377. Combination Sum IV

Medium <u>Array</u> <u>Dynamic Programming</u>

Problem Description

For example, if nums = [1, 2, 3] and target = 4, there are seven combinations that you could use to get a sum of 4:

In this problem, you are given an array of unique integers nums and a target integer target. Your task is to find out how many

different combinations of elements from nums can be added together to sum up to the target value. Importantly, combinations

• 1+1+1+1

may use the same element from **nums** multiple times if needed.

• 1+1+2 • 1+2+1

• 2+1+1 • 2+2 • 1+3

• 3+1 The results must fit within a 32-bit integer range, meaning they should be between $-(2^31)$ and $(2^31)-1$.

Intuition

The intuition for solving this problem comes from recognizing that it is similar to computing the number of ways to make change for a certain amount of money given different coin denominations. Here, each integer in nums can be thought of as a coin denomination.

To arrive at the solution, we apply <u>dynamic programming</u> because the problem asks for counting combinations, which suggests

The dp array is created where each index i represents the target sum, and the value at that index represents the number of ways to reach that sum using elements from $\frac{1}{2}$ nums. The base case is $\frac{dp[0]}{dp} = 1$, indicating there's one way to reach a sum of 0, which is using no elements. We iterate through each sub-target from 1 to the desired target, and for each, we look at all elements in nums. If the current

is finding the number of ways to reach a smaller target, which we can use to build up the solution to a larger target.

overlapping subproblems and the need for memorization of previous results to optimize the computation. Each subproblem here

element is less than or equal to the sub-target, we add to dp[i] the value of dp[i - x], where x is the value of the current element from $\frac{1}{1}$ nums. This signifies that all the ways to reach the sum $\frac{1}{1-x}$ can contribute to ways to reach the sum $\frac{1}{1}$ by adding $\frac{1}{x}$.

By the end of the iterations, dp[target] contains the desired number of combinations. **Solution Approach**

The solution approach utilizes dynamic programming, a method for solving complex problems by breaking them down into

simpler sub-problems. It is a helpful strategy when a problem has overlapping subproblems and optimal substructure, meaning

nums. Whenever the current number x is less than or equal to the sub-target i, we update dp[i] by adding the number of ways

to form the sum i-x. This is based on the assumption that if we have a way to arrive at a sum of i-x, then by adding x to it, we

In simpler terms, to solve for dp[i] which represents the number of combinations to reach a sum of i, we look at all numbers x

In our case, we use an array dp to store the number of combinations that sum up to each value up to target. We initialize dp [0] = 1 because there's exactly one combination to achieve a sum of 0: using no numbers from nums.

the problem can be broken down into smaller, simpler subproblems which can be solved independently.

The outer loop iterates through all sub-targets from 1 to target, and the inner loop goes through all the available numbers in

can get to i.

class Solution:

Example Walkthrough

in nums that could contribute to i and sum up all the combinations that make up the sum i-x (all of which are valid ways to reach i when adding x). This solution's algorithm can be summarized in the following steps:

Define an array dp of length target + 1 and initialize it with zeros. dp[0] is set to 1 since there's one combination that

 Inside this loop, iterate through each number x in nums. ■ If x is less than or equal to i, increment dp[i] by dp[i - x]. After the loops finish, dp[target] will hold the number of ways to combine the numbers in nums to sum up to the target. The Python code implementing the solution is as follows:

In this implementation, dp is initialized with size target+1 to accommodate sums from 0 to target inclusive. The two nested

dp[i] += dp[i - x]

Let's illustrate the solution approach with a smaller example.

Loop through each sub-target i from 1 to target.

For each sub-target i, consider each number x from nums.

 \circ We update dp[2] to be dp[2] + dp[2 - 1] = dp[2] + dp[1] = 0 + 1 = 1.

 \circ Update dp[4] with 1 to be dp[4] + dp[4 - 1] = dp[4] + dp[3] = 0 + 2 = 2.

