

Programming Abstractions

CS106B

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Today's Topics:

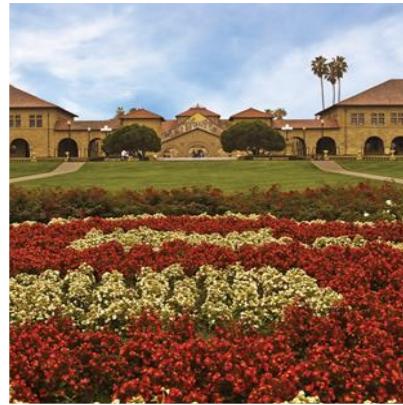
- Contrasting performance of 3 recursive algorithms
- Quantifying algorithm performance with Big-O analysis
- Getting a sense of scale in Big-O analysis

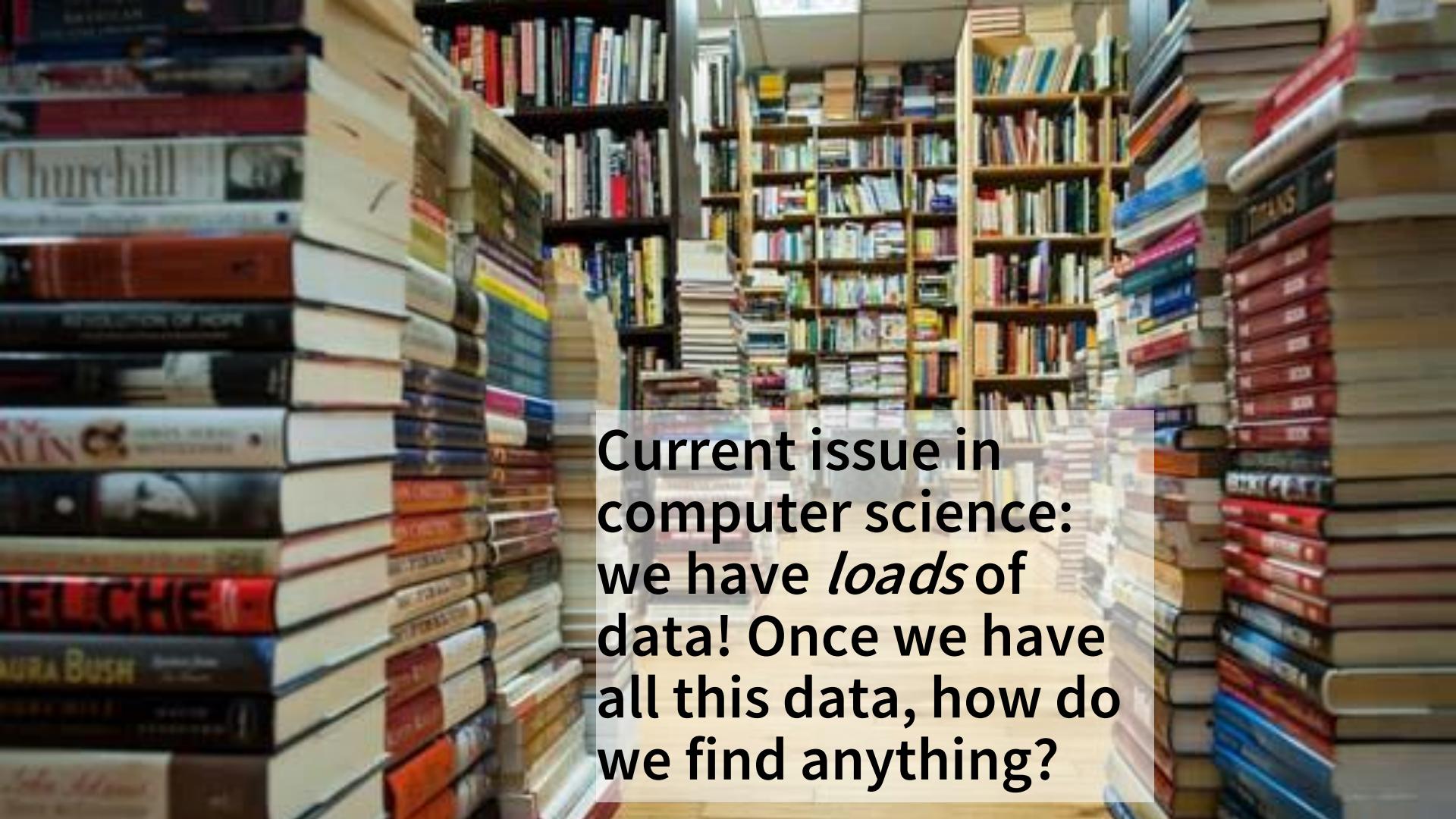
Announcements:

- Please don't start assignments on the deadline day! Yikes!
 - At Stanford, classes are supposed to be 3hrs/unit. CS106B is 5 units, so that's 15 hours of work per week. Of that, you should budget about 8-10 hours working on assignments.
- Remember that assignments are DUE FRIDAY.
 - The Sunday grace period is for emergencies such as injuries, illness, laptop died, etc. *only*.
 - If you plan on routine use of the grace period, you rob yourself of that safety buffer.
 - **Extension requests emailed to Neel are for issues whose scale and duration exceed the ability of the grace period to address**, not because something unexpectedly interfered with a *choice* to complete the assignment during the grace period.

Binary Search

AN ELEGANT SOLUTION TO
THE PROBLEM OF TOO MUCH
DATA





Current issue in
computer science:
we have *loads* of
data! Once we have
all this data, how do
we find anything?

The context:

- You have a **collection of numbers**
 - › Say product IDs for items in stock in a store
 - › We're going to store our collection of numbers in a Vector
 - › We're going to keep them *in sorted order*

0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- It's important to be able to **find out whether you have a particular number** in your collection or not
 - › A customer asks, "Do you have item 8 in stock?" (Yes.)
 - › A customer asks, "Do you have item 55 in stock?" (No.)
- **Key question: How long does this take?**

Does this list of numbers contain X?

Context: we have a collection of numbers in a Vector, in sorted order.

0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- **Basic approach: Start at the front and proceed forward until you find:**
 - › X (answer Yes)
 - › A number greater than X (answer No)
 - › End of the list (answer No)
- Key observation: each time you compare against the contents of a cell of the Vector and it's not X, you rule out *1* of the N cells in the Vector

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0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- **Efficiency Hack: Jump to the middle of the Vector and look there to find:**

- › X (answer Yes)
- › A number greater than X (rule out entire second half of Vector)

0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- › A number less than X (rule out entire first half of Vector)

0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- Key observation: with *one* comparison, you ruled out *N/2* of the N cells in the Vector!

