

Programming Abstractions

CS106B

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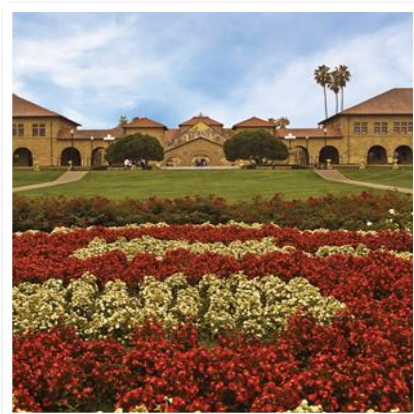
Today's Topics

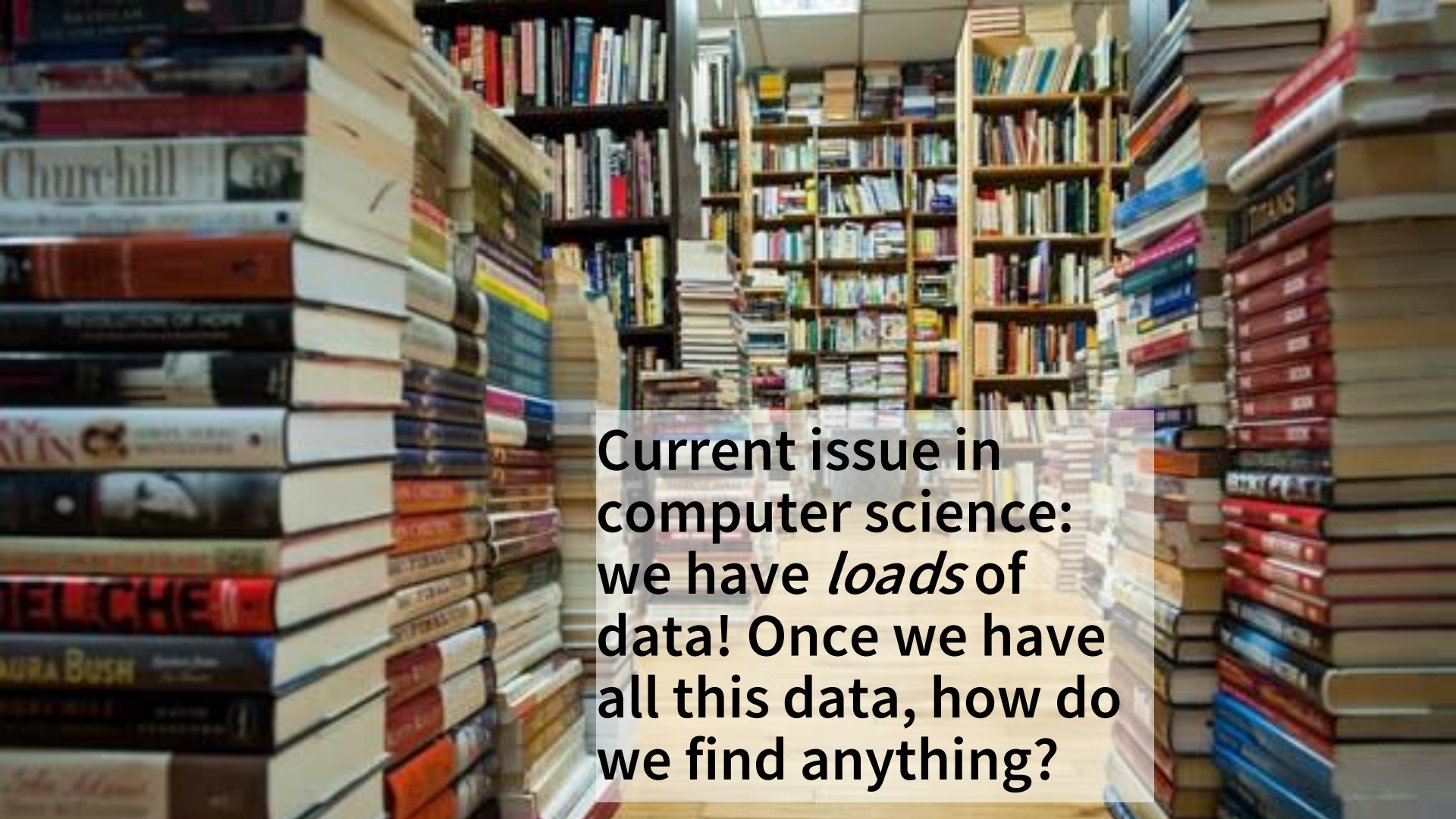
- Recursion!
 - › Algorithm performance analysis with Big-O
- Next time:
 - › More recursion! It's Recursion Week!
 - › Like Shark Week, but more nerdy
- For important announcements, be sure to see the weekly announcements post on the Ed Q&A board! <https://edstem.org>
- Also on Ed: live lecture Q&A with Chris & Jonathan



