



> StartBig

First Practical

2023

**MESA**

# WHO IS THIS GUY?



## Giovanni Manfredi

- MESA active Member since October 2022
- Participant at IT Sprint 2023 in Skopje, North Macedonia
- HO and creator of StartBig
- Tech savvy since I was 6 years old



# AGENDA

- **MESA** - What is MESA?
- What **does** MESA do?
- How do I **join**?
- What about **StartBig**?

# What is MESA?

## Local Association



Milan Engineering  
Student Association

## European Association



Electrical Engineering Student  
European Association

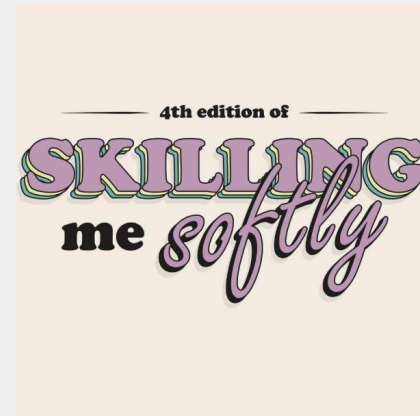
# What does MESA do?

## EVENTS!

Hard Skills



Soft Skills



# An International Network



**24** countries



**43** universities



**5000+** people



**5** regions

## ■ How do I join?



**Scan** the QR and  
**click** on the section Join us

## ■ What about StartBig?

```
>startBig
```


```
Initiating coding interview preparation in process...
```





## CareerService session

How do we get to do a coding interview?

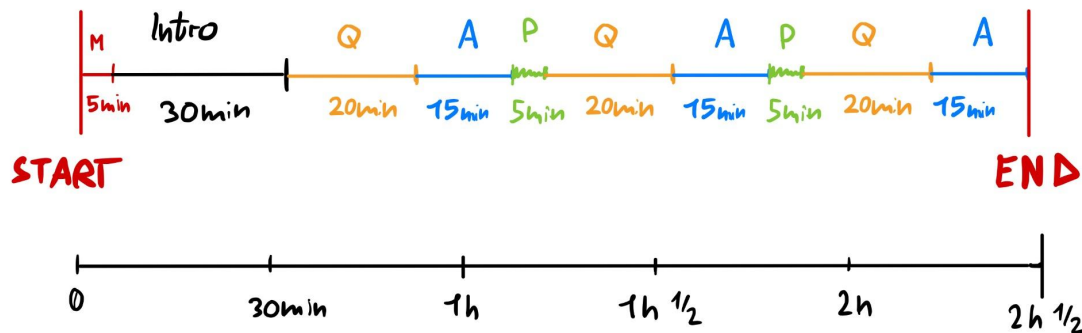
 **CareerService** is here to help us find that out!

## 3 Practical sessions

How can we prepare for a coding interview?

Content of the sessions:

- Introduction to coding interview questions - held by me
- Trees & Graphs (DSF & BSF algorithms) - held by researcher Davide Yi Xian Hu
- What is Dynamic Programming? - held by researcher Nicolò Felicioni





## Company session

How is the true experience behind coding interviews?

Andrea from **Oracle** will help us understand that!

There will be a small ice-breaker interview and then open questions from you!

# STAGE TO ...

**Giovanni Manfredi**

Yes, that's still me



■ Tutors here to help!

