Graph

1. Number of Islands

 ${\bf Pattern} \colon {\bf Graph\ Traversal\ (DFS/BFS\ on\ Grid)}$

Problem Statement

Given an m x n 2D binary grid grid which represents a map of '1's (land) and '0's (water), return the number of islands.

An **island** is surrounded by water and is formed by connecting adjacent lands horizontally or vertically. You may assume all four edges of the grid are all surrounded by water.

```
Input: grid = [
    ["1","1","1","1","0"],
    ["1","1","0","1","0"],
    ["0","0","0","0","0"]
]
Output: 3
Explanation: One contiguous island in top-left (9 cells),
one single-cell island at (0,3), rest is water.
```

```
Input: grid = [["1"]]
Output: 1
Explanation: Single land cell → one island.

Input: grid = [
    ["0","0","0"],
    ["0","0","0"]
]
Output: 0
Explanation: No land → zero islands.
```

```
from typing import List
class Solution:
    def numIslands(self, grid: List[List[str]]) -> int:
        # STEP 1: Initialize structures
        # - Track visited land via in-place mutation or visited set.
        # - Here, we mutate grid: mark visited '1' as '0' to avoid revisiting.
        if not grid or not grid[0]:
            return 0
        rows, cols = len(grid), len(grid[0])
        island_count = 0
        # Helper: DFS to sink the entire island
        def dfs(r, c):
            # Boundary check + water/visited check
            if (r < 0 \text{ or } r \ge rows \text{ or } c < 0 \text{ or } c \ge rols
                 or grid[r][c] == '0'):
                 return
            # Mark as visited by turning into water
            grid[r][c] = '0'
            # Visit all 4 neighbors
            dfs(r + 1, c)
            dfs(r - 1, c)
```

```
dfs(r, c + 1)
           dfs(r, c - 1)
       # STEP 2: Main loop / recursion
       # - Scan every cell. When we find unvisited land ('1'),
             trigger DFS to "sink" the whole island and increment count.
       for r in range(rows):
           for c in range(cols):
               if grid[r][c] == '1':
                   dfs(r, c)
                   island_count += 1
       # STEP 3: Update state / bookkeeping
       # - island_count is updated per DFS call.
       # - Grid is mutated in place (allowed per problem constraints).
       # STEP 4: Return result
       # - Handles empty grid above; otherwise returns count.
       return island_count
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   grid1 = [
       ["1","1","1","1","0"],
       ["1","1","0","1","0"],
       ["1","1","0","0","0"],
       ["0","0","0","0","0"]
   print(sol.numIslands(grid1)) # Expected: 1 (not 3! - all connected)
   # Actually one contiguous island. Earlier explanation was wrong.
   # Test 2: Edge case - single cell
   grid2 = [["1"]]
   print(sol.numIslands(grid2)) # Expected: 1
   # Test 3: Tricky/negative - all water
   grid3 = [
       ["0","0","0"],
       ["0","0","0"]
```

```
print(sol.numIslands(grid3)) # Expected: 0
```

Example Walkthrough

We'll walk through **Test 1** step by step:

Initial grid:

```
[
    ['1','1','1','1','0'],
    ['1','1','0','1','0'],
    ['1','1','0','0','0','0'],
    ['0','0','0','0','0']]
```

Step 1: r=0, c=0 \rightarrow grid[0][0] == '1' \rightarrow start DFS.

- Mark grid[0][0] = '0'.
- Recursively visit neighbors: (1,0), (-1,0) invalid, (0,1), (0,-1) invalid.

Step 2: From (0,1): mark as '0', visit (0,2), (1,1), etc.

- DFS continues, turning all connected '1's into '0's.
- Eventually, the entire top-left block (including (1,3)) becomes '0' because: (0,3) connects down to (1,3), which connects left to (1,2)? No (1,2) is '0'.
- But (0,3) is connected to $(0,2) \rightarrow \text{yes!}$ So it's all one island.

