

# Solutions to DSA Questions 111-141 (Graphs, DP, Bit Manipulation) For 1-2

Years Experience Roles at EPAM Compiled on September 27, 2025

## Introduction

This document provides detailed solutions for 31 Data Structures and Algorithms (DSA) problems (questions 111 to 141) from the Graphs, Dynamic Programming, and Bit Manipulation categories, tailored for candidates with 1-2 years of experience preparing for roles at EPAM Systems. Each problem includes a problem statement, dry run with test cases, algorithm, and a Python solution, formatted for clarity. Dynamic programming problems include both memoization and tabulation approaches where applicable.

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# 1 Pacific Atlantic Water Flow

## 1.1 Problem Statement

Given an  $m \times n$  matrix of heights, find all cells where water can flow to both Pacific and Atlantic oceans.

## 1.2 Dry Run on Test Cases

- **Test Case 1:** heights =  $\begin{bmatrix} [1,2,2,3,5], [3,2,3,4,4], [2,4,5,3,1], [6,7,1,4,5], [5,1,1,2,4] \end{bmatrix}$  → Output:  $\begin{bmatrix} [0,4], [1,3], [1,4], [2,2], [3,0], [3,1], [4,0] \end{bmatrix}$
- **Test Case 2:** heights =  $\begin{bmatrix} [1] \end{bmatrix}$  → Output:  $\begin{bmatrix} [0,0] \end{bmatrix}$
- **Test Case 3:** heights =  $\begin{bmatrix} [1,2], [2,1] \end{bmatrix}$  → Output:  $\begin{bmatrix} [0,0], [0,1], [1,0], [1,1] \end{bmatrix}$
- **Test Case 4:** heights =  $\begin{bmatrix} [] \end{bmatrix}$  → Output:  $\begin{bmatrix} [] \end{bmatrix}$

## 1.3 Algorithm

1. Use DFS from Pacific (top/left) and Atlantic (bottom/right) borders.

2. Mark reachable cells for each ocean.

3. Return cells reachable by both.

**Time Complexity:**  $O(m \cdot n)$     **Space Complexity:**  $O(m \cdot n)$

## 1.4 Python Solution

```
1 def pacific_atlantic(heights):
2     if not heights or not heights[0]:
3         return []
4
5     rows, cols = len(heights), len(heights[0])
6     pacific = set()
7     atlantic = set()
8
9     def dfs(i, j, visited, prev_height):
10        if i < 0 or i >= rows or j < 0 or j >= cols or (i, j) in
11            visited or heights[i][j] < prev_height:
12            return
13        visited.add((i, j))
14        for di, dj in [(1,0), (-1,0), (0,1), (0,-1)]:
15            dfs(i + di, j + dj, visited, heights[i][j])
16
17    # Pacific: top and left borders
18    for j in range(cols):
19        dfs(0, j, pacific, float('-inf'))
20    for i in range(rows):
21        dfs(i, 0, pacific, float('-inf'))
22
23    # Atlantic: bottom and right borders
24    for j in range(cols):
25        dfs(rows-1, j, atlantic, float('-inf'))
26    for i in range(rows):
27        dfs(i, cols-1, atlantic, float('-inf'))
28
29    return list(pacific & atlantic)
```

## 2 Surrounded Regions

### 2.1 Problem Statement

Given a 2D board of 'X' and 'O', flip all 'O's surrounded by 'X' to 'X'.

### 2.2 Dry Run on Test Cases

- **Test Case 1:** board = [["X","X","X","X"], ["X","O","O","X"], ["X","X","O","X"], ["X","O","X","X"]]  
→ Output: [[ "X", "X", "X", "X"], [ "X", "X", "X", "X"], [ "X", "X", "X", "X"], [ "X", "O", "X", "X"]]
- **Test Case 2:** board = [[ "O"]] → Output: [[ "O"]]

- **Test Case 3:** board = [] → Output: []
- **Test Case 4:** board = [["X"]] → Output: [["X"]]

## 2.3 Algorithm

1. Use DFS to mark 'O's connected to border as safe.
2. Flip unmarked 'O's to 'X'.
3. Restore safe 'O's.

**Time Complexity:**  $O(m \cdot n)$     **Space Complexity:**  $O(m \cdot n)$

## 2.4 Python Solution

```

1 def solve(board):
2     if not board or not board[0]:
3         return
4
5     rows, cols = len(board), len(board[0])
6
7     def dfs(i, j):
8         if i < 0 or i >= rows or j < 0 or j >= cols or board[i][j]
9             != '0':
10             return
11         board[i][j] = '#'
12         for di, dj in [(1,0), (-1,0), (0,1), (0,-1)]:
13             dfs(i + di, j + dj)
14
15     # Mark border-connected '0's
16     for i in range(rows):
17         if board[i][0] == '0':
18             dfs(i, 0)
19         if board[i][cols-1] == '0':
20             dfs(i, cols-1)
21     for j in range(cols):
22         if board[0][j] == '0':
23             dfs(0, j)
24         if board[rows-1][j] == '0':
25             dfs(rows-1, j)
26
27     # Flip '0' to 'X', '#' to '0'
28     for i in range(rows):
29         for j in range(cols):
30             if board[i][j] == '0':
31                 board[i][j] = 'X'
32             elif board[i][j] == '#':
33                 board[i][j] = '0'
```

# 3 Graph Valid Tree

## 3.1 Problem Statement

Given n nodes and edges, check if they form a valid tree (connected, no cycles).

## 3.2 Dry Run on Test Cases

- **Test Case 1:** n = 5, edges = [[0,1],[0,2],[0,3],[1,4]] → Output: True
- **Test Case 2:** n = 5, edges = [[0,1],[1,2],[2,3],[1,3],[1,4]] → Output: False
- **Test Case 3:** n = 1, edges = [] → Output: True
- **Test Case 4:** n = 2, edges = [] → Output: False

## 3.3 Algorithm

1. Use DFS to detect cycles and check connectivity.
2. Build adjacency list.
3. Ensure all nodes visited and no cycles.

**Time Complexity:**  $O(V + E)$     **Space Complexity:**  $O(V + E)$

## 3.4 Python Solution

```
1 from collections import defaultdict
2
3 def valid_tree(n, edges):
4     if len(edges) != n - 1:
5         return False
6
7     graph = defaultdict(list)
8     for u, v in edges:
9         graph[u].append(v)
10        graph[v].append(u)
11
12    visited = set()
13
14    def dfs(node, parent):
15        visited.add(node)
16        for neighbor in graph[node]:
17            if neighbor not in visited:
18                if not dfs(neighbor, node):
19                    return False
20            elif neighbor != parent:
21                return False
22    return True
23
```

24      `return dfs(0, -1) and len(visited) == n`

## 4 Minimum Spanning Tree (Kruskal's Algorithm)

### 4.1 Problem Statement

Given a weighted undirected graph, find the minimum spanning tree weight using Kruskal's algorithm.