Does this list of numbers contain X?

Context: we have a collection of numbers in a Vector, in sorted order.

0	1	2	3	4	5	6	7	8	9	10
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0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

Now we could do our Basic Approach, but in **half the time**.

Thanks, Efficiency Hack!!

- Key observation: with *one* comparison, you ruled out *N/2* of the Numbers in the Vector!

...but I have an even better idea...

Does this list of numbers contain X?

Context: we have a collection of numbers in a Vector, in sorted order.

0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- **Extreme Efficiency Hack: Keep jumping to the middle!**

- › Let's say our first jump to the middle found a number less than X, so we ruled out the whole first half:

0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- › Now jump to the middle of the remaining second half:

0	1	2	3	4	5	6	7	8	9	10
2	7	8	13	25	29	33	51	89	90	95

- Key observation: we do one piece of work, then delegate the rest. **Recursion!!**

Binary Search pseudocode

- We'll write the real C++ code together on Friday, but here's the outline/pseudocode of how it works:

```
bool binarySearch(Vector<int>& data, int key)
{
    if (data.size() == 0) {
        return false;
    }
    if (key == data[midpoint]) {
        return true;
    } else if (key < data[midpoint]) {
        return binarySearch(data[first half only], key);
    } else {
        return binarySearch(data[second half only], key);
    }
}
```

