



Intro to Algorithmic Problem Solving

Week 2

Housekeeping

- **Contact me:** yuhan@dal.ca
- **Language for assignments:** Any language you prefer
- **DDL extension for A1:** Sunday, Jun 15
- **Exam language & make-up exam:** Your choice (lmk asap)

How AI Do Reasoning?

- **Semantic Parsing:** Convert natural language input into structured concepts, intent, and implicit assumptions.
- **Logical Modeling:** Build an internal reasoning model (e.g. logic tree, flow graph, or function-like structure).
- **Inference Execution:** Run the model to derive conclusions, using rules, conditions, and consistency checks.
- **Natural Language Generation:** Translate the reasoning trace into coherent, human-readable output.





Then vs Now

Then (Pre-AI Era)	Now (AI Era)
Learn to solve problems by hand	Learn to design problem solvers and guide AI thinking
Focused on coding implementations	Focused on algorithmic thinking and strategy
Tested through whiteboard coding	Applied through prompt engineering and tool building
Algorithms as end-goal of computer science learning	Algorithms as meta-tools to control intelligent systems
Human solves → machine executes	Human defines → machine solves



Course Overview

Week 1

Algorithm Intro
Time/Space Complexity
Data Structures
Two pointers
Prefix Sum

Week 2

Recursion
Divide and Conquer
Sorting
Binary Search

Week 3

Math Problems
Bit Manipulation
Greedy
Dynamic Programming

Week 4

Tree & Graph
BFS/DFS
Backtracking
Union-Find



Recursion

Recursion

A method where a function calls itself to solve a problem by breaking it into smaller subproblems.



Russian
nesting
dolls

Factorial

$$n! = \begin{cases} 1, & \text{if } n = 0 \\ n \times (n - 1)!, & \text{if } n > 0 \end{cases}$$

```
def factorial(n):
    if n == 0:
        return 1 # base case
    else:
        return n * factorial(n - 1) # recursive call
```

Time complexity: O(n)

Space complexity: O(n)
(using stack with depth n)

```
factorial(3)
= 3 * factorial(2)
= 3 * 2 * factorial(1)
= 3 * 2 * 1 * factorial(0)
= 3 * 2 * 1 * 1
= 6
```

Time Complexity

$$T(n) = T(n - 1) + \mathcal{O}(1)$$

Using recursion:

$$\begin{aligned} T(n) &= T(n - 1) + 1 \\ &= T(n - 2) + 1 + 1 \\ &= T(n - 3) + 1 + 1 + 1 \\ &\quad \dots \\ &= T(0) + n \\ &= 1 + n = \mathcal{O}(n) \end{aligned}$$

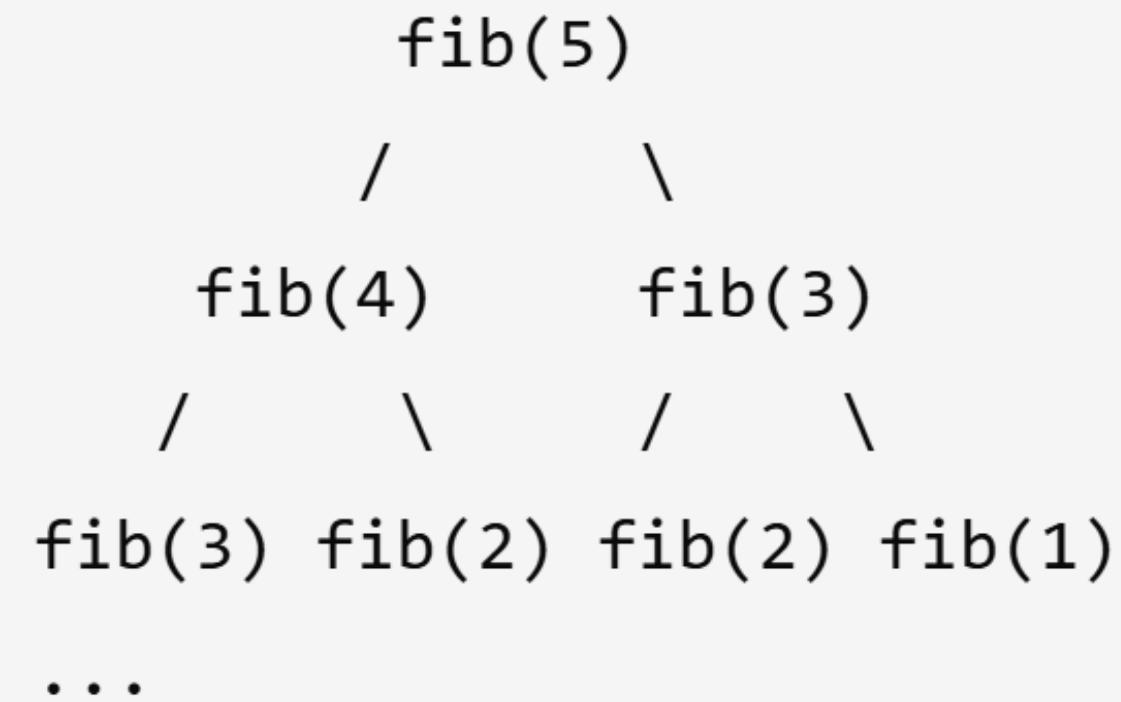
Fibonacci (Naive Recursion)