### 4.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 4$ , edges =  $[[0,1,1],[0,2,2],[1,2,5],[1,3,1],[2,3,4]] \rightarrow$  Output: 4
- **Test Case 2:**  $n = 1$ , edges =  $[] \rightarrow$  Output: 0
- **Test Case 3:**  $n = 2$ , edges =  $[[0,1,1]] \rightarrow$  Output: 1
- **Test Case 4:**  $n = 3$ , edges =  $[[0,1,2],[1,2,2]] \rightarrow$  Output: 4

### 4.3 Algorithm

1. Sort edges by weight.
2. Use Union-Find to avoid cycles.
3. Add edges to MST if they connect different components.

Time Complexity:  $O(E \log E)$    Space Complexity:  $O(V)$

### 4.4 Python Solution

```
1  class UnionFind:
2      def __init__(self, n):
3          self.parent = list(range(n))
4          self.rank = [0] * n
5
6      def find(self, x):
7          if self.parent[x] != x:
8              self.parent[x] = self.find(self.parent[x])
9          return self.parent[x]
10
11     def union(self, x, y):
12         px, py = self.find(x), self.find(y)
13         if px == py:
14             return False
15         if self.rank[px] < self.rank[py]:
16             px, py = py, px
17             self.parent[py] = px
18         if self.rank[px] == self.rank[py]:
19             self.rank[px] += 1
```

```

20         return True
21
22 def minimum_spanning_tree(n, edges):
23     edges.sort(key=lambda x: x[2])
24     uf = UnionFind(n)
25     total_weight = 0
26     edges_used = 0
27
28     for u, v, weight in edges:
29         if uf.union(u, v):
30             total_weight += weight
31             edges_used += 1
32
33     return total_weight if edges_used == n - 1 else -1

```

## 5 Dijkstra's Algorithm

### 5.1 Problem Statement

Given a weighted directed graph and source, find shortest paths to all vertices.

### 5.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 4$ , edges =  $[[0,1,1],[0,2,4],[1,2,2],[1,3,6],[2,3,3]]$ , source = 0  
→ Output: [0,1,3,6]
- **Test Case 2:**  $n = 1$ , edges =  $[]$ , source = 0 → Output: [0]
- **Test Case 3:**  $n = 2$ , edges =  $[[0,1,1]]$ , source = 0 → Output: [0,1]
- **Test Case 4:**  $n = 3$ , edges =  $[[0,1,2],[1,2,2]]$ , source = 0 → Output: [0,2,4]

### 5.3 Algorithm

1. Use min-heap to store (distance, node).
2. Update distances to neighbors if shorter path found.
3. Return distance array.

**Time Complexity:**  $O((V + E) \log V)$     **Space Complexity:**  $O(V + E)$

### 5.4 Python Solution

```

1 import heapq
2 from collections import defaultdict
3
4 def dijkstra(n, edges, source):
5     graph = defaultdict(list)
6     for u, v, w in edges:

```

```

7     graph[u].append((v, w))
8
9     distances = [float('inf')] * n
10    distances[source] = 0
11    heap = [(0, source)]
12
13    while heap:
14        dist, node = heapq.heappop(heap)
15        if dist > distances[node]:
16            continue
17        for neighbor, weight in graph[node]:
18            if dist + weight < distances[neighbor]:
19                distances[neighbor] = dist + weight
20                heapq.heappush(heap, (dist + weight,
21                                     neighbor))
22
23    return [d if d != float('inf') else -1 for d in
24            distances]

```

## 6 Climbing Stairs

### 6.1 Problem Statement

Given  $n$  stairs, find the number of ways to climb (1 or 2 steps at a time).

### 6.2 Dry Run on Test Cases

- \* **Test Case 1:**  $n = 2 \rightarrow$  Output: 2 ( $[1,1]$ ,  $[2]$ )
- \* **Test Case 2:**  $n = 3 \rightarrow$  Output: 3 ( $[1,1,1]$ ,  $[1,2]$ ,  $[2,1]$ )
- \* **Test Case 3:**  $n = 1 \rightarrow$  Output: 1 ( $[1]$ )
- \* **Test Case 4:**  $n = 4 \rightarrow$  Output: 5

### 6.3 Algorithm (Memoization)

1. Use recursive function with memoization.
2. Base cases:  $n=0$  (1 way),  $n<0$  (0 ways).
3. Return sum of ways for  $n-1$  and  $n-2$ .

### 6.4 Algorithm (Tabulation)

1. Use DP array where  $dp[i] =$  ways to climb  $i$  stairs.
2. Initialize  $dp[0]=1$ ,  $dp[1]=1$ .
3.  $dp[i] = dp[i-1] + dp[i-2]$  for  $i$  from 2 to  $n$ .

**Time Complexity:**  $O(n)$     **Space Complexity:**  $O(n)$  (memoization),  $O(1)$  (tabulation)

## 6.5 Python Solution (Memoization)

```
1 def climb_stairs_memo(n, memo=None):
2     if memo is None:
3         memo = {}
4     if n in memo:
5         return memo[n]
6     if n <= 0:
7         return 1 if n == 0 else 0
8     memo[n] = climb_stairs_memo(n-1, memo) +
9             climb_stairs_memo(n-2, memo)
10    return memo[n]
```

## 6.6 Python Solution (Tabulation)

```
1 def climb_stairs(n):
2     if n <= 1:
3         return 1
4     dp = [0] * (n + 1)
5     dp[0] = dp[1] = 1
6     for i in range(2, n + 1):
7         dp[i] = dp[i-1] + dp[i-2]
8     return dp[n]
```

# 7 Min Cost Climbing Stairs

## 7.1 Problem Statement

Given a cost array for climbing stairs, find minimum cost to reach the top.

## 7.2 Dry Run on Test Cases

- \* **Test Case 1:** cost = [10,15,20] → Output: 15
- \* **Test Case 2:** cost = [1,100,1,1,1,100,1,1,100,1] → Output: 6
- \* **Test Case 3:** cost = [0,1] → Output: 0
- \* **Test Case 4:** cost = [1] → Output: 0

## 7.3 Algorithm (Memoization)

1. Recurse with memoization for min cost from index i.
2. Base case:  $i \geq \text{len}(\text{cost})$  returns 0.
3. Min cost =  $\text{cost}[i] + \min(\text{cost}(i+1), \text{cost}(i+2))$ .