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    }
}
```

Base case: we shrank the search problem so tiny it no longer exists!

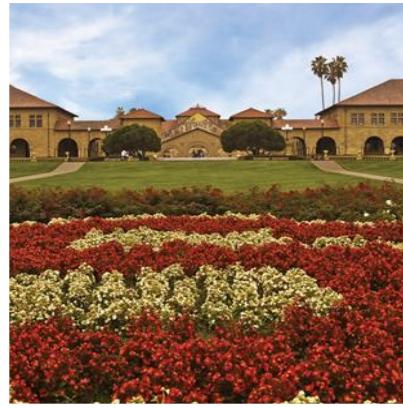
Recursive case:

Do one piece of work (comparison)

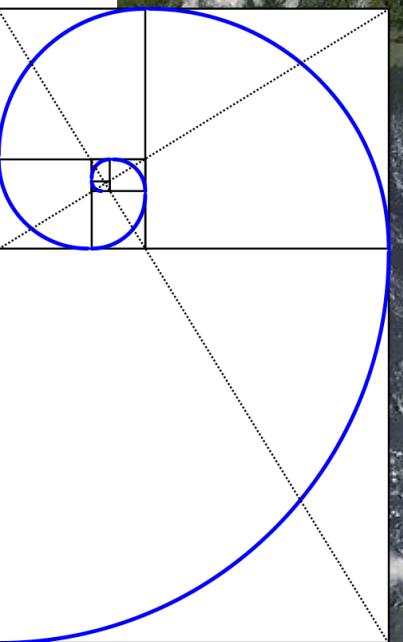
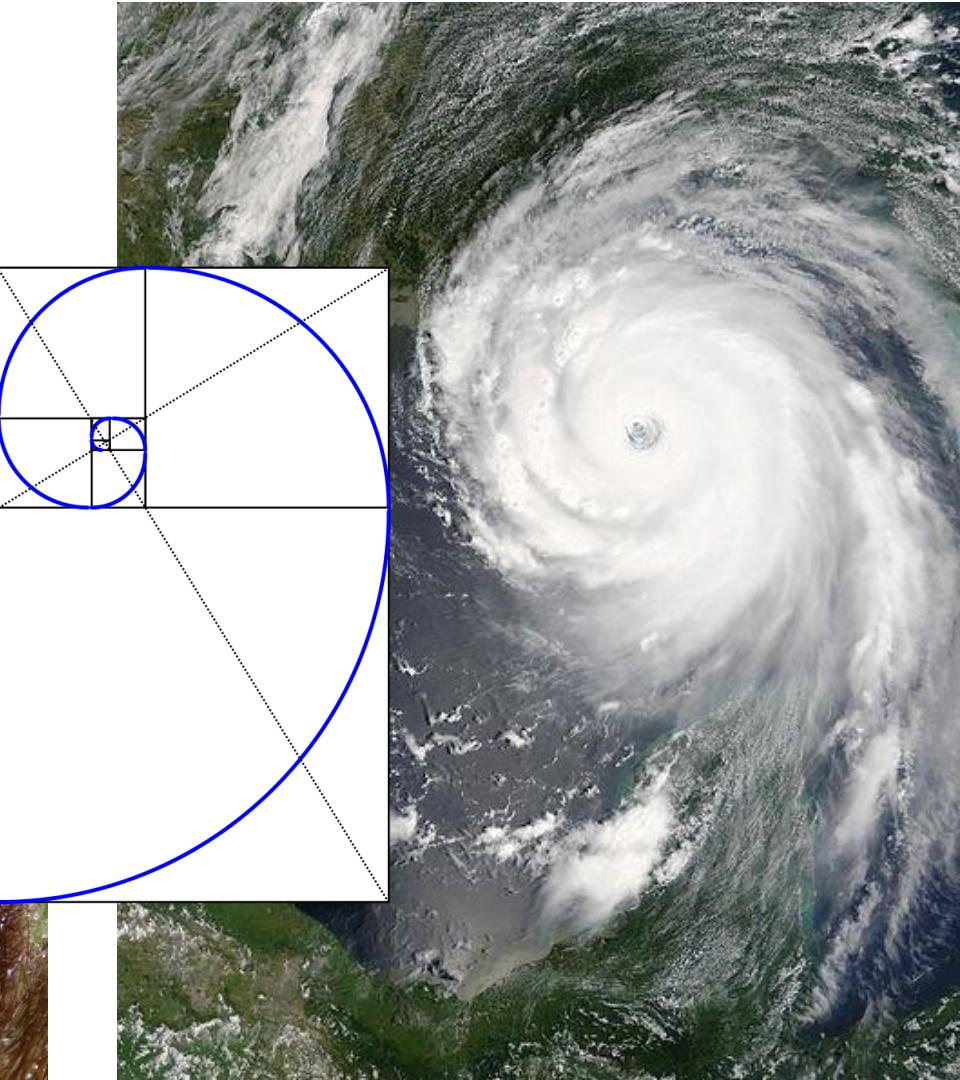
Delegate the rest of the work

The Fibonacci Sequence

*MATH NERD REJOICING
INTENSIFIES*



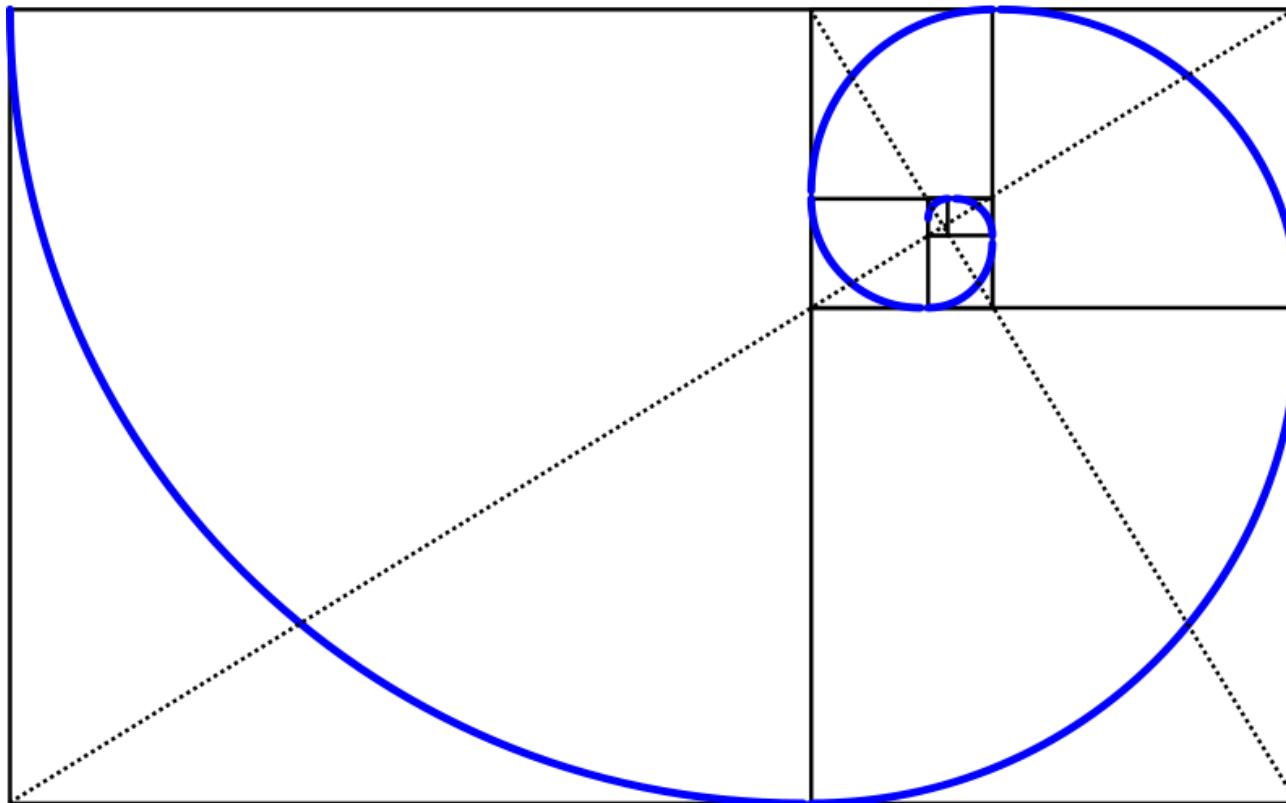
