

## A Possible Problem (Interview Question)

You got a list of items, where every item has a value and a weight. You got a bag that holds a maximum weight of X.

Write a program that maximizes the value of the items you put into the bag whilst ensuring that you don't exceed the maximum weight.

```
items = [  
  {id: 'a', val: 10, w: 3},  
  {id: 'b', val: 6, w: 8},  
  {id: 'c', val: 3, w: 3}  
]
```

```
maxWeight = 8
```

```
bag = ['a', 'c'] // solution
```

Knapsack problem

Value: 13  
Weight: 6 (< 8)

This is being asked to check your problem-solving skills.

# Algorithms: What and Why?

## An Algorithm

A sequence of steps (instructions) to solve a clearly defined problem

The same steps always lead to the same solution of a problem (given the same inputs)

Every program is an algorithm! Or:  
Every program consists of many smaller algorithms

As a programmer, you need to be able to solve problems (efficiently)!

# What is the “Best Possible Solution”?

Minimum amount of code?

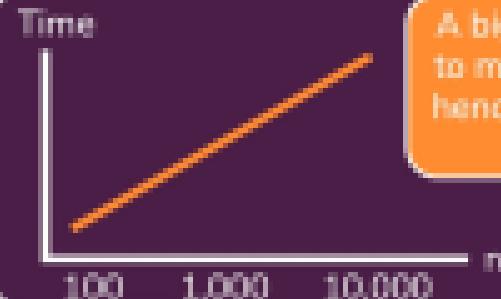
Best performance?

Least memory usage?

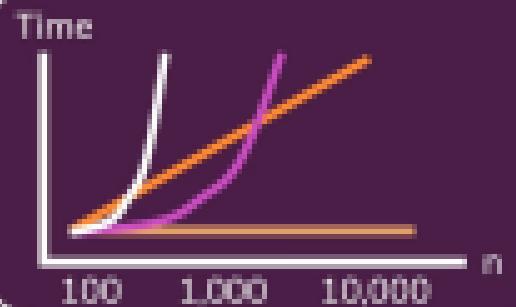
Personal preference?

# Measuring Performance (Time Complexity – Big O)

```
function sumUp(n) {  
    let result = 0;  
    for (let i = 1; i <= n; i++) {  
        result += i;  
    }  
    return result;  
}
```



A bigger number leads to more loop iterations, hence time increases in a linear way.



Linear Time

$O(n)$

Constant Time

$O(1)$

Quadratic Time

$O(n^2)$

Cubic Time

$O(n^3)$

We care about the trend/kind of function.

Big O Notation

# Deriving the Time Complexity Function

```
function sumUp(n) {  
    let result = 0;  
  
    for (let i = 1; i <= n; i++) {  
  
        result += i;  
  
    }  
    return result;  
}
```

n = 1

n = 3

n = 10

n = n

1

1

1

1

1

1

1

1

1

3

10

n

1

1

1

1

Count the number of expression executions.

$$T = 1 + 1 + n + 1 = 3 + n = 3 + 1 * n$$

# Deriving Big O (Asymptotic Analysis)



1

Define the function

$$T = \alpha n + b$$

2

Find the fastest growing term

$$T = \boxed{\alpha n} + b$$

3

Remove the coefficient

$$T = \cancel{\alpha} n$$

 $O(n)$ 

$$T = n$$

# Deriving Constant Time Complexity

```
function sumUp(n) {  
    return (n / 2) * (n + 1);  
}
```

n = 1

n = 3

n = 10

n = n

1

1

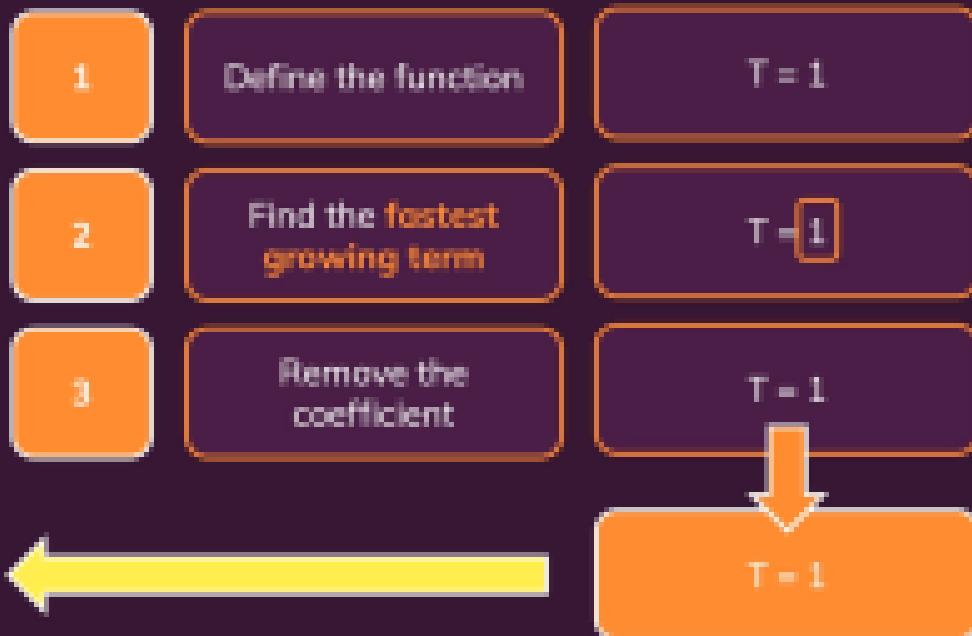
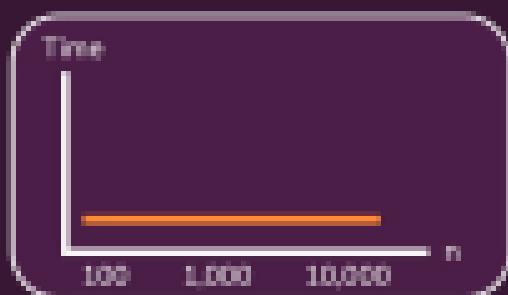
1

1

Count the number of expression executions.

T = 1

# Deriving Big O (Asymptotic Analysis)



# Using Big O to Compare Algorithms

$O(1)$



Constant Time Complexity

$n$  (number of input) has no effect on the time the algorithm takes

$O(\log n)$



Logarithmic Time Complexity

Execution time grows logarithmically with  $n$

$O(n)$



Linear Time Complexity

Execution time grows linearly with  $n$

$O(n^2)$



Quadratic Time Complexity

Execution time grows quadratically with  $n$

$O(2^n)$



Exponential Time Complexity

Execution time grows exponentially with  $n$

## Practice Time!

Write an algorithm that takes an **array of numbers** as input and calculates the sum of these numbers.

Define the Time Complexity of that algorithm and determine what the lowest possible Time Complexity is for this problem.

```
function sumNumbers(numbers) { ??? }
```

```
sumNumbers([1, 3, 10]) // should yield 14
```

Your task!

# About this Course

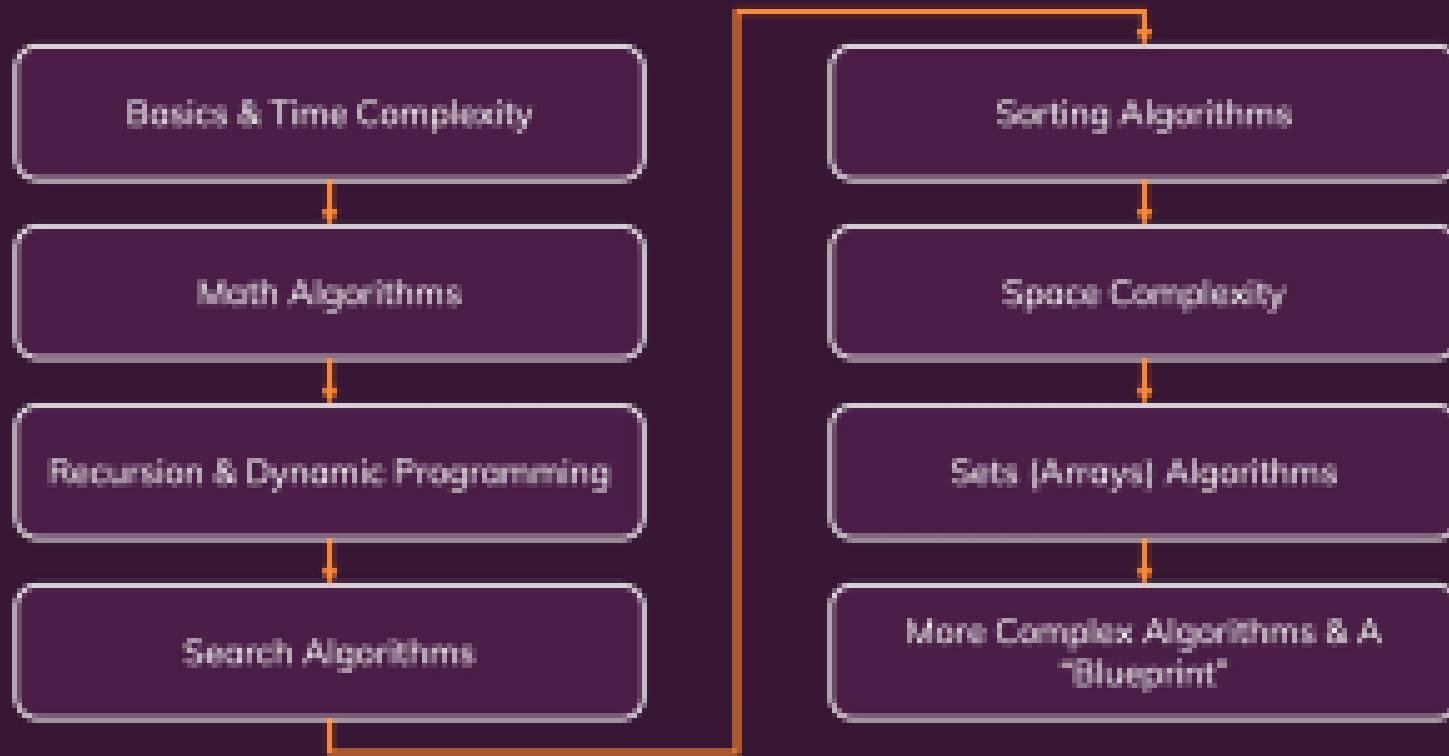
What & Why

Examples & Different Algorithms

Different Solution Approaches: Recursion,  
Dynamic Programming, Greedy  
Algorithms

A Solid Foundation & Plan

# Course Outline



# Math Algorithms

Fun With Numbers

# Module Content

Explore Math Algorithms

Get a Better Feeling for Big O Notation

How to Solve Problems

# Fibonacci Sequence

Starts with 1, 1

The Fibonacci Sequence

[1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, 144, ...]

