



# Intro to Algorithmic Problem Solving

Week 3

# Housekeeping

- **Assignments due:** every Sunday 11:59pm
- **Soft Skill Assessment:** Post on June 27 (Fri); Due on July 6 (Sun)
- **Exam:** July 3 (Thu), 18:00-20:00
- **Exam language & make-up exam:** last call



# Mid-course Survey

How do you feel about the pace of the course so far?

- 1. Too fast**
- 2. Just right**
- 3. Too slow**

# Mid-course Survey

Do you feel you have enough time to think during practice questions?

- 1. Yes, it's sufficient**
- 2. No, I need more time**
- 3. I would prefer less time and more questions**

# Mid-course Survey

How helpful are the in-class examples in understanding the concepts?

- 1. Very helpful**
- 2. Somewhat helpful**
- 3. Not very helpful**

# Mid-course Survey

Do you find the explanations clear and easy to follow?

- 1. Always**
- 2. Most of the time**
- 3. Sometimes**
- 4. Rarely**

# Mid-course Survey

Is the balance between theory and practice appropriate?

- 1. Yes, it's balanced**
- 2. Too much theory**
- 3. Too much practice**

# Mid-course Survey

How do you feel about the difficulty level of the content?

- 1. Too easy**
- 2. Just right**
- 3. Too difficult**

# Mid-course Survey

How do you feel about the amount of content provided?

- 1. I would prefer more content**
- 2. The current amount is just right**
- 3. I would prefer less content with more detailed explanation**



# Mid-course Survey

Do you have any suggestions to improve the rest of the course?



# 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



# Math Problems



## 69. Sqrt(x)

Solved

Easy

Topics

Companies

Hint

Given a non-negative integer `x`, return *the square root of `x` rounded down to the nearest integer*. The returned integer should be **non-negative** as well.

You **must not use** any built-in exponent function or operator.

- For example, do not use `pow(x, 0.5)` in c++ or `x ** 0.5` in python.

### Example 1:

**Input:** `x = 4`

**Output:** 2

**Explanation:** The square root of 4 is 2, so we return 2.

### Example 2:

**Input:** `x = 8`

**Output:** 2

**Explanation:** The square root of 8 is 2.82842..., and since we round it down to the nearest integer, 2 is returned.

# Binary Search

Uses binary search to find the greatest integer whose square is less than or equal to x.

```
1 class Solution:
2     def mySqrt(self, x: int) -> int:
3         low, high = 1, x
4         while low <= high:
5             mid = (low + high)//2
6             if mid*mid == x:
7                 return mid
8             elif mid*mid > x:
9                 high = mid-1
10            else:
11                low = mid+1
12        return high
```

Time complexity:  $O(\log n)$

Space complexity:  $O(1)$

# Newton's Method

Applies Newton's iterative formula to quickly converge on the square root of x using integer approximation.

$$r_{n+1} = \left\lfloor \frac{r_n + \frac{x}{r_n}}{2} \right\rfloor$$

```
1 class Solution:  
2     def mySqrt(self, x: int) -> int:  
3         r = x  
4         while r*r > x:  
5             r = (r + x//r)//2  
6         return r
```

Time complexity:  $O(\log n)$   
\*faster than binary search

Space complexity:  $O(\log n)$



## 50. Pow(x, n)

Medium

Topics

Companies

Implement `pow(x, n)`, which calculates  $x$  raised to the power  $n$  (i.e.,  $x^n$ ).

### Example 1:

**Input:**  $x = 2.00000$ ,  $n = 10$

**Output:** 1024.00000

### Example 2:

**Input:**  $x = 2.10000$ ,  $n = 3$

**Output:** 9.26100

### Example 3:

**Input:**  $x = 2.00000$ ,  $n = -2$

**Output:** 0.25000

**Explanation:**  $2^{-2} = 1/2^2 = 1/4 = 0.25$

# Exponentiation by Squaring

Recursively computes  $x^n$  using divide-and-conquer by halving the exponent and squaring the base, which reduces time complexity to  $O(\log n)$ .

```
1 class Solution:
2     def myPow(self, x: float, n: int) -> float:
3         if not n:
4             return 1
5         if n < 0:
6             return 1 / self.myPow(x, -n)
7         if n % 2:
8             return x * self.myPow(x, n-1)
9         return self.myPow(x*x, n//2)
```