$$F(n) = \begin{cases} 0, & \text{if } n = 0 \\ 1, & \text{if } n = 1 \\ F(n - 1) + F(n - 2), & \text{if } n \geq 2 \end{cases}$$

```
def fib(n):
    if n == 0: return 0
    if n == 1: return 1
    return fib(n - 1) + fib(n - 2)
```

Time complexity: $O(2^n)$ <-Disaster!!!

Space complexity: $O(n)$
(using stack with depth n)



Fibonacci (Memoization)

$$F(n) = \begin{cases} 0, & \text{if } n = 0 \\ 1, & \text{if } n = 1 \\ F(n - 1) + F(n - 2), & \text{if } n \geq 2 \end{cases}$$

```
def fib(n, memo={}):
    if n in memo:
        return memo[n]
    if n == 0:
        return 0
    if n == 1:
        return 1
    memo[n] = fib(n - 1, memo) + fib(n - 2, memo)
    return memo[n]
```

Time complexity: O(n)
Space complexity: O(n)
(at most n elements in memory)

This is a Top-Down recursion.

Top-Down vs. Bottom-Up

- **Top-Down:**
 - Start from the original problem and recursively break it down into subproblems. Often combined with memoization to avoid redundant work.
- **Bottom-Up:**
 - Start from the base cases and iteratively build up to the final answer. This is typically implemented using loops and often called Dynamic Programming (DP). → Week 3



Sorting

Bubble Sort

Repeatedly swaps adjacent elements if they are in the wrong order.

```
def bubble_sort(arr):
    n = len(arr)
    for i in range(n):
        for j in range(n - 1 - i):
            if arr[j] > arr[j + 1]:
                arr[j], arr[j + 1] = arr[j + 1], arr[j]
```

Time complexity:
Worst/Avg: $O(n^2)$,
Best: $O(n)$ (already sorted, **with optimization**)

Space complexity: $O(1)$



Bubble Sort (with optimization)

```
def bubble_sort(arr):
    n = len(arr)
    for i in range(n):
        swapped = False # flag to check if any swap happened
        for j in range(n - 1 - i):
            if arr[j] > arr[j + 1]:
                arr[j], arr[j + 1] = arr[j + 1], arr[j]
                swapped = True
        if not swapped:
            break # exit early if already sorted
```

Selection Sort

Repeatedly selects the smallest element and places it at the beginning.

```
def selection_sort(arr):
    n = len(arr)
    for i in range(n):
        min_idx = i
        for j in range(i + 1, n):
            if arr[j] < arr[min_idx]:
                min_idx = j
        arr[i], arr[min_idx] = arr[min_idx], arr[i]
```

Time complexity:
Worst/Avg/Best: $O(n^2)$

Space complexity: $O(1)$

Insertion Sort

Builds the sorted array one element at a time by inserting into the correct position.

```
def insertion_sort(arr):
    for i in range(1, len(arr)):
        key = arr[i]
        j = i - 1
        while j >= 0 and arr[j] > key:
            arr[j + 1] = arr[j]
            j -= 1
        arr[j + 1] = key
```