Every other element is the sum of its two preceding elements.

Popular Interview Question

Return the  $n$ th element of the Fibonacci sequence.  
e.g.  $\text{fib}(4)$  yields 5

Algorithm needs to do two things:

- (1) Calculate the sequence up to the element we're looking for
- (2) Return that element

# Primality Test

Determine whether an input number is a prime number

```
isPrime(9); // false  
isPrime(5); // true
```

Algorithm needs to do one thing:

- (1) Try dividing the number by all smaller numbers and return true if it's only divisible by itself and 1.

# Big O – Best Case, Worst Case, Average Case

For some problems (e.g. sorting array items), you have different cases with different time complexities

Example: "Sort the numbers in array from small to big"  
Algorithm used: "Bubble Sort" (covered later)

Best Case: Already sorted  
nums = [1, 2, 3]

Avg. Case: Random order  
nums = [2, 3, 1]

Worst Case: Inverse order  
nums = [3, 2, 1]

$O(n)$

$O(n^2)$

$O(n^2)$

# Primality Test - Improved!

## Two Examples:

5 is a prime: Divisible by itself and by 1

9 is NOT a prime: Divisible by itself, by 1 and by 3

Every number, that's NOT a prime has a product that consists of two factors a & b that are both neither 1 nor the number itself.

$$9 = 3 \times 3$$

At least one factor is smaller or equal to the square root of the number.

$\sqrt{9} = 3 \rightarrow 9 = 3 \times 3 \rightarrow$  Both factors are equal to the square root

## Practice Time!

Write two algorithms:

- (1) The first algorithm should take an array of numbers as input and return the minimum value in the array (i.e. the smallest number)
- (2) The second (independent) algorithm should take a number as input and return "true" if it's an even number, "false" for odd numbers

Also define the time complexities and possible cases for both algorithms!

## Is Power Of Two

Determine whether an input number is a power of two

```
isPowerOfTwo(8); // true  
isPowerOfTwo(5); // false
```

Algorithm needs to do one thing:

- (1) Divide number and future division results by two, until 1 is reached and check if the remainder is always 0

# Is Power Of Two - Improved!

## Bitwise Magic!

(Unsigned) Powers of two, in binary form, always have just one bit:

1: 1

2: 10

4: 100

A bitwise operation can be used to check if a number is power of two:

```
number & (number - 1) === 0 // true: it's power of two
```

# Determining the “Nature of an Algorithm”

$O(n)$

Higher  $n$  leads to a linear increase in runtime

Look for (single) loops

$O(1)$

Higher  $n$  does not affect runtime

Look for functions without loops and without any function calls

$O(\log n)$

Runtime grows with  $n$  but at a much slower pace

Look for functions where  $n$  is split (divided) into smaller “chunks”

# Factorial

Calculate the factorial of a number

```
fact(3); // 3 * 2 * 1 = 6  
fact(5); // 5 * 4 * 3 * 2 * 1 = 120
```

Algorithm needs to do one thing:

- (1) Go through all smaller numbers and multiply them with each other (and with the input number)

# Recursion & Dynamic Programming

Beyond Iterations

## Module Content

What is "Recursion"?

Recursion in Algorithms (and why it's not always better!)

Dynamic Programming

# Factorial – Recursive Solution

Base Case

```
function fact(n) {  
    if (n <= 1) {  
        return 1;  
    }  
  
    return n * fact(n - 1);  
}
```

Recursive Step

fact(4)

4 \* fact(3)

3 \* fact(2)

2 \* fact(1)