Time complexity:  $O(\log n)$

Space complexity:  $O(\log n)$



# Bit Manipulation

# Bit Manipulation

- A technique that works directly on the binary representation of integers using bitwise operators.
- **Bitwise operators:**

Operator	Name	Example	Explanation
&	Bitwise AND	$5 \& 3 \rightarrow 1$	$0101 \& 0011 = 0001 \rightarrow$ Only 1 if both bits are 1
	Bitwise OR	$5   3 \rightarrow 7$	$0101   0011 = 0111 \rightarrow$ 1 if either bit is 1
^	Bitwise XOR	$5 ^ 3 \rightarrow 6$	$0101 ^ 0011 = 0110 \rightarrow$ 1 if bits are different
<<	Shift Left	$3 << 1 \rightarrow 6$	$0011 << 1 = 0110 \rightarrow$ Shifts bits left (multiply by 2)
>>	Shift Right	$8 >> 2 \rightarrow 2$	$1000 >> 2 = 0010 \rightarrow$ Shifts bits right (divide by 4)
~	Bitwise NOT (Complement)	$\sim 5 \rightarrow -6$	$\sim 0101 = 1010 \rightarrow$ Inverts all bits (result is $-x - 1$ )



## 191. Number of 1 Bits

Easy

Topics

Companies

Given a positive integer  $n$ , write a function that returns the number of [set bits](#) in its binary representation (also known as the [Hamming weight](#)).

**Example 1:**

Input:  $n = 11$

Output: 3

Explanation:

The input binary string **1011** has a total of three set bits.

**Example 2:**

Input:  $n = 128$

Output: 1

Explanation:

The input binary string **10000000** has a total of one set bit.

**Example 3:**

Input:  $n = 2147483645$

Output: 30

Explanation:

The input binary string **111111111111111111111111111101** has a total of thirty set bits.

### Set Bit

d

A set bit refers to a bit in the binary representation of a number that has a value of  $1$ .

### Constraints:

- $1 \leq n \leq 2^{31} - 1$

# Bit Manipulation

Counts the number of 1s in the binary representation of a 32-bit integer by checking each bit with a right shift and bitwise AND.

```
1 class Solution:
2     def hammingWeight(self, n: int) -> int:
3         count = 0
4         for i in range(32):
5             if (n>>i & 1) == 1:
6                 count += 1
7         return count
```

Time complexity:  $O(32) = O(1)$

Space complexity:  $O(1)$

# Brian Kernighan's Algorithm

Repeatedly removes the lowest set bit from n, counting how many times it can do so until n becomes zero.

```
1 class Solution:  
2     def hammingWeight(self, n: int) -> int:  
3         count = 0  
4         while n:  
5             n &= n - 1  
6             count += 1  
7         return count
```

Time complexity:  $O(k)$   
\*k is the number of 1s

Space complexity:  $O(1)$



## 136. Single Number

Easy

Topics

Companies

Hint

Given a **non-empty** array of integers `nums`, every element appears *twice* except for one. Find that single one.

You must implement a solution with a linear runtime complexity and use only constant extra space.

### Example 1:

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

**Output:** 1

### Example 2:

**Input:** `nums = [4,1,2,1,2]`

**Output:** 4

### Example 3:

**Input:** `nums = [1]`

**Output:** 1

# Bit Manipulation

Uses the bitwise XOR operation to cancel out all numbers that appear twice, leaving only the unique number that appears once.

```
1 class Solution:
2     def singleNumber(self, nums: List[int]) -> int:
3         ans = 0
4         for i in nums:
5             ans ^= i
6         return ans
```

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



# Greedy Algorithm

# Greedy Algorithm

- A greedy algorithm makes a sequence of choices, each of which looks best at the moment, aiming for a local optimum **in hopes of** reaching a global optimum.
- **Key Characteristics:**
  - Makes locally optimal choices at each step (One strategy for the whole process)
  - No backtracking or reconsideration



# Activity Selection

- **Shiftkey Labs** is hosting 5 events on the same day. Each event has a **start time** and an **end time**, but some events overlap in schedule.
- Your goal is to attend as many non-overlapping events as possible – that is, you want to **maximize the number of events you can attend** without time conflicts.
- You can only attend **one event at a time**, and you must be present for the entire duration of any event you choose to attend.

