

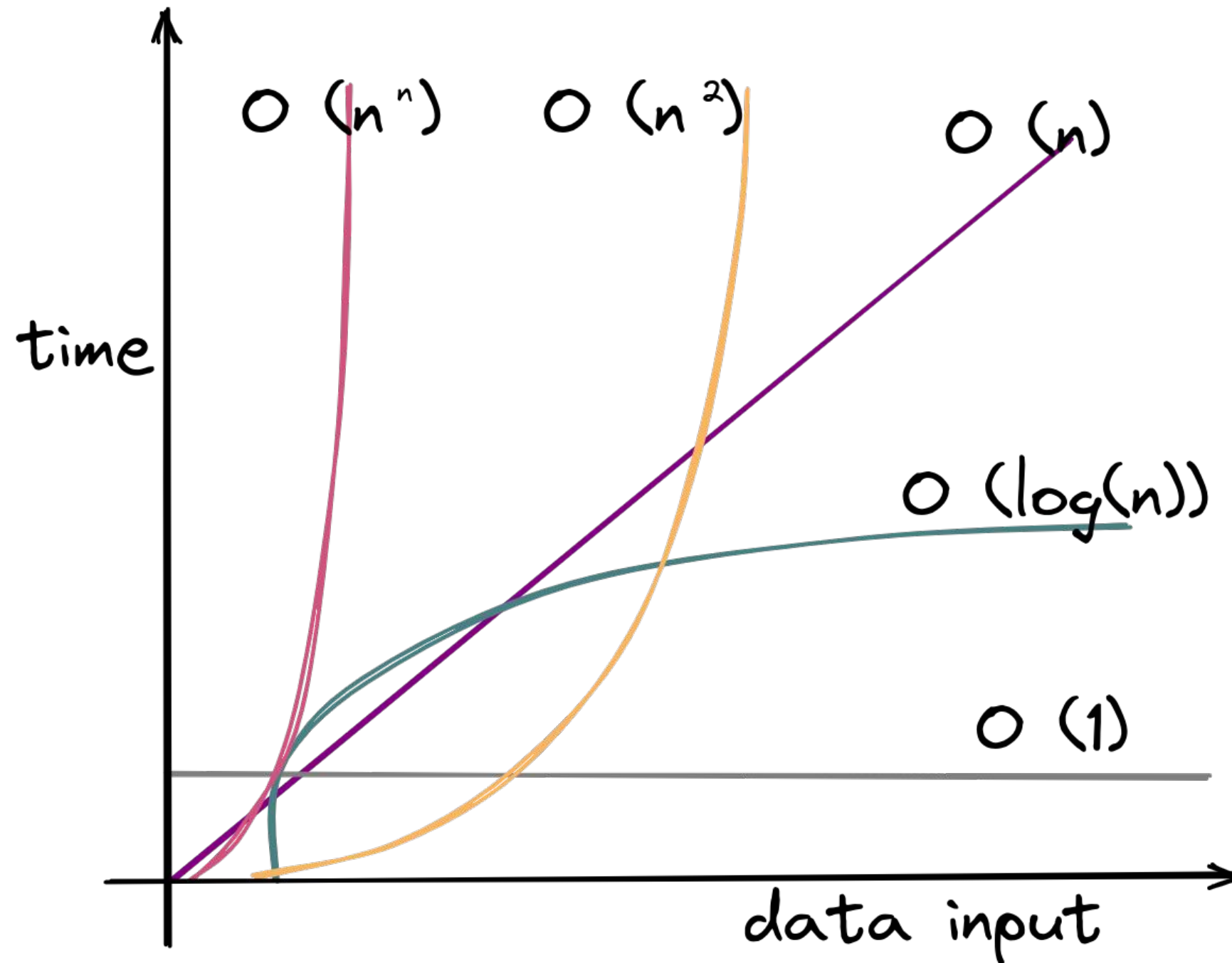
# Data Structures and Algorithms

Week 3

Introduction to Algorithms

Middlesex University Dubai; Winter22  
CST4050; Instructor: Ivan Reznikov

# Algorithm time complexity



- $O(1)$  – array access
- $O(\log(n))$  – binary search
- $O(n)$  – linear search
- $O(n^2)$  – bubble sort
- $O(n^n)$  – Hanoi tower problem

# Algorithm time complexity

$$O(100) \rightarrow O(1)$$

$$O(2 * \log(n)) \rightarrow O(\log(n))$$

$$O(5 * n + 6) \rightarrow O(n)$$

$$O(n^2 + 2 * n) \rightarrow O(n^2)$$

$$O(n^n + 2 * n^3 + 5 \log(n)) \rightarrow O(n^n)$$

# Algorithm time complexity

def return\_sum(N):

result\_sum = 0

$O(1)$

for i in range(N+1):

result\_sum = result\_sum + i

$O(n)$

return result\_sum

$O(1)$

# return\_sum(3) → 3+2+1 → 6

$O(1 + n + 1) = O(n)$

# return\_sum(10) → 55

# Algorithm time complexity

Given an array of integers *nums* and an integer *target*, return *indices* of the two numbers such that they add up to *target*. You may assume that each input would have exactly one solution, and you may not use the same element twice.

