

# 377. Combination Sum IV

Medium   Array   Dynamic Programming

## Problem Description

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

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  $(2^{31})-1$ .

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

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`.

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

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 outer loop iterates through all sub-targets from 1 to `target`, and the inner loop goes through all the available numbers in `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 can get to `i`.

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`).

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

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

```
1 class Solution:
2     def combinationSum4(self, nums: List[int], target: int) -> int:
3         dp = [1] + [0] * target
4         for i in range(1, target + 1):
5             for x in nums:
6                 if i >= x:
7                     dp[i] += dp[i - x]
8         return dp[target]
```

In this implementation, `dp` is initialized with size `target+1` to accommodate sums from 0 to `target` inclusive. The two nested loops are used to populate the `dp` array according to the [dynamic programming](#) strategy outlined above.

## Example Walkthrough

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 elements).

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

2. Loop through each sub-target `i` from 1 to `target`.
3. For each sub-target `i`, consider each number `x` from `nums`.

Let's fill the `dp` array:

- For `i = 1`: The possible number from `nums` to use is 1.
  - 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.
  - We update `dp[2]` to be `dp[2] + dp[2 - 1] = dp[2] + dp[1] = 0 + 1 = 1`.
- For `i = 3`: We can use 1 and 3 from `nums`.
  - 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`.
  - 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.
  - 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.
  - Update `dp[5]` with 1: `dp[5] + dp[5 - 1] = dp[5] + dp[4] = 0 + 3 = 3`.
  - Update `dp[5]` with 3: `dp[5] + dp[5 - 3] = dp[5] + dp[2] = 3 + 1 = 4`.
  - 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
- 1 + 1 + 3
- 1 + 3 + 1
- 3 + 1 + 1
- 1 + 4

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

## Python Solution

```
1 from typing import List
2
3 class Solution:
4     def combinationSum4(self, nums: List[int], target: int) -> int:
5         # Initialize a list to hold the count of combinations for each value up to the target.
6         # combinations[0] is 1 because there's one way to have a total of 0: by choosing nothing.
7         combinations = [1] + [0] * target
8
9         # Iterate through each value from 1 to target to find the combinations.
10        for i in range(1, target + 1):
11            # For each number in the nums list, check whether it can be used in a combination.
12            for num in nums:
13                # If the current number can be used in a combination for the current total (i),
14                # add the number of combinations without this number (i.e., combinations[i - num]).
15                if i >= num:
16                    combinations[i] += combinations[i - num]
17
18        # Return the number of combinations that add up to the target value.
19        return combinations[target]
20
21 # Example usage:
22 # solution = Solution()
23 # print(solution.combinationSum4([1,2,3], 4)) # Output: 7
24
```

## Java Solution

```
1 class Solution {
2     public int combinationSum4(int[] nums, int target) {
3         // dp represents the number of combinations to make up each value from 0 up to target
4         int[] dp = new int[target + 1];
5
6         // There is exactly one combination to make up the target 0, which is to choose nothing
7         dp[0] = 1;
8
9         // Iterate over all values from 1 to target to find combinations
10        for (int i = 1; i <= target; ++i) {
11            // Iterate through all numbers in the given array
12            for (int num : nums) {
13                // If the number is less than or equal to the current target (i)
14                // then we can use it to form a combination
15                if (i >= num) {
16                    // Add the number of combinations from the previous value (i - num)
17                    // to the current number of combinations
18                    dp[i] += dp[i - num];
19                }
20            }
21        }
22
23        // Return the number of combinations to form the target
24        return dp[target];
25    }
26 }
27
```

## C++ Solution

```
1 #include <vector>
2 #include <climits>
3
4 class Solution {
5 public:
6     int combinationSum4(vector<int>& nums, int target) {
7         // Create a dp vector to store the number of combinations for each value up to the target
8         // dp[i] will represent the number of ways to reach the sum i
9         vector<int> dp(target + 1, 0);
10
11        // There is one combination to reach the sum of 0 which is by choosing no element
12        dp[0] = 1;
13
14        // Compute the number of combinations for each sum from 1 to target
15        for (int i = 1; i <= target; ++i) {
16            // Check each number in the array to see if it can be used to reach the current sum (i)
17            for (int num : nums) {
18                // If the current sum minus the current number is non-negative,
19                // and adding would not cause integer overflow
20                if (i >= num && dp[i - num] < INT_MAX - dp[i]) {
21                    // Increment the current sum's combination count by
22                    // the combination count of the (current sum - current number)
23                    dp[i] += dp[i - num];
24                }
25            }
26        }
27
28        // Return the total number of combinations to reach the target sum
29        return dp[target];
30    }
31 };
32
```

## Typescript Solution

```
1 function combinationSum4(nums: number[], target: number): number {
2     // Initialize an array of length target + 1 to store the number of ways
3     // to reach every number up to target using the given numbers in nums
4     const combinationCounts: number[] = new Array(target + 1).fill(0);
5
6     // There is one way to reach 0, which is by using no numbers
7     combinationCounts[0] = 1;
8
9     // Loop through all numbers from 1 to target
10    for (let currentTarget = 1; currentTarget <= target; ++currentTarget) {
11        // Check each number in nums to see if it can be used to reach currentTarget
12        for (const num of nums) {
13            // If currentTarget is at least as large as num, it means that
14            // num can contribute to a combination that adds up to currentTarget
15            if (currentTarget >= num) {
16                // Increase the number of combinations for currentTarget by the number of combinations
17                // that result in the value (currentTarget - num), since num can nicely top up this value
18                combinationCounts[currentTarget] += combinationCounts[currentTarget - num];
19            }
20        }
21    }
22
23    // Return the total number of ways to reach the target using numbers from nums
24    return combinationCounts[target];
25 }
26
```

## Time and Space Complexity

### Time Complexity

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

We have an outer loop:

```
1 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:

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

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.

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  $O(\text{target} * n)$ .

### Space Complexity

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  $O(\text{target})$ .

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.