 SHIFTKEY LABS	
Activity 1	8:00-9:00
Activity 2	9:30-11:00
Activity 3	10:00-11:30
Activity 4	12:00-14:00
Activity 5	13:30-14:30



# Activity Selection

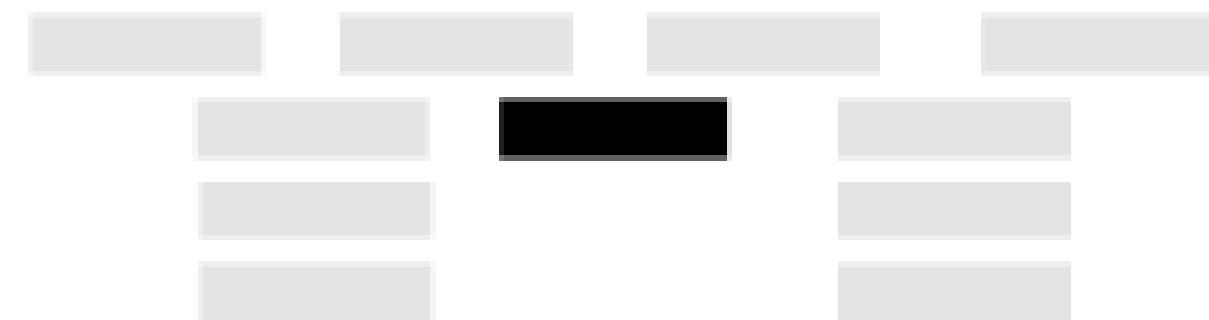
- Select activity with the shortest duration 