1

4 \* 6

3 \* 2

2 \* 1

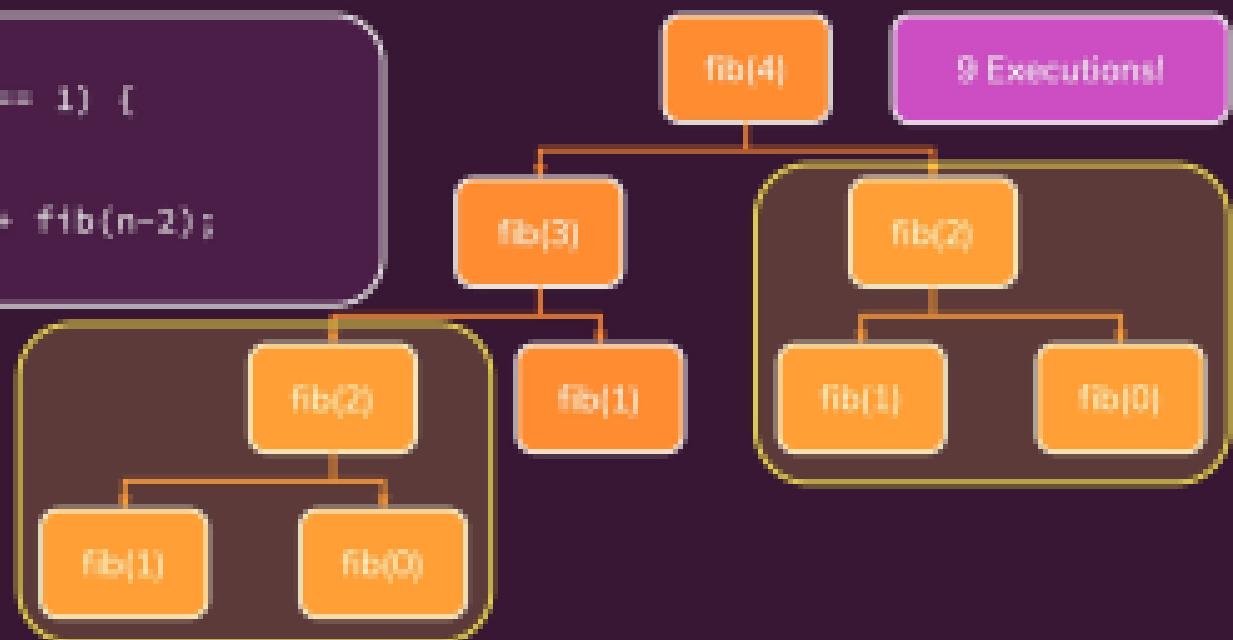
1

24

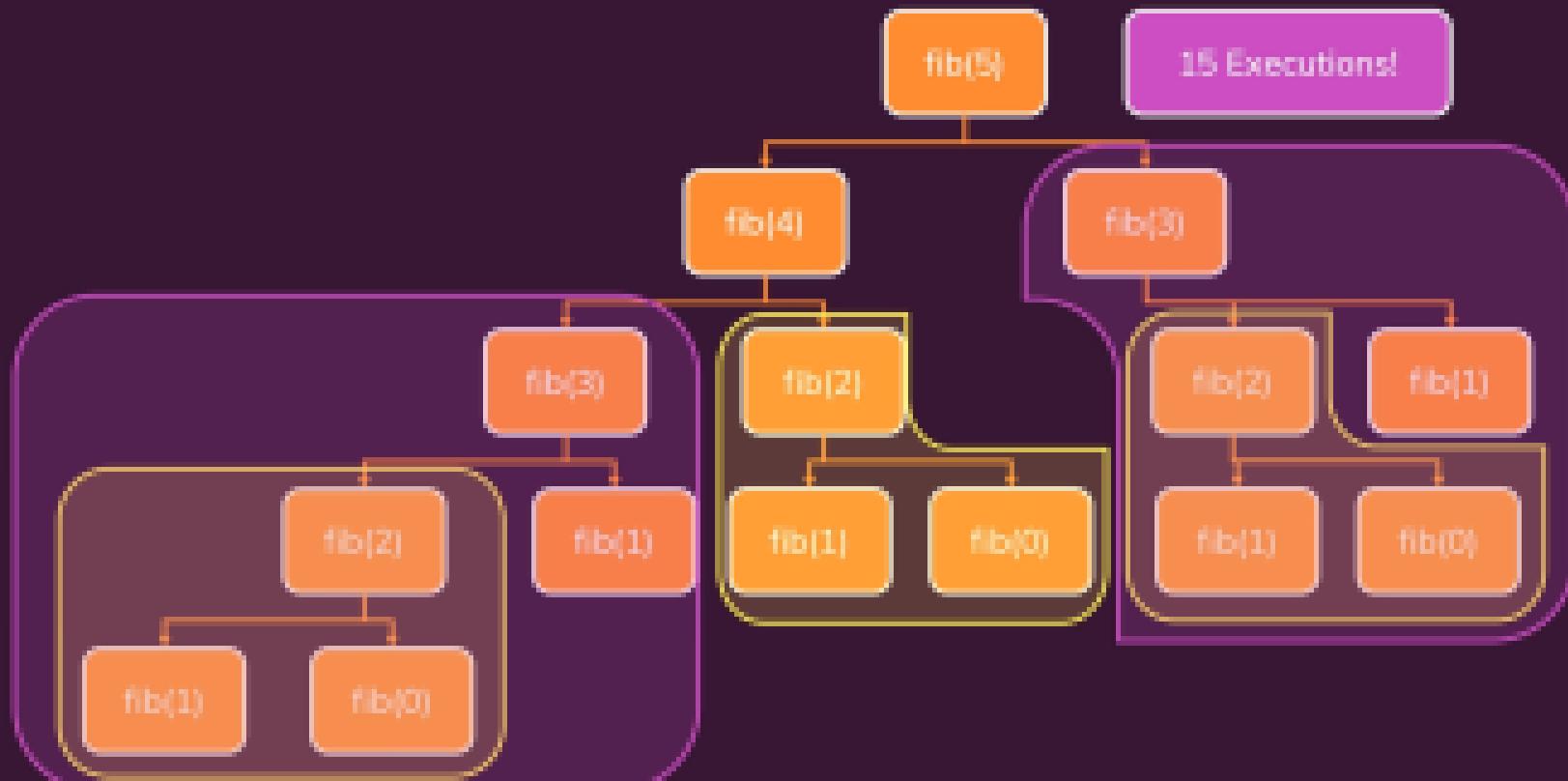
# Recursion Is Not Always Best

## Recursive Fibonacci

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



# Recursion Is Not Always Best

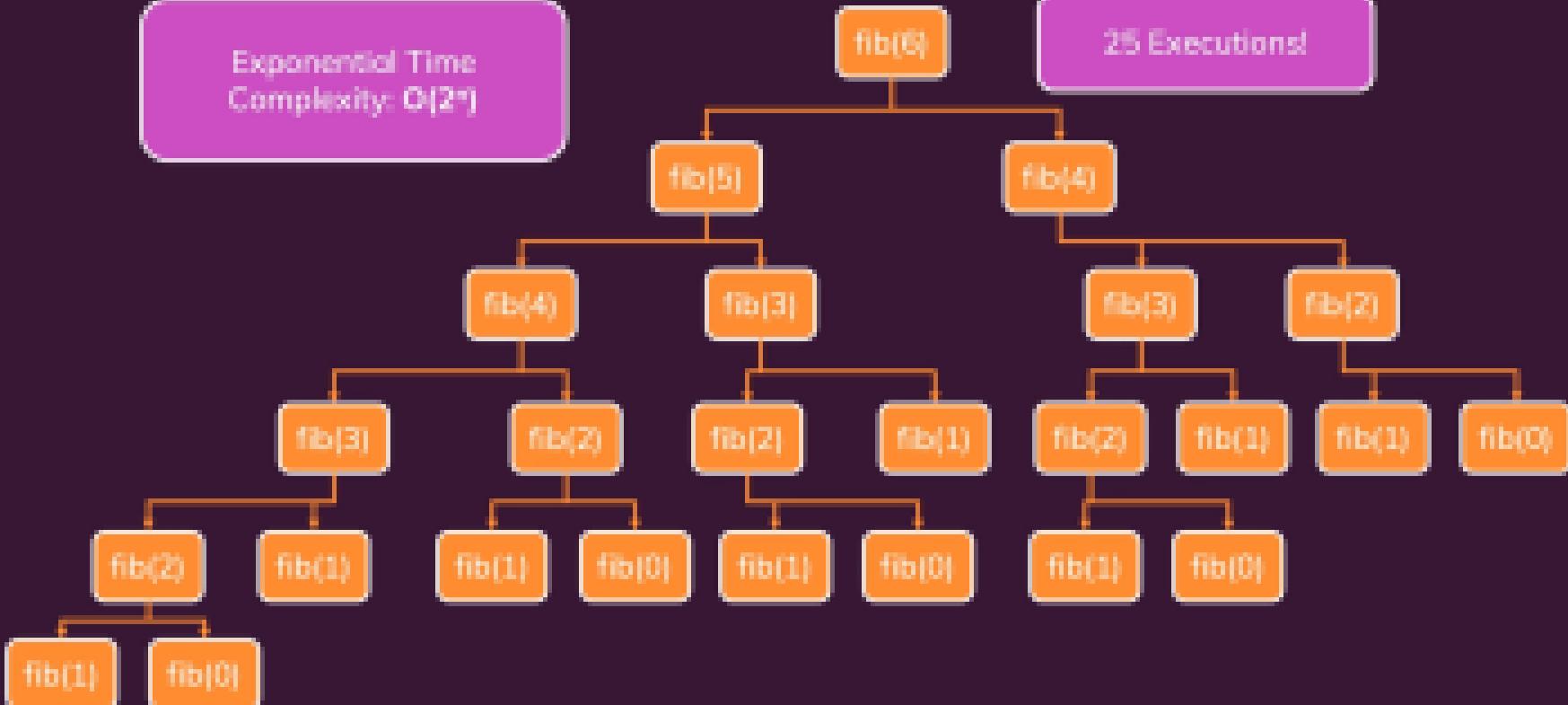


# Recursion Is Not Always Best

Exponential Time  
Complexity:  $O(2^n)$

fib(6)

25 Executions!



# What is "Dynamic Programming"?

Recursion



Stored Data ("Memoization")

Great for repeated computations

Avoid unnecessary recursive steps by storing data

Can lead to duplicate work though

Intermediate results are stored and re-used

Dynamic Programming

# What is "Dynamic Programming"?

Recursion



Stored Data ("Memoization")

Great for repeated computations

Avoid unnecessary recursive steps by storing data

Can lead to duplicate work though

Intermediate results are stored and re-used

Dynamic Programming

## Or: Use the "Bottom-Up Approach"

Memo



Build it up "from the bottom"

```
function fib(n) {  
    const numbers = [1, 1];  
    for (let i = 2; i < n + 1; i++) {  
        numbers.push(numbers[i - 2] + numbers[i - 1]);  
    }  
    return numbers[n];  
}
```

# Search Algorithms

Finding Things (Efficiently)

## Module Content

What are "Search Algorithms" all about?

Linear Search

Binary Search & The "Master Theorem"

## Search Algorithms?

3

10

-3

40

5

33

90

Looking for

5

# Search Algorithms?



You can use different algorithms for different kinds of lists – with different time complexities

# Linear Search



Linear search works on ordered and unordered lists and checks all items until it finds the element you're searching.

# Linear Search – Time Complexity

## Best Case

The item we're looking for is the very first item in the list.

 $O(1)$ 

## Average Case

Random order, we don't know where the item is.

Tends to be  $O(n)$

## Worst Case

The item we're looking for is the very last item in the list.

 $O(n)$

# Binary Search



Looking for

5

Doesn't work on unordered lists! You need  
to sort the list first!

# Binary Search



Looking for

5

Find median and compare it to the element you're trying to find

Is it the element you're looking for?



If element wasn't found, take the half in which must be inside

Repeat!

# Binary Search – Time Complexity

## Best Case

The item we're looking for is right in the middle

$O(1)$

## Average Case

We don't know where the item will be

Tends to be  $O(\log n)$

## Worst Case

The item we're looking for is at the beginning or end

$O(\log n)$

Because we split the array in half in every iteration

# Recursion & Big O (The Master Theorem)

How do you derive Big O for more complex recursive algorithms?

## Master Theorem

Runtime of recursion:  $\Theta(n^{\log_b a})$

Overall algorithm runtime (time complexity) - three cases:

Recursion does more work

$$\Theta(n^{log_b a})$$

Same work inside and outside of recursion

$$\Theta(n^{\log_b a} \log n)$$

Non-recursive part does more work

$$\Theta(f(n))$$

where

a equals the number of subproblems (number of recursion splits)

b equals the relative subproblem size (input reduction per split)

f(n) equals the runtime outside of the recursion

# Sorting Algorithms

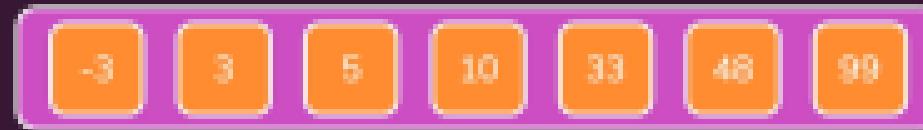
Everything Needs To Be In Order

## Module Content

What is "Sorting" all about?

Various Sorting Algorithms

# What is "Sorting"?



Goal: Sort items – for example to then apply search algorithms

## Bubble Sort

Compare two items at a time and sort them. Go through the entire array multiple times until all pairs were compared and sorted.

3

10

-3

48

5

33

99

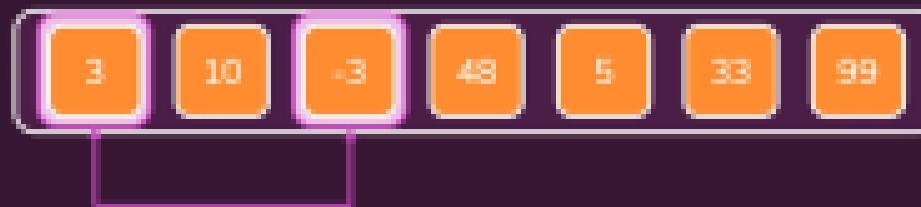
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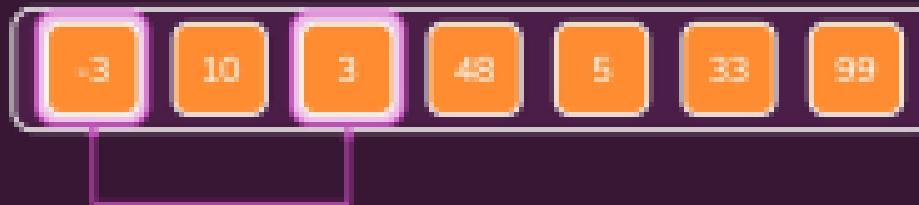
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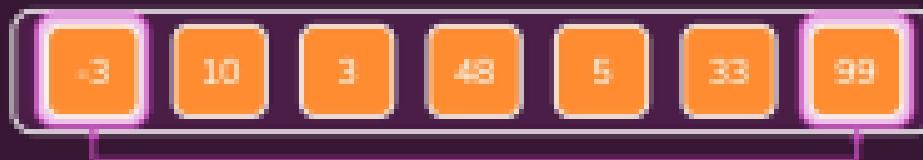
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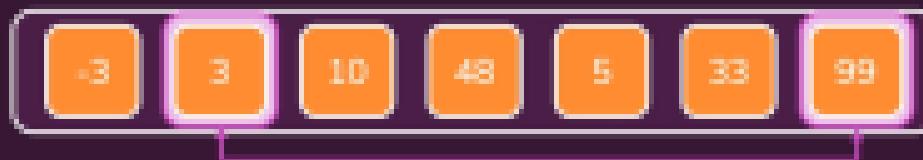
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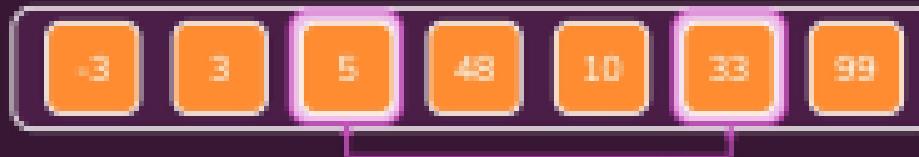
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## Bubble Sort

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-3

3

5

10

33

48

99

# Bubble Sort – Time Complexity

Best Case

Average Case

Worst Case

Items are already sorted

Random order, we don't know where the item is

Items are sorted in wrong order

$O(n)$

Tends to be  $O(n^2)$

$O(n^2)$

# Quicksort

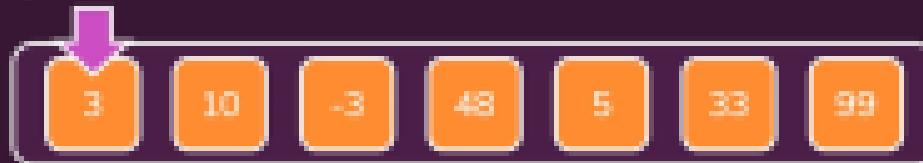
Use pivot elements to split array into smaller chunks – elements bigger, smaller and equal than the pivot element. Repeat that process for all chunks and concat the sorted chunks.



