377. Combination Sum IV Medium Array **Dynamic Programming**

Leetcode Link

Problem Description

different combinations of elements from nums can be added together to sum up to the target value. Importantly, combinations may use the same element from nums multiple times if needed.

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

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

```
• 1+1+1+1
• 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 $(2^31)-1$.

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

The intuition for solving this problem comes from recognizing that it is similar to computing the number of ways to make change for a

Intuition

To arrive at the solution, we apply dynamic programming because the problem asks for counting combinations, which suggests overlapping subproblems and the need for memorization of previous results to optimize the computation. Each subproblem here is

certain amount of money given different coin denominations. Here, each integer in nums can be thought of as a coin denomination.

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 nums. The base case is dp[0] = 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 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 nums. This signifies that all the ways to reach the sum i-x can contribute to ways to reach the sum i by adding x.

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 the problem

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

because there's exactly one combination to achieve a sum of 0: using no numbers from nums.

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

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

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 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).

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

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 can get to i.

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

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

2. Loop through each sub-target i from 1 to target. Inside this loop, iterate through each number x in nums.

3. 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:

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

return dp[target]

Example Walkthrough

elements).

Let's fill the dp array:

■ If x is less than or equal to i, increment dp[i] by dp[i - x].

This solution's algorithm can be summarized in the following steps:

```
class Solution:
    def combinationSum4(self, nums: List[int], target: int) -> int:
        dp = [1] + [0] * target
        for i in range(1, target + 1):
            for x in nums:
                if i >= x:
```

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

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

```
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:
 1. 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
```

After initialization, our dp array looks like this: dp = [1, 0, 0, 0, 0, 0]

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

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

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

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

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

```
\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.
```

```
\circ We can update dp[3] to be dp[3] + dp[3 - 1] (using 1) = dp[3] + dp[2] = 0 + 1 = 1.
```

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

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

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

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

// dp represents the number of combinations to make up each value from 0 up to target

// There is exactly one combination to make up the target 0, which is to choose nothing

// Create a dp vector to store the number of combinations for each value up to the target

// Check each number in the array to see if it can be used to reach the current sum (i)

// There is one combination to reach the sum of 0 which is by choosing no element

// if the current sum minus the current number is non-negative,

// the combination count of the (current sum - current number)

// Increment the current sum's combination count by

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]).

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

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

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

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

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

      \circ Update dp[4] with 4: dp[4] + dp[4 - 4] = dp[4] + dp[0] = 2 + 1 = 3.
  • For i = 5: We can use 1, 3, and 4.
      \circ Update dp[5] with 1: dp[5] + dp[5 - 1] = dp[5] + dp[4] = 0 + 3 = 3.
      \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.
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 of 5
using elements from nums [1, 3, 4] with repetition allowed.
Those combinations are:
  • 1 + 1 + 1 + 1 + 1
  \bullet 1 + 1 + 3
  \bullet 1 + 3 + 1
  \bullet 3 + 1 + 1
  \bullet 1 + 4
```

combinations[0] is 1 because there's one way to have a total of 0: by choosing nothing. combinations = [1] + [0] * target # Iterate through each value from 1 to target to find the combinations. for i in range(1, target + 1): 10

```
Java Solution
```

class Solution {

21 # Example usage:

22 # solution = Solution()

Python Solution

class Solution:

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1 from typing import List

for num in nums:

if i >= num:

return combinations[target]

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

public int combinationSum4(int[] nums, int target) {

int combinationSum4(vector<int>& nums, int target) {

dp[i] += dp[i - num];

vector<int> dp(target + 1, 0);

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

for (int num : nums) {

dp[0] = 1;

return dp[target];

// dp[i] will represent the number of ways to reach the sum i

// Compute the number of combinations for each sum from 1 to target

// and adding would not cause integer overflow

if $(i \ge num \&\& dp[i - num] < INT_MAX - dp[i]) {$

// Return the total number of combinations to reach the target sum

int[] dp = new int[target + 1];

```
dp[0] = 1;
           // Iterate over all values from 1 to target to find combinations
 9
           for (int i = 1; i <= target; ++i) {</pre>
10
               // Iterate through all numbers in the given array
                for (int num : nums)
13
                   // If the number is less than or equal to the current target (i)
                   // then we can use it to form a combination
14
                   if (i >= num) {
15
                       // Add the number of combinations from the previous value (i - num)
16
                       // to the current number of combinations
17
                        dp[i] += dp[i - num];
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23
           // Return the number of combinations to form the target
24
            return dp[target];
26
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C++ Solution
 1 #include <vector>
   #include <climits>
   class Solution {
```

```
31 };
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```

Typescript Solution

public:

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```
function combinationSum4(nums: number[], target: number): number {
       // Initialize an array of length target + 1 to store the number of ways
       // to reach every number up to target using the given numbers in nums
       const combinationCounts: number[] = new Array(target + 1).fill(0);
 6
       // There is one way to reach 0, which is by using no numbers
       combinationCounts[0] = 1;
 8
       // Loop through all numbers from 1 to target
 9
       for (let currentTarget = 1; currentTarget <= target; ++currentTarget) {</pre>
10
           // Check each number in nums to see if it can be used to reach currentTarget
11
12
           for (const num of nums) {
13
               // If currentTarget is at least as large as num, it means that
               // num can contribute to a combination that adds up to currentTarget
14
               if (currentTarget >= num) {
15
                   // Increase the number of combinations for currentTarget by the number of combinations
16
                   // that result in the value (currentTarget - num), since num can nicely top up this value
17
                   combinationCounts[currentTarget] += combinationCounts[currentTarget - num];
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23
       // Return the total number of ways to reach the target using numbers from nums
       return combinationCounts[target];
24
25 }
26
Time and Space Complexity
Time Complexity
```

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:

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

We have an outer loop:

outer loop.

Space Complexity

```
for x in nums:
    if i >= x:
        f[i] += f[i - x]
```

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

```
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 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.
```

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

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

Analyzing space complexity involves looking at how much additional memory the algorithm uses as a function of the input size. The space complexity is driven by the list f that has target + 1 elements:

```
1 f = [1] + [0] * target
This means we need a space proportional to (and linear with) target. Hence, the space complexity of the algorithm is 0(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.