- Select the activity that starts earliest 



- Select the activity with the least number of conflicts 

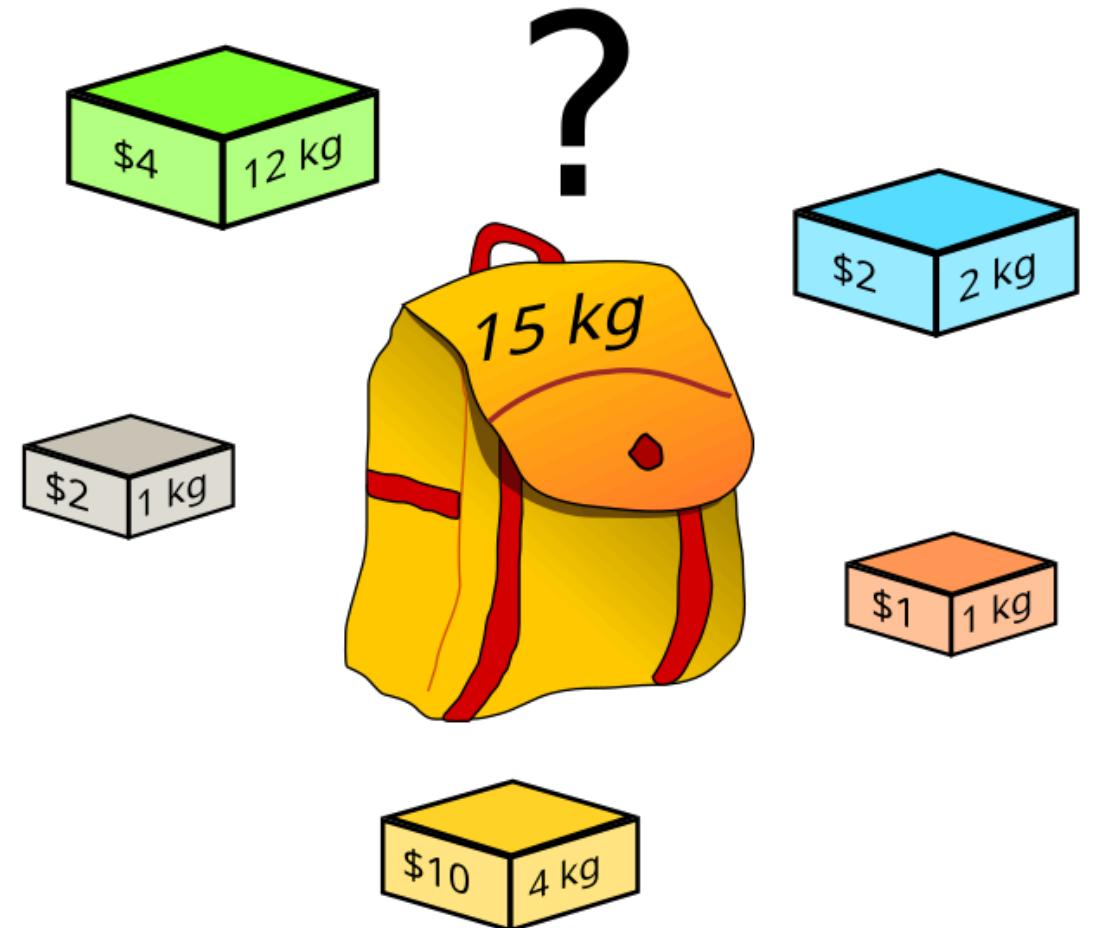


# Activity Selection

- **Select the activity that finishes earliest among the remaining compatible activities** 
- **Why it works:**
  - **Greedy-choice:** A globally optimal solution can be arrived at by choosing the local optimal solution.
  - **Optimal substructure:** The problem can be broken into subproblems that are themselves activity selection problems.

# Fractional Knapsack Problem

- You're preparing some snacks for a Shiftkey Labs' event.
- You have a limited backpack with **a weight capacity of 15 kg**, and there are several boxes of snacks you can choose from.
- Each box has a total **weight** and a **value**.
- You want to **maximize the total value** of snacks you bring – but here's the catch:
  - You're allowed to take partial boxes. That means you can open a box and only pack part of it, proportionally gaining part of its value.



# Fractional Knapsack Problem

- Always pack the item with the highest value-to-weight ratio first, and take as much of it as possible. 

$$\max_i \left( \frac{v_i}{w_i} \right)$$

- What if the items are not fractional?
  - Greedy may make an early choice that blocks a better combination later.  
→ In this case, you need **dynamic programming**