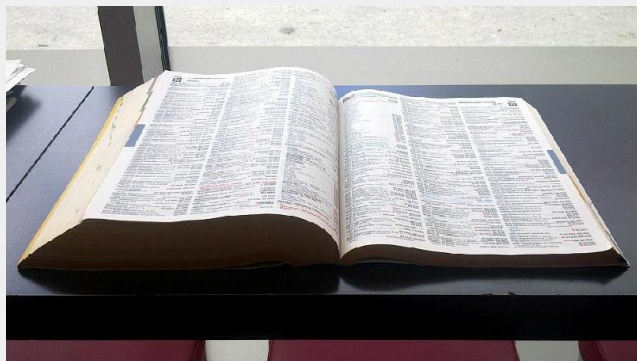
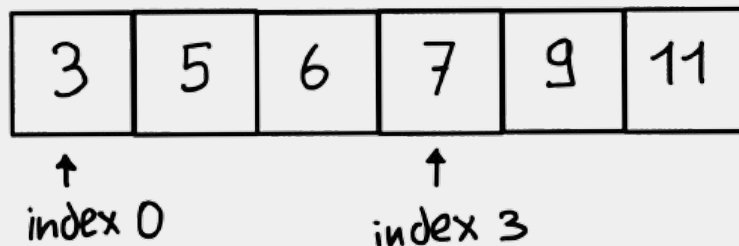


# First Practical Meeting AGENDA

1. **How** do I **solve** a **coding interview** question?
2. **How** can I **optimise** my solution?
3. What's **LeetCode** and how does it work?
4. Let's **start coding**!
5. Let's see a **solution** together
6. **Repeat** 4. three times
7. **End** of the session!

# Draw out the problem!

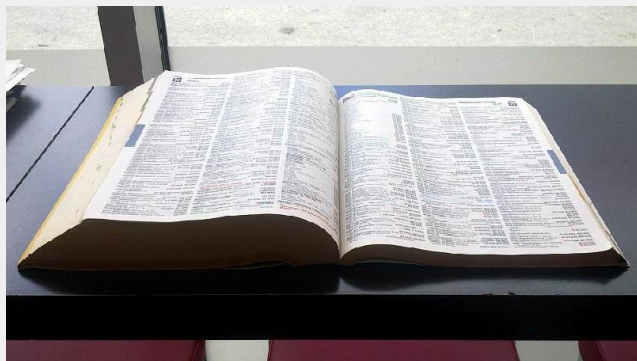
**Example:** Search in an ordered Array



# How would you solve it by hand?

**Example:** How do you search into a address book? You use the index!

If you open the book at random, you will turn the pages right or left **depending** on how the letter you're searching for **compares** to the one present in the address book!