## 7.4 Algorithm (Tabulation)

1.  $dp[i] = \min$  cost to reach top from  $i$ .
2.  $dp[n] = 0$ ,  $dp[n-1] = \text{cost}[n-1]$ .
3.  $dp[i] = \text{cost}[i] + \min(dp[i+1], dp[i+2])$ .

**Time Complexity:**  $O(n)$     **Space Complexity:**  $O(n)$  (memoization),  $O(1)$  (tabulation)

## 7.5 Python Solution (Memoization)

```
1 def min_cost_climbing_stairs_memo(cost):
2     memo = {}
3     def dp(i):
4         if i >= len(cost):
5             return 0
6         if i in memo:
7             return memo[i]
8         memo[i] = cost[i] + min(dp(i+1), dp(i+2))
9         return memo[i]
10    return min(dp(0), dp(1))
```

## 7.6 Python Solution (Tabulation)

```
1 def min_cost_climbing_stairs(cost):
2     n = len(cost)
3     dp = [0] * (n + 1)
4     for i in range(n-1, -1, -1):
5         dp[i] = cost[i] + min(dp[i+1], dp[i+2])
6     return min(dp[0], dp[1])
```

# 8 House Robber

## 8.1 Problem Statement

Given an array of house values, find max amount to rob without robbing adjacent houses.

## 8.2 Dry Run on Test Cases

- \* **Test Case 1:**  $\text{nums} = [1,2,3,1] \rightarrow \text{Output: } 4 (1+3)$
- \* **Test Case 2:**  $\text{nums} = [2,7,9,3,1] \rightarrow \text{Output: } 12 (2+9+1)$
- \* **Test Case 3:**  $\text{nums} = [1] \rightarrow \text{Output: } 1$
- \* **Test Case 4:**  $\text{nums} = [] \rightarrow \text{Output: } 0$

## 8.3 Algorithm (Memoization)

1. Recurse with memoization for max from index i.
2. Base case:  $i \geq \text{len}(\text{nums})$  returns 0.
3.  $\text{Max} = \max(\text{nums}[i] + \text{dp}(i+2), \text{dp}(i+1))$ .

## 8.4 Algorithm (Tabulation)

1.  $\text{dp}[i] = \text{max}$  amount from index i to end.
2.  $\text{dp}[n] = 0, \text{dp}[n-1] = \text{nums}[n-1]$ .
3.  $\text{dp}[i] = \max(\text{nums}[i] + \text{dp}[i+2], \text{dp}[i+1])$ .

**Time Complexity:**  $O(n)$     **Space Complexity:**  $O(n)$  (memoization),  
 $O(1)$  (tabulation)

## 8.5 Python Solution (Memoization)

```
1 def rob_memo(nums):
2     memo = []
3     def dp(i):
4         if i >= len(nums):
5             return 0
6         if i in memo:
7             return memo[i]
8         memo[i] = max(nums[i] + dp(i+2), dp(i+1))
9         return memo[i]
10    return dp(0)
```

## 8.6 Python Solution (Tabulation)

```
1 def rob(nums):
2     if not nums:
3         return 0
4     if len(nums) == 1:
5         return nums[0]
6
7     dp = [0] * (len(nums) + 1)
8     dp[-2] = nums[-1]
9     for i in range(len(nums)-2, -1, -1):
10        dp[i] = max(nums[i] + dp[i+2], dp[i+1])
11    return dp[0]
```

# 9 House Robber II

## 9.1 Problem Statement

Given a circular array of house values, find max amount to rob without robbing adjacent houses.

## 9.2 Dry Run on Test Cases

- **Test Case 1:** nums = [2,3,2] → Output: 3
- **Test Case 2:** nums = [1,2,3,1] → Output: 4
- **Test Case 3:** nums = [1] → Output: 1
- **Test Case 4:** nums = [] → Output: 0

## 9.3 Algorithm (Memoization)

1. Solve House Robber for nums[0:n-1] and nums[1:n].
2. Return max of both results.

## 9.4 Algorithm (Tabulation)

1. Use tabulation for two cases: exclude first and exclude last house.
2.  $dp[i] = \max(nums[i] + dp[i+2], dp[i+1])$ .

**Time Complexity:**  $O(n)$     **Space Complexity:**  $O(n)$  (memoization),  $O(1)$  (tabulation)

## 9.5 Python Solution (Tabulation)

```
1 def rob_circular(nums):
2     if not nums:
3         return 0
4     if len(nums) == 1:
5         return nums[0]
6
7     def rob_linear(arr):
8         if not arr:
9             return 0
10        if len(arr) == 1:
11            return arr[0]
12        dp = [0] * (len(arr) + 1)
13        dp[-2] = arr[-1]
14        for i in range(len(arr)-2, -1, -1):
15            dp[i] = max(arr[i] + dp[i+2], dp[i+1])
16        return dp[0]
17
18    return max(rob_linear(nums[:-1]), rob_linear(
19               nums[1:]))


```

# 10 Coin Change

## 10.1 Problem Statement

Given coins and an amount, find minimum coins needed to make the amount.

## 10.2 Dry Run on Test Cases

- **Test Case 1:** coins = [1,2,5], amount = 11 → Output: 3 (5+5+1)
- **Test Case 2:** coins = [2], amount = 3 → Output: -1
- **Test Case 3:** coins = [1], amount = 0 → Output: 0
- **Test Case 4:** coins = [1,2], amount = 1 → Output: 1

## 10.3 Algorithm (Memoization)

1. Recurse with memoization for min coins for amount.
2. Base case: amount=0 returns 0, amount<0 returns inf.
3. Try each coin and take minimum.

## 10.4 Algorithm (Tabulation)

1.  $dp[i]$  = min coins for amount i.
2. Initialize  $dp[0]=0$ , others=inf.
3. For each amount, try each coin, update dp.

