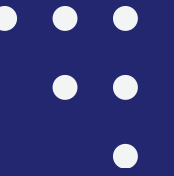


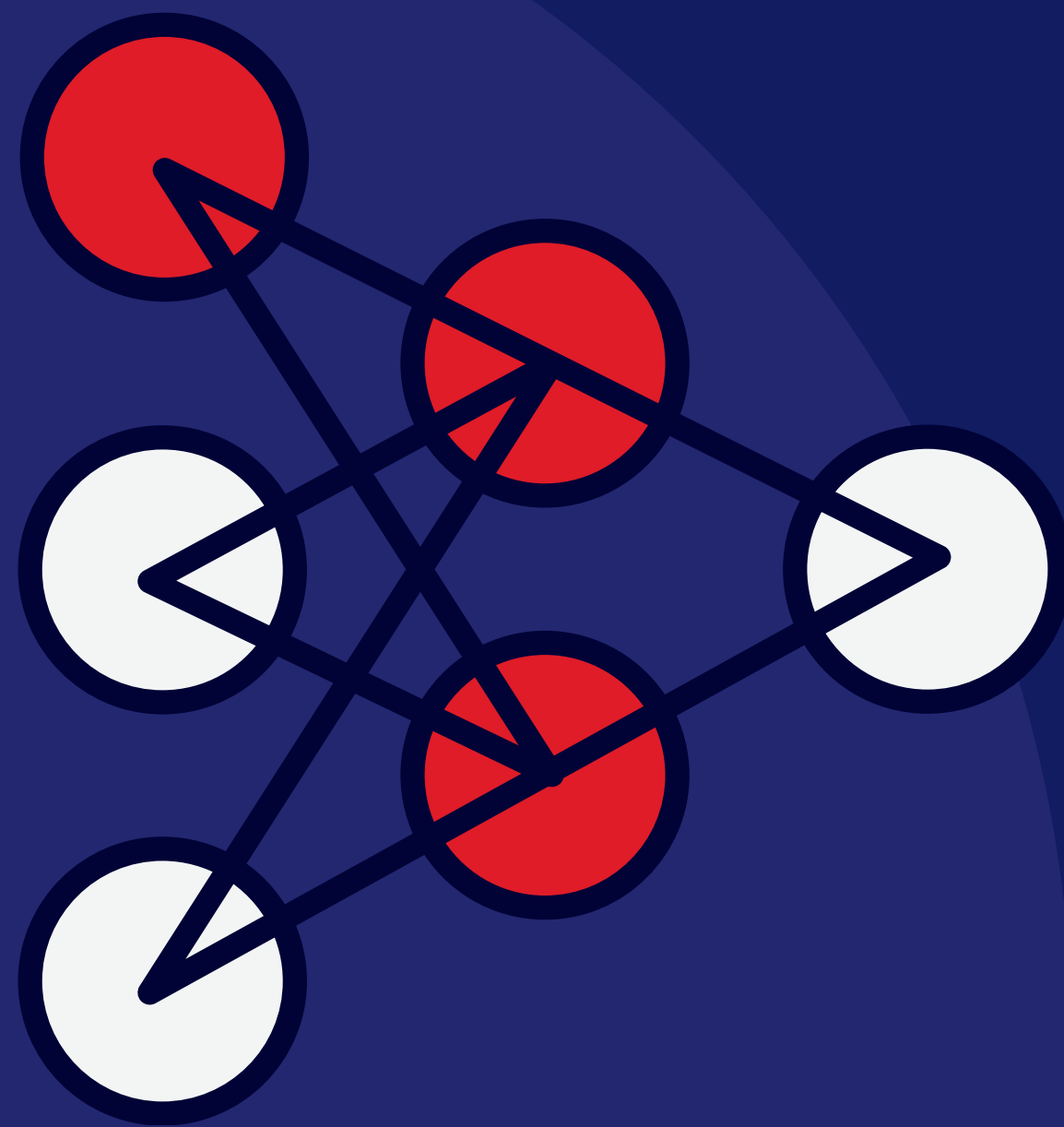


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WAYS TO REACH TARGET ELEMENTS USING ARRAY ELEMENTS





Introduction

LET'S CONSIDER WE ARE GIVEN SOME ELEMENTS SUCH AS ARRAY { 5, 3, -6, 2 } WE NEED TO FIND WAYS TO ACHIEVE OUR DESIRED OUTPUT OF NUMBER 6 FOR EXAMPLE AND FROM THE ARRAY WE CAN TAKE ANY NUMBER OF ELEMENTS

