

## ■ Documentation: Fibonacci Number (Leetcode 509)

### 1. Problem Statement

Given an integer  $n$ , return the  $n$ th number in the Fibonacci sequence.

The Fibonacci sequence is defined as:

- $F(0) = 0$
- $F(1) = 1$
- $F(n) = F(n - 1) + F(n - 2)$ , for  $n > 1$

Constraints:

- $0 \leq n \leq 30$

### 2. Intuition

The Fibonacci sequence builds on the idea of recurrence—each number is the sum of the two before it. Since the value at  $F(n)$  depends only on  $F(n-1)$  and  $F(n-2)$ , we can compute values iteratively using just two variables, making the solution both time and space efficient.

### 3. Key Observations

- The sequence starts from 0 and 1.
- Every next number is determined solely by the previous two numbers.
- We don't need the entire sequence stored—just the last two values.
- Brute-force recursive solutions are highly inefficient due to overlapping subproblems.

#### 4. Approach

We use an iterative (bottom-up) dynamic programming approach with two variables:

- If  $n$  is 0, return 0.
- If  $n$  is 1, return 1.
- Start from the third number, compute each Fibonacci value up to  $n$  using two variables that store the last two computed values.
- Return the final computed value.

#### 5. Edge Cases

- $n = 0 \rightarrow$  Output: 0
- $n = 1 \rightarrow$  Output: 1
- The function handles all values from 0 to 30 as per constraints.

#### 6. Complexity Analysis

⌚ Time Complexity

- $O(n)$ : We loop through from 2 to  $n$  once.

📦 Space Complexity

- $O(1)$ : Only two variables are used, regardless of  $n$ .

## 7. Alternative Approaches

### a) Recursive Solution (Brute-force)

```
def fib(n):  
    if n <= 1:  
        return n  
    return fib(n-1) + fib(n-2)
```

- ✗ Time Complexity:  $O(2^n)$  — exponential
- ✗ Inefficient for large  $n$  due to repeated calculations

### b) Memoization (Top-down DP)

```
def fib(n, memo = {}):  
    if n in memo:  
        return memo[n]  
    if n <= 1:  
        return n  
    memo[n] = fib(n-1, memo) + fib(n-2, memo)  
    return memo[n]
```

- ✓ Efficient with  $O(n)$  time
- 📦 Uses  $O(n)$  space due to the dictionary

### c) Bottom-Up with Array

```
def fib(n):  
    if n <= 1:  
        return n  
    dp = [0] * (n + 1)  
    dp[1] = 1  
    for i in range(2, n+1):  
        dp[i] = dp[i-1] + dp[i-2]
```

```
return dp[n]
```

- ✓ Time:  $O(n)$
- ✗ Space:  $O(n)$

## 8. Test Cases

Input	Expected Output	Explanation
0	0	$F(0) = 0$
1	1	$F(1) = 1$
2	1	$F(2) = F(1) + F(0) = 1 + 0$
3	2	$F(3) = F(2) + F(1) = 1 + 1$
4	3	$F(4) = F(3) + F(2) = 2 + 1$
10	55	Computed iteratively
30	832040	Edge case at upper constraint

## 9. Final Thoughts

- The iterative method is best for small  $n$  with minimal memory use.
- Recursive solutions without memoization should be avoided due to performance issues.
- For larger Fibonacci problems (e.g.,  $n > 10^5$ ), matrix exponentiation or Binet's Formula may be considered.
- This problem is a classic example of optimizing recursive relationships using dynamic programming principles.