Time complexity:
Worst/Avg: $O(n^2)$,
Best: $O(n)$

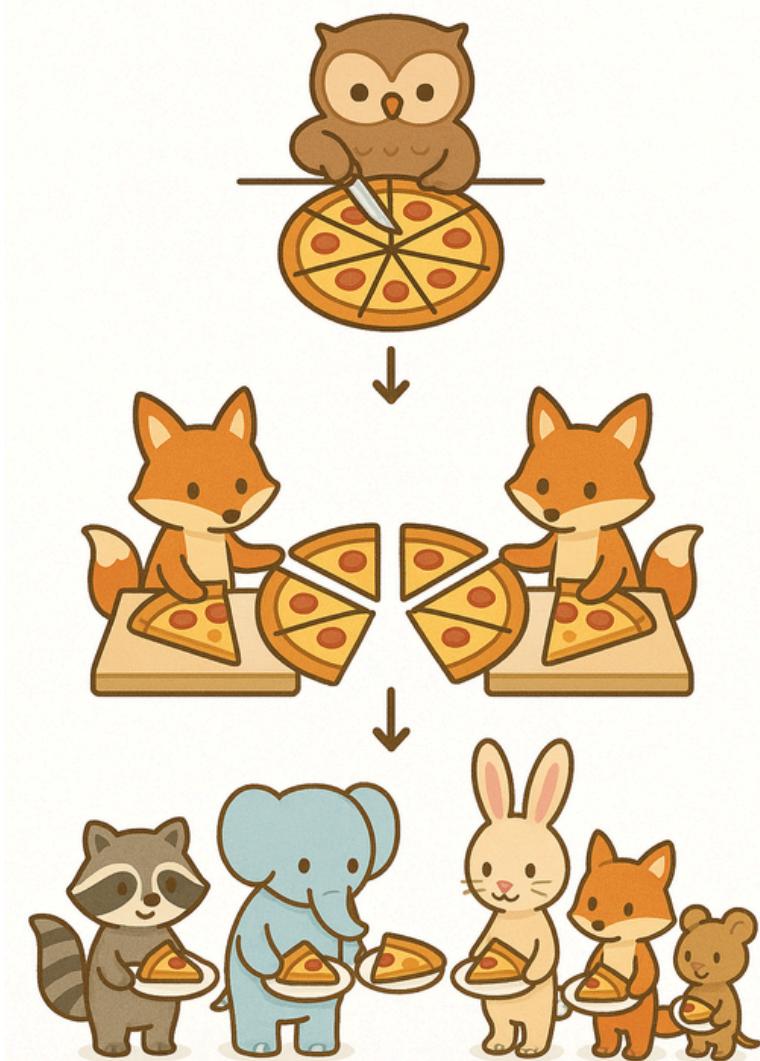
Space complexity: $O(1)$



Divide and Conquer

Divide and Conquer

- Divide-and-conquer is a common use case of recursion, but not all recursive algorithms divide problems (e.g., factorial is recursive but not divide-and-conquer).
- **Structure:**
 - **Divide:** Split the problem into smaller parts
 - **Conquer:** Solve each part recursively
 - **Combine:** Merge the solutions of subproblems
- **Common use cases:** Merge Sort, Quick Sort
- **Time Complexity:** Depends on the problem, often analyzed using recurrence relations



**QUICK
INSIGHTS**



First Principles

"First principles is a physics way of looking at the world. You boil things down to the most fundamental truths and then reason up from there.

(For example) people may say 'battery packs are really expensive and that is the way they will always be'. No, that's pretty dumb. If you apply that reasoning to anything new, you wouldn't ever be able to get to that new thing."

- Elon Musk

Merge Sort

Divides the array into halves, recursively sorts, and merges them.

```
def merge_sort(arr):
    if len(arr) <= 1:
        return arr
    mid = len(arr) // 2
    left = merge_sort(arr[:mid])
    right = merge_sort(arr[mid:])
    return merge(left, right)
```

```
def merge(left, right):
    res = []
    i = j = 0
    while i < len(left) and j < len(right):
        if left[i] <= right[j]:
            res.append(left[i])
            i += 1
        else:
            res.append(right[j])
            j += 1
    return res + left[i:] + right[j:]
```

Time complexity:
Worst/Avg/Best: $O(n \log n)$

Space complexity: $O(n)$

Time Complexity of Merge Sort

$$T(n) = 2T(n/2) + \mathcal{O}(n)$$

Two sub-problem of size $n/2$ Merge

$$T(n) = aT(n/b) + f(n)$$

Mater Theorem

Theorem 4.1 (Master theorem)