 \circ Update dp[4] with 4: dp[4] + dp[4 - 4] = dp[4] + dp[0] = 2 + 1 = 3.

 \circ Update dp[5] with 1: dp[5] + dp[5 - 1] = dp[5] + dp[4] = 0 + 3 = 3.

For i = 1: The possible number from nums to use is 1.

dp = [1] + [0] * target

for x in nums:

return dp[target]

if i >= x:

for i in range(1, target + 1):

results in a sum of zero, which is not using any numbers.

def combinationSum4(self, nums: List[int], target: int) -> int:

Loop through each sub-target i from 1 to target.

Suppose nums = [1, 3, 4] and target = 5. We want to find out how many different combinations of elements from nums can be added together to sum up to 5. Here's how we would complete the task step by step: Initialize a dp array with target + 1 zeros and set dp[0] to 1, because there is exactly one way to achieve the sum of 0 (using no elements). After initialization, our dp array looks like this: dp = [1, 0, 0, 0, 0, 0]

loops are used to populate the dp array according to the dynamic programming strategy outlined above.

\circ We update dp[1] to be dp[1] + dp[1 - 1] = dp[1] + dp[0] = 0 + 1 = 1. For i = 2: The possible number from nums to use is 1.

 \bullet 1 + 1 + 3

 \bullet 1 + 3 + 1

• 3 + 1 + 1

• 1 + 4

Python

from typing import List

class Solution:

Example usage:

class Solution {

Java

C++

solution = Solution()

Let's fill the dp array:

For i = 3: We can use 1 and 3 from nums.

For i = 5: We can use 1, 3, and 4.

• We can update dp[3] to be dp[3] + dp[3 - 1] (using 1) = dp[3] + dp[2] = 0 + 1 = 1. • We can also update dp[3] using 3 to be dp[3] + dp[3 - 3] = dp[3] + dp[0] = 1 + 1 = 2. For i = 4: We can use 1, 3, and 4 from nums.

The final dp array is [1, 1, 1, 2, 3, 5], and dp[5] is 5, meaning there are five different combinations to reach the target sum

 \circ Update dp[5] with 3: dp[5] + dp[5 - 3] = dp[5] + dp[2] = 3 + 1 = 4. \circ Update dp[5] with 4: dp[5] + dp[5 - 4] = dp[5] + dp[1] = 4 + 1 = 5.

○ Update dp[4] with 3: No change, since dp[4 - 3] is zero.

Those combinations are: • 1 + 1 + 1 + 1 + 1

of 5 using elements from nums [1, 3, 4] with repetition allowed.

No other combinations are possible without exceeding the target, so the final answer is 5. Solution Implementation

def combinationSum4(self, nums: List[int], target: int) -> int:

combinations = [1] + [0] * target

for i in range(1, target + 1):

if i >= num:

return combinations[target]

print(solution.combinationSum4([1,2,3], 4)) # Output: 7

for (int i = 1; i <= target; ++i) {</pre>

for (int num : nums) {

for num in nums:

public int combinationSum4(int[] nums, int target) { // dp represents the number of combinations to make up each value from 0 up to target int[] dp = new int[target + 1]; // There is exactly one combination to make up the target 0, which is to choose nothing

Initialize a list to hold the count of combinations for each value up to the target.

Iterate through each value from 1 to target to find the combinations.

combinations[i] += combinations[i - num]

// Iterate over all values from 1 to target to find combinations

// If the number is less than or equal to the current target (i)

// Iterate through all numbers in the given array

// then we can use it to form a combination

// Return the number of combinations to form the target

function combinationSum4(nums: number[], target: number): number {

// There is one way to reach 0, which is by using no numbers

def combinationSum4(self, nums: List[int], target: int) -> int:

// Loop through all numbers from 1 to target

if (currentTarget >= num) {

combinations = [1] + [0] * target

for i in range(1, target + 1):

if i >= num:

return combinations[target]

