



## Experiment- 6A

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**Semester: 6<sup>TH</sup>**

**Date of Performance: 20/02/25**

**Subject Name: AP Lab-2**

**Subject Code: 22CSH-352**

### 1. TITLE:

Climbing Stairs.

### 2. AIM:

You are climbing a staircase. It takes  $n$  steps to reach the top.

### 3. Algorithm

- **Define a DP array**  $dp$  where  $dp[i]$  represents the number of distinct ways to reach the  $i$ -th stair.

- Initialize base cases:

$dp[0] = 1 \rightarrow$  There is 1 way to stay at the ground without climbing.

$dp[1] = 1 \rightarrow$  There is 1 way to reach the first stair (taking a single step).

- **Iterate from  $i = 2$  to  $n$**  and use the recurrence relation:

$dp[i] = dp[i-1] + dp[i-2]$

- **Return  $dp[n]$** , which contains the total number of ways to reach the  $n$ -th stair.

### Implementation/Code

```
class Solution {
public:
    int climbStairs(int n) {
        // dp[i] := the number of ways to climb to the i-th stair
        vector<int> dp(n + 1);
        dp[0] = 1;
        dp[1] = 1;

        for (int i = 2; i <= n; ++i)
            dp[i] = dp[i - 1] + dp[i - 2];
    }
};
```

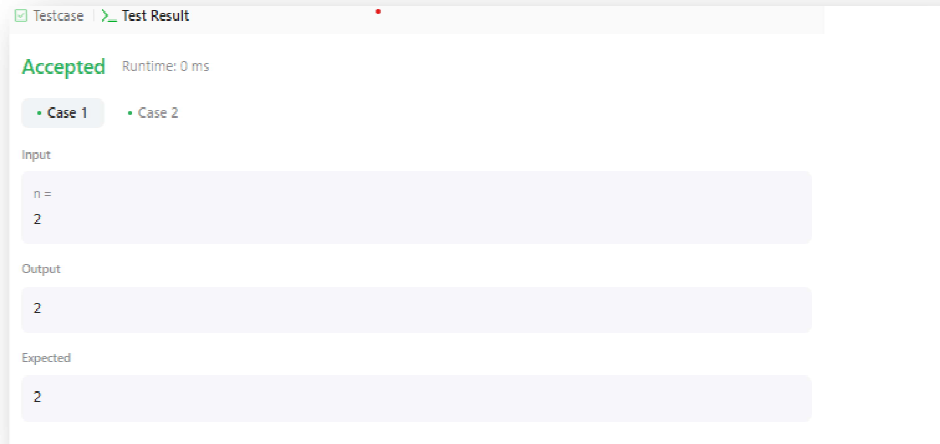


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```
return dp[n];  
}  
};
```

## Output:



**Time Complexity :  $O(n)$**

**Space Complexity :  $O(n)$**

## Learning Outcomes:-

- The given solution is a **bottom-up DP** approach, where smaller subproblems are solved first.
- This makes it an ideal candidate for **DP** rather than a naive recursive approach (which has exponential complexity).



## Experiment - 6B

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Semester: 6<sup>TH</sup>

Date of Performance: 20/02/25

Subject Name: AP Lab-2

Subject Code: 22CSH-352

### 1. TITLE:

Maximum Subarray.

### 2. AIM:

Given an integer array `nums`, find the `subarray` with the largest sum, and return *its sum*.

### 3. Algorithm

- Define DP array `dp[i]`, where:

`dp[i]` represents the **maximum sum subarray that ends at index `i`**.

This ensures that every subarray considered **includes `nums[i]`**

- Base Case:  
`dp[0] = nums[0]` → The maximum sum subarray ending at index 0 is the element itself.
- State Transition (Recurrence Relation):  
For each `i` from 1 to `n-1`, compute:  
$$dp[i] = \max(nums[i], dp[i-1] + nums[i])$$
- Final Answer:  
The overall maximum sum subarray is obtained by computing:  $\max(dp[0], dp[1], \dots, dp[n-1])$

### Implementation/Code:

```
class Solution {
public:
    int maxSubArray(vector<int>& nums) {
        // dp[i] := the maximum sum subarray ending in i
        vector<int> dp(nums.size());

        dp[0] = nums[0];
```



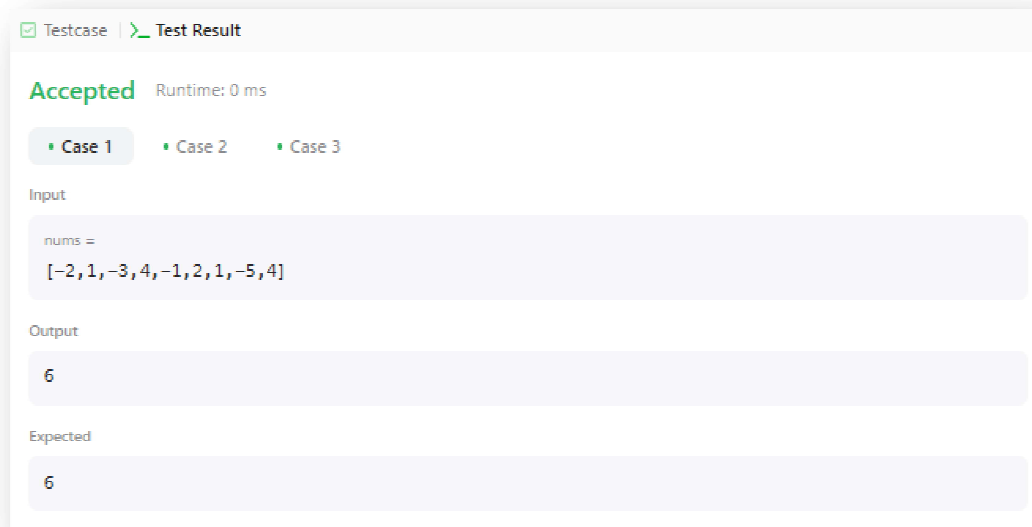
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```
for (int i = 1; i < nums.size(); ++i)
    dp[i] = max(nums[i], dp[i - 1] + nums[i]);

return ranges::max(dp);
}
```

## Output:



**Time Complexity :  $O(N)$**

**Space Complexity :  $O(1)$**

## Learning Outcomes:-

- Recognizing **overlapping subproblems** (each subarray solution builds on the previous).
- The optimized approach shows the **power of greedy techniques** in reducing space complexity.



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