# Dynamic Programming

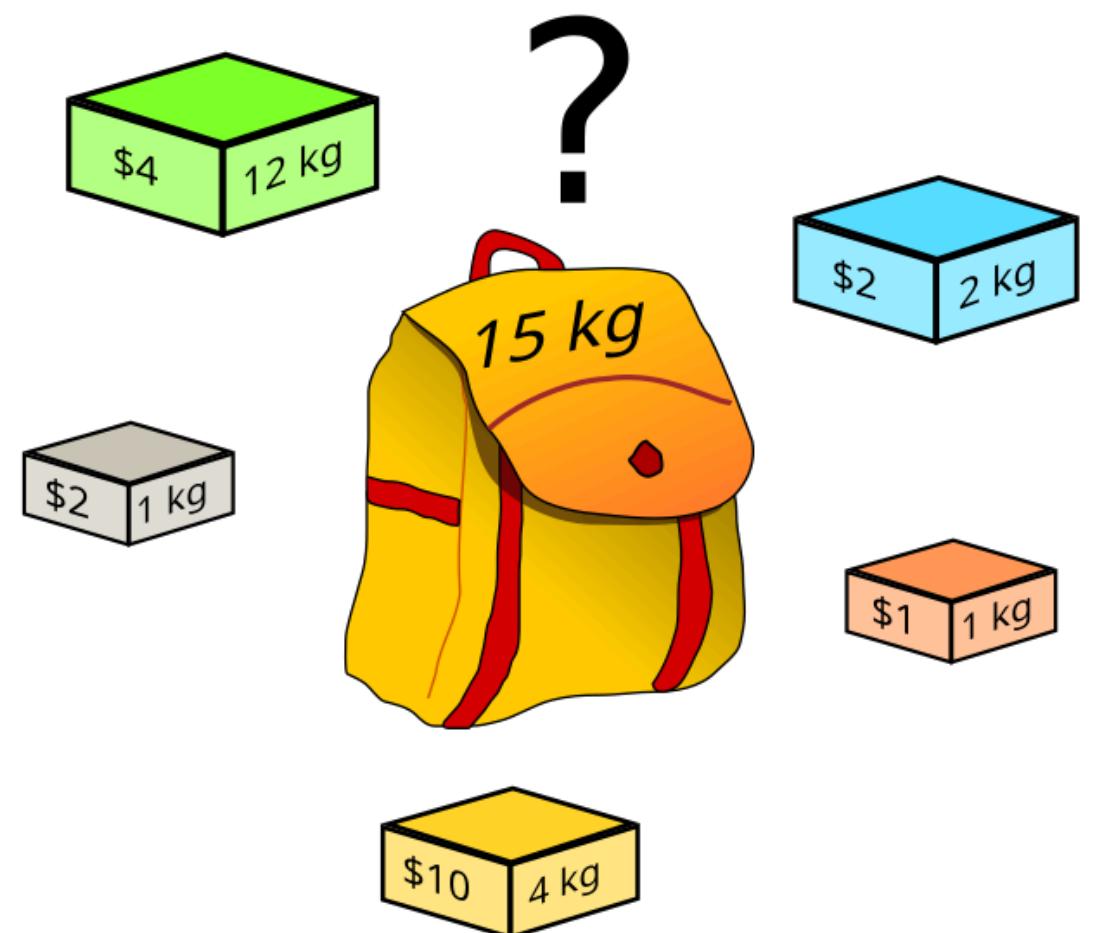
# Dynamic Programming

- Dynamic Programming solves problems by breaking them down into overlapping subproblems, storing results to avoid redundant work.
- **Key Characteristics:**
  - Solves each subproblem once, then reuses the answer
  - Requires optimal substructure and overlapping subproblems
  - Usually involves recursion + memoization or bottom-up tabulation



# 0-1 Knapsack Problem

- In this case, you cannot divide the items – you must either take the entire item or leave it.
- Your goal remains the same: to **maximize the total value** without exceeding the weight capacity of the knapsack.



# 0-1 Knapsack Problem

- Let  $dp[i][w]$  represent the maximum value achievable by using the first  $i$  items with a knapsack of capacity  $w$ .
- Base case:**

$$dp[0][w] = 0 \quad \text{for all } w$$

$$dp[i][0] = 0 \quad \text{for all } i$$

- Recurrence relation:**
  - If the current item's weight  $w_i$  is greater than the capacity  $w$ :

$$dp[i][w] = dp[i - 1][w]$$

- Otherwise:

Don't take the item

$$dp[i][w] = \max (dp[i - 1][w], dp[i - 1][w - w_i] + v_i)$$

Take the item

# 0-1 Knapsack Problem

```

1  capacity = 15
2  n = len(items)
3
4  # Initialize DP table
5  dp = [[0 for w in range(capacity + 1)] for i in range(n + 1)]
6
7  # Fill DP table
8  for i in range(1, n + 1):
9      value, weight = items[i - 1]
10     for w in range(capacity + 1):
11         if weight > w:
12             dp[i][w] = dp[i - 1][w]
13         else:
14             dp[i][w] = max(dp[i - 1][w], dp[i - 1][w - weight] + value)
15
16 # Final result
17 print("Maximum value:", dp[n][capacity])

```

Item	Value	Weight
1	4	12
2	2	2
3	2	1
4	1	1
5	10	4

Time complexity:  $O(n \cdot W)$   
 \*filling the  $n \times W$  table

Space complexity:  $O(n \cdot W)$

# 0-1 Knapsack Problem

Table dp

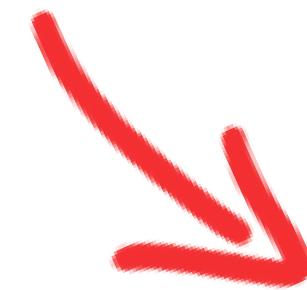
	W=0	W=1	W=2	W=3	W=4	W=5	W=6	W=7	W=8	W=9	W=10	W=11	W=12	W=13	W=14	W=15
Item 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Item 1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
Item 2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
Item 3	0	2	2	4	4	4	4	4	4	4	4	4	4	6	6	8
Item 4	0	2	3	4	5	5	5	5	5	5	5	5	5	6	7	8
Item 5	0	2	3	4	10	12	13	14	15	15	15	15	15	15	15	15