## Example 1:

Input: `nums = [2,7,11,15]`, `target = 9`

Output: `[0,1]`

Explanation: Because `nums[0] + nums[1] == 9`, we return `[0, 1]`.

## Example 2:

Input: `nums = [3,2,4]`, `target = 6`

Output: `[1,2]`

<https://leetcode.com/problems/two-sum>

# Algorithm time complexity

```
def sum_of_two(arr, target):
```

```
    for i, x in enumerate(arr):
```

$O(N)$

```
        for i2, y in enumerate(arr[i+1:]):
```

$O(N)$

```
            if x+y == target:
```

```
                return i, i2+i+1
```

$O(1)$

# Algorithm time complexity

```
def sum_of_two(arr, target):
```

```
    passed_values = {}  $O(1)$ 
```

```
    for i in range(len(arr)):  $O(N)$ 
```

```
        if target-arr[i] in passed_values:
```

```
            return [passed_values[target-arr[i]], i]
```

```
    passed_values[arr[i]] = i  $O(1)$ 
```

# Basic search: $O(n)$

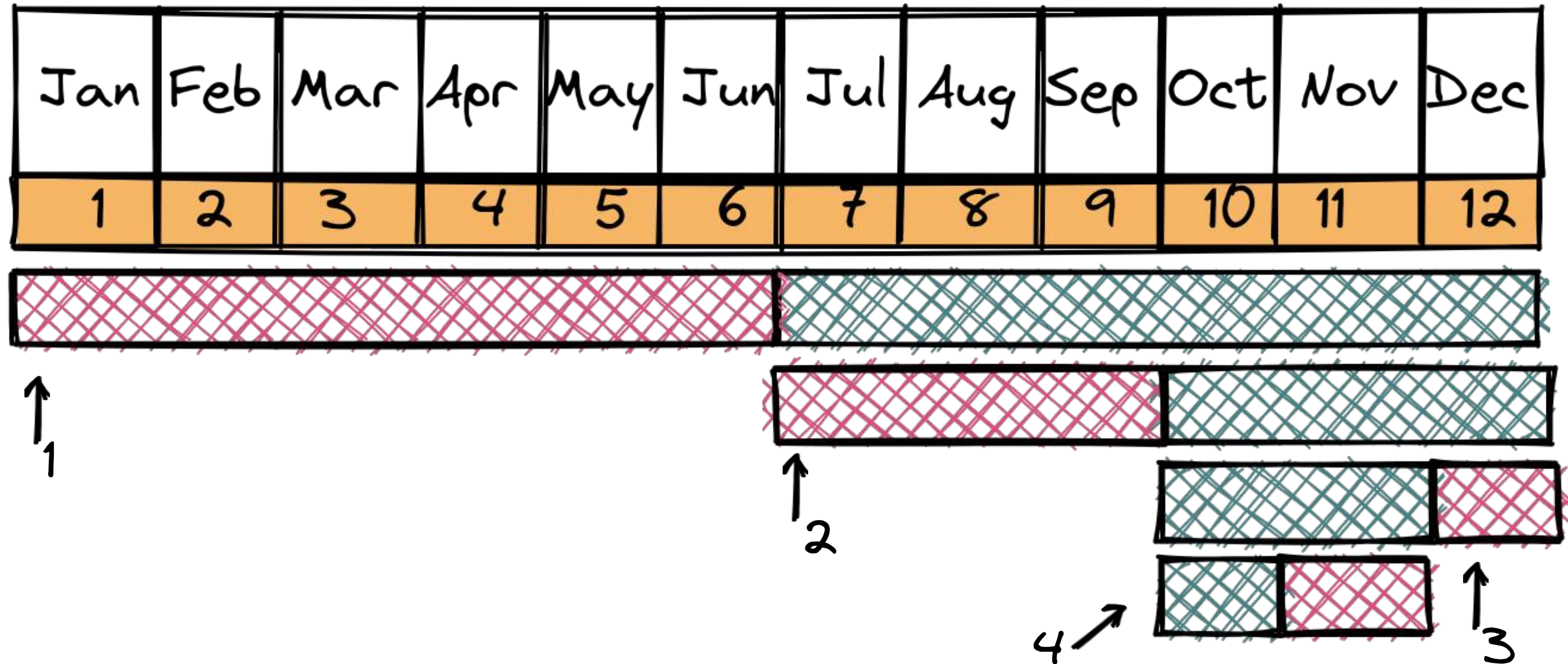
Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
1	2	3	4	5	6	7	8	9	10	11	12