## What are some more examples?

**Example:** think of other inputs to our program, does it still work with negative numbers? With zero numbers?

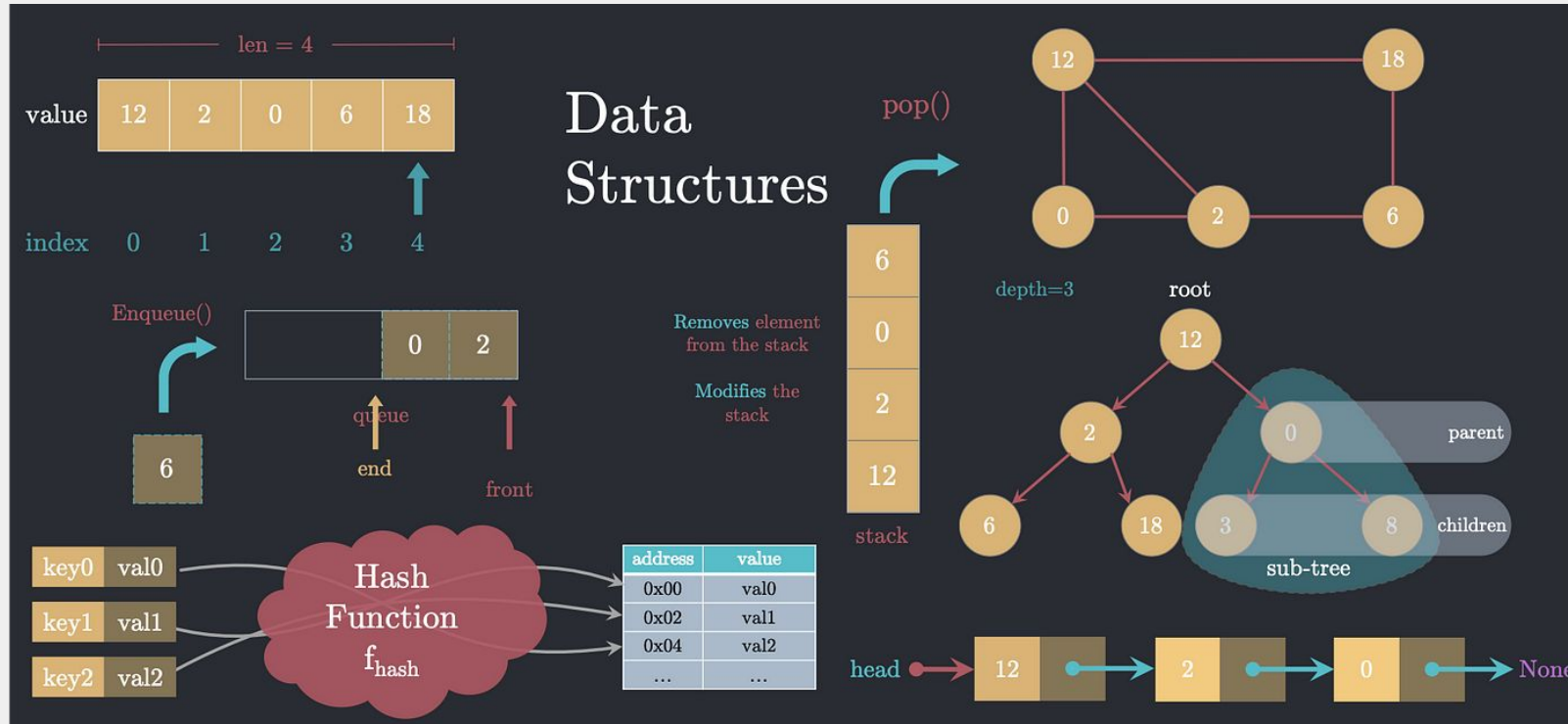
This also greatly helps with **test cases**!

# ■ Cut the elephant into pieces

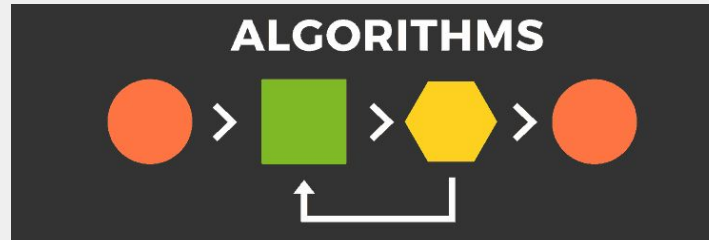
**Divide** your main **problem into** (ideally independent) **subproblems**



# Remember and use your tools



# Remember and use your tools



## Common algorithms and approaches

- Sorting
- Binary Search
- Sliding window technique
- Two pointers approach
- Union find
- Breadth First Search (2° Practical meeting topic)
- Depth First Search (2° Practical meeting topic)
- Topological sorting

# Optimising your solution

## 1. Time complexity

the time complexity is the **computational complexity** that describes the **amount of computer time** it takes to run an algorithm

- a. Typically **more relevant** than space complexity in a coding interview
- b. We will use the **Big-O notation**  $\rightarrow O(N)$ , etc.

## 2. Space complexity

The space complexity of an algorithm or a computer program is the **amount of memory** space required to solve an instance of the computational problem as a function of characteristics of the input. It is the **memory required** by an algorithm until it executes **completely**

- a. Typically **less relevant** than time complexity in a coding interview
- b. We will use the **Big-O notation**  $\rightarrow O(N)$ , etc.

# Time complexity

We need to identify the **Best Theoretical Time Complexity** (from now on, **BTTC**) of the solution.

The **BTTC** is the the **time complexity that you cannot beat**.

## Example:

The **BTTC** of finding the **sum of numbers** in array is  $O(N)$  because you have to **look at every value** in the array at least once.

