



# SOME PRACTICE WITH BIG-OH: LIMITATIONS

Both Selection Sort and Bubble Sort are  $O(N^2)$  algorithms. Does this mean they are equivalent in terms of performance?

- A. Yes
- B. No
- C. I don't know
- D. Why so many questions today ???

# BIG-OH LIMITATIONS

is a useful but **coarse** measure.

It hides the constants and lower order terms that can make a difference.

# BACK TO OUR INITIAL PROBLEM



**V.S.**





# NARUTO'S IDEA: INSERTION SORT

```
int n = array.length;
for (int j = 1; j < n; j++) {
    int key = array[j];
    int i = j-1;
    while ( (i > -1) && ( array [i] > key ) ) {
        array [i+1] = array [i];
        i--;
    }
    array[i+1] = key;
}
```

What is the worst case running time for insertion sort?

- A.  $O(n)$
- B.  $O(n^2)$
- C.  $O(2^n)$
- D. Boss says  $O(n!)$





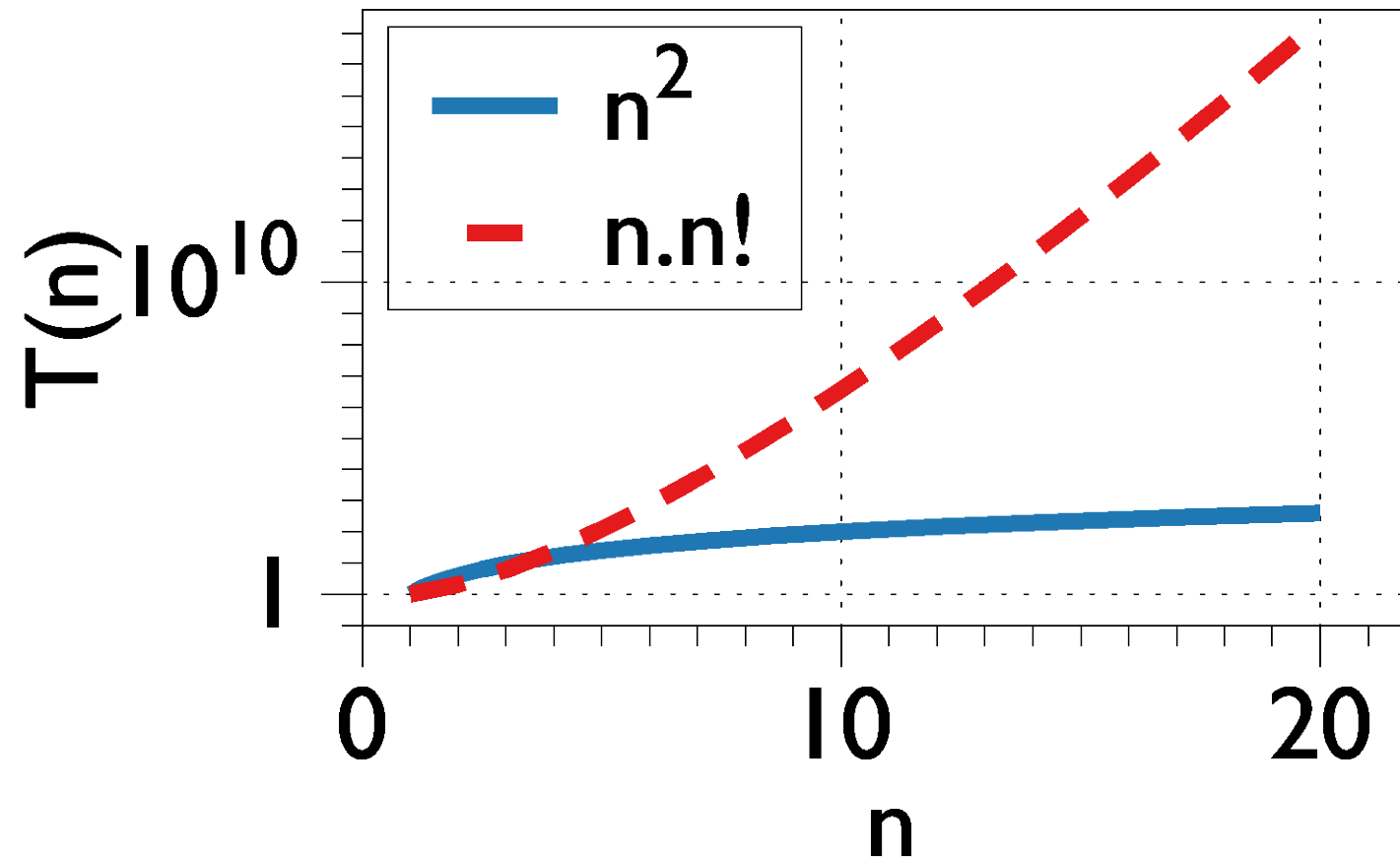
# BOSS'S IDEA: BOGOSORT

```
while items is not sorted  
    permute(items)
```

What is the worst case running time for bogosort?