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 $\frac{1}{2}$ 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!

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

Now we could do our Basic Approach, but in **half the time.**
Thanks, Efficiency Hack!!

...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!!**

Basic Recursive Function Design Pattern

Always two parts:

Base case:

- This problem is so tiny, it's hardly a problem anymore! Just give answer.

Recursive case:

- This problem is still a bit large, let's (1) bite off just one piece, and (2) delegate the remaining work to 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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    } else if (key < data[midpoint]) {  
        return binarySearch(data[first half only], key);  
    } else {  
        return binarySearch(data[second half only], key);  
    }  
}
```

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

Recursive case:

Do one piece of work (comparison)

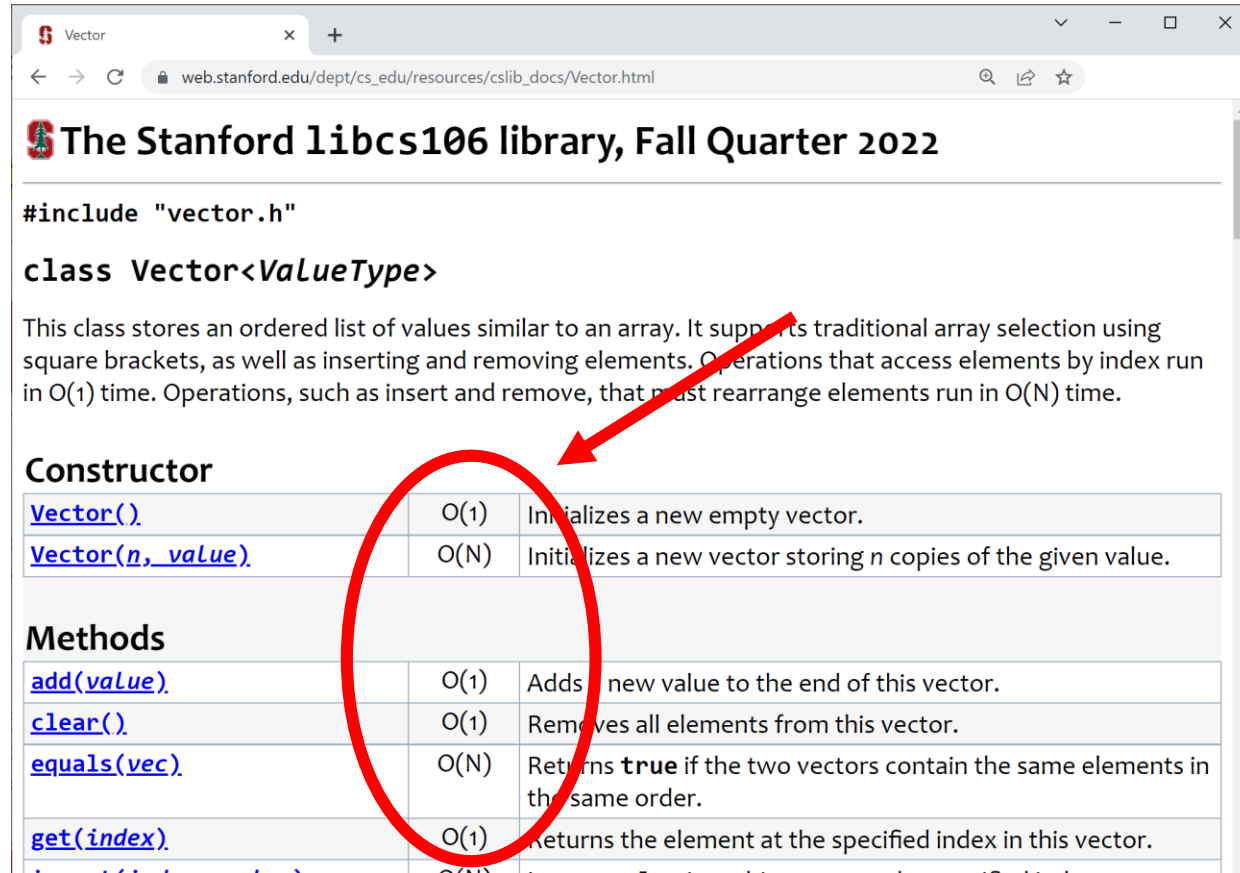
Delegate the rest of the work

