



Experiment- 6A

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Semester:6TH

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];
    }
};
```

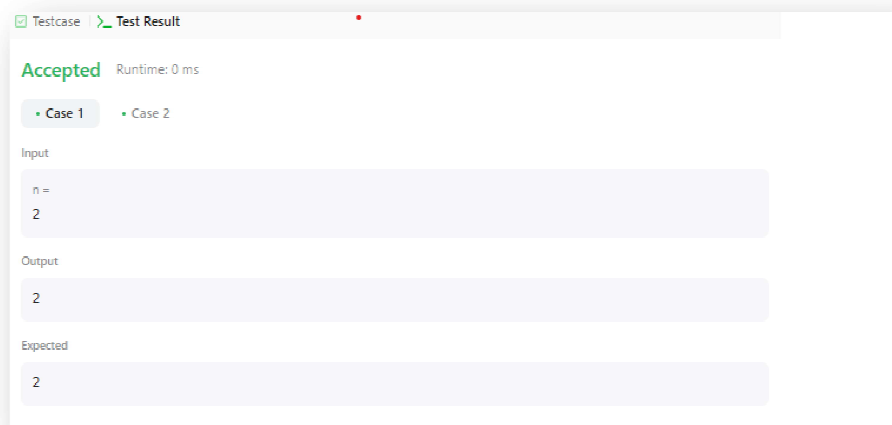


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```
return dp[n];  
}  
};
```

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).



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Experiment - 6B

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Semester: 6TH

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 largest sum, & 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:

$dp[0] = nums[0] \rightarrow$ 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:

$dp[i] = \max(nums[i], dp[i-1] + nums[i])$
 $dp[i] = \max(nums[i], dp[i-1] + nums[i])$

- 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:

The screenshot shows a test result interface with the following details:

- Testcase | Test Result
- Accepted Runtime: 0 ms
- Case 1 Case 2 Case 3
- Input: nums = [-2, 1, -3, 4, -1, 2, 1, -5, 4]
- Output: 6
- Expected: 6

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.