**NB!** The **BTTC doesn't always correspond to the total number of elements in a data structure**. Think at the **Binary search** that uses the fact that the set is ordered to not look at every single element (time complexity  $O(\log N)$ ).

# Where am I losing time?

## 1. Identify overlapping and repeated computation

If your algorithm is doing something **repetitive**, that **you wouldn't do** if you were to solve it by hand, think of another approach that it's faster and doesn't waste that time.

## 2. Try different data structures

Knowing your data structures means to be able to understand when you need them and when you don't. **Example:** if you're struggling with lookup times, you might want to use a **hashMap**

# Space complexity

## 1. Changing data in-place/overwriting input data

If your solution creates a **support data structure** for the input, you can save some space by instead of creating it, **modifying directly the input**. This is discouraged in software engineering (hard to maintain), but **can be used in coding interviews** to reduce space complexity.

## 2. Change the data structure

As for time complexity, also for space complexity **selecting the correct data structure is crucial** to reduce the complexity to the minimum.



# ■ LeetCode, your new best friend



A perfect platform for **practicing coding problems** and to master coding interviews!

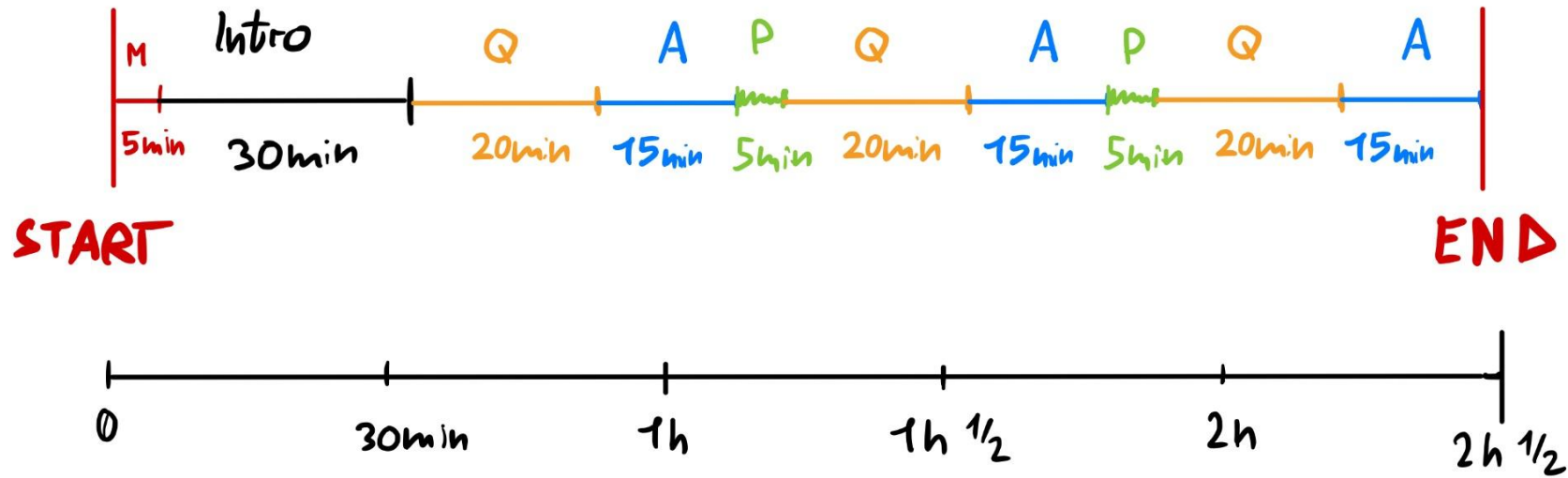
## **Problem difficulty:**

As in a videogame, start easy, then go to higher difficulties

## General advices

1. Try to **find a solution** even if not efficient
2. **Don't look at solutions** straight away
3. **Tips** only **after trying**
4. Practice makes perfect
5. Learn from your mistakes
6. It's more **fun with friends**

■ Let's do some questions together!