- A.  $O(n)$
- B.  $O(n^2)$
- C.  $O(2^n)$
- D.  $O(n \cdot n!)$
- E. Boss says  $O(1)$

# THE DIFFERENCE?



# WHO HAS THE BETTER IDEA?



**V.S.**





# BOSS'S NEXT IDEA: RANDOMIZED BOGOSORT

I'M A  
WINNER!

```
while items is not sorted  
    randomShuffle(items)
```



# BOSS'S NEXT IDEA: RANDOMIZED BOGOSORT

I'M A  
WINNER!

```
while items is not sorted  
    randomShuffle(items)
```

What is the worst case running time for randomized bogosort?

- A.  $O(n)$
- B.  $O(n^2)$
- C.  $O(2^n)$
- D. Boss says  $O(1)$
- E. Unbounded

# WHO HAS THE BETTER IDEA?



**V.S.**



**QUESTIONS?**





## BOSS'S NEXT IDEA: RANDOMIZED BOGOSORT

BUT...

```
while items is not sorted  
    randomShuffle(items)
```

What is the average case running time for randomized bogosort?

- A.  $O(n)$
- B.  $O(2^n)$
- C.  $O(n \cdot n!)$
- D. Boss still says  $O(1)$
- E. Unbounded



# Sorting the Slow Way: An Analysis of Perversely Awful Randomized Sorting Algorithms

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**Abstract.** This paper is devoted to the “Discovery of Slowness.” The archetypical perversely awful algorithm bogo-sort, which is sometimes referred to as Monkey-sort, is analyzed with elementary methods. Moreover, practical experiments are performed.

## 1 Introduction

To our knowledge, the analysis of perversely awful algorithms can be tracked back at least to the seminal paper on pessimal algorithm design in 1984 [2]. But what’s a perversely awful algorithm? In the “The New Hacker’s Dictionary” [7] one finds the following entry:

**bogo-sort:** /boh‘goh-sort’/ /n./ (var. ‘stupid-sort’) The archetypical perversely awful algorithm (as opposed to → **bubble sort**, which is merely the generic \*bad\* algorithm). Bogo-sort is equivalent to repeatedly throwing a deck of cards in the air, picking them up at random, and then testing whether they are in order. It serves as a sort of canonical example of awfulness. Looking at a program and seeing a dumb algorithm, one might say ”Oh, I see, this program uses bogo-sort.” Compare → **bogus**, → **brute force**, → **Lasherism**.



# LEARNING OUTCOMES

By the end of this session, you should be able to:

- **Determine the computational complexity** of an algorithm under the standard sequential computation model
- **Use Big-Oh Notation** to describe algorithm performance

## OTHER TAKE AWAYS

Big-Oh is a good but coarse measurement (use it wisely)

You can use  $O(g(n))$  to measure the performance of other kinds of resource use



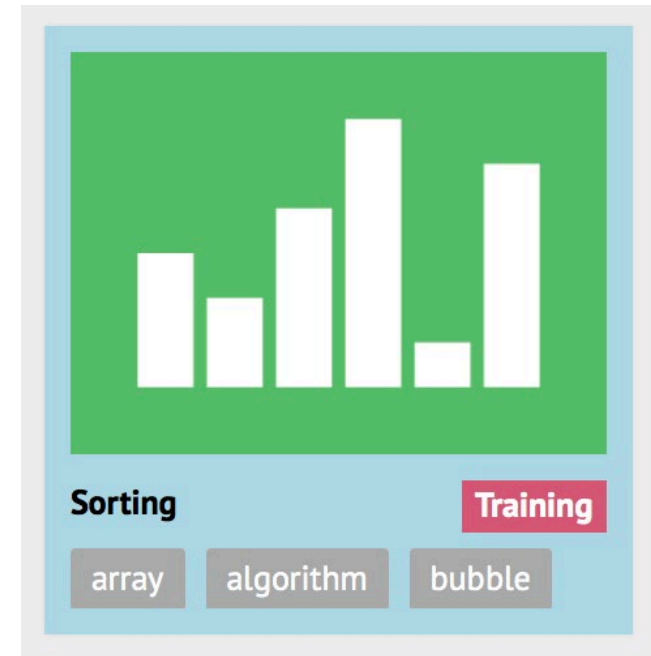
# BEFORE NEXT WEEK'S LECTURE

Go to Visualgo.net and do the  
Sorting Module:

<https://visualgo.net/en/sorting>

Review: 1 1 (Quick Sort)

Optional: 1 2 onwards



**QUESTIONS?**