Answer



# 0-1 Knapsack Problem

```
for w in range(capacity + 1):
    if weight > w:
        dp[i][w] = dp[i - 1][w]
    else:
        dp[i][w] = max(dp[i - 1][w], dp[i - 1][w - weight] + value)
```



```
for w in range(capacity + 1):
    if weight > w:
        dp[i][w] = dp[i-1][w]          # cannot take item i
    else:
        without = dp[i-1][w]
        with_it = dp[i-1][w-weight] + value
        if with_it > without:         # choose item i
            dp[i][w] = with_it
            S[i][w] = 1
        else:                         # skip item i
            dp[i][w] = without
            S[i][w] = 0
```

- To keep tracking our choice.
- The table  $S[i][w]$  stores a 1 when item  $i$  is chosen for capacity  $w$ , and 0 otherwise.

# 0-1 Knapsack Problem

Table S

	W=0	W=1	W=2	W=3	W=4	W=5	W=6	W=7	W=8	W=9	W=10	W=11	W=12	W=13	W=14	W=15
item 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
item 1	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1
item 2	0	0	1	1	1	1	1	1	1	1	1	0	0	1	1	1
item 3	0	1	0	1	1	1	1	1	1	1	1	1	1	0	1	1
item 4	0	0	1	0	1	1	1	1	1	1	1	1	0	1	1	0
item 5	0	0	0	0	1	1	1	1	1	1	1	1	1	1	1	1

Answer: select item 2, item 3, item 4, item 5

# Fibonacci (Top-Down)

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

```
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)

# Fibonacci (Bottom-Up)

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