Let $a \geq 1$ and $b > 1$ be constants, let $f(n)$ be a function, and let $T(n)$ be defined on the nonnegative integers by the recurrence

$$T(n) = aT(n/b) + f(n),$$

where we interpret n/b to mean either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then $T(n)$ has the following asymptotic bounds:

1. If $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \lg n)$.
3. If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$, and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$. ■

Quick Sort

Picks a pivot and partitions the array into smaller and larger parts, then recursively sorts.

```
def quick_sort(arr):
    if len(arr) <= 1:
        return arr
    pivot = arr[0]
    less = [x for x in arr[1:] if x <= pivot]
    greater = [x for x in arr[1:] if x > pivot]
    return quick_sort(less) + [pivot] + quick_sort(greater)
```

Time complexity:
Avg/Best: $O(n \log n)$, Worst: $O(n^2)$

Space complexity: $O(\log n)$ (due to recursion)

Built-in Python Sort (Timsort)

Hybrid of merge sort and insertion sort,
optimized for real-world data.

```
nums = [5, 2, 9, 1]
nums.sort()
print(nums) # Output: [1, 2, 5, 9]
```

Time complexity:
Best: O(n), Avg/Worst: O(n log n)

```
nums = [5, 2, 9, 1]
sorted_nums = sorted(nums)
print(sorted_nums) # Output: [1, 2, 5, 9]
print(nums) # Output: [5, 2, 9, 1] (original unchanged)
```

Space complexity: O(log n) (due
to recursion)

For string, use sorted() only.



Binary Search

Binary Search

- **Idea**

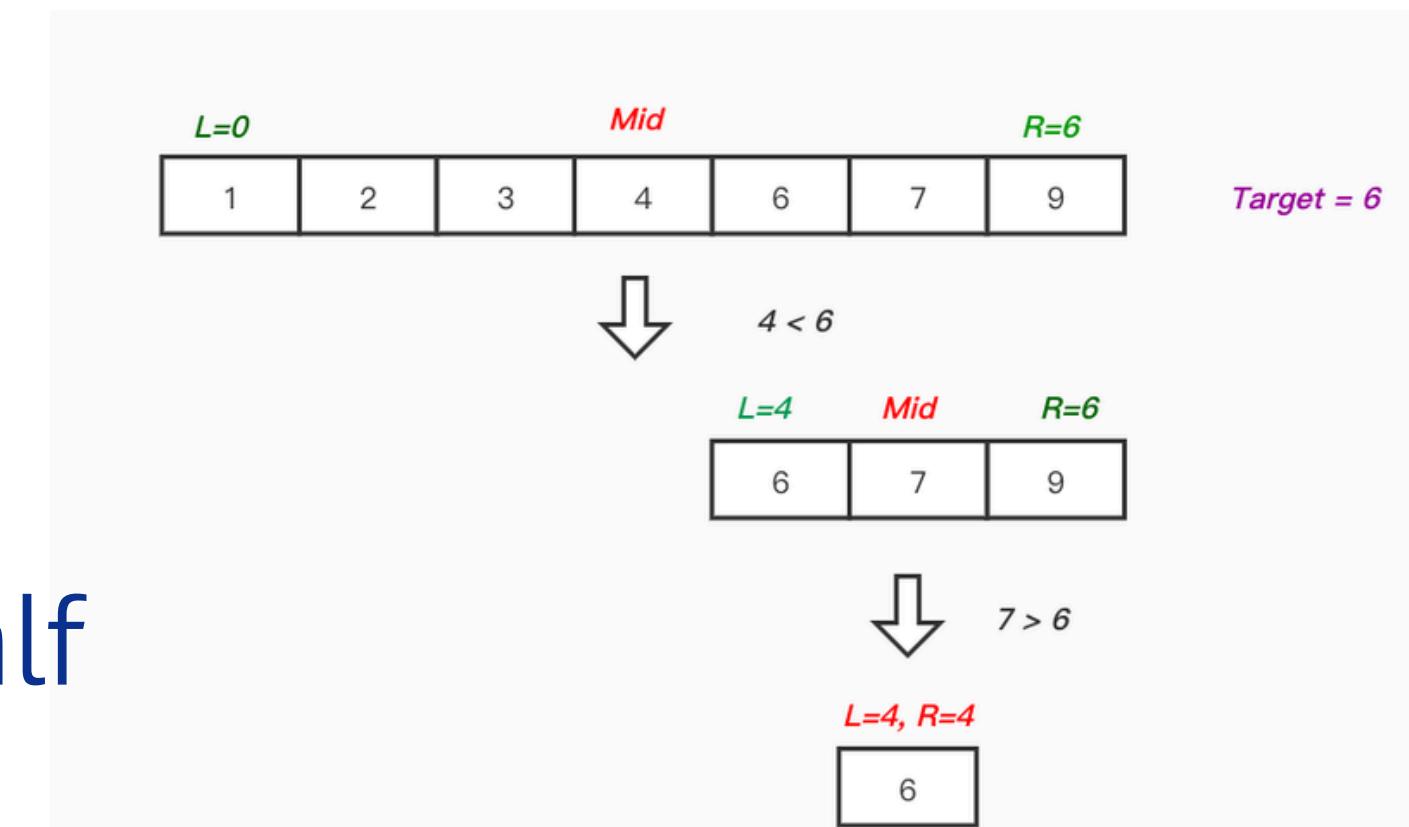
- Find an element in a sorted list by repeatedly dividing the search interval in half.