Fibonacci in nature



These files are, respectively: public domain (hurricane) and licensed under the [Creative Commons Attribution 2.0 Generic license](#) (fibonacci and fern).

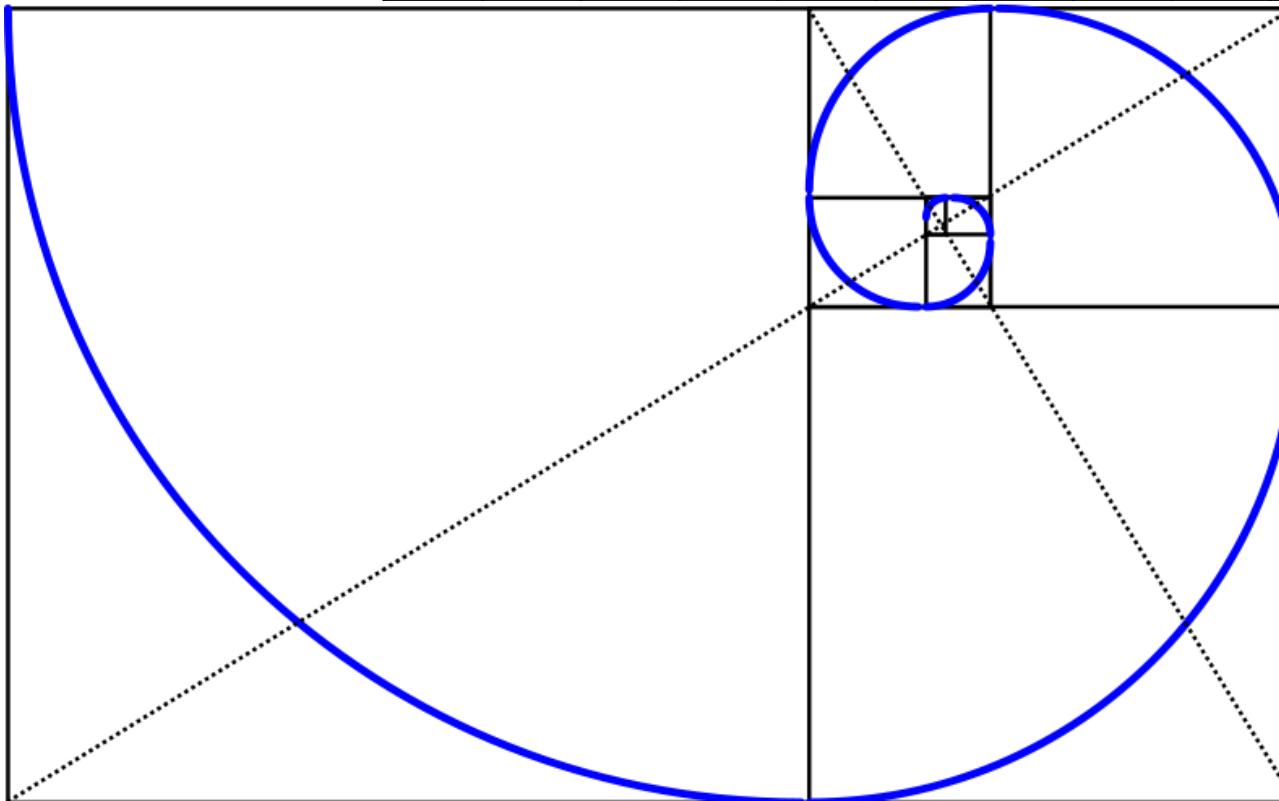
Fibonacci

0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89,



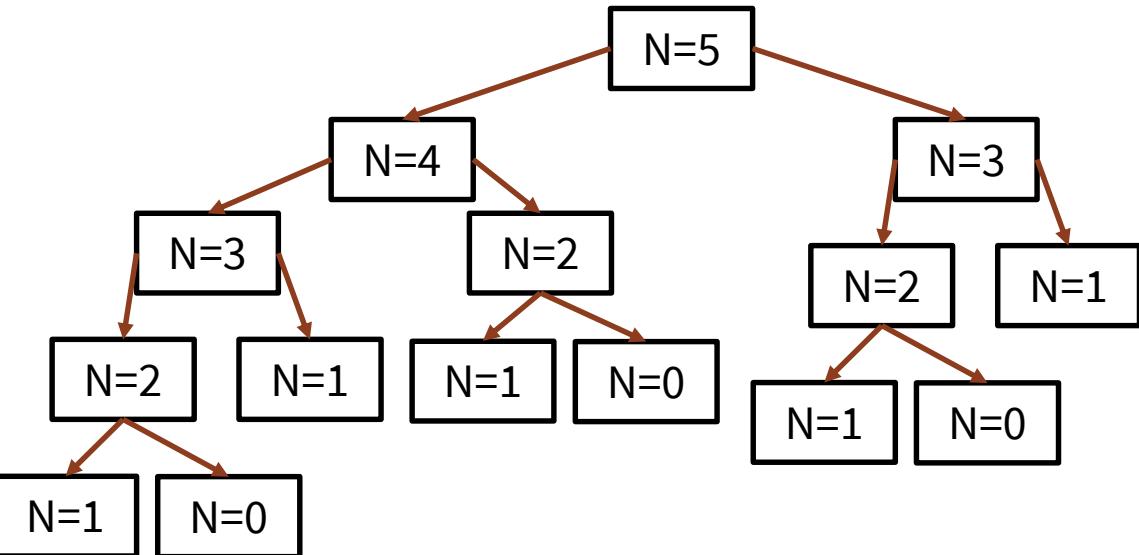
Fibonacci

0	1	2	3	4	5	6	7	8	9	10	11
0	1	1	2	3	5	8	13	21	34	55	89



Fibonacci