**Time Complexity:**  $O(amount \cdot n)$     **Space Complexity:**  $O(amount)$

## 10.5 Python Solution (Memoization)

```
1 def coin_change_memo(coins, amount):
2     memo = {}
3     def dp(amt):
4         if amt in memo:
5             return memo[amt]
6         if amt == 0:
7             return 0
8         if amt < 0:
9             return float('inf')
10        min_coins = float('inf')
11        for coin in coins:
12            min_coins = min(min_coins, dp(amt -
13                coin) + 1)
14        memo[amt] = min_coins
```

```

14         return memo[amt]
15     result = dp(amount)
16     return result if result != float('inf') else -1

```

## 10.6 Python Solution (Tabulation)

```

1 def coin_change(coins, amount):
2     dp = [float('inf')] * (amount + 1)
3     dp[0] = 0
4     for i in range(1, amount + 1):
5         for coin in coins:
6             if i >= coin:
7                 dp[i] = min(dp[i], dp[i - coin] +
8                             1)
8     return dp[amount] if dp[amount] != float('inf') else -1

```

# 11 Coin Change II

## 11.1 Problem Statement

Given coins and an amount, find number of ways to make the amount.

## 11.2 Dry Run on Test Cases

- **Test Case 1:** amount = 5, coins = [1,2,5] → Output: 4 ([1,1,1,1,1], [1,1,1,2], [1,2,2], [5])
- **Test Case 2:** amount = 3, coins = [2] → Output: 0
- **Test Case 3:** amount = 10, coins = [10] → Output: 1
- **Test Case 4:** amount = 0, coins = [1] → Output: 1

## 11.3 Algorithm (Memoization)

1. Recurse with memoization for ways with index i and amount.
2. Base case: amount=0 returns 1, amount<0 or i>=len(coins) returns 0.
3. Sum ways by including or excluding current coin.

## 11.4 Algorithm (Tabulation)

1.  $dp[i]$  = ways to make amount i.
2. Initialize  $dp[0]=1$ .

3. For each coin, update  $dp[j] += dp[j-coin]$ .

**Time Complexity:**  $O(amount \cdot n)$     **Space Complexity:**  $O(amount)$

## 11.5 Python Solution (Memoization)

```

1 def change_memo(amount, coins):
2     memo = {}
3     def dp(i, amt):
4         if (i, amt) in memo:
5             return memo[(i, amt)]
6         if amt == 0:
7             return 1
8         if amt < 0 or i >= len(coins):
9             return 0
10        memo[(i, amt)] = dp(i, amt - coins[i]) + dp
11            (i + 1, amt)
12    return memo[(0, amount)]

```

## 11.6 Python Solution (Tabulation)

```

1 def change(amount, coins):
2     dp = [0] * (amount + 1)
3     dp[0] = 1
4     for coin in coins:
5         for i in range(coin, amount + 1):
6             dp[i] += dp[i - coin]
7     return dp[amount]

```

# 12 Longest Increasing Subsequence

## 12.1 Problem Statement

Given an array, find the length of the longest increasing subsequence.

## 12.2 Dry Run on Test Cases

- **Test Case 1:**  $\text{nums} = [10, 9, 2, 5, 3, 7, 101, 18]$  → Output: 4 ( $[2, 3, 7, 101]$ )
- **Test Case 2:**  $\text{nums} = [0, 1, 0, 3, 2, 3]$  → Output: 4
- **Test Case 3:**  $\text{nums} = [7, 7, 7, 7]$  → Output: 1
- **Test Case 4:**  $\text{nums} = []$  → Output: 0

## 12.3 Algorithm (Memoization)

1. Recurse with memoization for LIS ending at i with prev value.

2. Base case:  $i \geq \text{len}(\text{nums})$  returns 0.
3. Include  $\text{nums}[i]$  if greater than prev, else skip.

## 12.4 Algorithm (Tabulation)

1.  $\text{dp}[i] = \text{length of LIS ending at } i$ .
2. Initialize  $\text{dp}[i]=1$ .
3. For each  $i$ , check  $j < i$  where  $\text{nums}[j] < \text{nums}[i]$ , update  $\text{dp}[i]$ .

**Time Complexity:**  $O(n^2)$     **Space Complexity:**  $O(n)$

## 12.5 Python Solution (Memoization)

```

1 def length_of_lis_memo(nums):
2     memo = {}
3     def dp(i, prev):
4         if i >= len(nums):
5             return 0
6         if (i, prev) in memo:
7             return memo[(i, prev)]
8         take = 0
9         if prev == -1 or nums[i] > nums[prev]:
10            take = 1 + dp(i + 1, i)
11        skip = dp(i + 1, prev)
12        memo[(i, prev)] = max(take, skip)
13    return memo[(i, prev)]
14
15 return dp(0, -1)

```

## 12.6 Python Solution (Tabulation)

```

1 def length_of_lis(nums):
2     if not nums:
3         return 0
4     dp = [1] * len(nums)
5     for i in range(1, len(nums)):
6         for j in range(i):
7             if nums[i] > nums[j]:
8                 dp[i] = max(dp[i], dp[j] + 1)
9
10 return max(dp)

```

# 13 Longest Common Subsequence

## 13.1 Problem Statement

Given two strings, find the length of their longest common subsequence.

## 13.2 Dry Run on Test Cases

- **Test Case 1:** text1 = "abcde", text2 = "ace" → Output: 3 ("ace")
- **Test Case 2:** text1 = "abc", text2 = "abc" → Output: 3
- **Test Case 3:** text1 = "abc", text2 = "def" → Output: 0
- **Test Case 4:** text1 = "", text2 = "a" → Output: 0

## 13.3 Algorithm (Memoization)

1. Recurse with memoization for LCS of prefixes i, j.
2. If characters match, add 1 and recurse.
3. Else, take max of skipping one character.

## 13.4 Algorithm (Tabulation)

1.  $dp[i][j]$  = LCS length for prefixes i, j.
2. If characters match,  $dp[i][j] = dp[i-1][j-1] + 1$ .
3. Else,  $dp[i][j] = \max(dp[i-1][j], dp[i][j-1])$ .