```
def fib(n):
    if n <= 1:
        return n
    dp = [0, 1]
    for i in range(2, n+1):
        dp.append(dp[i-1] + dp[i-2])
    return dp[n]
```

Time complexity: O(n)  
Space complexity: O(n)  
Optimization?

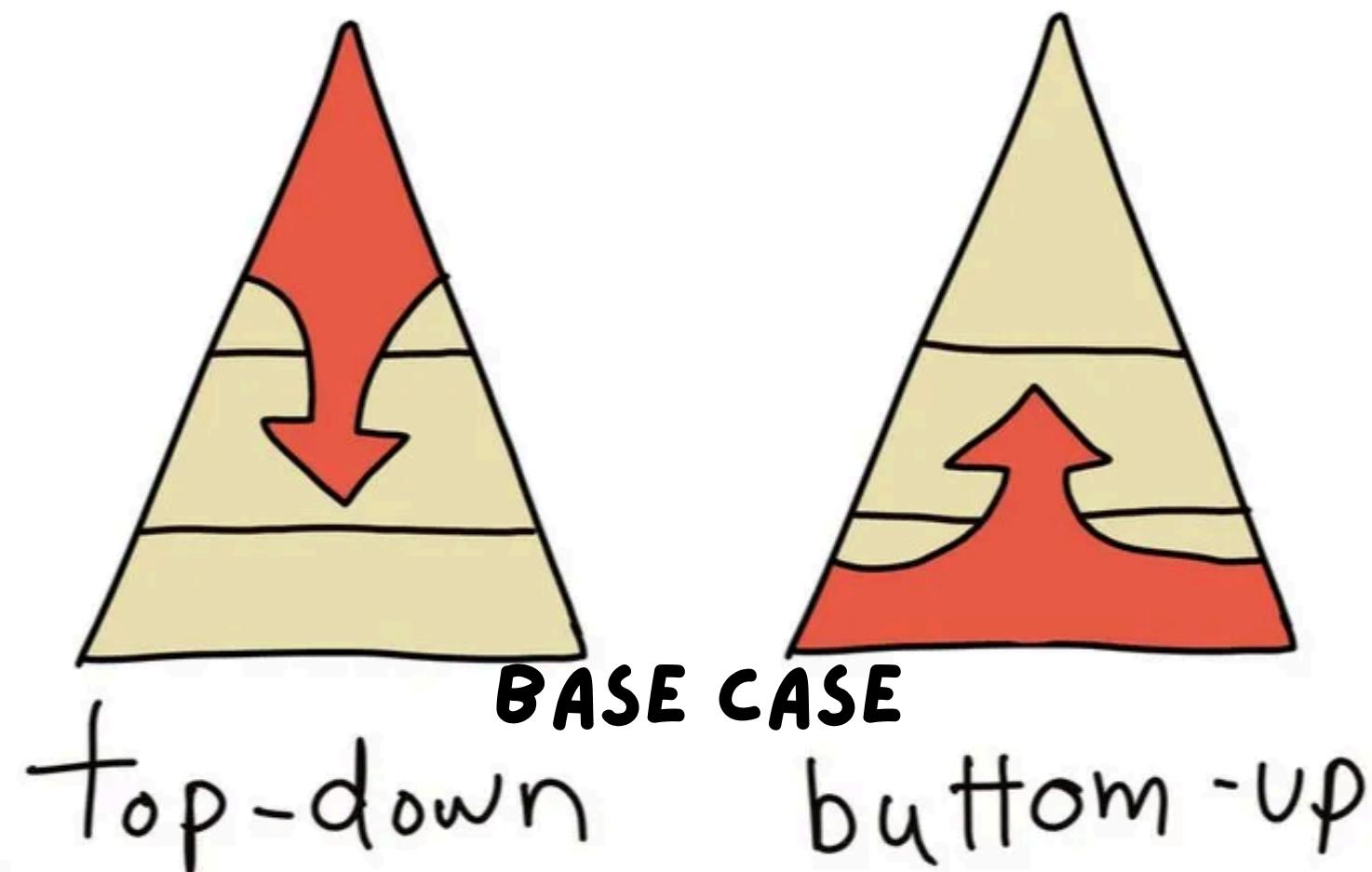
Only keep last two values → O(1)  
space

# 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).

# Top-Down vs. Bottom-Up

OUR PROBLEM



Q: Are Top-Down and Bottom-Up Opposites?

A: **Not exactly.** Even though recursive Top-Down algorithms call down, they solve up.

# Greedy VS. Dynamic Programming

## GREEDY



## DYNAMIC PROGRAMMING



	<b>Greedy Algorithm</b>	<b>Dynamic Programming</b>
<b>Core Idea</b>	Makes the best local choice at each step	Breaks the problem into overlapping subproblems, solves each optimally, and combines them for the global solution
<b>Decision Strategy</b>	Irrevocable choices; only considers the current step	Considers all possible subproblem combinations to ensure the best overall decision
<b>Overlapping Subproblems</b>	Typically none or minimal	Yes; subproblems recur and benefit from reuse
<b>Guarantee of Optimality</b>	Not always; may lead to suboptimal solutions	Always guarantees optimal solution
<b>Time Complexity</b>	Often faster, usually $O(n)$ or $O(n \log n)$	Usually slower, often $O(n^2)$ or worse
<b>Space Complexity</b>	Low; often $O(1)$ or $O(n)$	Higher; requires additional memory for subproblem solutions



# Practice Questions



## 55. Jump Game

Solved

Medium

Topics

Companies

You are given an integer array `nums`. You are initially positioned at the array's **first index**, and each element in the array represents your maximum jump length at that position.

Return `true` if you can reach the last index, or `false` otherwise.

### Example 1:

**Input:** `nums = [2,3,1,1,4]`

**Output:** `true`

**Explanation:** Jump 1 step from index 0 to 1, then 3 steps to the last index.

### Example 2:

**Input:** `nums = [3,2,1,0,4]`

**Output:** `false`

**Explanation:** You will always arrive at index 3 no matter what. Its maximum jump length is 0, which makes it impossible to reach the last index.

# Greedy

Greedily tracks the farthest index reachable at each step and returns False if the current position is beyond that reach.

```
1 class Solution:
2     def canJump(self, nums: List[int]) -> bool:
3         farthest = 0
4         n = len(nums)
5         for i in range(n):
6             if i > farthest:
7                 return False
8             farthest = max(farthest, i + nums[i])
9         return farthest >= n - 1
```

Time complexity:  $O(n)$

Space complexity:  $O(1)$



## 121. Best Time to Buy and Sell Stock

Solved

Easy Topics Companies

You are given an array `prices` where `prices[i]` is the price of a given stock on the  $i^{\text{th}}$  day.

You want to maximize your profit by choosing a **single day** to buy one stock and choosing a **different day in the future** to sell that stock.

Return *the maximum profit you can achieve from this transaction*. If you cannot achieve any profit, return `0`.

### Example 1:

**Input:** `prices = [7,1,5,3,6,4]`

**Output:** 5

**Explanation:** Buy on day 2 (price = 1) and sell on day 5 (price = 6), profit =  $6-1 = 5$ .

Note that buying on day 2 and selling on day 1 is not allowed because you must buy before you sell.

### Example 2:

**Input:** `prices = [7,6,4,3,1]`

**Output:** 0

**Explanation:** In this case, no transactions are done and the max profit = 0.

# Dynamic Programming

Finds the maximum profit from a single stock buy-sell transaction by tracking the lowest price seen so far and calculating the best profit at each step.

