



# Dynamic Programming

Memoization, Tabulation &  
Space Optimization

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```
filterByOrg = filterByOrg ? study.lead_organization === filterByOrg : true  
!Status = filterByStatus ? study.status === filterByStatus : true  
matchStatus) {  
    function filterStudies({ studies, filterByOrg, filterByStatus }) {  
        return studies.filter(study => {  
            return filterByOrg === study.lead_organization &&  
                filterByStatus === study.status &&  
                !filterByStatus || study.status === filterByStatus  
        })  
    }  
}
```

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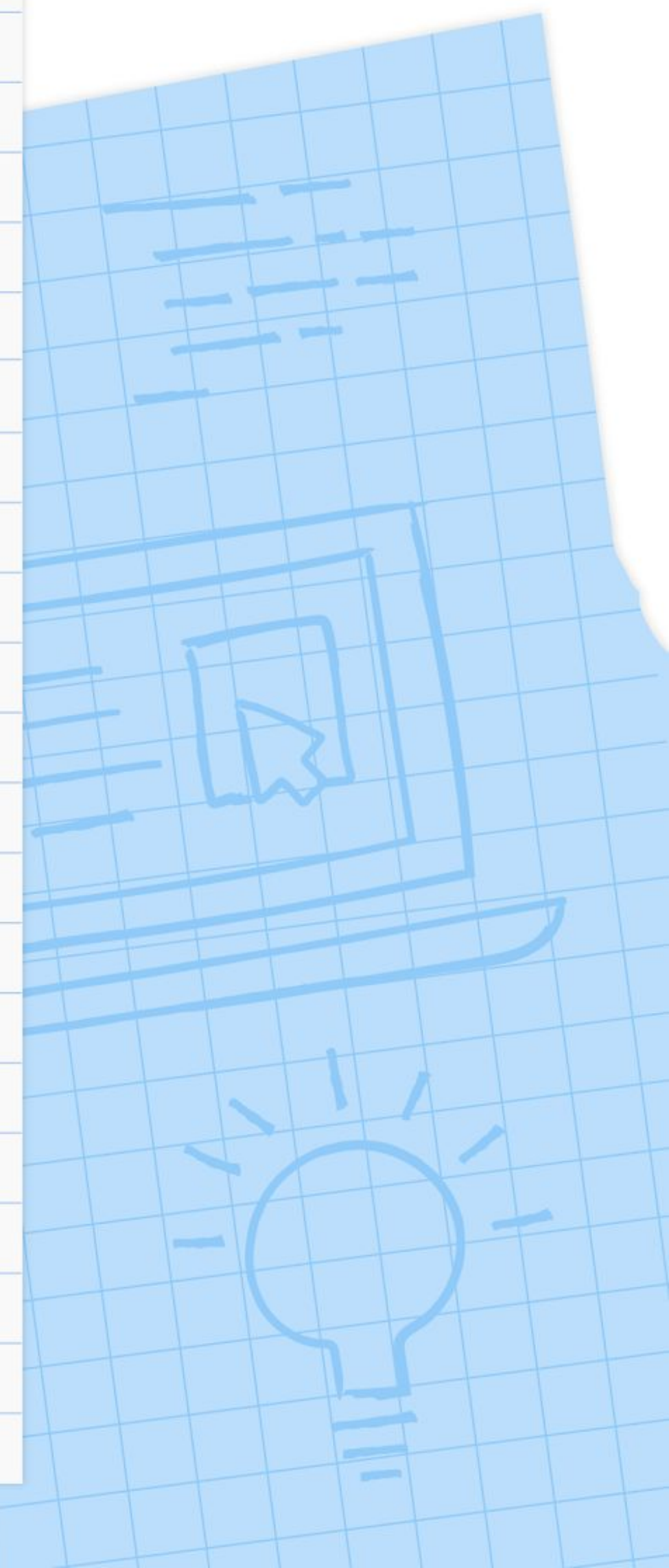
- Introduction to Dynamic Programming (DP)