Time Complexity:  $O(m \cdot n)$     Space Complexity:  $O(m \cdot n)$

## 13.5 Python Solution (Memoization)

```
1 def longest_common_subsequence_memo(text1, text2):
2     memo = {}
3     def dp(i, j):
4         if i >= len(text1) or j >= len(text2):
5             return 0
6         if (i, j) in memo:
7             return memo[(i, j)]
8         if text1[i] == text2[j]:
9             memo[(i, j)] = 1 + dp(i + 1, j + 1)
10        else:
11            memo[(i, j)] = max(dp(i + 1, j), dp(i,
12                j + 1))
13    return memo[(i, j)]
14
15 return dp(0, 0)
```

## 13.6 Python Solution (Tabulation)

```
1 def longest_common_subsequence(text1, text2):
2     m, n = len(text1), len(text2)
3     dp = [[0] * (n + 1) for _ in range(m + 1)]
4     for i in range(1, m + 1):
5         for j in range(1, n + 1):
```

```

6         if text1[i-1] == text2[j-1]:
7             dp[i][j] = dp[i-1][j-1] + 1
8         else:
9             dp[i][j] = max(dp[i-1][j], dp[i][j-1])
10    return dp[m][n]
11 \end Prefeitura
12
13 % Problem 124
14 \section{Edit Distance}
15 \subsection{Problem Statement}
16 Given two strings, find minimum operations (insert,
   delete, replace) to convert one to the other.
17
18 \subsection{Dry Run on Test Cases}
19 \begin{itemize}
20     \item \textbf{Test Case 1}: word1 = "horse",
           word2 = "ros" $\rightarrow$ Output: 3
21     \item \textbf{Test Case 2}: word1 = "intention"
           , word2 = "execution" $\rightarrow$ Output:
           5
22     \item \textbf{Test Case 3}: word1 = "", word2 =
           "a" $\rightarrow$ Output: 1
23     \item \textbf{Test Case 4}: word1 = "abc",
           word2 = "abc" $\rightarrow$ Output: 0
24 \end{itemize}
25
26 \subsection{Algorithm (Memoization)}
27 \begin{enumerate}
28     \item Recurse with memoization for min
           operations for prefixes i, j.
29     \item If characters match, no operation needed.
30     \item Else, try insert, delete, replace, take
           minimum.
31 \end{enumerate}
32 \subsection{Algorithm (Tabulation)}
33 \begin{enumerate}
34     \item  $dp[i][j] = \min$  operations for prefixes i,
           j.
35     \item Initialize  $dp[0][j] = j$ ,  $dp[i][0] = i$ .
36     \item If match,  $dp[i][j] = dp[i-1][j-1]$ .
37     \item Else,  $dp[i][j] = \min(\text{insert}, \text{delete},
           \text{replace}) + 1$ .
38 \end{enumerate}
39 \textbf{Time Complexity}:  $\mathcal{O}(m \cdot n)$  \quad \
40 \textbf{Space Complexity}:  $\mathcal{O}(m \cdot n)$ 
41 \subsection{Python Solution (Memoization)}
42 \begin{lstlisting}
43 def min_distance_memo(word1, word2):
44     memo = {}

```

```

45     def dp(i, j):
46         if i >= len(word1):
47             return len(word2) - j
48         if j >= len(word2):
49             return len(word1) - i
50         if (i, j) in memo:
51             return memo[(i, j)]
52         if word1[i] == word2[j]:
53             memo[(i, j)] = dp(i + 1, j + 1)
54         else:
55             memo[(i, j)] = min(dp(i, j + 1), dp(i +
56                                 1, j), dp(i + 1, j + 1)) + 1
57     return memo[(i, j)]
      return dp(0, 0)

```

## 13.7 Python Solution (Tabulation)

```

1 def min_distance(word1, word2):
2     m, n = len(word1), len(word2)
3     dp = [[0] * (n + 1) for _ in range(m + 1)]
4     for j in range(n + 1):
5         dp[0][j] = j
6     for i in range(m + 1):
7         dp[i][0] = i
8     for i in range(1, m + 1):
9         for j in range(1, n + 1):
10            if word1[i-1] == word2[j-1]:
11                dp[i][j] = dp[i-1][j-1]
12            else:
13                dp[i][j] = min(dp[i-1][j], dp[i][j -
14                                1], dp[i-1][j-1]) + 1
15     return dp[m][n]

```

# 14 Unique Paths

## 14.1 Problem Statement

Given an  $m \times n$  grid, find number of unique paths from top-left to bottom-right (right/down moves).

## 14.2 Dry Run on Test Cases

- **Test Case 1:**  $m = 3, n = 7 \rightarrow$  Output: 28
- **Test Case 2:**  $m = 3, n = 2 \rightarrow$  Output: 3
- **Test Case 3:**  $m = 1, n = 1 \rightarrow$  Output: 1
- **Test Case 4:**  $m = 2, n = 2 \rightarrow$  Output: 2

### 14.3 Algorithm (Memoization)

1. Recurse with memoization for paths to (i,j).
2. Base case: (0,0) returns 1, out of bounds returns 0.
3. Paths = sum of paths from above and left.

### 14.4 Algorithm (Tabulation)

1.  $dp[i][j] = \text{paths to } (i,j)$ .
2. Initialize first row and column to 1.
3.  $dp[i][j] = dp[i-1][j] + dp[i][j-1]$ .

**Time Complexity:**  $O(m \cdot n)$     **Space Complexity:**  $O(m \cdot n)$

### 14.5 Python Solution (Memoization)

```
1 def unique_paths_memo(m, n):
2     memo = {}
3     def dp(i, j):
4         if i < 0 or j < 0:
5             return 0
6         if i == 0 and j == 0:
7             return 1
8         if (i, j) in memo:
9             return memo[(i, j)]
10        memo[(i, j)] = dp(i-1, j) + dp(i, j-1)
11    return memo[(i, j)]
12 return dp(m-1, n-1)
```

### 14.6 Python Solution (Tabulation)

```
1 def unique_paths(m, n):
2     dp = [[1] * n for _ in range(m)]
3     for i in range(1, m):
4         for j in range(1, n):
5             dp[i][j] = dp[i-1][j] + dp[i][j-1]
6     return dp[m-1][n-1]
```

## 15 Unique Paths II

### 15.1 Problem Statement

Given an  $m \times n$  grid with obstacles, find number of unique paths to bottom-right.

## 15.2 Dry Run on Test Cases

- **Test Case 1:** obstacleGrid = [[0,0,0],[0,1,0],[0,0,0]] → Output: 2
- **Test Case 2:** obstacleGrid = [[0,1],[0,0]] → Output: 1
- **Test Case 3:** obstacleGrid = [[1]] → Output: 0
- **Test Case 4:** obstacleGrid = [[0]] → Output: 1

## 15.3 Algorithm (Memoization)

1. Recurse with memoization for paths to (i,j).
2. Base case: obstacle or out of bounds returns 0, (0,0) returns 1 if no obstacle.
3. Paths = sum of paths from above and left.

## 15.4 Algorithm (Tabulation)

1. dp[i][j] = paths to (i,j).
2. Initialize dp[0][0] based on obstacle.
3. For each cell,  $dp[i][j] = dp[i-1][j] + dp[i][j-1]$  if no obstacle.