THE TOTAL NUMBER OF WAYS TO ACHIEVE A TARGET SUM OF 6 USING ONLY + AND - OPERATORS IS 4 AS:

$$(-) - 6 = 6$$

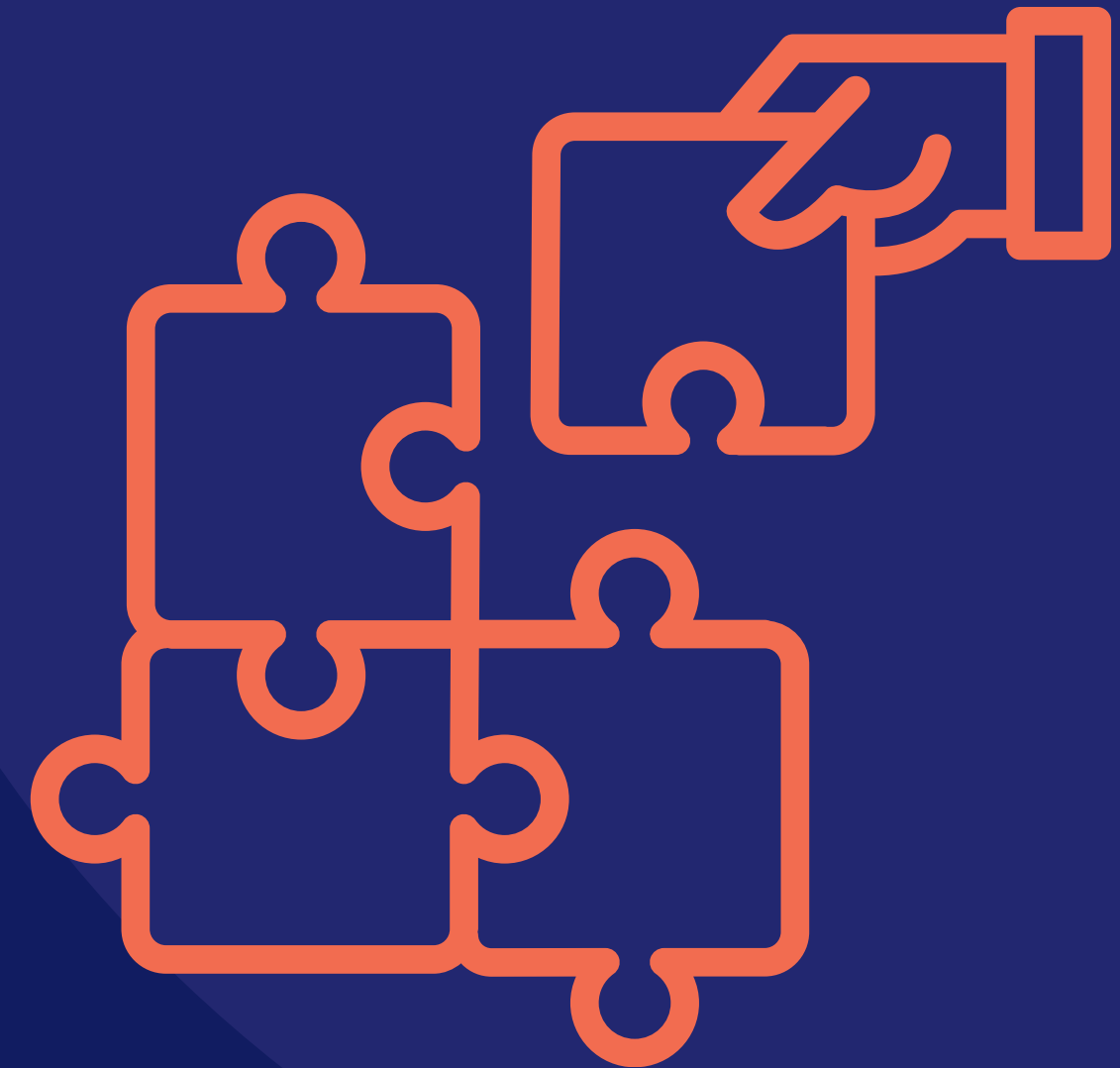
$$(+) 5 (+) 3 (-) 2 = 6$$

$$(+) 5 (-) 3 (-) - 6 (-) 2 = 6$$

$$(-) 5 (+) 3 (-) - 6 (+) 2 = 6$$

THERE ARE MAINLY FOUR WAYS TO SOLVE THIS PROBLEM

- Brute Force (Recursion) – Tries all possible + and - combinations.
- Recursion + Memoization – Caches results to avoid recomputation.
- Dynamic Programming (Subset Sum) – Converts problem into a subset sum variation.
- Optimized DP – Reduces space complexity for larger inputs.