- DP Techniques: Memoization, Tabulation, Space-Optimization
- Applications of DP
- Exercises



# Introduction to DP

When can DP be applied?

- **Optimal Substructure:** The solution to the main problem can be constructed from the optimal solutions of its subproblems
- **Overlapping Subproblems:** If there is redundancy in solving the same subproblems, DP can be applied to store & retrieve the intermediate results from an in-memory cache

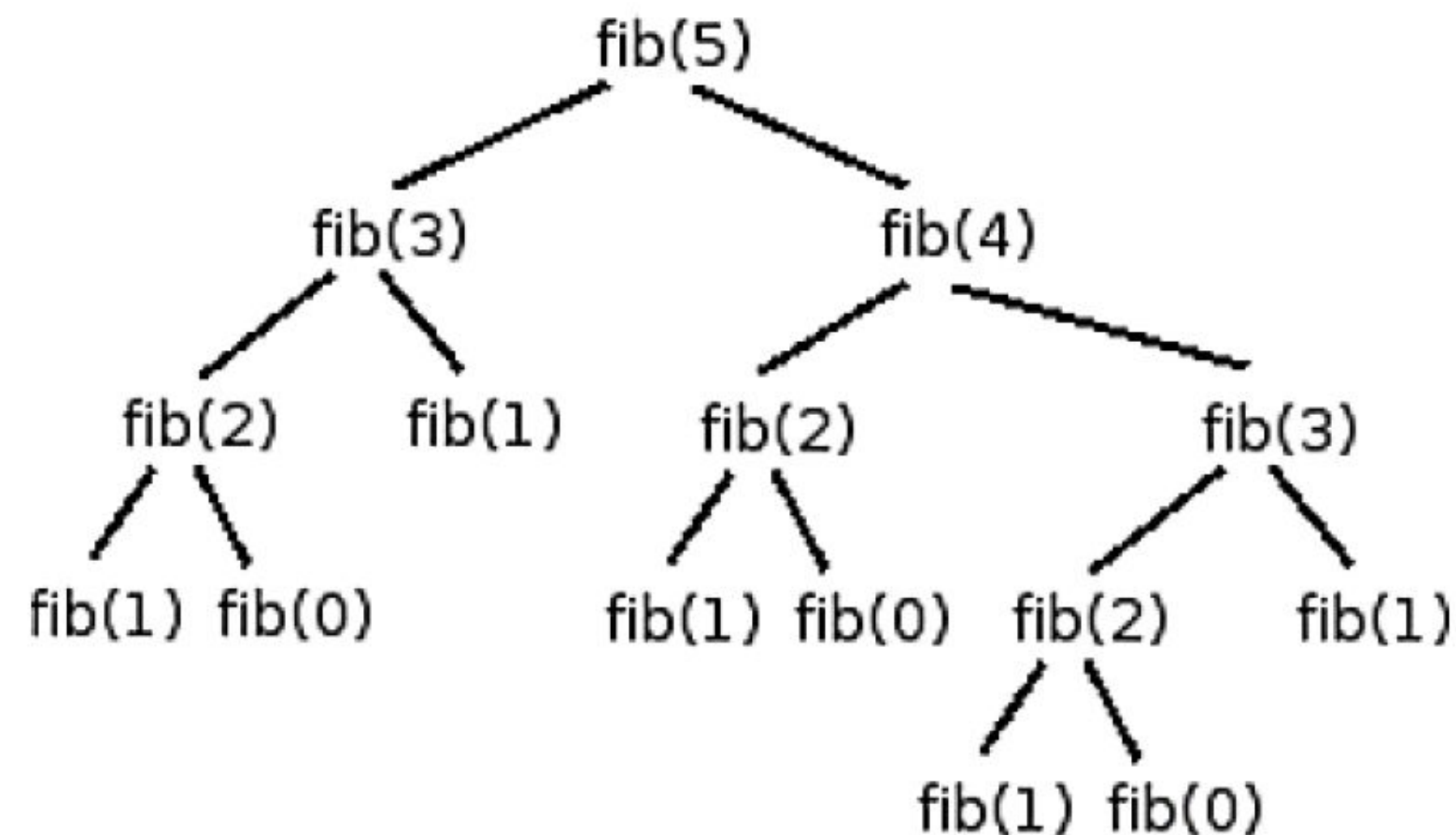


# Fibonacci Sequence

## Using Divide & Conquer

We know that  $\text{fib}(n) = \text{fib}(n-1) + \text{fib}(n-2)$ ;  $\text{fib}(0) = 0$ ,  $\text{fib}(1) = 1$

Now, compute  $\text{fib}(5)$ :



15 calls (nodes)  
7 additions (parents)



# Fibonacci Sequence

## Using DP Bottom-Up Approach (Tabulation)

Compute fib(5):