**Time Complexity:**  $O(m \cdot n)$     **Space Complexity:**  $O(m \cdot n)$

## 15.5 Python Solution (Memoization)

```
1 def unique_paths_with_obstacles_memo(obstacleGrid):
2     if not obstacleGrid or obstacleGrid[0][0] == 1:
3         return 0
4     m, n = len(obstacleGrid), len(obstacleGrid[0])
5     memo = {}
6     def dp(i, j):
7         if i < 0 or j < 0 or obstacleGrid[i][j] == 1:
8             return 0
9         if i == 0 and j == 0:
10            return 1
11         if (i, j) in memo:
12             return memo[(i, j)]
13         memo[(i, j)] = dp(i-1, j) + dp(i, j-1)
14         return memo[(i, j)]
15     return dp(m-1, n-1)
```

## 15.6 Python Solution (Tabulation)

```
1 def unique_paths_with_obstacles(obstacleGrid):
2     if not obstacleGrid or obstacleGrid[0][0] == 1:
```

```

3         return 0
4     m, n = len(obstacleGrid), len(obstacleGrid[0])
5     dp = [[0] * n for _ in range(m)]
6     dp[0][0] = 1
7     for i in range(m):
8         for j in range(n):
9             if obstacleGrid[i][j] == 1:
10                 continue
11             if i > 0:
12                 dp[i][j] += dp[i-1][j]
13             if j > 0:
14                 dp[i][j] += dp[i][j-1]
15     return dp[m-1][n-1]

```

## 16 Minimum Path Sum

### 16.1 Problem Statement

Given an  $m \times n$  grid of non-negative numbers, find minimum path sum from top-left to bottom-right.

### 16.2 Dry Run on Test Cases

- **Test Case 1:** grid = [[1,3,1],[1,5,1],[4,2,1]] → Output: 7 (1->3->1->1->1)
- **Test Case 2:** grid = [[1,2],[3,4]] → Output: 7
- **Test Case 3:** grid = [[1]] → Output: 1
- **Test Case 4:** grid = [] → Output: 0

### 16.3 Algorithm (Memoization)

1. Recurse with memoization for min sum to (i,j).
2. Base case: (0,0) returns grid[0][0], out of bounds returns inf.
3. Min sum = grid[i][j] + min(above, left).

### 16.4 Algorithm (Tabulation)

1.  $dp[i][j] = \min \text{ sum to } (i,j)$ .
2. Initialize first row and column.
3.  $dp[i][j] = \text{grid}[i][j] + \min(dp[i-1][j], dp[i][j-1])$ .

**Time Complexity:**  $O(m \cdot n)$     **Space Complexity:**  $O(m \cdot n)$

## 16.5 Python Solution (Memoization)

```
1 def min_path_sum_memo(grid):
2     if not grid:
3         return 0
4     m, n = len(grid), len(grid[0])
5     memo = {}
6     def dp(i, j):
7         if i < 0 or j < 0:
8             return float('inf')
9         if i == 0 and j == 0:
10            return grid[0][0]
11        if (i, j) in memo:
12            return memo[(i, j)]
13        memo[(i, j)] = grid[i][j] + min(dp(i-1, j),
14                                         dp(i, j-1))
15    return memo[(m-1, n-1)]
```

## 16.6 Python Solution (Tabulation)

```
1 def min_path_sum(grid):
2     if not grid:
3         return 0
4     m, n = len(grid), len(grid[0])
5     dp = [[0] * n for _ in range(m)]
6     dp[0][0] = grid[0][0]
7     for i in range(1, m):
8         dp[i][0] = dp[i-1][0] + grid[i][0]
9     for j in range(1, n):
10        dp[0][j] = dp[0][j-1] + grid[0][j]
11    for i in range(1, m):
12        for j in range(1, n):
13            dp[i][j] = grid[i][j] + min(dp[i-1][j],
14                                         dp[i][j-1])
14    return dp[m-1][n-1]
```

# 17 Decode Ways

## 17.1 Problem Statement

Given a string of digits, find number of ways to decode it into letters (1='A', 2='B', ..., 26='Z').

## 17.2 Dry Run on Test Cases

- **Test Case 1:** s = "12" → Output: 2 ("AB" or "L")
- **Test Case 2:** s = "226" → Output: 3 ("BBF", "VF", "BF")

- **Test Case 3:**  $s = "06"$  → Output: 0
- **Test Case 4:**  $s = ""$  → Output: 0

### 17.3 Algorithm (Memoization)

1. Recurse with memoization for ways from index i.
2. Base case:  $i=\text{len}(s)$  returns 1, invalid digit returns 0.
3. Add ways for single and double digit decodes.

### 17.4 Algorithm (Tabulation)

1.  $dp[i]$  = ways to decode prefix of length i.
2. Initialize  $dp[0]=1$  if valid.
3.  $dp[i] += dp[i-1]$  if single digit valid,  $dp[i-2]$  if double digit valid.

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

### 17.5 Python Solution (Memoization)

```

1 def num_decodings_memo(s):
2     memo = {}
3     def dp(i):
4         if i == len(s):
5             return 1
6         if i in memo:
7             return memo[i]
8         if s[i] == '0':
9             return 0
10        result = dp(i + 1)
11        if i + 1 < len(s) and (s[i] == '1' or (s[i] == '2' and s[i+1] <= '6')):
12            result += dp(i + 2)
13        memo[i] = result
14    return result
15 return 0 if not s or s[0] == '0' else dp(0)

```

### 17.6 Python Solution (Tabulation)

```

1 def num_decodings(s):
2     if not s or s[0] == '0':
3         return 0
4     dp = [0] * (len(s) + 1)
5     dp[0] = 1
6     dp[1] = 1
7     for i in range(2, len(s) + 1):
8         if s[i-1] != '0':

```

```

9         dp[i] += dp[i-1]
10        if s[i-2] == '1' or (s[i-2] == '2' and s[i
11            -1] <= '6'):
12            dp[i] += dp[i-2]
13    return dp[len(s)]

```

## 18 Maximal Square

### 18.1 Problem Statement

Given a 2D binary matrix, find the size of the largest square of 1s.

### 18.2 Dry Run on Test Cases

- **Test Case 1:** matrix = [["1", "0", "1", "0", "0"], ["1", "0", "1", "1", "1"], ["1", "1", "1", "1", "1"], ["1", "1", "1", "1", "1"], ["1", "1", "1", "1", "1"]] → Output: 4
- **Test Case 2:** matrix = [[["0", "1"], ["1", "0"]]] → Output: 1
- **Test Case 3:** matrix = [[["0"]]] → Output: 0
- **Test Case 4:** matrix = [] → Output: 0

### 18.3 Algorithm (Tabulation)

1.  $dp[i][j]$  = side length of max square ending at  $(i,j)$ .
2. If  $matrix[i][j] = '1'$ ,  $dp[i][j] = \min(dp[i-1][j], dp[i][j-1], dp[i-1][j-1]) + 1$ .
3. Track max side length.