- **Key Requirement**

- Input must be sorted

- **Process**

- Check the middle element
- If it's too small, search the right half
- If it's too big, search the left half



Binary Search

```
def binary_search(arr, target):
    left, right = 0, len(arr) - 1
    while left <= right:
        mid = left + (right - left) // 2
        if arr[mid] == target:
            return mid
        elif arr[mid] < target:
            left = mid + 1
        else:
            right = mid - 1
    return -1
```

Time complexity: $O(\log n)$
Space complexity: $O(1)$

Q: Why we use
 $mid = left + (right - left) // 2$
instead of
 $mid = (left + right) // 2$?

A: Using $mid = left + (right - left) // 2$ prevents potential integer overflow that can occur with $(left + right) // 2$, especially in languages with fixed-size integers like Java or C++.

Even though Python is safe, this is considered a best practice in algorithm writing.



Practice Questions



167. Two Sum II - Input Array Is Sorted

Medium

Topics

Companies

Given a **1-indexed** array of integers `numbers` that is already **sorted in non-decreasing order**, find two numbers such that they add up to a specific `target` number. Let these two numbers be `numbers[index1]` and `numbers[index2]` where `1 <= index1 < index2 <= numbers.length`.

Return the *indices of the two numbers, `index1` and `index2`, added by one as an integer array `[index1, index2]` of length 2*.

The tests are generated such that there is **exactly one solution**. You may not use the same element twice.

Your solution must use only constant extra space.

Example 1:

Input: `numbers = [2,7,11,15]`, `target = 9`
Output: `[1,2]`

Explanation: The sum of 2 and 7 is 9. Therefore, `index1 = 1`, `index2 = 2`. We return `[1, 2]`.

Example 2:

Input: `numbers = [2,3,4]`, `target = 6`
Output: `[1,3]`

Explanation: The sum of 2 and 4 is 6. Therefore `index1 = 1`, `index2 = 3`. We return `[1, 3]`.

Example 3:

Input: `numbers = [-1,0]`, `target = -1`
Output: `[1,2]`

Explanation: The sum of -1 and 0 is -1. Therefore `index1 = 1`, `index2 = 2`. We return `[1, 2]`.

Binary Search

Use binary search to find the complement of each element, achieving better efficiency than brute force by using the sorted order of the array.

```
1 # binary search
2 class Solution:
3     def twoSum(self, numbers: List[int], target: int) -> List[int]:
4         for i in range(len(numbers)):
5             l, r = i+1, len(numbers)-1
6             tmp = target - numbers[i]
7             while l <= r:
8                 mid = l + (r-1)//2
9                 if numbers[mid] == tmp:
10                     return [i+1, mid+1]
11                 elif numbers[mid] < tmp:
12                     l = mid+1
13                 else:
14                     r = mid-1
```