```
- fib(0) = 0
  fib(1) = 1
  ✓ fib(2) = fib(1) + fib(0)
    = 0 + 1
    = 1
  ✓ fib(3) = fib(2) + fib(1)
    = 1 + 1
    = 2
  ✓ fib(4) = fib(3) + fib(2)
    = 2 + 1
    = 3
  ✓ fib(5) = fib(4) + fib(3)
    = 3 + 2
    = 5
```

4 calls  
4 additions

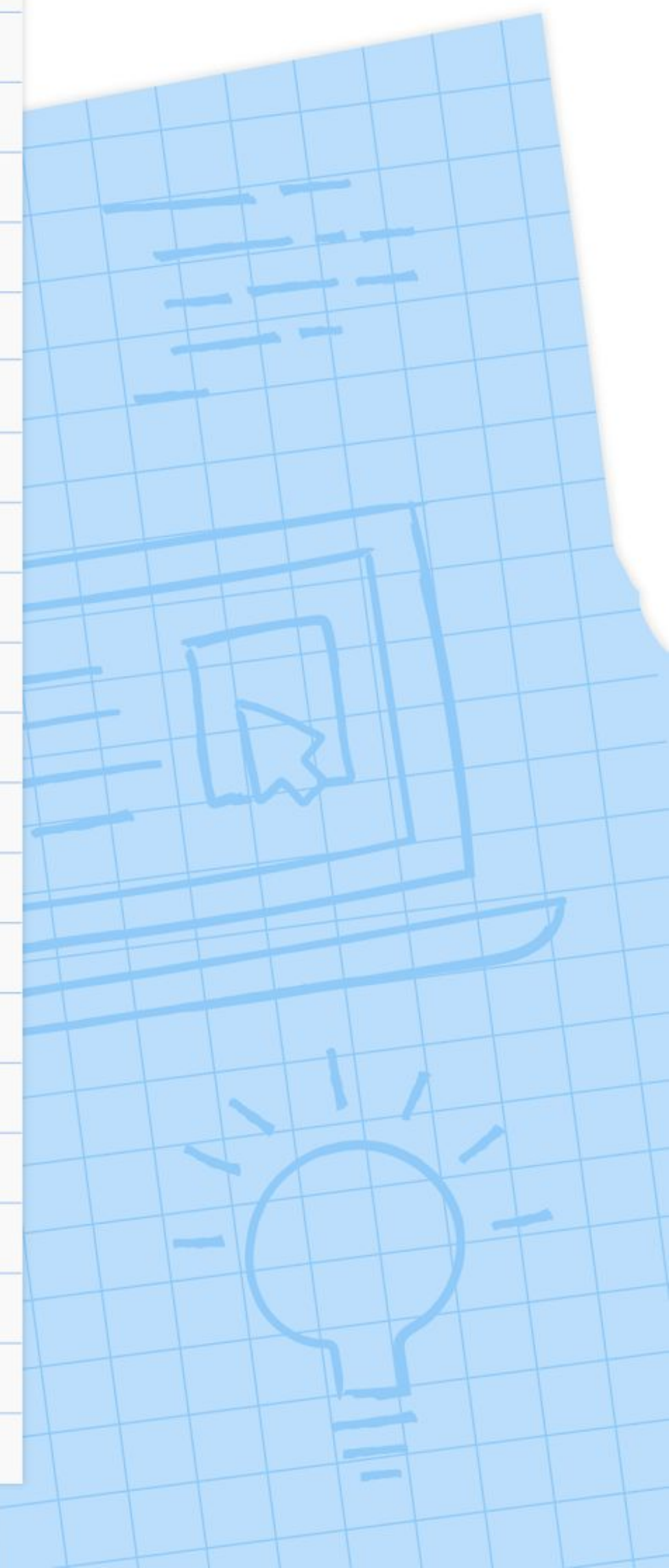
# Observation

There is significant reduction in **CPU time** and **function call overhead** along with the decrease in the **no. of basic operations** performed.

This not only enhances **computational efficiency** but also contributes to a more **streamlined** and **optimized execution** of the algorithm.

# DP Techniques

- **Memoization:** Top-Down, Recursive formulation, typically uses a helper function
- **Tabulation:** Bottom-Up, Iterative approach, reducing the usage of stack space
- **Space-Optimized Tabulation:** Similar to Tabulation, with more efficient space usage





# Fibonacci Sequence

## Memoization

**Direction:** main problem to subproblem

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

**Space:**  $O(n)$  (dp array) +  $O(n)$  (recursion stack)