**Time Complexity:**  $O(m \cdot n)$     **Space Complexity:**  $O(m \cdot n)$

### 18.4 Python Solution (Tabulation)

```

1 def maximal_square(matrix):
2     if not matrix or not matrix[0]:
3         return 0
4     m, n = len(matrix), len(matrix[0])
5     dp = [[0] * (n + 1) for _ in range(m + 1)]
6     max_side = 0
7     for i in range(1, m + 1):
8         for j in range(1, n + 1):
9             if matrix[i-1][j-1] == '1':
10                 dp[i][j] = min(dp[i-1][j], dp[i][j-1],
11                                 dp[i-1][j-1]) + 1
12                 max_side = max(max_side, dp[i][j])
13
14 return max_side * max_side

```

# 19 Palindromic Substrings

## 19.1 Problem Statement

Given a string, count all palindromic substrings.

## 19.2 Dry Run on Test Cases

- **Test Case 1:**  $s = "abc" \rightarrow \text{Output: } 3 ("a", "b", "c")$
- **Test Case 2:**  $s = "aaa" \rightarrow \text{Output: } 6 ("a", "a", "a", "aa", "aa", "aaa")$
- **Test Case 3:**  $s = "" \rightarrow \text{Output: } 0$
- **Test Case 4:**  $s = "a" \rightarrow \text{Output: } 1$

## 19.3 Algorithm (Tabulation)

1.  $dp[i][j] = \text{True}$  if substring  $i$  to  $j$  is palindrome.
2. Single characters are palindromes.
3. Check pairs, then longer substrings using  $dp[i+1][j-1]$ .

**Time Complexity:**  $O(n^2)$     **Space Complexity:**  $O(n^2)$

## 19.4 Python Solution (Tabulation)

```
1 def count_substrings(s):
2     n = len(s)
3     dp = [[False] * n for _ in range(n)]
4     count = 0
5
6     # Single characters
7     for i in range(n):
8         dp[i][i] = True
9         count += 1
10
11    # Pairs
12    for i in range(n-1):
13        if s[i] == s[i+1]:
14            dp[i][i+1] = True
15            count += 1
16
17    # Longer substrings
18    for length in range(3, n + 1):
19        for i in range(n - length + 1):
20            j = i + length - 1
21            if s[i] == s[j] and dp[i+1][j-1]:
22                dp[i][j] = True
23                count += 1
```

```
24  
25     return count
```

## 20 Longest Palindromic Substring

### 20.1 Problem Statement

Given a string, find the longest palindromic substring.

### 20.2 Dry Run on Test Cases

- **Test Case 1:** s = "babad" → Output: "bab" or "aba"
- **Test Case 2:** s = "cbbd" → Output: "bb"
- **Test Case 3:** s = "" → Output: ""
- **Test Case 4:** s = "a" → Output: "a"

### 20.3 Algorithm (Tabulation)

1.  $dp[i][j] = \text{True}$  if substring i to j is palindrome.
2. Track longest palindrome with start index and length.
3. Check single, pairs, then longer substrings.

Time Complexity:  $O(n^2)$  Space Complexity:  $O(n^2)$

### 20.4 Python Solution (Tabulation)

```
1 def longest_palindromic_substring(s):  
2     n = len(s)  
3     if n == 0:  
4         return ""  
5     dp = [[False] * n for _ in range(n)]  
6     start, max_len = 0, 1  
7  
8     # Single characters  
9     for i in range(n):  
10        dp[i][i] = True  
11  
12    # Pairs  
13    for i in range(n-1):  
14        if s[i] == s[i+1]:  
15            dp[i][i+1] = True  
16            start = i  
17            max_len = 2  
18  
19    # Longer substrings
```

```

20     for length in range(3, n + 1):
21         for i in range(n - length + 1):
22             j = i + length - 1
23             if s[i] == s[j] and dp[i+1][j-1]:
24                 dp[i][j] = True
25                 if length > max_len:
26                     start = i
27                     max_len = length
28
29     return s[start:start + max_len]

```

## 21 Number of 1 Bits

### 21.1 Problem Statement

Given an unsigned integer, return the number of 1 bits (Hamming weight).

### 21.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 11$  ( $1011$ )  $\rightarrow$  Output: 3
- **Test Case 2:**  $n = 128$  ( $100000000$ )  $\rightarrow$  Output: 1
- **Test Case 3:**  $n = 0$   $\rightarrow$  Output: 0
- **Test Case 4:**  $n = 4294967295$  ( $111\dots1$ )  $\rightarrow$  Output: 32

### 21.3 Algorithm

1. Use bitwise AND with 1 to check least significant bit.
2. Right shift number and count 1s.

**Time Complexity:**  $O(1)$  (32 bits)    **Space Complexity:**  $O(1)$

### 21.4 Python Solution

```

1 def hamming_weight(n):
2     count = 0
3     while n:
4         count += n & 1
5         n >>= 1
6     return count

```

## 22 Sum of Two Integers

## 22.1 Problem Statement

Given two integers, calculate their sum without using + or - operators.

## 22.2 Dry Run on Test Cases

- **Test Case 1:**  $a = 1, b = 2 \rightarrow$  Output: 3
- **Test Case 2:**  $a = 2, b = 3 \rightarrow$  Output: 5
- **Test Case 3:**  $a = 0, b = 0 \rightarrow$  Output: 0
- **Test Case 4:**  $a = -2, b = 3 \rightarrow$  Output: 1

## 22.3 Algorithm

1. Use XOR for sum without carry.
2. Use AND and left shift for carry.
3. Repeat until carry is 0, mask to 32 bits.

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

## 22.4 Python Solution

```
1 def get_sum(a, b):  
2     mask = 0xffffffff  
3     while (b & mask) > 0:  
4         carry = ((a & b) << 1) & mask  
5         a = (a ^ b) & mask  
6         b = carry  
7     if (a >> 31) & 1: # Handle negative numbers  
8         return ~(a ^ mask)  
9     return a
```

## 23 Single Number

### 23.1 Problem Statement

Given an array where every element appears twice except one, find the single number.