Time and Space Complexity

print(solution.combinationSum4([1,2,3], 4)) # Output: 7

for num in nums:

combinationCounts[0] = 1;

for (const num of nums) {

// Initialize an array of length target + 1 to store the number of ways

for (let currentTarget = 1; currentTarget <= target; ++currentTarget) {</pre>

// Check each number in nums to see if it can be used to reach currentTarget

// num can contribute to a combination that adds up to currentTarget

// Increase the number of combinations for currentTarget by the number of combinations

combinationCounts[currentTarget] += combinationCounts[currentTarget - num];

Initialize a list to hold the count of combinations for each value up to the target.

Iterate through each value from 1 to target to find the combinations.

combinations[i] += combinations[i - num]

Return the number of combinations that add up to the target value.

combinations[0] is 1 because there's one way to have a total of 0: by choosing nothing.

For each number in the nums list, check whether it can be used in a combination.

If the current number can be used in a combination for the current total (i).

add the number of combinations without this number (i.e., combinations[i - num]).

// that result in the value (currentTarget - num), since num can nicely top up this value

// If currentTarget is at least as large as num, it means that

// to reach every number up to target using the given numbers in nums

const combinationCounts: number[] = new Array(target + 1).fill(0);

Return the number of combinations that add up to the target value.

combinations[0] is 1 because there's one way to have a total of 0: by choosing nothing.

For each number in the nums list, check whether it can be used in a combination.

If the current number can be used in a combination for the current total (i),

add the number of combinations without this number (i.e., combinations[i - num]).

if (i >= num) { // Add the number of combinations from the previous value (i - num) // to the current number of combinations dp[i] += dp[i - num];

return dp[target];

dp[0] = 1;

#include <vector> #include <climits> class Solution { public: int combinationSum4(vector<int>& nums, int target) { // Create a dp vector to store the number of combinations for each value up to the target // dp[i] will represent the number of ways to reach the sum i vector<int> dp(target + 1, 0); // There is one combination to reach the sum of 0 which is by choosing no element dp[0] = 1;// Compute the number of combinations for each sum from 1 to target for (int i = 1; i <= target; ++i) {</pre> // Check each number in the array to see if it can be used to reach the current sum (i) for (int num : nums) { // if the current sum minus the current number is non-negative, // and adding would not cause integer overflow if $(i \ge num \&\& dp[i - num] < INT MAX - dp[i]) {$ // Increment the current sum's combination count by // the combination count of the (current sum — current number) dp[i] += dp[i - num];// Return the total number of combinations to reach the target sum return dp[target]; **}**; **TypeScript**

// Return the total number of ways to reach the target using numbers from nums return combinationCounts[target];

from typing import List

class Solution:

Example usage:

solution = Solution()

Time Complexity

for x in nums:

if i >= x:

f[i] += f[i - x]

We have an outer loop:

for i in range(1, target + 1): This loop runs once for every integer from 1 to target, inclusive, so it runs target times. Inside the outer loop, we have an inner loop:

The time complexity of the given code can be analyzed based on the nested loops.

This loop iterates over all elements in nums. If n is the number of elements in nums, the inner loop runs n times for each iteration of the outer loop.

the worst case, though, where i is always greater than or equal to every element x in nums, the body of the inner loop will execute.

The space complexity is driven by the list f that has target + 1 elements: f = [1] + [0] * target

Space Complexity

This means we need a space proportional to (and linear with) target. Hence, the space complexity of the algorithm is

O(target). No other data structures that scale with the input size is used, so the fixed-size variables and inputs do not affect the overall space complexity significantly compared to the list f which dominates the space usage.

However, not all iterations of the inner loop will execute the update f[i] += f[i - x]. The condition i >= x needs to be met. In

Therefore, the worst-case time complexity is the product of the number of iterations of both loops, which is 0(target * n).

Analyzing space complexity involves looking at how much additional memory the algorithm uses as a function of the input size.