```
int fib(int n)
{
    if (n == 0) {
        return 0;
    } else if (n == 1)
        return 1;
    } else {
        return fib(n - 1) + fib(n - 2);
    }
}
```

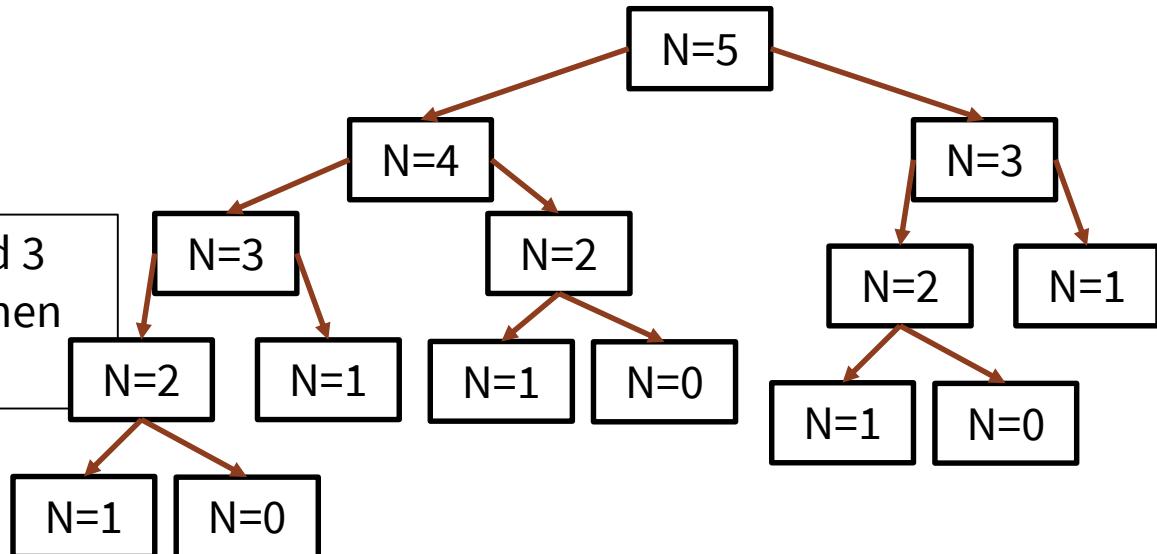


Work is duplicated throughout the call tree

- fib(2) is calculated 3 separate times when calculating fib(5)!
- 15 function calls in total for fib(5)!

Fibonacci

fib(2) is calculated 3 separate times when calculating fib(5)!



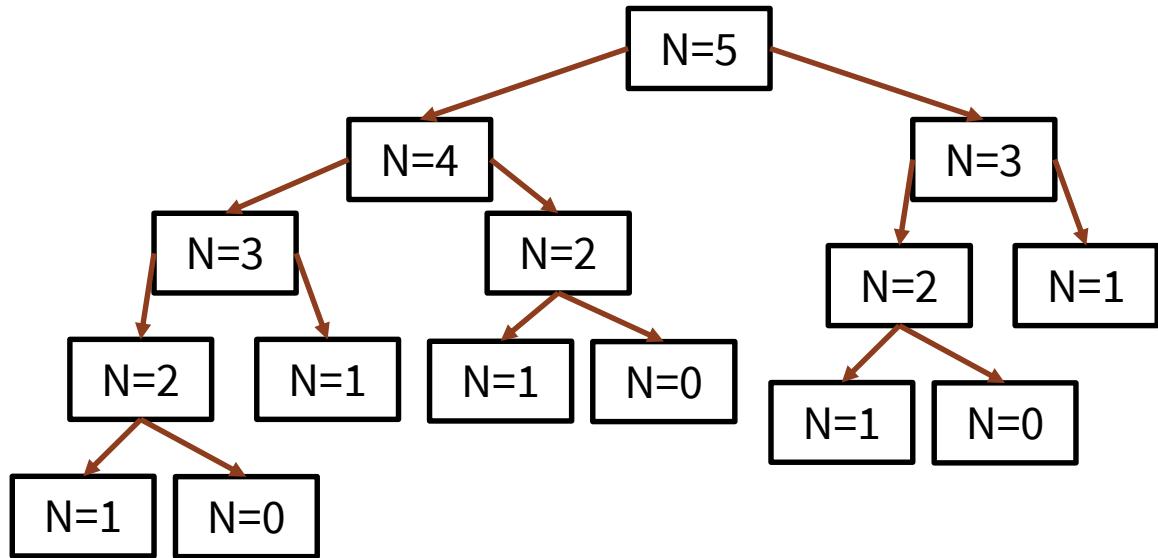
How many times would we calculate $\text{fib}(2)$ while calculating $\text{fib}(6)$?

See if you can just “read” it off the chart above.

- A. 4 times
- B. 5 times
- C. 6 times
- D. Other/none/more

Fibonacci

N	$\text{fib}(N)$	# of calls to $\text{fib}(2)$
2	1	1
3	2	1
4	3	2
5	5	3
6	8	5
7	13	8
8	21	13
9	34	21
10	55	34



Efficiency of naïve Fibonacci implementation

When we **added 1** to the input N, the number of times we had to calculate fib(2) **nearly doubled** ($\sim 1.6^*$ times)

- Ouch!

* This number is called the “Golden Ratio” in math—cool!

Goal: predict how much time it will take to compute for arbitrary input N.

Calculation: “approximately” $(1.6)^N$

Big-O Performance Analysis

A WAY TO COMPARE THE
NUMBER OF STEPS TO RUN
THESE FUNCTIONS



Big-O analysis in computer science

The Stanford libcs106 library, Fall Quarter 2022