BRUTE FORCE(RECURSION)

- ◆ Concept: Try all + and - combinations using recursion.
- ◆ Example (nums = [1, 2, 3], target = 3):
 - +1 +2 -3 = 0 ✗,
 - +1 -2 +3 = 2 ✗,
 - 1 +2 +3 = 4 ✗, ... (Total 2^N cases!)
- ◆ Code:

```
if (index == nums.size()) return target == 0 ? 1 : 0;  
return countWays(nums, index + 1, target - nums[index]) +  
    countWays(nums, index + 1, target + nums[index]);
```
- ◆ Time Complexity: $O(2^N)$ ✗ (Slow for large N)
- ✓ Good for $N \leq 15$, ✗ Use DP for larger inputs!

Recursive + Memoization

Concept:

- Stores results of subproblems to avoid redundant calculations (Top-Down DP).
- Improves efficiency over brute force recursion.

◆ Optimized Recursive Code (C++):

```
unordered_map<string, int> dp;
int countWays(vector<int>& nums, int i, int target) {
    string key = to_string(i) + "," + to_string(target);
    if (dp.count(key)) return dp[key];
    if (i == nums.size()) return target == 0 ? 1 : 0;
    return dp[key] = countWays(nums, i+1, target - nums[i]) +
        countWays(nums, i+1, target + nums[i]);
}
```

◆ Time Complexity: $O(N \times S)$ ✓ (Faster than brute force)

✓ Best for $15 < N \leq 30$, ✗ Use DP for even larger inputs!



DYNAMIC PROG (BOTTOM UP)

◆ Concept:

- Uses a 2D DP table to store solutions iteratively.
- Converts problem into a subset sum variation.

◆ Formula:

$$dp[i][j] = dp[i-1][j] + dp[i-1][j - \text{nums}[i-1]]$$

```
vector<vector<int>> dp(n+1, vector<int>(S+1, 0));
dp[0][0] = 1;
for (int i = 1; i <= n; i++) {
    for (int j = 0; j <= S; j++) {
        dp[i][j] = dp[i-1][j];
        if (j >= nums[i-1]) dp[i][j] += dp[i-1][j - nums[i-1]];
    }
}
```

◆ Time Complexity: $O(N \times S)$ ✓ (Faster for large N)

✓ Best for $N > 30$, ✗ Uses extra memory



Space-Optimized DP Approach

Space-Optimized DP Approach

- ◆ Concept:
 - Reduces 2D DP table to 1D array (since we only need the previous row).
 - Uses rolling array technique to save space.

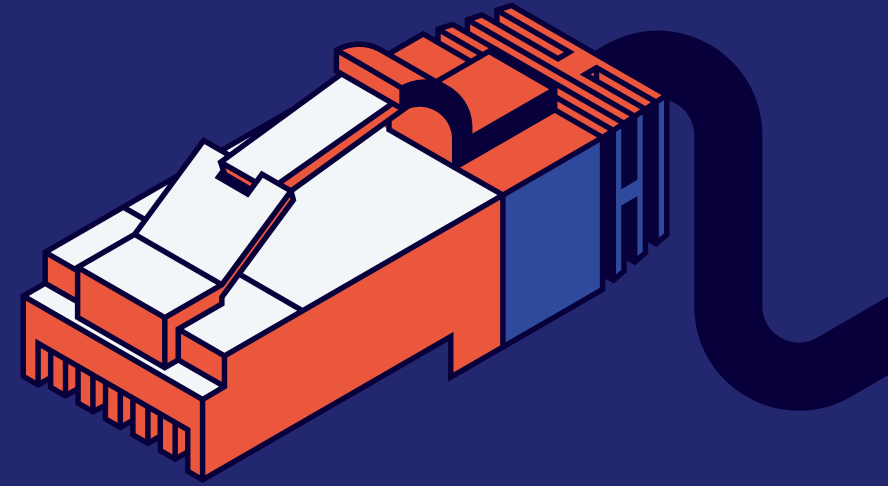
- ◆ Formula:

$dp[j] = dp[j] + dp[j - \text{nums}[i-1]]$ (Iterate backward to avoid overwriting).

```
vector<int> dp(S+1, 0);
dp[0] = 1;
for (int num : nums) {
    for (int j = S; j >= num; j--) {
        dp[j] += dp[j - num];
    }
}
```

- ◆ Time Complexity: $O(N \times S)$ ✓ (Same as DP)
- ◆ Space Complexity: $O(S)$ ✓ (Much better than 2D DP)
- ✓ Best for $N > 30$, ✗ Still not optimal for very large S





Optimization Techniques



Brute Force

- 📌 When to Use?
- ✓ Small N (≤ 20).
- ✗ Inefficient for large N .