# Quicksort

Use pivot elements to split array into smaller chunks – elements bigger, smaller and equal than the pivot element. Repeat that process for all chunks and concat the sorted chunks.

First pivot element



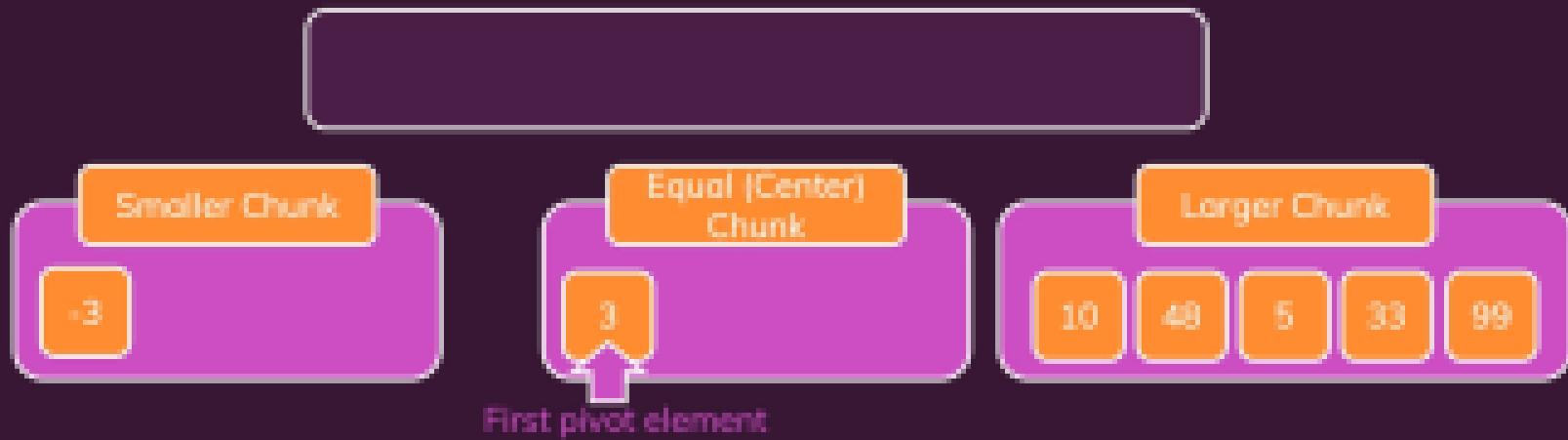
Smaller Chunk

Equal (Center)  
Chunk

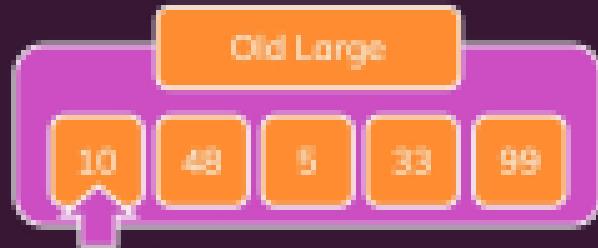
Longer Chunk

# Quicksort

Use pivot elements to split array into smaller chunks – elements bigger, smaller and equal than the pivot element. Repeat that process for all chunks and concat the sorted chunks.



# Quicksort



New pivot element

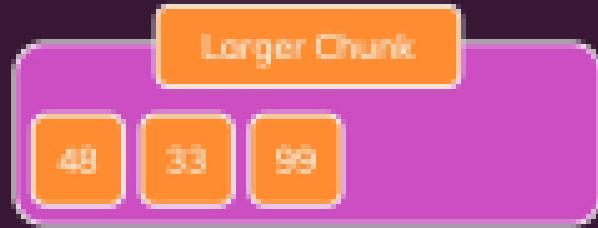
Repeat for every chunk (= array)



# Quicksort



Repeat for every chunk (= array)



New pivot element

# Quicksort

-3

3

5

10

33

48

99

# Quicksort



# Quicksort – Time Complexity

## Best Case

Items are sorted randomly  
(NOT in right or wrong  
order)

$O(n * \log n)$

## Average Case

Items are sorted randomly  
(NOT in right or wrong  
order)

$O(n * \log n)$

## Worst Case

Items are already sorted  
(order does not matter)

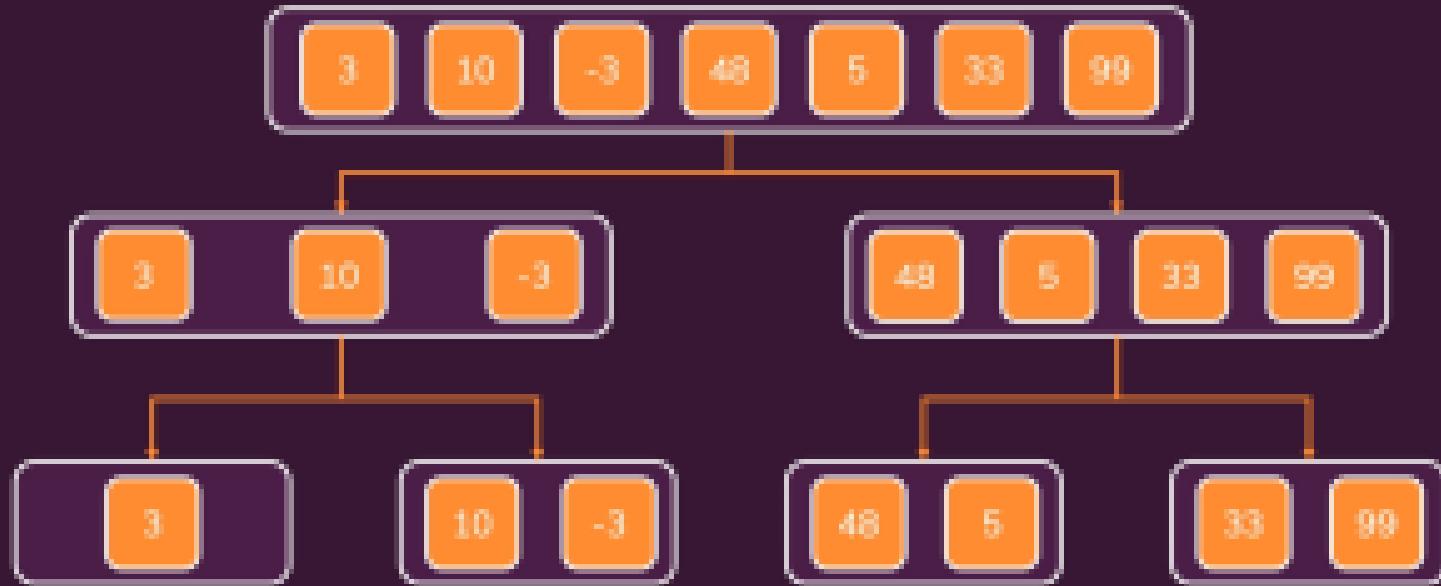
$O(n^2)$

# Merge Sort

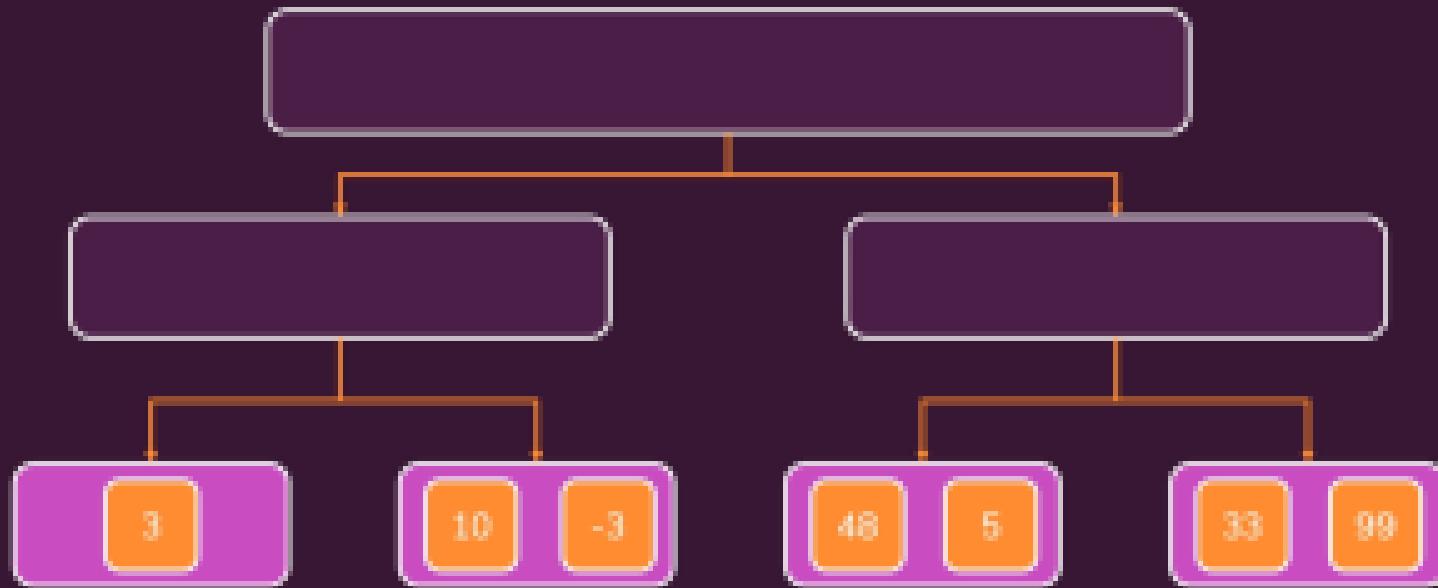
Split array multiple times until we have only 2-element arrays left – sort those arrays and merge them back together.