Time complexity: $O(n \log n)$

Space complexity: $O(1)$

21. Merge Two Sorted Lists

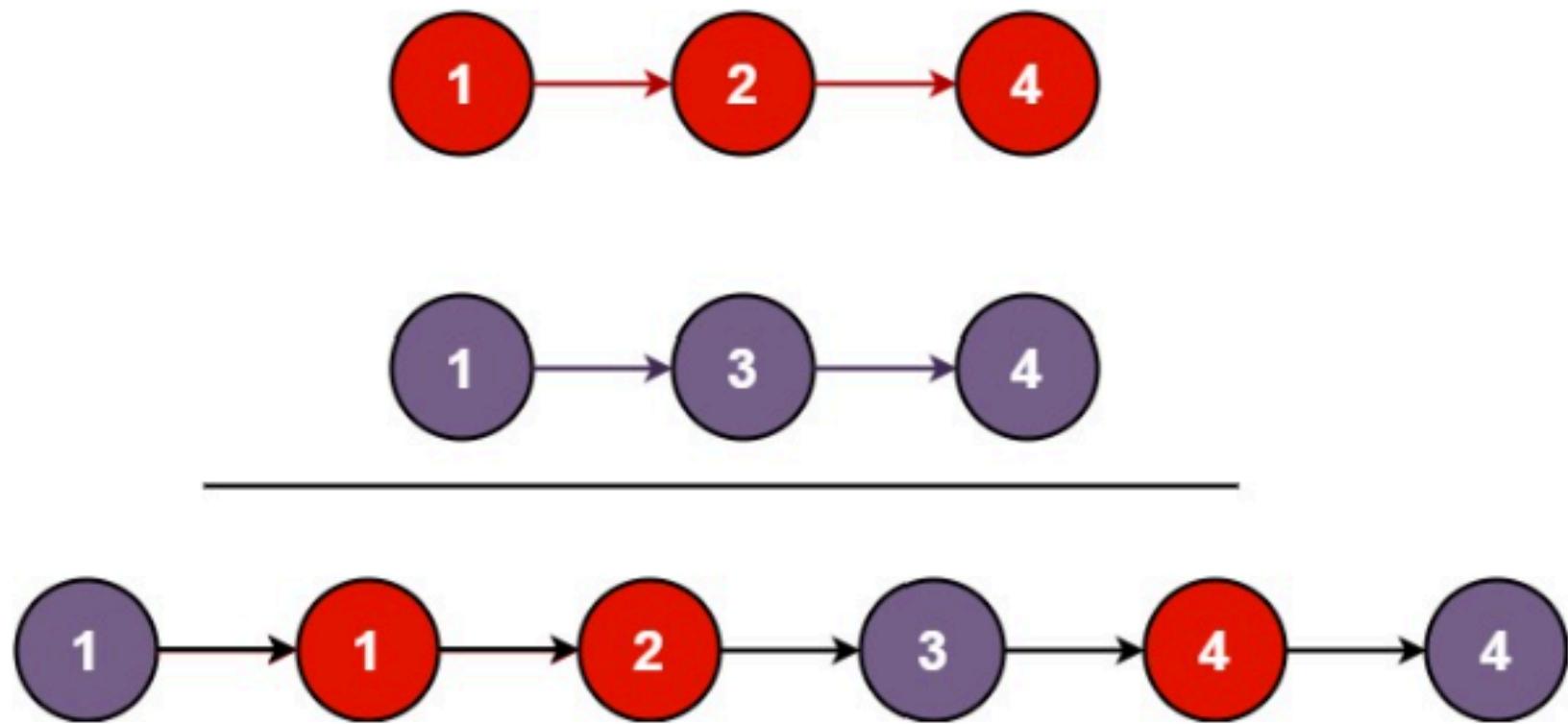
Easy Topics Companies

You are given the heads of two sorted linked lists `list1` and `list2`.

Merge the two lists into one **sorted** list. The list should be made by splicing together the nodes of the first two lists.

Return *the head of the merged linked list*.