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

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

Methods

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

Big-O analysis in computer science

W Binary search algorithm - Wikipe x +

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



Class	Search algorithm
Data structure	Array
Worst-case performance	$O(\log n)$
Best-case performance	$O(1)$
Average performance	$O(\log n)$

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

$$f(n) \leq c \cdot g(n) \text{ for all } n \geq n_0.$$

c says “we don’t care
about constant
coefficients”

n_0 says “we only care
about performance on
big data sets”

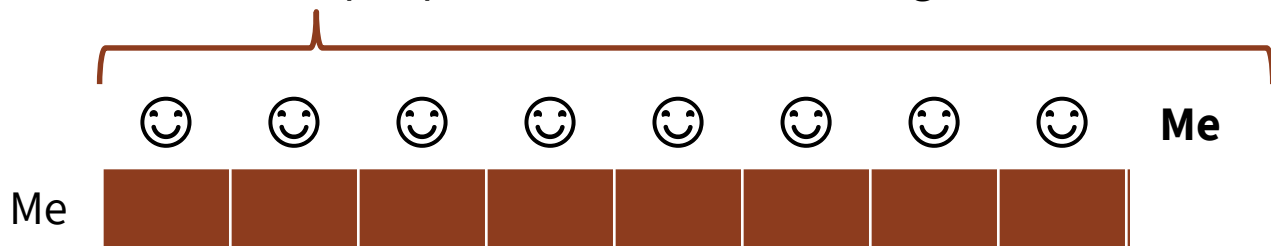
Before we start, let's get introduced

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?

There are **N** people in the room including me



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:

$$O(N - 1) \rightarrow O(N)$$

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:

$$O(3N - 3) \rightarrow O(N)$$

Before we start, let's get introduced