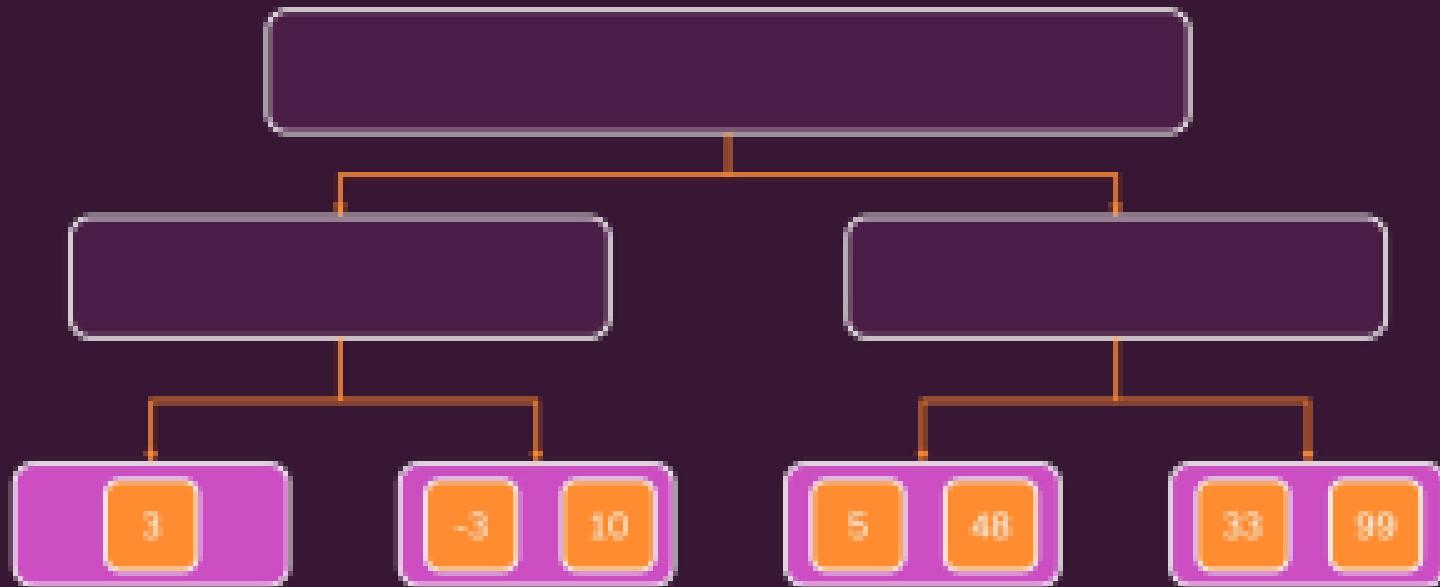
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# Merge Sort



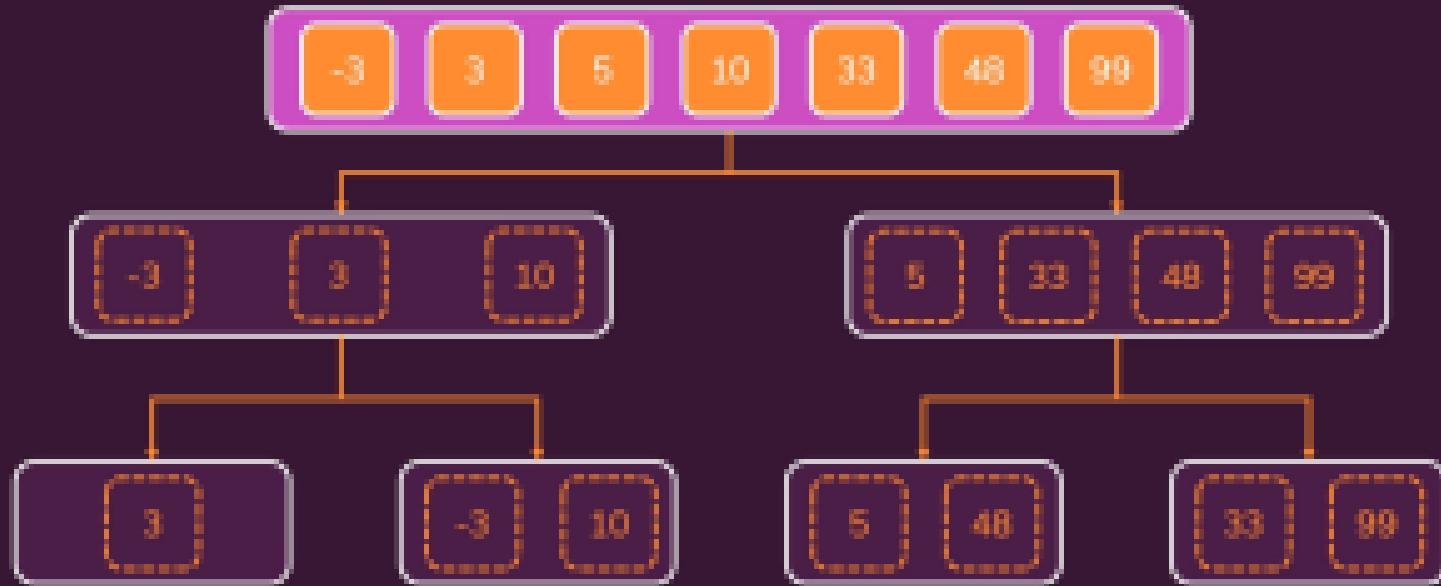
# Merge Sort



# Merge Sort



# Merge Sort



# Merge Sort – Time Complexity

Best Case

Average Case

Worst Case

Items are sorted randomly

Items are sorted randomly

Items are sorted randomly

$O(n * \log n)$

$O(n * \log n)$

$O(n * \log n)$

# Space Complexity

Not just Runtime Performance

## Module Content

What is Space Complexity?

Deriving Space Complexity

Examples

## What is "Space Complexity"?

How much space in memory does your algorithm occupy?

All values in JavaScript are stored in memory

Especially arrays and objects can take up a bit more space

Generally, in JavaScript, you won't need to worry about space complexity and memory too much though

# Deriving Space Complexity

Find places where data (values) is stored "permanently" in your algorithm



Count how often such "permanently" stored values are being created (and kept around)



Determine how the number of values depends on your "n"

→  $O(n)$ ,  $O(1)$  etc. exists for space complexity as well

# Examples

## Algorithm

## Space Complexity

## Reason

Factorial (Loop)

$O(1)$

We operate on one and the same number, no new ("permanent") value is created per iteration

Factorial (Recursive)

$O(n)$

A new value is created for every nested function call (the parameter received)

Linear Search

$O(1)$

No new "permanent" values are created during the iteration

Binary Search

$O(1)$

No new "permanent" values are created during the iteration

## More Examples

### Algorithm

Bubble Sort

### Quicksort

### Merge Sort

### Space Complexity

$O(1)$

$O(n)$   
 $(O(n \log n) \text{ is possible})$

$O(n)$

### Reason

No new "permanent" values are created during the iteration

Nested function calls with new values being created

Nested function calls with new values being created

# Sets (Array) Algorithms

Getting Tricky

## Module Content

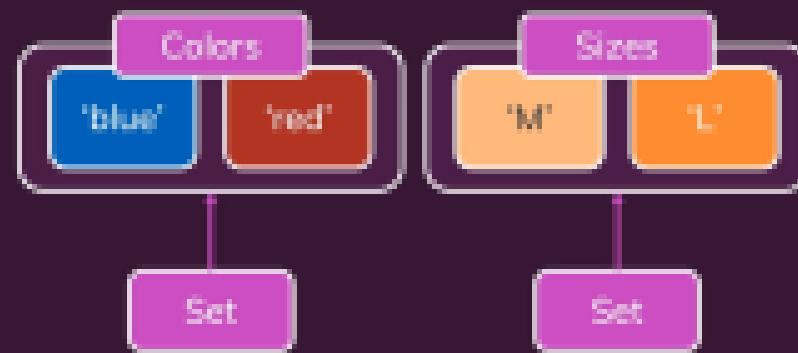
What are "Set (Array) Algorithms" All About?

Examples!

# What are Set (Array) Algorithms About?

A Set is a collection of values (objects, numbers) which forms an entity itself.

Example: Shirt manufacturing

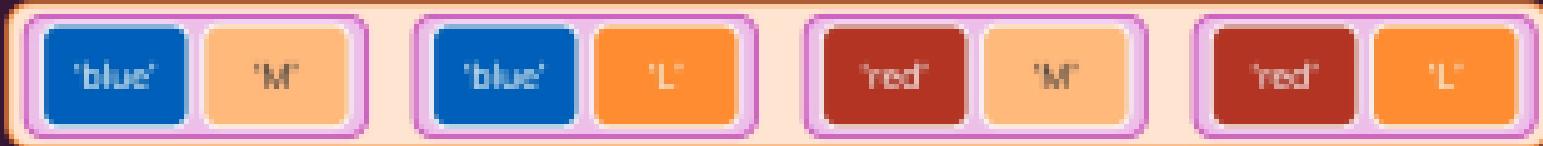


# Cartesian Product

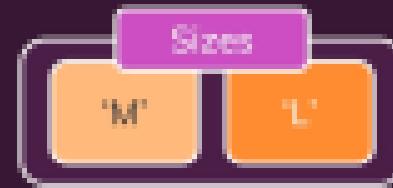
Example: Shirt manufacturing



New Set: All variants that should be manufactured



# Cartesian Product – More Than 2 Sets



# Permutations (With & Without Repetition)

An ordered combination of values.

## Without Repetition

Example: Todo list items

["Walk dog", "Clean toilet", "Order food"]

["Walk dog", "Clean toilet", "Order food"]  
["Clean toilet", "Walk dog", "Order food"]  
["Walk dog", "Order food", "Clean toilet"]

...

## With Repetition

Example: Safe combination you want to set

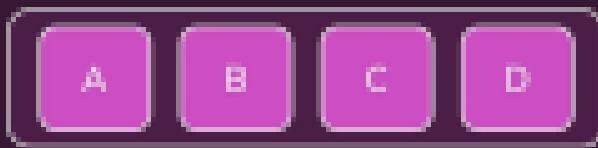
[1, 5, 9]

[1, 1, 9]  
[1, 1, 1]  
[9, 5, 9]

...