```
1  int fun(int n, vector<int> &dp)
2  {
3      // Base case
4      if (n <= 1)
5          return n;
6
7      // Lookup
8      if (dp[n] != -1)
9          return dp[n];
10
11     // Compute, store and return
12     return dp[n] = fun(n - 1, dp) + fun(n - 2, dp);
13 }
14
15 int fib(int n)
16 {
17     vector<int> dp(n + 1, -1); // dp[i] = fib(i)
18     return fun(n, dp);
19 }
```



# Fibonacci Sequence

## Tabulation

**Direction:** subproblem to main problem

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

**Space:**  $O(n)$  (for dp array)

```
1  int fib2(int n)
2  {
3      vector<int> dp(n + 1, -1);
4
5      dp[0] = 0;
6      dp[1] = 1;
7      for (int i = 2; i <= n; i++)
8          dp[i] = dp[i - 1] + dp[i - 2];
9
10     return dp[n];
11 }
```

# Fibonacci Sequence

## Space-Optimized Tabulation

**Direction:** subproblem to main problem

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

**Space:**  $O(1)$  (no dp array)

```
1  int fib3(int n)
2  {
3      // Base case
4      if (n <= 1)
5          return n;
6
7      int first = 0;
8      int second = 1;
9      int third;
10     for (int i = 2; i <= n; i++)
11     {
12         third = first + second;
13         first = second;
14         second = third;
15     }
16
17     return second; // OR return third;
18 }
```



# Applications of DP

- **Reduce TLE** (Time Limit Exceeded) in Competitive Programming
- Path Finding & Shortest Paths questions. Eg: Shortest Sum in grid
- Majority of the **Optimization problems**. Eg: 0/1 Knapsack

# Exercise 1: Frog Jump

## 1-dimensional DP Question

Given the no. of stairs and a frog.

The frog wants to climb from the **0<sup>th</sup> stair** to the **(N-1)<sup>th</sup> stair**.

At a time the frog can climb either **one or two steps**.

A **height[N] array** is also given. Whenever the frog jumps from a stair *i* to stair *j*, the energy consumed in the jump is **abs(height[i] - height[j])**, where abs() means the absolute difference.

Return the **min energy** that can be used by the frog to jump from stair 0 to stair N-1.





# Exercise 1: Frog Jump

## Intuition

- **Optimization Problem:** Find the min energy to go from 0th to  $(N-1)^{\text{th}}$  stair
- **Optimal Substructure & Overlapping Subproblems:** The  $i^{\text{th}}$  stair, can be reached from the  $(i-1)^{\text{th}}$  or  $(i-2)^{\text{th}}$  stair.



# Exercise 1: Frog Jump

## Memoization

### Main function

Initializing the DP array