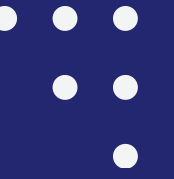
Memoization

- 📌 When to Use?
- ✓ Medium-sized N (≤ 30).
- ✓ Saves time compared to brute force.



DP

- 📌 When to Use?
- ✓ Large N (≤ 1000).
- ✓ Best for efficiency but requires extra space.



C++ IMPLEMENTATION



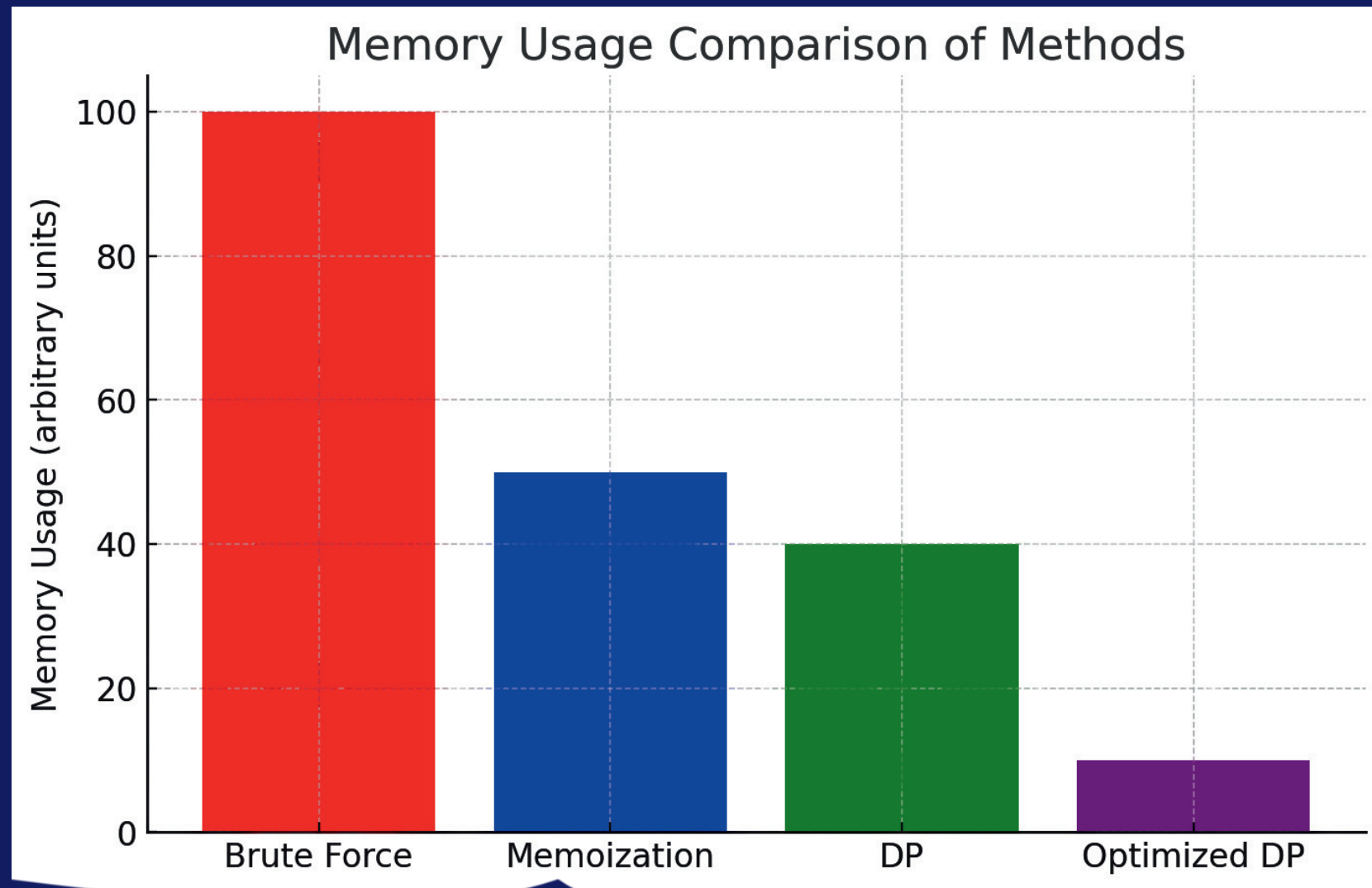
Explanation of Code