# Recursive Permutations

Main function call



1. Recursive Step



Combine with

2. Recursive Step



3. Recursive Step



Combine with



# Recursive Permutations



# A Structured Problem Solving Approach

Solving ANY Problem

## Module Content

How to approach problems

Examples!

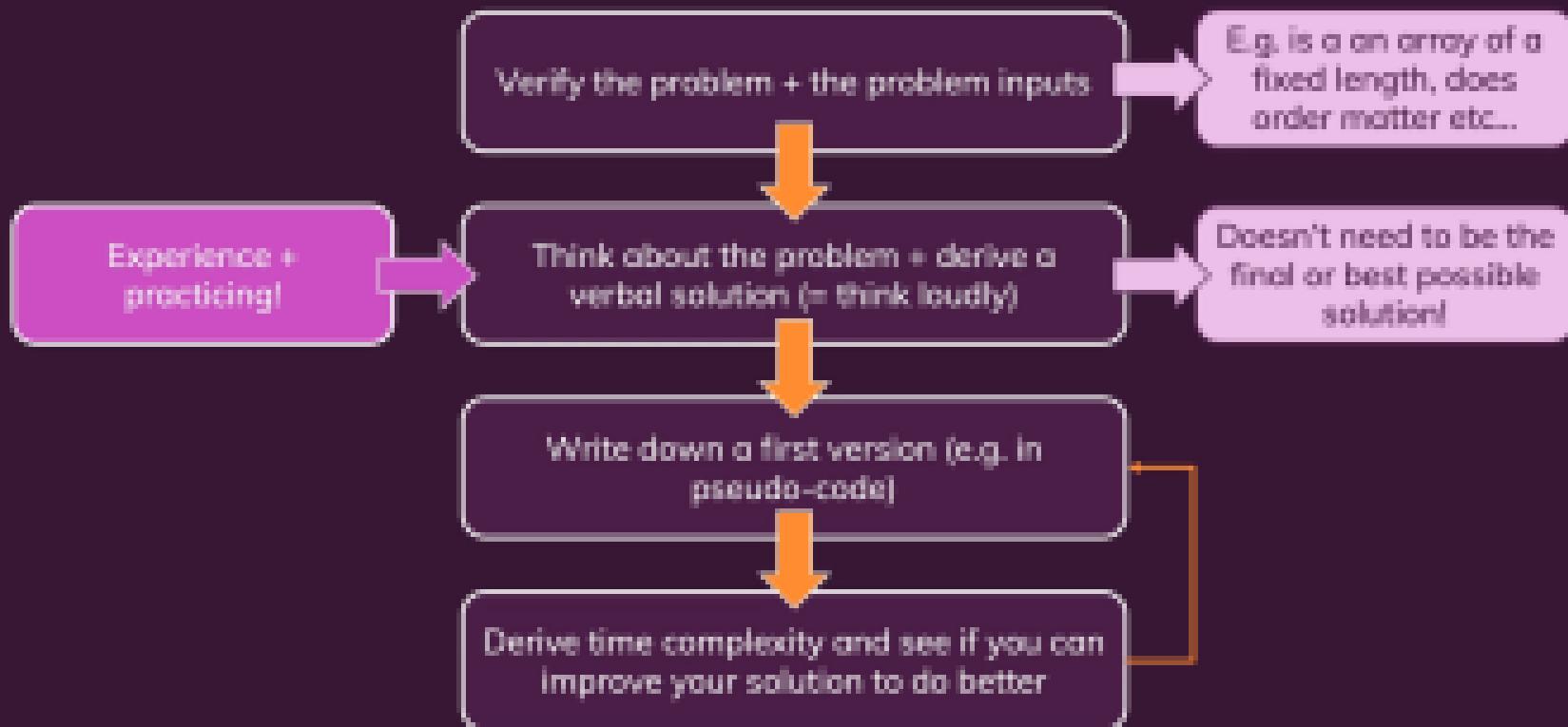
## Complex Algorithms Are Complex...

Coming up with the best possible algorithm can be very hard for tricky problems

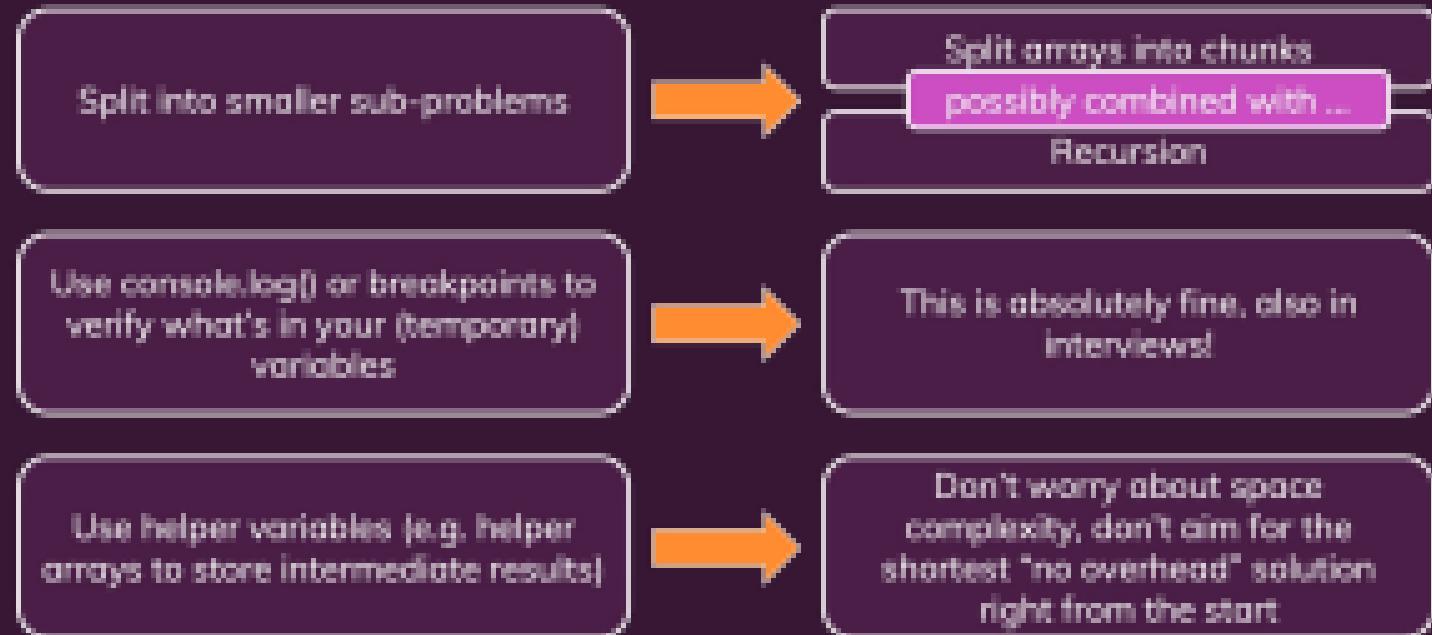
That's why algorithms are popular in interviews: People want to see if you can solve problems.

Good news: It's NOT the best possible solution that counts. It's your ability to come up with solutions

# Solving Problems / Coming Up With Algorithms



# Ways of Simplifying a Problem



## Practice Makes Perfect!

Finding good approaches to solve a problem takes practice – there is no simple “blueprint” that you can apply to every problem.



Practice by diving into common algorithms and interview questions

# The Knapsack Problem

You got a list of items, where every item has a value and a weight. You got a bag that holds a maximum weight of X.

Write a program that maximizes the value of the items you put into the bag whilst ensuring that you don't exceed the maximum weight.

```
items = [  
  {id: 'a', val: 3, w: 3},  
  {id: 'b', val: 6, w: 8},  
  {id: 'c', val: 10, w: 3}  
]  
  
maxWeight = 8  
  
bag = ['a', 'c'] // solution
```

# Solving the Knapsack Problem

Verify inputs: Can items be used multiple times?

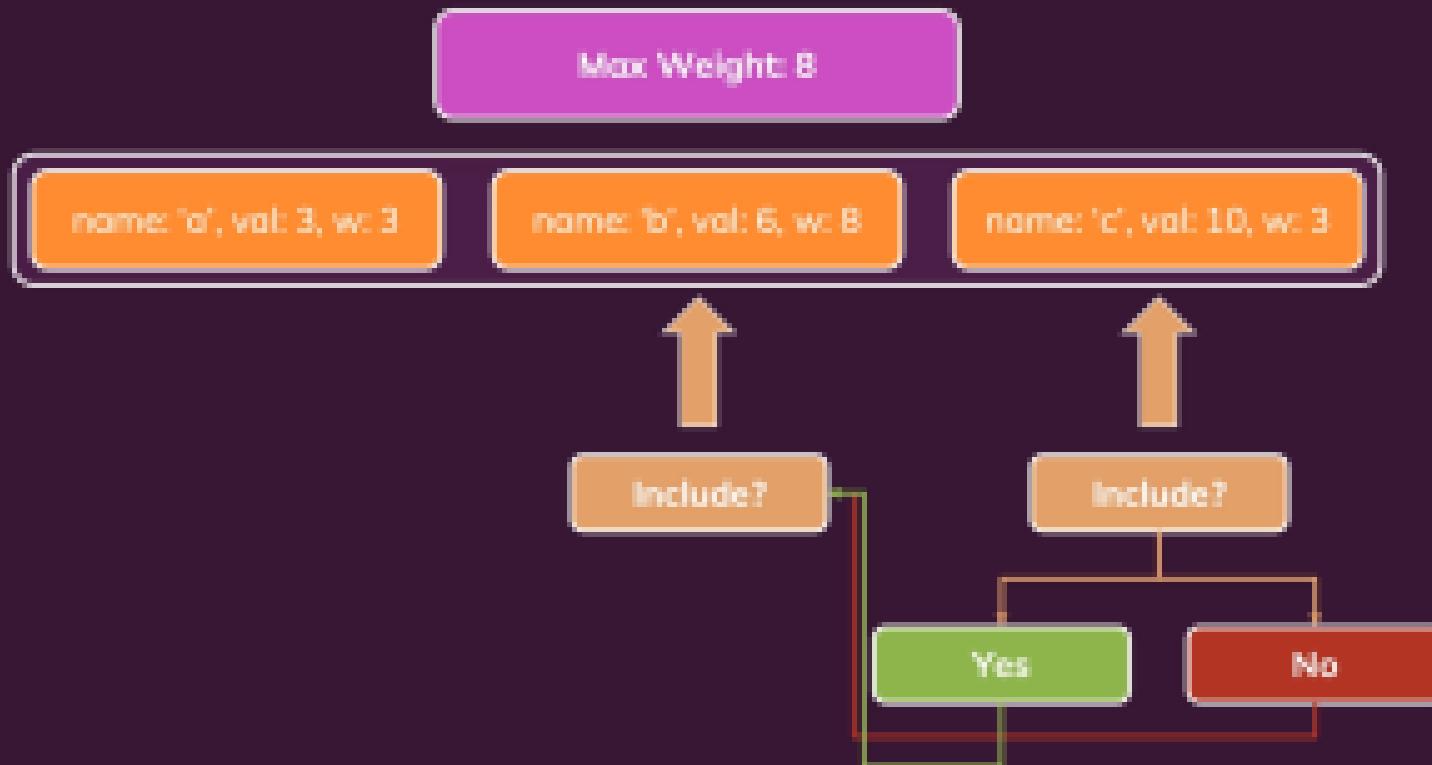


Derive a first (verbal) solution: We could derive all possible combinations and find the one with highest value and fitting weight