```
1  int minEnergy(int n, vector<int> &height)
2  {
3      vector<int> dp(n, -1); // 0 to n-1 stair
4      return fun(n - 1, height, dp);
5  }
```





# Exercise 1: Frog Jump

## Memoization

### Helper function

Recursive formulation  
of logic

```
1  int fun(int ind, vector<int> &height, vector<int> &dp)
2  {
3      // Base Case
4      if (ind == 0) // 0th stair
5          return 0;
6
7      // Lookup
8      if (dp[ind] != -1)
9          return dp[ind];
10
11     // Recursive Case
12
13     // Energy to reach current stair from 1 stair before
14     int jump1 = abs(height[ind] - height[ind - 1]) + fun(ind - 1, height, dp);
15
16     // Energy to reach current stair from 2 stairs before
17     int jump2 = INT_MAX;
18     if (ind >= 2) // valid index
19         jump2 = abs(height[ind] - height[ind - 2]) + fun(ind - 2, height, dp);
20
21     return dp[ind] = min(jump1, jump2);
22 }
```

# Exercise 1: Frog Jump

## Tabulation

**Iteratively** filling  
the dp array in  
bottom-up fashion

```
1  int minEnergy2(int n, vector<int> &height)
2  {
3      vector<int> dp(n, -1); // 0 to n-1 stair
4
5      dp[0] = 0;             // 0th stair
6      for (int i = 1; i < n; i++) // 1 to n-1
7      {
8          int jump1 = abs(height[i] - height[i - 1]) + dp[i - 1];
9          int jump2 = INT_MAX;
10         if (i >= 2)
11             jump2 = abs(height[i] - height[i - 2]) + dp[i - 2];
12
13         dp[i] = min(jump1, jump2);
14     }
15
16     return dp[n - 1];
17 }
```





# Exercise 1: Frog Jump

## Space-Optimized Tabulation

**Reducing** DP array  
to two variables  
(constant space)

```
1  int minEnergy3(int n, vector<int> &height)
2  {
3      int first = 0;
4      int second = 0;
5      for (int i = 1; i < n; i++) // 1 to n-1
6      {
7          int jump1 = abs(height[i] - height[i - 1]) + second;
8          int jump2 = INT_MAX;
9          if (i >= 2)
10             jump2 = abs(height[i] - height[i - 2]) + first;
11
12             first = second;
13             second = min(jump1, jump2);
14     }
15
16     return second;
17 }
```

# Exercise 2: House Robber

## 1-dimensional DP Question

A thief needs to rob money in a street.

The houses in the street are arranged in a **circular manner**.

Therefore the first and the last house are adjacent to each other.

The security system in the street is such that if **adjacent houses** are robbed, the **police** will get **notified**.

Given an array of integers "arr" which represents money at each house.

Return the **max amount of money** that the thief can rob without alerting the police.





# Exercise 2: House Robber

## Approach

- Make **2 reduced arrays**: arr1 (w/o last element) and arr2 (w/o first element).
- Find the **max sum of non-adjacent elements** arr1 and arr2 separately.
- Return **max(ans1, ans2)**



# Exercise 2: House Robber

## Tabulation

### Main function