```
#include "vector.h"

class Vector<ValueType>
```

This class stores an ordered list of values similar to an array. It supports traditional array selection using square brackets, as well as inserting and removing elements. Operations that access elements by index run in O(1) time. Operations, such as insert and remove, that must rearrange elements run in O(N) time.

Constructor

<u>Vector()</u>	O(1)	Initializes a new empty vector.
<u>Vector(n, value)</u>	O(N)	Initializes a new vector storing n copies of the given value.

Methods

<u>add(value)</u>	O(1)	Adds new value to the end of this vector.
<u>clear()</u>	O(1)	Removes all elements from this vector.
<u>equals(vec)</u>	O(N)	Returns true if the two vectors contain the same elements in the same order.
<u>get(index)</u>	O(1)	Returns the element at the specified index in this vector.
<u>insert(index, value)</u>	O(N)	Inserts a new value at the specified index in this vector.
<u>remove(index)</u>	O(N)	Removes the element at the specified index in this vector.

Big-O analysis in computer science

Binary search algorithm - Wikipedia

en.wikipedia.org/wiki/Binary_search_algorithm

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Binary search algorithm

From Wikipedia, the free encyclopedia

This article is about searching a finite sorted array. For searching continuous function values, see [bisection method](#).

In computer science, **binary search**, also known as **half-interval search**,^[1] **logarithmic search**,^[2] or **binary chop**,^[3] is a search algorithm that finds the position of a target value within a [sorted array](#).^{[4][5]} Binary search compares the target value to the middle element of the array. If they are not equal, the half in which the target cannot lie is eliminated and the search continues on the remaining half, again taking the middle element to compare to the target value, and repeating this until the target value is found. If the search ends with the remaining half being empty, the target is not in the array.

Binary search algorithm

Visualization of the binary search algorithm where 7 is the target value

1	3	4	6	7	8	10	13	14	18	19	21	24	37	40	45	75
---	---	---	---	---	---	----	----	----	----	----	----	----	----	----	----	----

Class
Data structure
Worst-case performance
Best-case performance
Average performance

Search algorithm
Array
 $O(\log n)$
 $O(1)$
 $O(\log n)$



A red arrow points from the text "Binary search compares the target value to the middle element of the array." to the "Data structure" section of the sidebar. A red circle highlights the "Average performance" section of the sidebar.

Formal definition of big-O

We say a function $f(n)$ is “big-O” of another function $g(n)$
(written $f(n) = O(g(n))$)
if and only if

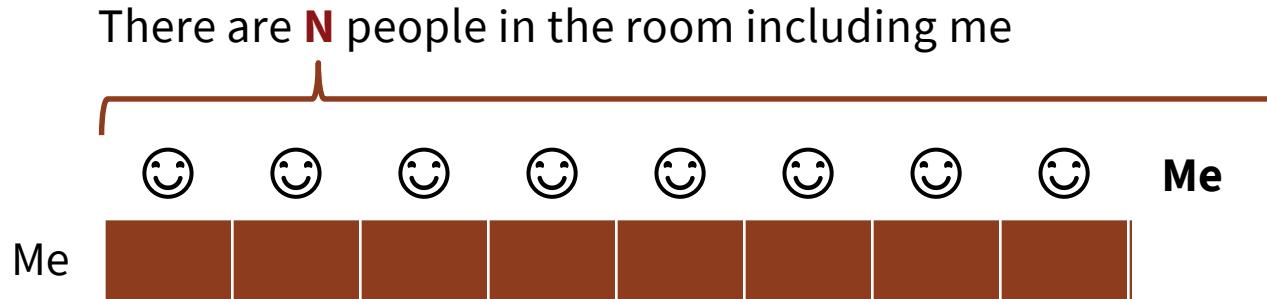
there exist positive constants c and n_0 such that
$$f(n) \leq c \cdot g(n) \text{ for all } n \geq n_0.$$

Before we start, let's get introduced

Before we start, let's get introduced

Lets say I want to meet each of you today with a handshake and *you tell me* your name...

How many introductions need to happen?



But do I need to shake hands with myself, or tell myself my name?

N-1 introductions

Putting this in Big-O terms

Big-O is a way of categorizing amount of work to be done in general terms, with a focus on:

- ***Rate of growth*** as a function of the problem size N
- What that rate looks like ***on the horizon*** (i.e., for large N)

Therefore, we don't really care about an insignificant ± 1



Putting this in Big-O terms

For the first handshake problem, the rate N is important and the -1 constant is not, so **N - 1** introductions becomes:

$$O(N-1)$$

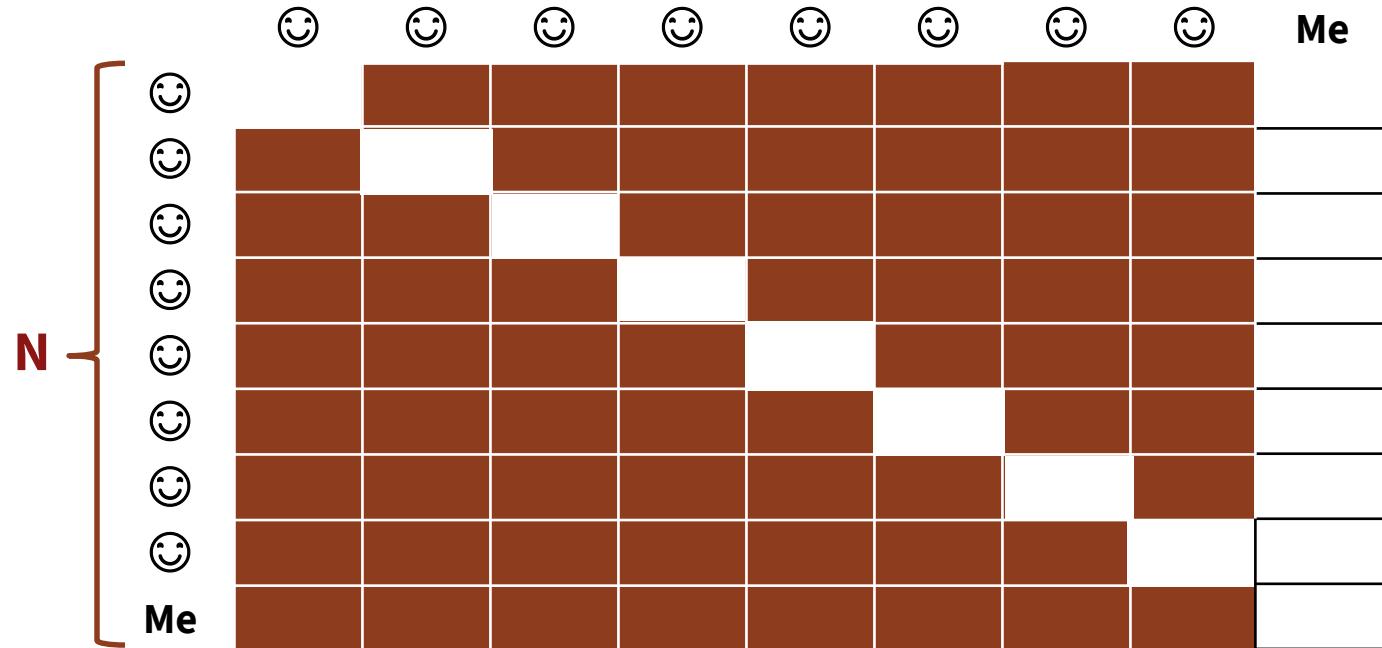
Similarly, if we said that each introduction **takes 3 seconds**, the amount of time is **$3(N - 1) = 3N - 3$** , but we disregard the constant 3s:

$$O(3N - 3)$$

Before we start, let's get introduced

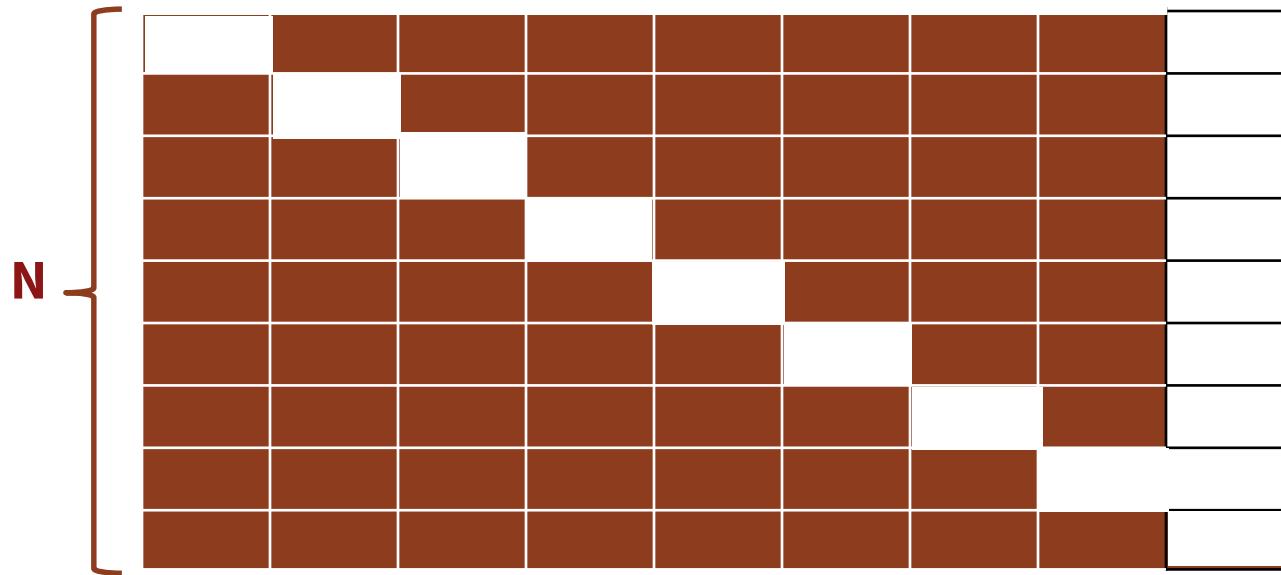
What if I not only want you to be introduced to me, but to each other?

Now how many introductions? N^2



Before we start, let's get introduced

What if I not only want you to be introduced to me, but to each other?
Now how many introductions? $N^2 - 2N + 1$



Putting this in Big-O terms

For the second handshake problem, the introductions was $N^2 - N$:

$$O(N^2 - 2N + 1)$$

But wait, didn't we just say that a term of $+/- N$ was important?

For Big-O, we only care about the **largest term** of the polynomial

Big-O and Binary Search

SPOILER: FAST!!



Binary search



Jump right to the middle of the region to search, then repeat this process of roughly cutting the array in half again and again until we either find the item or (worst case) cut it down to nothing.

Worst case cost is number of times we can divide length in half:

$$O(\log_2 N)$$

Putting it all together

Binary search

Handshake #1

Handshake #2

MANY important
optimization and
other problems

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	
3	8	24	64	256
4	16	64	256	65,536
5	32	160	1,024	4,294,967,296
6	64			
7	128			
8	256			
9	512			
10	1,024			
30	2,700,000,000			

Naïve
Recursive
Fibonacci
($O(1.6^n)$)

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
3	8	24	64	256
4	16	64	256	65,536
5	32	160	1,024	4,294,967,296
6	64			2.4s
7	128			Easy!
8	256			
9	512			
10	1,024			
30	2,700,000,000			



Traveling Salesperson Problem:
We have a bunch of cities to visit. In what order should we visit them to minimize total travel distance?



Traveling Salesperson Problem:

We have a bunch of cities to visit. In what order should we visit them to minimize total travel distance?





Exhaustively try all orderings: $O(n!)$

Use current best known algorithm: $O(n^2n)$

Maybe we could invent an algorithm that fits in our rightmost column: $O(2^n)$





So let's say we come up with a way to solve Traveling Salesperson Problem in $O(2^n)$.

It would take 4 days to solve Traveling Salesperson Problem on 50 state capitals.

Two *tiny* little updates

Imagine we approve statehood for US territory Puerto Rico

- Add San Juan, the capital city

Also add Washington, DC



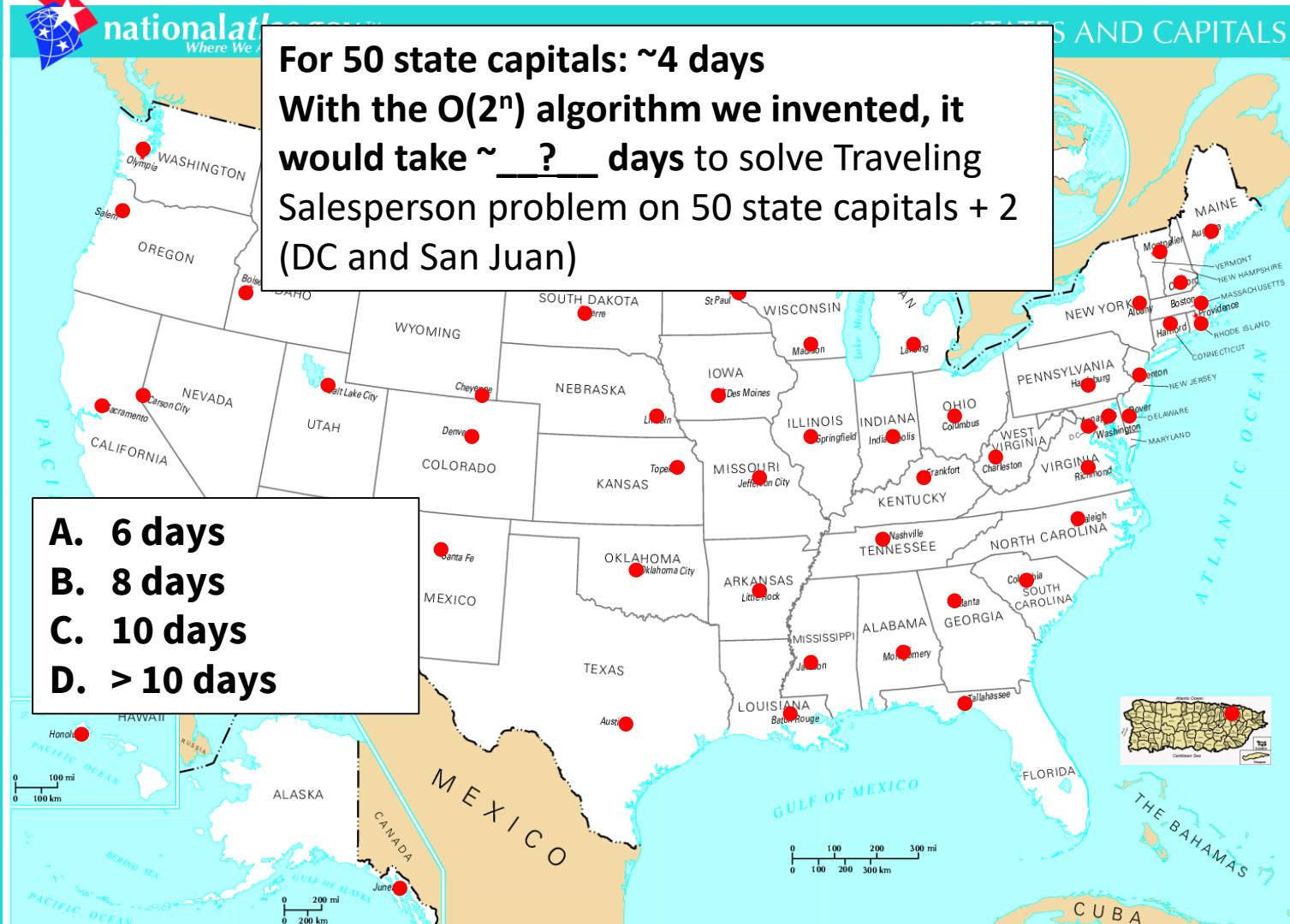
This work has been released into the [public domain](#) by its author, [Madden](#).
This applies worldwide.

Now 52 capital cities instead of 50



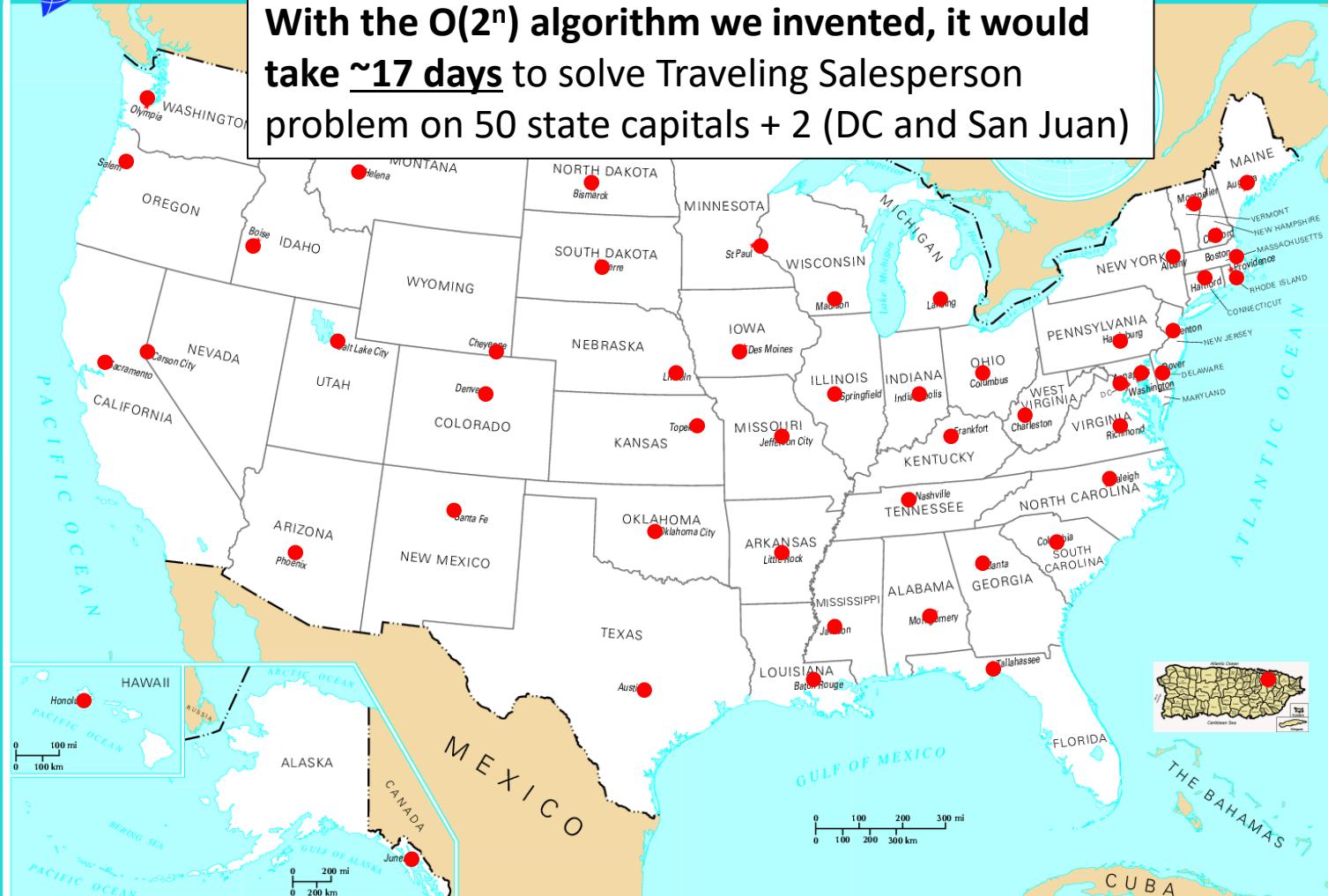
For 50 state capitals: ~4 days
With the $O(2^n)$ algorithm we invented, it would take ~ ? days to solve Traveling Salesperson problem on 50 state capitals + 2 (DC and San Juan)

- A. 6 days
- B. 8 days
- C. 10 days
- D. > 10 days





With the $O(2^n)$ algorithm we invented, it would take ~17 days to solve Traveling Salesperson problem on 50 state capitals + 2 (DC and San Juan)