# ■ 3, 2, ... 1, CODE!

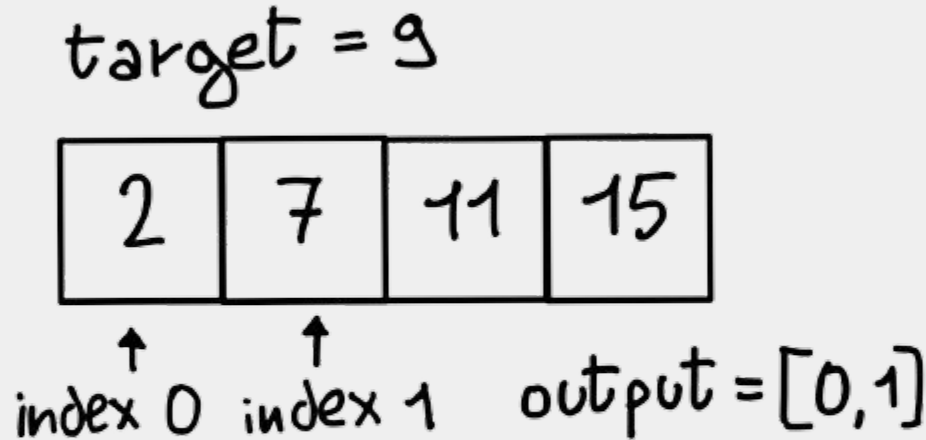
Your first exercise is **TWO SUM** (See GitHub repo)



Start coding now.

## Two sum solution

1. Visualise the problem: draw the array



2. Solving it by hand: comparison between one element and each other element

# First implementation

- Time complexity:  $O(N^2)$
- Space complexity:  $O(1)$

```
int* twoSum(int* nums, int numsSize, int target, int* returnSize) {  
    int *arr = malloc(2*sizeof(int));  
    *returnSize = 2;  
    for(int i=0; i < numsSize - 1; ++i) {  
        for(int j=i+1; j < numsSize; ++j) {  
            if(nums[i] + nums[j] == target) {  
                arr[0] = i;  
                arr[1] = j;  
                return arr;  
            }  
        }  
    }  
    return arr;  
}
```

Language: C

## Two sum solution

3. Some more examples: let us make other sample arrays
4. Cut elephant into pieces: here we could think to **divide the lookup from the comparison part**
5. Remember and use you tools: the lookup part in the previous implementation took a lot of time. Maybe we could use a **HashMap** to reduce the time complexity

## ■ Second implementation

- Time complexity:  **$O(N)$**
- Space complexity:  **$O(N)$**

```
class Solution {  
    public int[] twoSum(int[] nums, int target) {  
        Map<Integer, Integer> numToIndex = new HashMap<>();  
        // For each number we verify if in the HashMap there is  
        // its "complement" to get to target  
        for (int i = 0; i < nums.length; i++) {  
            if (numToIndex.containsKey(target - nums[i])) {  
                return new int[] {numToIndex.get(target - nums[i]), i};  
            }  
            numToIndex.put(nums[i], i);  
        }  
        return new int[] {};  
    }  
}
```