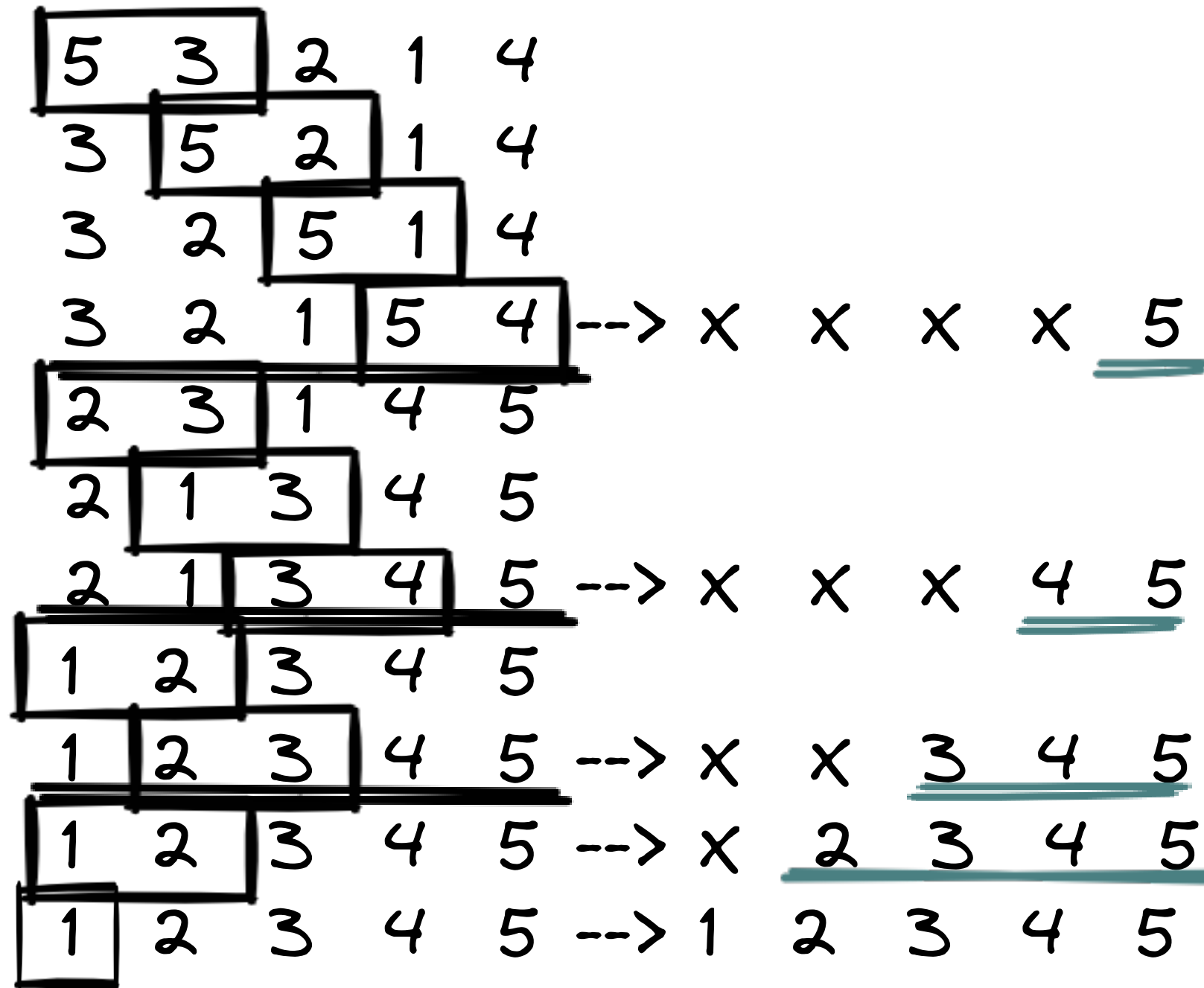
↑	↑	↑	↑	↑	↑	↑	↑	↑	↑		
1	2	3	4	5	6	7	8	9	10		