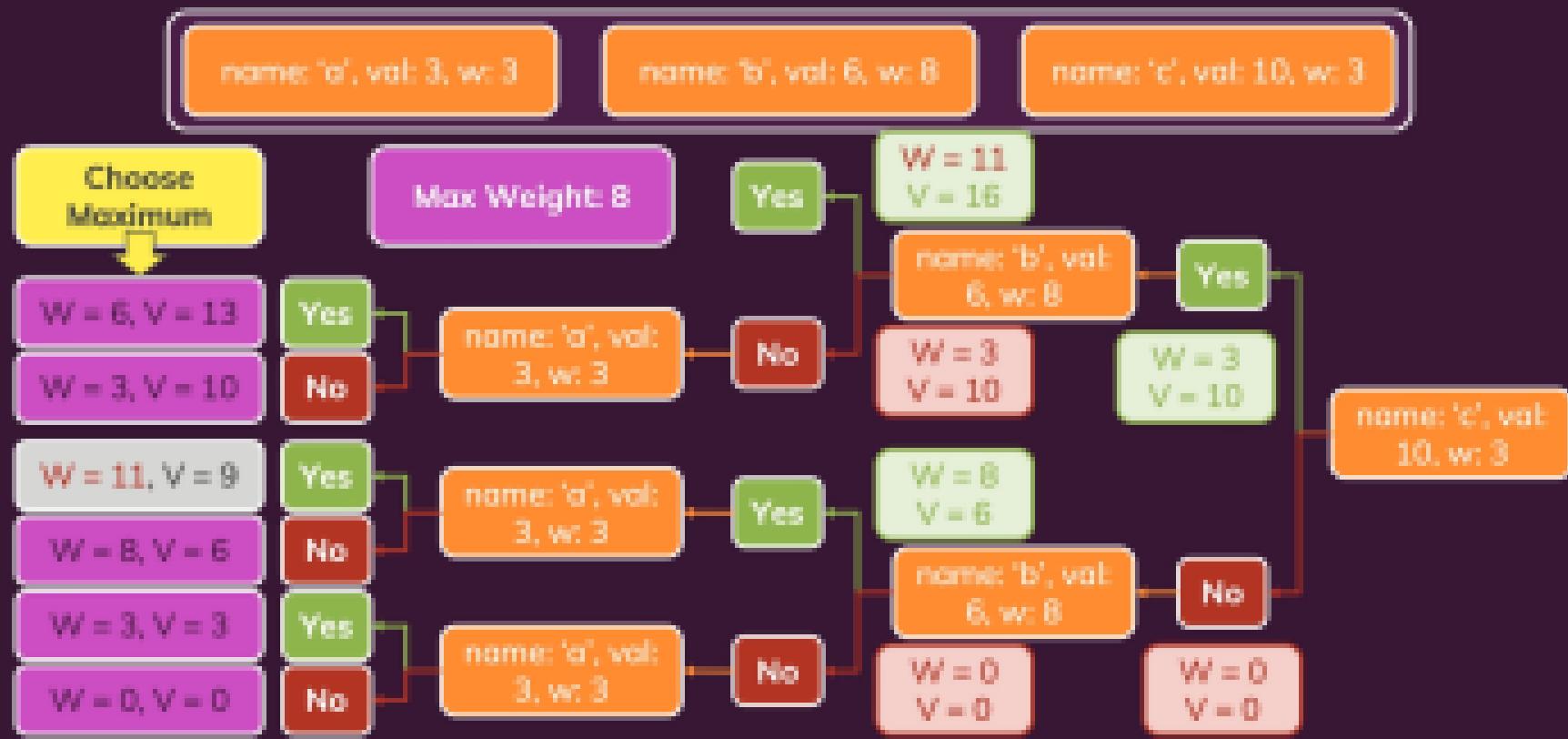


Write down a first version!

# Let's Rethink!



# We Evaluate All Possible Cases / Combinations



# Greedy vs Dynamic Algorithms / Solutions

## Greedy

Make best possible decision in every step and hope that it leads to the overall best solution

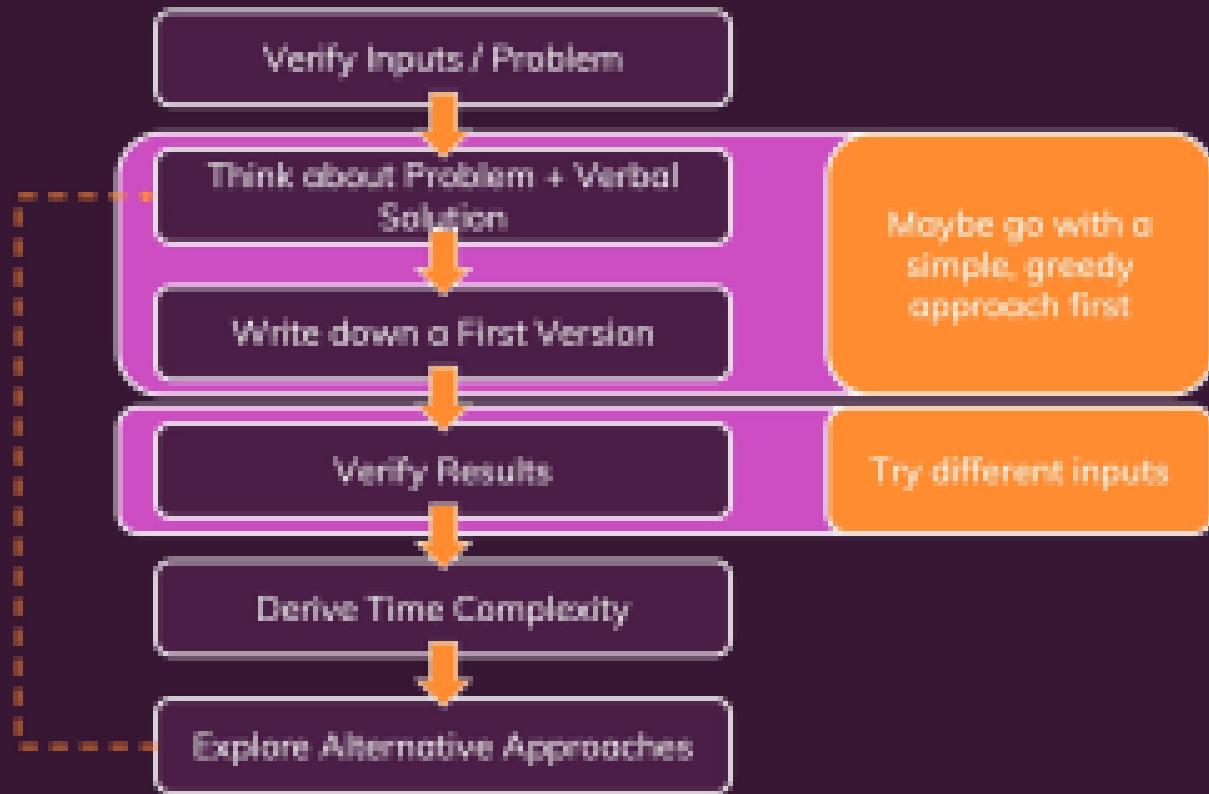
Greedy algorithms often are faster to set up and come up with but they don't necessarily provide the best runtime and/or result

## Dynamic

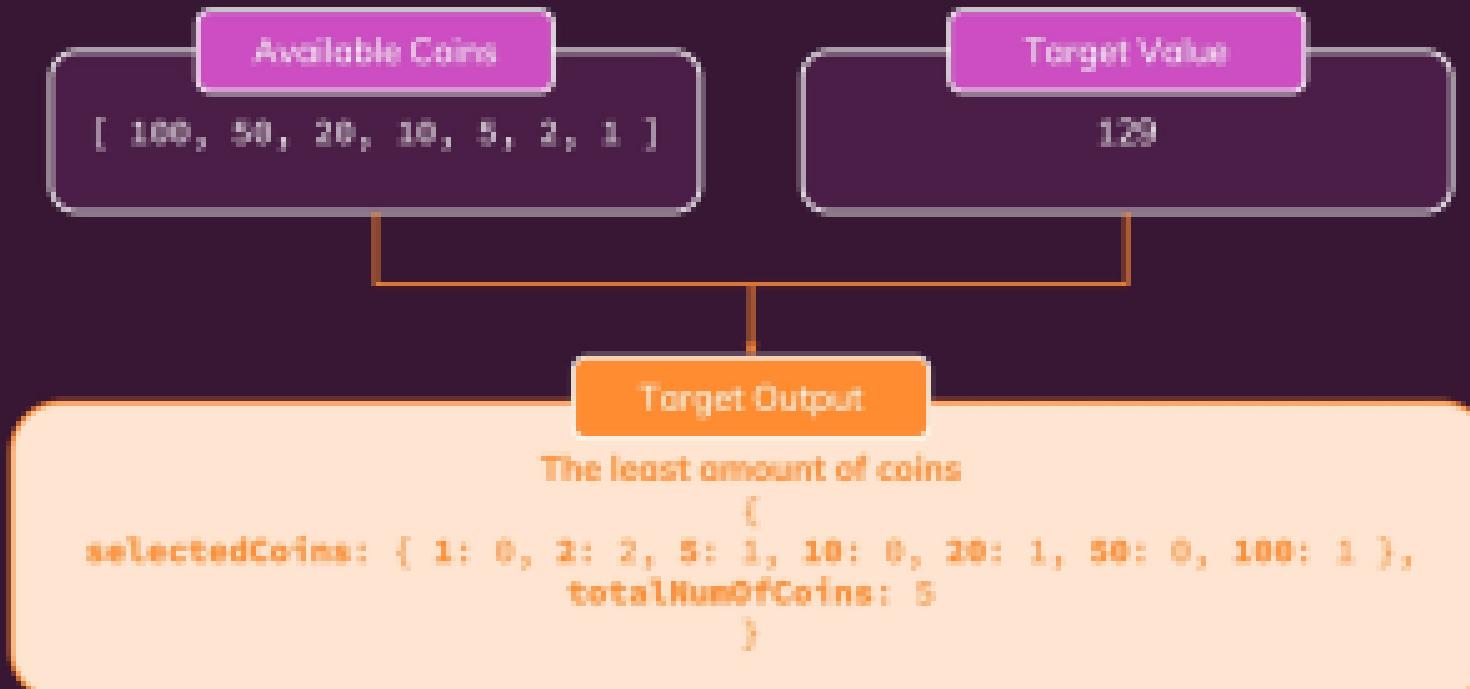
Evaluate all possible solutions and find overall best solution via comparison