```
1 class Solution:  
2     def maxProfit(self, prices: List[int]) -> int:  
3         lsf = inf  
4         op = 0  
5         for i in range(len(prices)):  
6             lsf = min(lsf, prices[i])  
7             op = max(op, prices[i]-lsf)  
8         return op
```

Time complexity:  $O(n)$

Space complexity:  $O(1)$

## 62. Unique Paths

Solved 

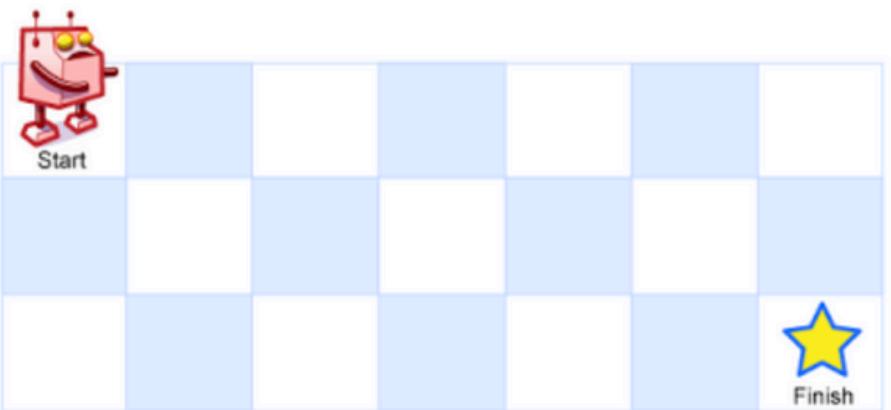
Medium  Topics  Companies

There is a robot on an  $m \times n$  grid. The robot is initially located at the **top-left corner** (i.e., `grid[0][0]`). The robot tries to move to the **bottom-right corner** (i.e., `grid[m - 1][n - 1]`). The robot can only move either down or right at any point in time.

Given the two integers  $m$  and  $n$ , return *the number of possible unique paths that the robot can take to reach the bottom-right corner*.

The test cases are generated so that the answer will be less than or equal to  $2 * 10^9$ .

**Example 1:**



**Input:**  $m = 3$ ,  $n = 7$

**Output:** 28

**Example 2:**

**Input:**  $m = 3$ ,  $n = 2$

**Output:** 3

**Explanation:** From the top-left corner, there are a total of 3 ways to reach the bottom-right corner:

1. Right  $\rightarrow$  Down  $\rightarrow$  Down
2. Down  $\rightarrow$  Down  $\rightarrow$  Right
3. Down  $\rightarrow$  Right  $\rightarrow$  Down

# Dynamic Programming(2D)

Use a 2D array  $dp[i][j]$  to store the number of unique paths to reach cell  $(i, j)$ .

```
1 class Solution:
2     def uniquePaths(self, m: int, n: int) -> int:
3         dp = [[1]*n for _ in range(m)]
4         for i in range(1,m):
5             for j in range(1,n):
6                 dp[i][j] = dp[i-1][j] + dp[i][j-1]
7         return dp[m-1][n-1]
```

Time complexity:  $O(n*m)$

Space complexity:  $O(n*m)$

# Dynamic Programming(1D)

Reduce space to 1D since each row only depends on the previous row.

```
1 class Solution:  
2     def uniquePaths(self, m: int, n: int) -> int:  
3         dp = [1]*n  
4         for i in range(1,m):  
5             for j in range(1,n):  
6                 dp[j] += dp[j-1]  
7         return dp[-1]
```

Time complexity:  $O(n*m)$

Space complexity:  $O(n)$



# Quiz



# Reference

[https://github.com/changgyhub/leetcode\\_101](https://github.com/changgyhub/leetcode_101)

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

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



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