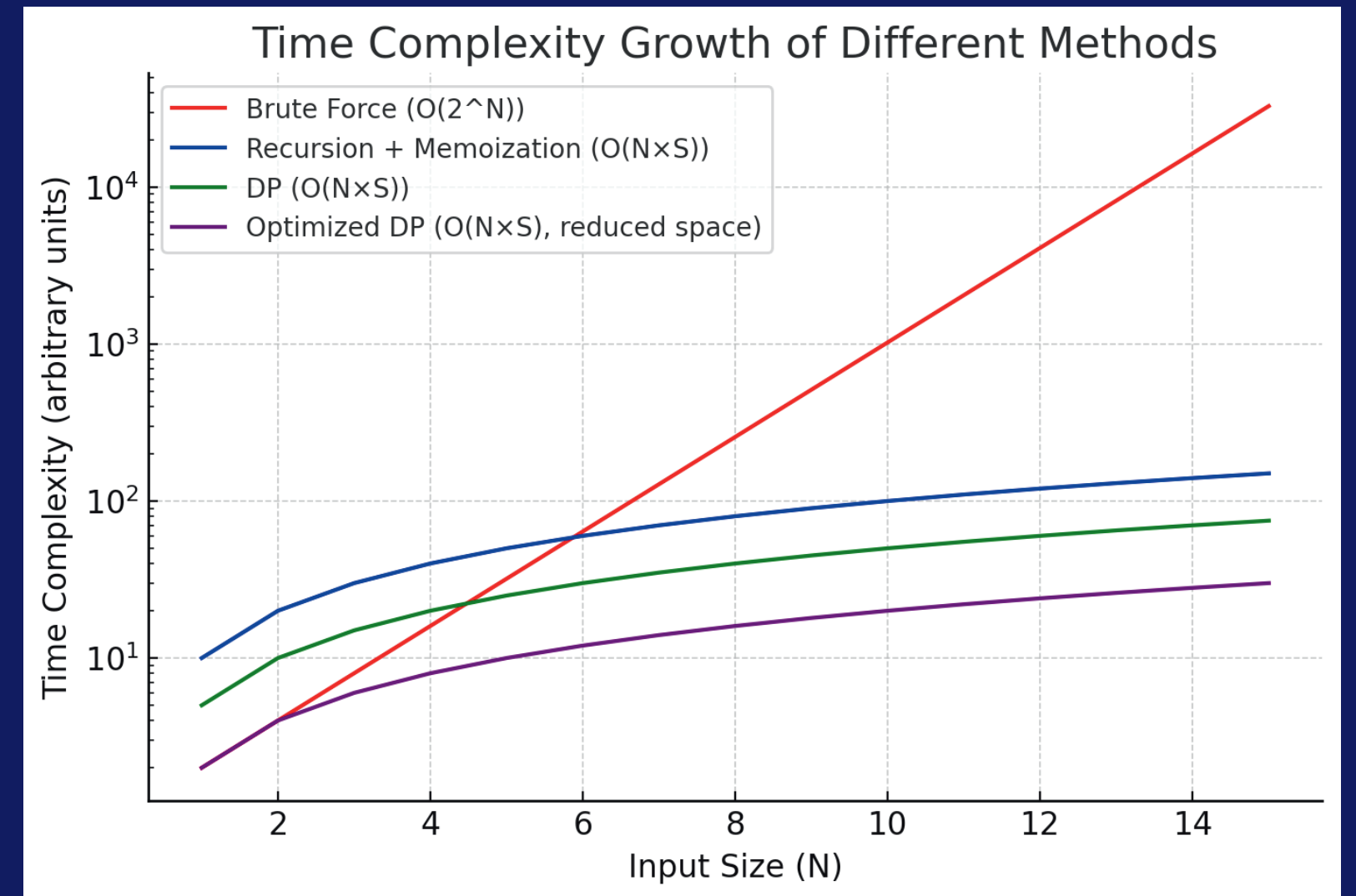
KEY TAKEAWAY

- For small inputs → Use recursion.
- For medium inputs → Use recursion with memoization.
- For large inputs → Use Dynamic Programming.
- Optimized DP is the best for handling large data.

Space Complexity



Time Complexity





THANK YOU

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