**Language: JAVA**



# 3, 2, ... 1, CODE!

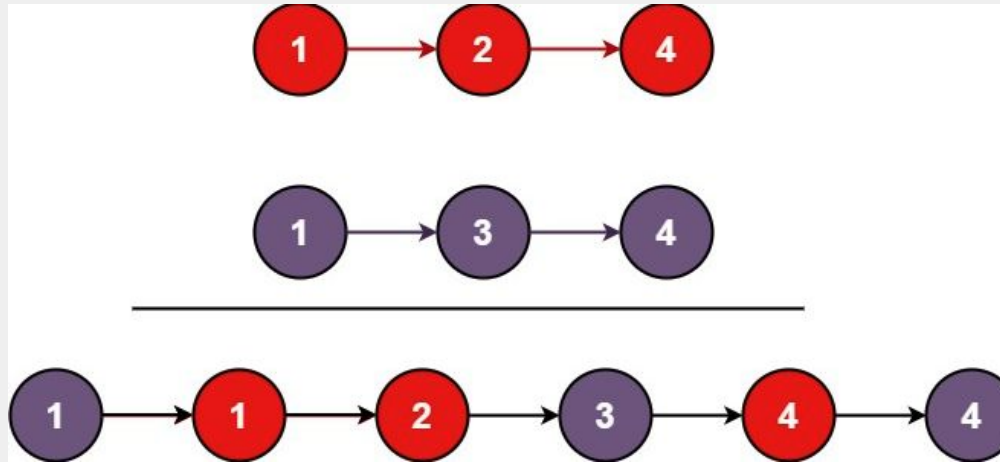
Your first exercise is **Merge two sorted lists** (See GitHub repo)



Start coding now.

# Merge two sorted lists solution

1. Visualise the problem: draw the two linked lists



2. Solving it by hand: going through each element of the two lists I would take the lowest for the new list

# First implementation

- Time complexity:  **$O(N+M)$**
- Space complexity:  **$O(1)$**

```
struct ListNode* mergeTwoLists(struct ListNode* list1, struct ListNode* list2) {  
    if(list1 == NULL) {  
        return list2;  
    }  
    if(list2 == NULL) {  
        return list1;  
    }  
    if(list1 -> val <= list2 -> val) {  
        list1 -> next = mergeTwoLists(list1 -> next, list2);  
        return list1;  
    } else {  
        list2 -> next = mergeTwoLists(list1, list2 -> next);  
        return list2;  
    }  
}
```

**Language: C**

## Merge two sorted lists solution

3. Some more examples: let us make other sample lists
4. Cut elephant into pieces: here we could think to **keep track of the element of the two lists by having two pointers**, one per each list
5. Remember and use you tools: by using two pointers (references) instead of one we are much faster

## ■ Second implementation

- Time complexity:  **$O(N)$**
- Space complexity:  **$O(1)$**

```
class Solution:
    def mergeTwoLists(self, list1: Optional[ListNode], list2: Optional[ListNode])
-> Optional[ListNode]:
    cur = dummy = ListNode()
    while list1 and list2:
        if list1.val < list2.val:
            cur.next = list1
            list1, cur = list1.next, list1
        else:
            cur.next = list2
            list2, cur = list2.next, list2
    if list1 or list2:
        cur.next = list1 if list1 else list2
    return dummy.next
```

**Language: Python**

# ■ 3, 2, ... 1, CODE!

Your first exercise is **Top K frequent elements** (See GitHub repo)



Start coding now.

# ■ Top K frequent elements solution

1. Visualise the problem: draw the array
2. Solving it by hand: writing down on a sheet of paper how many times a number appears and at the end of the array get our best K
3. Some more examples: let us make other sample arrays
4. Cut elephant into pieces: here we could think to **first go through the array and saving the frequencies**, then **calculate the top K**
5. Remember and use you tools: the best tool for easy access is a **HashMap** (so we can use it to save the information), then we can do a **Heap** to make the top K. Then the element in the heap will be our solution

# First implementation

- Time complexity:  **$O(N \log K)$**
- Space complexity:  **$O(1)$**

```
class Solution:
    def topKFrequent(self, nums: List[int], k: int) -> List[int]:
        freq = {}
        for num in nums:
            freq[num] = freq.get(num, 0) + 1
        heap = []
        for num, count in freq.items():
            if len(heap) < k:
                heapq.heappush(heap, (count, num))
            elif count > heap[0][0]:
                heapq.heappushpop(heap, (count, num))
        return [num for count, num in heap]
```

**Language: Python**



■ T.HANKS everyone!



 T.HANKS A LOT!