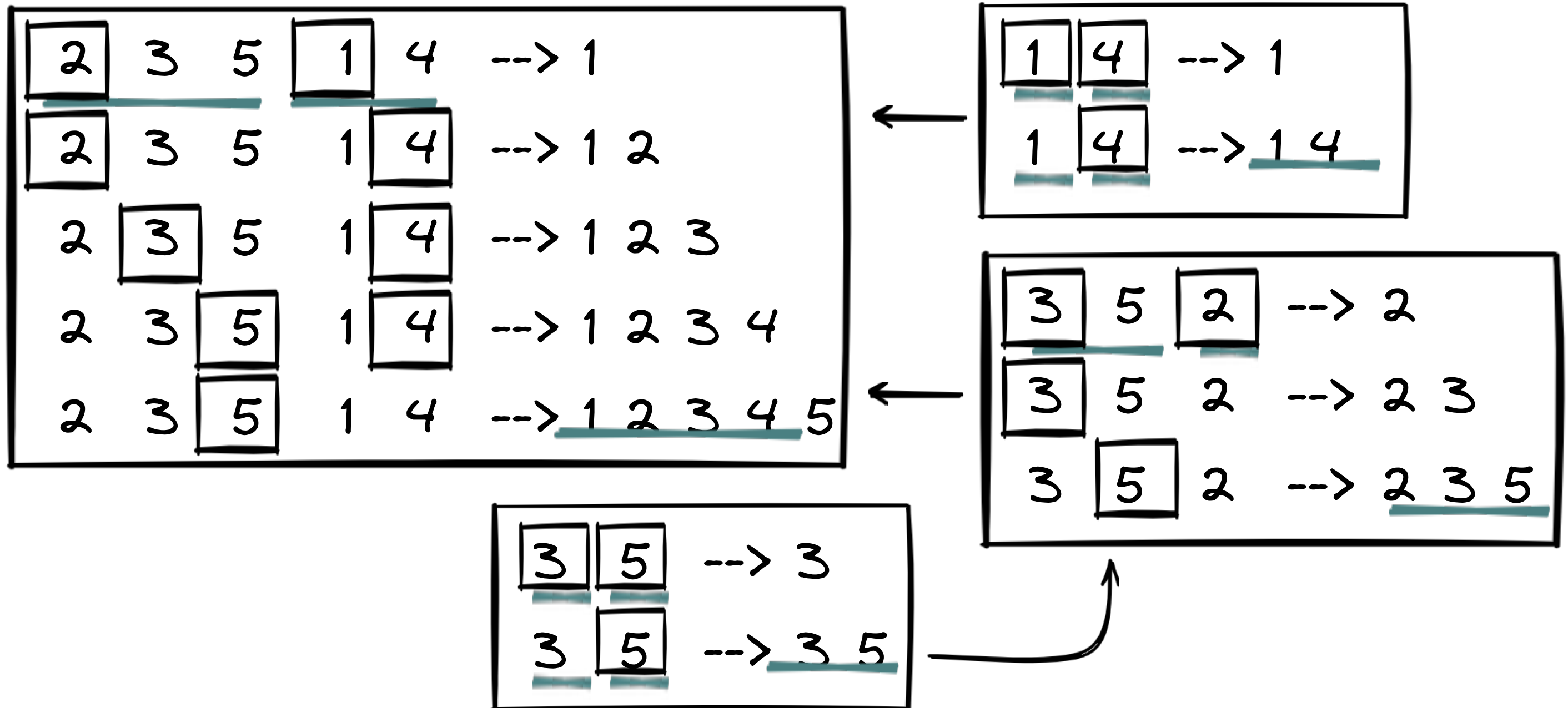
# Binary search: $O(\log(n))$



# Bubble sort: $O(n^2)$



# Merge sort: $O(n \times \log(n))$



# Time Space Complexity

$O(1)$  – constant time

$O(N)$  – time proportional to  $N$

## Common Data Structure Operations

Data Structure	Time Complexity								Space Complexity
	Average				Worst				Worst
	Access	Search	Insertion	Deletion	Access	Search	Insertion	Deletion	
<u>Array</u>	$\theta(1)$	$\theta(n)$	$\theta(n)$	$\theta(n)$	$\theta(1)$	$\theta(n)$	$\theta(n)$	$\theta(n)$	$\theta(n)$
<u>Stack</u>	$\theta(n)$	$\theta(n)$	$\theta(1)$	$\theta(1)$	$\theta(n)$	$\theta(n)$	$\theta(1)$	$\theta(1)$	$\theta(n)$
<u>Queue</u>	$\theta(n)$	$\theta(n)$	$\theta(1)$	$\theta(1)$	$\theta(n)$	$\theta(n)$	$\theta(1)$	$\theta(1)$	$\theta(n)$
<u>Singly-Linked List</u>	$\theta(n)$	$\theta(n)$	$\theta(1)$	$\theta(1)$	$\theta(n)$	$\theta(n)$	$\theta(1)$	$\theta(1)$	$\theta(n)$

# P/ NP Problems

P is the complexity class of decision problems that can be solved in polynomial time. (*multiplication*)

NP (nondeterministic polynomial) is a complexity class of all decision problems for which the problem solution can be **verified** in polynomial time (*Sudoku solver*).

NP Hard