```
1  int robStreet(int n, vector<int> &arr)
2  {
3      if (n == 1) // Only 1 house
4          return arr[0];
5
6      // Reduced Arrays
7      vector<int> arr1(arr.begin(), arr.end() - 1); // w/o last element
8      vector<int> arr2(arr.begin() + 1, arr.end()); // w/o first element
9
10     // Find the max sum of non-adjacent elements arr1 and arr2 separately
11     int ans1 = maxSumNonAdj(n - 1, arr1);
12     int ans2 = maxSumNonAdj(n - 1, arr2);
13     return max(ans1, ans2);
14 }
```



# Exercise 2: House Robber

## Tabulation

Helper function to find the **max sum of non-adjacent elements** in an array

```
1  int maxSumNonAdj(int n, vector<int> &arr)
2  {
3      vector<int> dp(n, -1); // 0 to n-1
4
5      dp[0] = arr[0];          // Base Case (Only 1 element)
6      for (int i = 1; i < n; i++) // 1 to n-1
7      {
8          int notPick = dp[i - 1];
9
10         int pick = arr[i];
11         if (i >= 2)
12             pick += dp[i - 2];
13
14         dp[i] = max(pick, notPick);
15     }
16
17     return dp[n - 1];
18 }
```

# Exercise 3: Unique Paths

## 2-dimensional DP Question

Given two values  $M$  and  $N$ , which represent a matrix  $matrix[M][N]$ .  
Find the **total unique paths**  
from the **top-left cell** ( $matrix[0][0]$ )  
to the **rightmost cell** ( $matrix[M-1][N-1]$ ).

At any cell we are allowed to move in only **two directions**:- **bottom** and **right**.





# Exercise 3: Unique Paths

## Approach

- Make a **2D dp array** of size  $M \times N$ .
- **$dp[i][j]$**  = total unique paths from the top-left cell to the cell  $(i, j)$ .
- **Dependency Direction:** up & left



# Exercise 3: Unique Paths

## Memoization

### Main function

```
1  int uniquePaths(int m, int n)
2  {
3      // rows: index (0 to m-1), cols: index (0 to n-1)
4      vector<vector<int>> dp(m, vector<int>(n, -1));
5      return fun(m - 1, n - 1, dp);
6  }
```





# Exercise 3: Unique Paths

## Memoization

Helper function to  
**calculate dp[i][j]**

```
1  int fun(int i, int j, vector<vector<int>> &dp)
2  {
3      // Base Cases
4      if (i == 0 && j == 0) // top-left cell
5          return 1;
6      if (i < 0 || j < 0) // out of bounds
7          return 0;
8
9      // Lookup
10     if (dp[i][j] != -1)
11         return dp[i][j];
12
13     // Recursive Case
14
15     // Move up
16     int up = fun(i - 1, j, dp);
17
18     // Move Left
19     int left = fun(i, j - 1, dp);
20
21     return dp[i][j] = up + left;
22 }
```

# Exercise 3: Unique Paths

## Tabulation

Iteratively filling the  
dp matrix in  
bottom-up fashion

```
1  int uniquePaths2(int m, int n)
2  {
3      // rows: index (0 to m-1), cols: index (0 to n-1)
4      vector<vector<int>> dp(m, vector<int>(n, 0));
5
6      for (int i = 0; i < m; i++) // rows
7      {
8          for (int j = 0; j < n; j++) // cols
9          {
10             if (i == 0 && j == 0) // top-left cell
11             {
12                 dp[i][j] = 1;
13                 continue;
14             }
15
16             // Move up
17             int up = 0;
18             if (i > 0)
19                 up = dp[i - 1][j];
20
21             // Move left
22             int left = 0;
23             if (j > 0)
24                 left = dp[i][j - 1];
25
26             dp[i][j] = up + left;
27         }
28     }
29
30     return dp[m - 1][n - 1];
31 }
```



# Thank You!

“

```
function filterStudies({ studies, filterByOrg = false, filterByStudy = false }) {
  return studies.filter(study => {
    // Filter by organization
    if (filterByOrg) {
      return study.organization === 'United States'
    }
    // Filter by study
    if (filterByStudy) {
      return study.study === 'Study 1'
    }
    // No filters applied
    return true
  })
}
```