"Divide and conquer" approach:  
Divide the problem into smaller, easy-to-solve subproblems

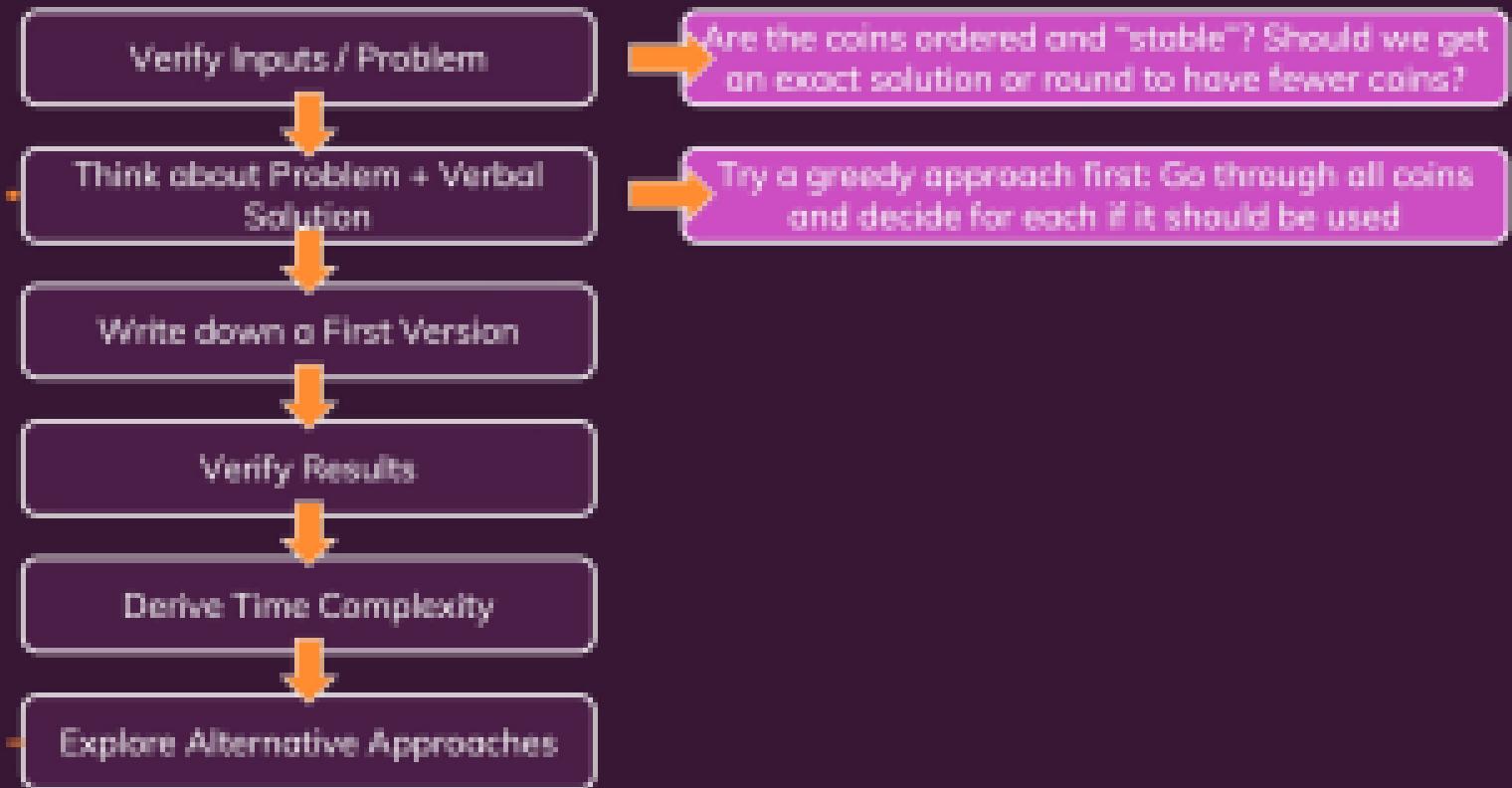
# Our Final "Problem-Solving" Plan



# The Change Making Problem



# Change Making Problem: Our Plan



# The More Difficult Change Making Problem

Available Coins

```
[ 0, 6, 5, 1 ]
```

Target Value

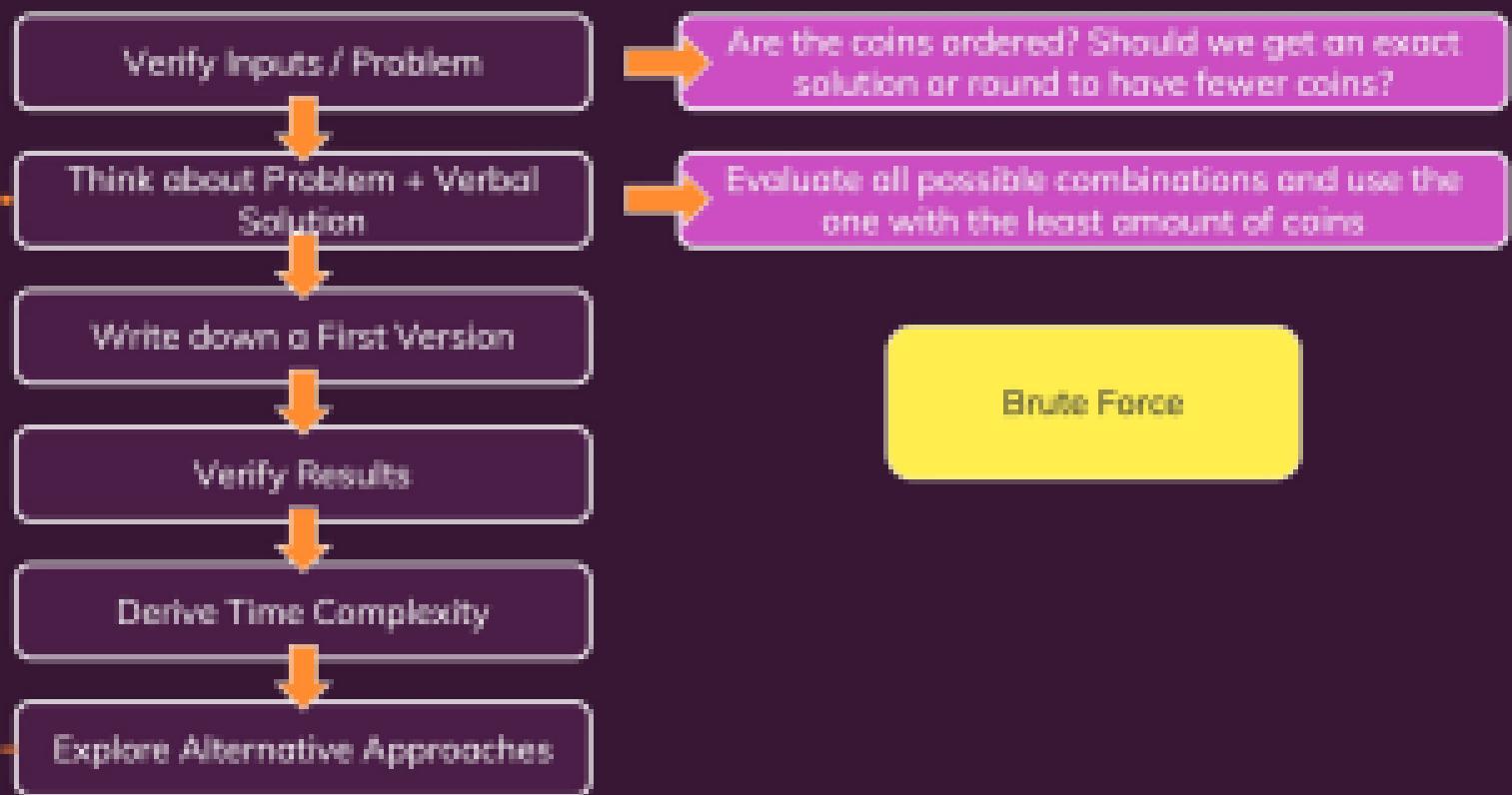
11

Target Output

The least amount of coins

```
{  
    selectedCoins: { 0: 0, 6: 1, 5: 1, 1: 0 },  
    totalNumOfCoins: 2  
}
```

# Change Making Problem: Adjusted Plan



# How To Continue

Next Steps

## Practice Algorithms!

You only get better by solving a lot problems and understanding a lot of algorithms

Good thing is: You find plenty of examples and explanations online – and with this course, you'll be able to understand them

Don't try to learn algorithms by heart! Understand common patterns, approaches etc. instead

## Next Step: Data Structures

- "Data structures" are things like arrays and objects but typically, custom-built data structures are meant
- Custom data structures are built by combining default data structures like arrays and objects and enriching them with functionalities
- Such custom data structures can make solving some problems easier and/or more efficient
- If you understand algorithms, you'll have an easy time understanding data structures