Example 1:



```
Input: list1 = [1,2,4], list2 = [1,3,4]
Output: [1,1,2,3,4,4]
```

Example 2:

```
Input: list1 = [], list2 = []
Output: []
```

Example 3:

```
Input: list1 = [], list2 = [0]
Output: [0]
```

Iterative

Iteratively merges two sorted lists by comparing heads and building a new list using a dummy node.

```
1 class Solution:
2     def mergeTwoLists(self, list1: Optional[ListNode], list2: Optional[ListNode]) -> Optional[ListNode]:
3         curr = dummy = ListNode()
4         while list1 and list2:
5             if list1.val <= list2.val:
6                 curr.next = list1
7                 list1, curr = list1.next, list1
8             else:
9                 curr.next = list2
10                list2, curr = list2.next, list2
11         if list1 or list2:
12             curr.next = list1 if list1 else list2
13         return dummy.next
```

Time complexity: $O(m+n)$ Space complexity: $O(m+n)$
*m and n are the lengths of the two linked lists

Recursive

Recursively merges two sorted lists by always taking the smaller head and advancing its pointer.

```
1 class Solution:
2     def mergeTwoLists(self, list1: Optional[ListNode], list2: Optional[ListNode]) -> Optional[ListNode]:
3         if not list1:
4             return list2
5         if not list2:
6             return list1
7         if list1.val <= list2.val:
8             list1.next = self.mergeTwoLists(list1.next, list2)
9             return list1
10        else:
11            list2.next = self.mergeTwoLists(list1, list2.next)
12            return list2
```

Time complexity: $O(m+n)$

Space complexity: $O(1)$



169. Majority Element

Solved

Easy

Topics

Companies

Given an array `nums` of size `n`, return *the majority element*.

The majority element is the element that appears more than $\lfloor n / 2 \rfloor$ times. You may assume that the majority element always exists in the array.

Example 1:

Input: `nums = [3,2,3]`

Output: 3

Example 2:

Input: `nums = [2,2,1,1,1,2,2]`

Output: 2

Sorting

Sort the array and return the middle element, which must be the majority due to its frequency.

```
1 class Solution:  
2     def majorityElement(self, nums: List[int]) -> int:  
3         nums.sort()  
4         return nums[len(nums) // 2]
```

Time complexity: $O(n \log n)$

Space complexity: $O(1)$

Hash Map (Frequency Counting)

Use a hash map to count occurrences of each element and return the one that appears more than $\lfloor n / 2 \rfloor$ times.

```
1 class Solution:
2     def majorityElement(self, nums: List[int]) -> int:
3         count = {}
4         for num in nums:
5             count[num] = count.get(num, 0) + 1
6             if count[num] > len(nums) // 2:
7                 return num
```

Time complexity: O(n)

Space complexity: O(n)

Moore's Voting Algorithm

Maintain a candidate and a counter, incrementing or decrementing based on matches to efficiently find the majority in one pass.

```
1 class Solution:
2     def majorityElement(self, nums: List[int]) -> int:
3         candidate = 0
4         count = 0
5         for num in nums:
6             if count == 0:
7                 candidate = num
8             if candidate == num:
9                 count +=1
10            else:
11                count -=1
12        return candidate
```

Time complexity: $O(n)$

Space complexity: $O(1)$



278. First Bad Version

Solved

Easy

Topics

Companies

You are a product manager and currently leading a team to develop a new product. Unfortunately, the latest version of your product fails the quality check. Since each version is developed based on the previous version, all the versions after a bad version are also bad.

Suppose you have `n` versions `[1, 2, ..., n]` and you want to find out the first bad one, which causes all the following ones to be bad.

You are given an API `bool isBadVersion(version)` which returns whether `version` is bad. Implement a function to find the first bad version. You should minimize the number of calls to the API.

Example 1:

Input: `n = 5, bad = 4`

Output: `4`

Explanation:

`call isBadVersion(3) -> false`

`call isBadVersion(5) -> true`

`call isBadVersion(4) -> true`

Then `4` is the first bad version.

Example 2:

Input: `n = 1, bad = 1`

Output: `1`

Binary Search

Use binary search to efficiently find the first bad version by narrowing down the search range based on `isBadVersion(mid)`.

```
1 // The API isBadVersion is defined for you.
2 // bool isBadVersion(int version);
3
4 class Solution {
5 public:
6     int firstBadVersion(int n) {
7         int low = 0;
8         int high = n;
9         while (low <= high) {
10             int mid = low + (high - low)/2;
11             if (isBadVersion(mid) == false) {
12                 low = mid + 1;
13             }
14             else{
15                 high = mid -1;
16             }
17         }
18         return low;
19     }
20 };
```

Time complexity: $O(\log n)$
Space complexity: $O(1)$



Quiz



Reference

https://github.com/changgyhub/leetcode_101

<https://github.com/halfrost/LeetCode-Go>

<https://github.com/krahets/hello-algo>



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