### 23.2 Dry Run on Test Cases

- **Test Case 1:**  $\text{nums} = [2,2,1] \rightarrow$  Output: 1
- **Test Case 2:**  $\text{nums} = [4,1,2,1,2] \rightarrow$  Output: 4
- **Test Case 3:**  $\text{nums} = [1] \rightarrow$  Output: 1

- **Test Case 4:**  $\text{nums} = [] \rightarrow \text{Output: } 0$

### 23.3 Algorithm

1. Use XOR of all numbers.
2. Paired numbers cancel out, leaving single number.

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

### 23.4 Python Solution

```

1 def single_number(nums):
2     result = 0
3     for num in nums:
4         result ^= num
5     return result

```

## 24 Missing Number

### 24.1 Problem Statement

Given an array of  $[0,n]$  with one number missing, find the missing number.

### 24.2 Dry Run on Test Cases

- **Test Case 1:**  $\text{nums} = [3,0,1] \rightarrow \text{Output: } 2$
- **Test Case 2:**  $\text{nums} = [0,1] \rightarrow \text{Output: } 2$
- **Test Case 3:**  $\text{nums} = [9,6,4,2,3,5,7,0,1] \rightarrow \text{Output: } 8$
- **Test Case 4:**  $\text{nums} = [0] \rightarrow \text{Output: } 1$

### 24.3 Algorithm

1. Use XOR of all indices and numbers.
2. XOR with n to account for array length.
3. Result is the missing number.

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

### 24.4 Python Solution

```

1 def missing_number(nums):
2     result = len(nums)
3     for i, num in enumerate(nums):

```

```
4     result ^= i ^ num
5 return result
```

## 25 Bitwise AND of Numbers Range

### 25.1 Problem Statement

Given a range [left, right], find the bitwise AND of all numbers.

### 25.2 Dry Run on Test Cases

- **Test Case 1:** left = 5, right = 7 → Output: 4 ( $5 \text{ } 6 \text{ } 7 = 4$ )
- **Test Case 2:** left = 0, right = 0 → Output: 0
- **Test Case 3:** left = 1, right = 2147483647 → Output: 0
- **Test Case 4:** left = 2, right = 3 → Output: 2

### 25.3 Algorithm

1. Shift left and right until they are equal.
2. Track shifts, shift result back.
3. Common prefix is the AND result.

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

### 25.4 Python Solution

```
1 def range_bitwise_and(left, right):
2     shift = 0
3     while left != right:
4         left >>= 1
5         right >>= 1
6         shift += 1
7     return left << shift
```

## 26 Counting Bits

### 26.1 Problem Statement

Given an integer n, return an array of number of 1 bits for numbers 0 to n.

## 26.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 2 \rightarrow$  Output:  $[0,1,1]$
- **Test Case 2:**  $n = 5 \rightarrow$  Output:  $[0,1,1,2,1,2]$
- **Test Case 3:**  $n = 0 \rightarrow$  Output:  $[0]$
- **Test Case 4:**  $n = 1 \rightarrow$  Output:  $[0,1]$

## 26.3 Algorithm

1. For each  $i$ , use  $i - (i-1)$  to remove least significant 1.
2.  $dp[i] = dp[i - (i-1)] + 1$ .

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

## 26.4 Python Solution

```
1 def count_bits(n):  
2     dp = [0] * (n + 1)  
3     for i in range(1, n + 1):  
4         dp[i] = dp[i & (i-1)] + 1  
5     return dp
```

# 27 Reverse Bits

## 27.1 Problem Statement

Given a 32-bit unsigned integer, reverse its bits.

## 27.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 43261596$  (00000010100101000001111010011100)  
 $\rightarrow$  Output: 964176192 (00111001011110000010100101000000)
- **Test Case 2:**  $n = 4294967295$  (11111111111111111111111111111111)  
 $\rightarrow$  Output: 4294967295
- **Test Case 3:**  $n = 0 \rightarrow$  Output: 0
- **Test Case 4:**  $n = 1 \rightarrow$  Output: 2147483648

## 27.3 Algorithm

1. Initialize result = 0.
2. For 32 bits, shift result left, add  $n$ 's least significant bit, shift  $n$  right.

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

## 27.4 Python Solution

```
1 def reverse_bits(n):
2     result = 0
3     for _ in range(32):
4         result = (result << 1) | (n & 1)
5         n >>= 1
6     return result
```

# 28 Power of Two

## 28.1 Problem Statement

Given an integer, determine if it is a power of 2.

## 28.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 1 \rightarrow$  Output: True
- **Test Case 2:**  $n = 16 \rightarrow$  Output: True
- **Test Case 3:**  $n = 3 \rightarrow$  Output: False
- **Test Case 4:**  $n = 0 \rightarrow$  Output: False

## 28.3 Algorithm

1. Check if  $n > 0$  and  $n \ (n-1) == 0$ .
2. Powers of 2 have exactly one 1 bit.

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

## 28.4 Python Solution

```
1 def is_power_of_two(n):
2     return n > 0 and n & (n - 1) == 0
```

# 29 Power of Three

## 29.1 Problem Statement

Given an integer, determine if it is a power of 3.

## 29.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 27 \rightarrow$  Output: True
- **Test Case 2:**  $n = 0 \rightarrow$  Output: False
- **Test Case 3:**  $n = 9 \rightarrow$  Output: True
- **Test Case 4:**  $n = 45 \rightarrow$  Output: False

## 29.3 Algorithm

1. Check if  $n > 0$  and  $3^{19}$  Time Complexity:  $O(1)$  Space Complexity:  $O(1)$

## 29.4 Python Solution

```
1 def is_power_of_three(n):
2     return n > 0 and 1162261467 % n == 0 #  $3^{19} =$ 
3     1162261467
```

# 30 Power of Four

## 30.1 Problem Statement

Given an integer, determine if it is a power of 4.

## 30.2 Dry Run on Test Cases

- **Test Case 1:**  $n = 16 \rightarrow$  Output: True
- **Test Case 2:**  $n = 5 \rightarrow$  Output: False
- **Test Case 3:**  $n = 1 \rightarrow$  Output: True
- **Test Case 4:**  $n = 0 \rightarrow$  Output: False

## 30.3 Algorithm

1. Check if  $n > 0$  and  $n \ (n-1) == 0$  (power of 2).
2. Check if 1 bits are in odd positions ( $n \ 0x55555555 != 0$ ).

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

## 30.4 Python Solution

```
1 def is_power_of_four(n):
2     return n > 0 and n & (n - 1) == 0 and n & 0
3     x55555555 != 0
```