What if I not only want you to be introduced to me, but to each other? (Note: I don't need to tell anyone my name, and nobody needs to tell themselves their own name, so subtract those out.)

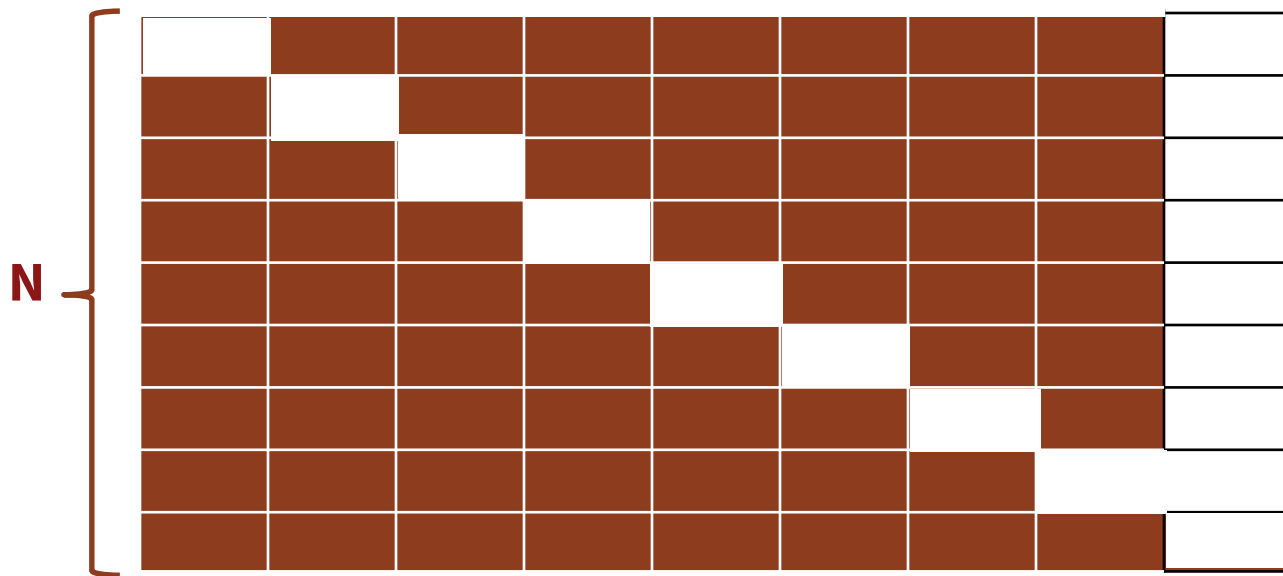
Now how many introductions? N^2

	😊	😊	😊	😊	😊	😊	😊	😊	Me
😊									
😊									
😊									
😊									
😊									
😊									
😊									
😊									
Me									

Before we start, let's get introduced

What if I not only want you to be introduced to me, but to each other? (Note: I don't need to tell anyone my name, and nobody needs to tell themselves their own name, so subtract those out.)

Now how many introductions? $(N - 1)^2 = N^2 - 2N + 1$



Putting this in Big-O terms

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

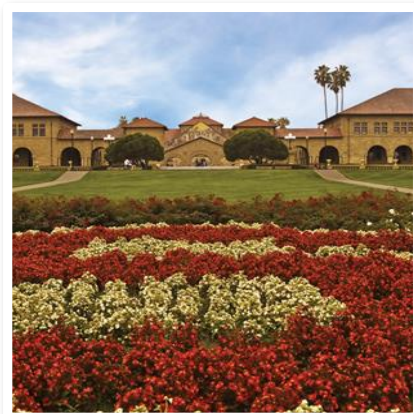
$$O(N^2 - 2N + 1) \rightarrow O(N^2)$$

But wait, didn't we just say that a term of $\pm N$ was important?

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

Big-O and Binary Search

SPOILER: FAST!!



Binary search

2	7	8	13	25	29	33	51	89	90	95
---	---	---	----	----	----	----	----	----	----	----

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:

$O(n^2 2^n)$

Maybe we could invent an algorithm that fits in our
rightmost column:

$O(2^n)$





Two *tiny* little updates

Imagine we approve statehood for US territory Puerto Rico

- Add San Juan, the capital city

Also add Washington, DC

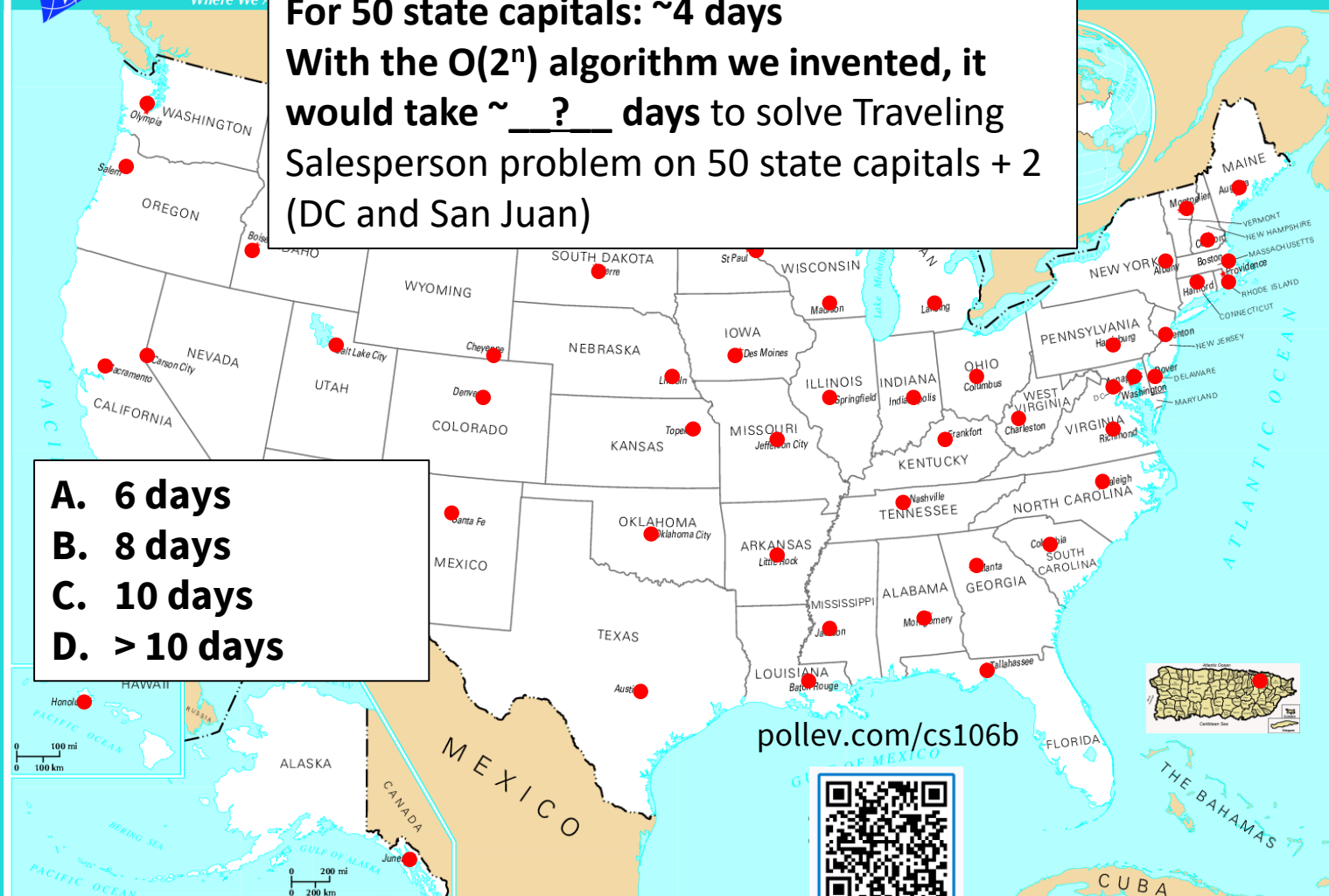


This work has been released into the [public domain](#) by its author, [Madden](#).
This applies worldwide.

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)

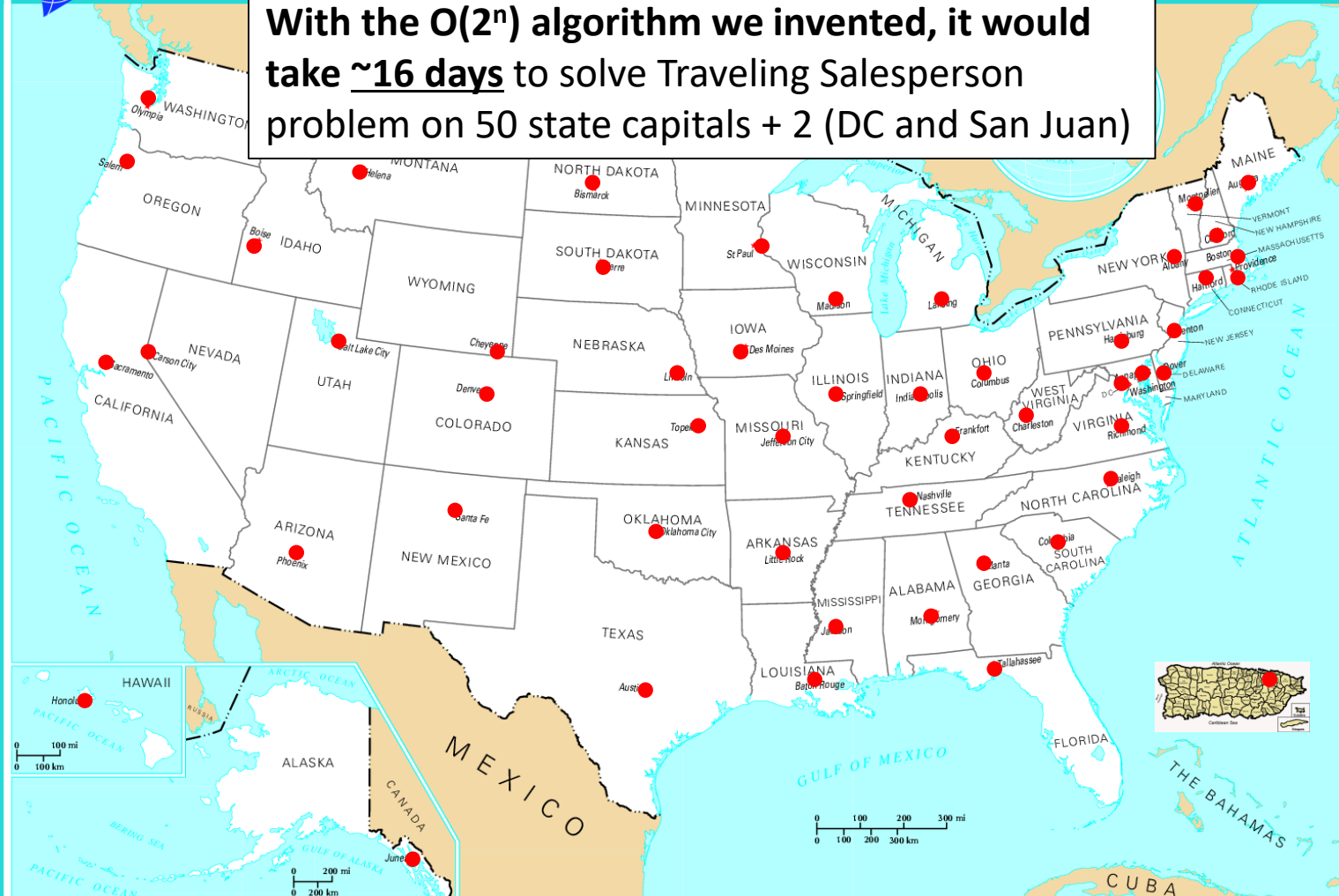
- A. 6 days
- B. 8 days
- C. 10 days
- D. > 10 days



pollev.com/cs106b



With the $O(2^n)$ algorithm we invented, it would take ~16 days to solve Traveling Salesperson problem on 50 state capitals + 2 (DC and San Juan)





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			
9	512			
10	1,024			
30	2,700,000,000			

3.59E+21 YEARS

$\log_2 n$	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			
9	512			
10	1,024			
30	2,700,000,000			

**3,590,000,000,000,000,000,000
YEARS**

$\log_2 n$	n	$n \log_2 n$	n^2	2^n
2	4	8	16	16
3	8	24	64	256
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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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9	512	4,608	262,144	1.34×10^{154}
10	1,024			
30	2,700,000,000			

1.42E+137 YEARS

$\log_2 n$	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
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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)	7,290,000,000,000,000,000 (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
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31	2,700,000,000	84,591,843,105 (28s)	7,290,000,000,000,000, 000 (77 years)	$1.962227 \times 10^{812,780,998}$

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30	2,700,000,000	84,591,843,105 (28s)	7,290,000,000,000,000,000 (77 years)	$1.962227 \times 10^{812,780,998}$

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!!!!
- **COULD 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, just focus on writing clean, easy-to-read code!!)*
- (Also remember that efficiency is not necessarily a virtue—first and foremost focus on correctness, both technical correctness and ethical/moral correctness)