Sacramento is not exactly the most interesting or important city in California (sorry, Sacramento).

What if we add the 12 biggest non-capital cities in the United States to our map?





With the $O(2^n)$ algorithm we invented,
It would take **194 YEARS** to solve Traveling
Salesman problem on 64 cities (state capitals +
DC + San Juan + 12 biggest non-capital cities)



$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
3	8	24	64	256
4	16	64	256	65,536
5	32	160	1,024	4,294,967,296
6	64	384	4,096	1.84×10^{19}
7	128			194 YEARS
8	256			
9	512			
10	1,024			
30	2,700,000,000			

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
3	8	24	64	256
4	16	64	256	65,536
5	32	160	1,024	4,294,967,296
6	64	384	4,096	1.84×10^{19}
7	128	896	16,384	3.40×10^{38}
8	256			3.59E+21 YEARS
9	512			
10	1,024			
30	2,700,000,000			

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
3	8	24	64	256
4	16	64	256	65,536
5	32	160	1,024	4,294,967,296
6	64	384	4,096	1.84×10^{19}
7	128	896	16,384	3.40×10^{38}
8	256	3,590,000,000,000,000,000,000 YEARS		
9	512			
10	1,024			
30	2,700,000,000			

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
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8	256	2,048	65,536	1.16×10^{77}
9	512			
10	1,024			
30	2,700,000,000			

For comparison: there are about 10^{80} atoms in the universe. No big deal.

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2	4	8	16	16
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8	256	2,048	65,536	1.16×10^{77}
9	512	4,608	262,144	1.34×10^{154}
10	1,024			1.42E+137 YEARS
30	2,700,000,000			

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
3	8	24	64	256
4	16	64	256	65,536
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9	512	4,608	262,144	1.34×10^{154}
10	1,024	10,240 (.000003s)	1,048,576 (.0003s)	1.80×10^{308}
30	2,700,000,000	84,591,843,105 (28s)	7,290,000,000,000,000 (77 years)	LOL

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
3	8	24	64	256
4	16	64	256	65,536
5	32	160	1,024	4,294,967,296
6	64	384	4,096	1.84×10^{19}
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10	1,024	10,240 (.000003s)	1,048,576 (.0003s)	1.80×10^{308}
31	2,700,000,000	84,591,843,105 (28s)	7,290,000,000,000,000 00 (77 years)	$1.962227 \times 10^{812,780,998}$

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2	4	8	16	16
3	8	24	64	256
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9	512	4,608	262,144	1.34×10^{154}
30	2,700,000,000	84,591,843,105 (28s)	7,290,000,000,000,000 (77 years)	1.962227 x $10^{812,780,998}$

2^n is clearly infeasible, but look at
 $\log_2 n$ —only a tiny fraction of a second!

In Conclusion

- **NOT worth doing:** Optimization of your code that **just trims** a bit
 - › Like that +/-1 handshake—we don't need to worry ourselves about it!
 - › Just write clean, easy-to-read code!!!!
- **MAY be worth doing:** Optimization of your code that **changes Big-O**
 - › If performance of a particular function is important, focus on this!
 - › (*but if performance of the function is not very important, for example it will only run on small inputs, focus on just writing clean, easy-to-read code!!*)
- (Also remember that efficiency is not necessarily a virtue—first and foremost focus on correctness, both technical and ethical/moral/societal justice)