Result: Only one DFS is triggered \rightarrow island_count = 1.

Correction: The initial sample explanation mistakenly said "3 islands", but the grid actually forms **one** connected island. The correct LeetCode example with output 3 uses a different grid (e.g., with disconnected land at (0,3) and (1,3) not connected). Our test uses a fully connected top region.

Final outputs:

1 1 0

Complexity Analysis

• Time Complexity: O(m * n)

Each cell is visited at most once. DFS visits each land cell once, and outer loops scan all cells. Total operations scale linearly with grid size.

• Space Complexity: O(m * n)

In worst case (all '1's), recursion depth = m * n (e.g., snake-like DFS path). This is the call stack size. No extra visited set used — we mutate input.

2. Flood Fill

Pattern: Graph Traversal (DFS/BFS)

Problem Statement

An image is represented by an m x n integer grid image where image[i][j] represents the pixel value of the image.

You are also given three integers sr, sc, and color. You should perform a flood fill on the image starting from the pixel image[sr][sc].

To perform a flood fill, consider the starting pixel, plus any pixels connected 4-directionally to the starting pixel of the same color as the starting pixel, plus any pixels connected 4-directionally to those pixels (and so on), and change their color to color.

Return the modified image after performing the flood fill.

```
Input: image = [[1,1,1],[1,1,0],[1,0,1]], sr = 1, sc = 1, color = 2
Output: [[2,2,2],[2,2,0],[2,0,1]]
Explanation: From the center (1,1), all connected 1s are changed to 2.

Input: image = [[0,0,0],[0,0,0]], sr = 0, sc = 0, color = 0
Output: [[0,0,0],[0,0,0]]
Explanation: No change - new color is same as original.

Input: image = [[1]], sr = 0, sc = 0, color = 3
Output: [[3]]
Explanation: Single pixel updated.
```

```
from typing import List
class Solution:
    def floodFill(
        self, image: List[List[int]], sr: int, sc: int, color: int
    ) -> List[List[int]]:
        # STEP 1: Initialize structures
           - Store original color to avoid infinite loops
            - If new color == original, return early (no work needed)
        original_color = image[sr][sc]
        if original color == color:
            return image
        rows, cols = len(image), len(image[0])
        # STEP 2: Main loop / recursion
        # - Use DFS via recursion to visit 4-directional neighbors
        # - Invariant: only pixels with original_color are processed
        def dfs(r, c):
            # Base case: out of bounds or wrong color
            if (
                r < 0 or r >= rows or
                c < 0 or c >= cols or
```

```
image[r][c] != original_color
           ):
               return
           # STEP 3: Update state / bookkeeping
           # - Paint current pixel
           # - Recurse to neighbors (up, down, left, right)
           image[r][c] = color
           dfs(r - 1, c) # up
           dfs(r + 1, c) # down
           dfs(r, c - 1) # left
           dfs(r, c + 1) # right
       dfs(sr, sc)
       # STEP 4: Return result
       # - Image is modified in-place; return it
       return image
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   img1 = [[1,1,1],[1,1,0],[1,0,1]]
   result1 = sol.floodFill(img1, 1, 1, 2)
   expected1 = [[2,2,2],[2,2,0],[2,0,1]]
   assert result1 == expected1, f"Test 1 failed: got {result1}"
   print(" Test 1 passed")
   # Test 2: Edge case - same color
   img2 = [[0,0,0],[0,0,0]]
   result2 = sol.floodFill(img2, 0, 0, 0)
   expected2 = [[0,0,0],[0,0,0]]
   assert result2 == expected2, f"Test 2 failed: got {result2}"
   print(" Test 2 passed")
   # Test 3: Tricky/negative - single pixel
   img3 = [[1]]
   result3 = sol.floodFill(img3, 0, 0, 3)
   expected3 = [[3]]
   assert result3 == expected3, f"Test 3 failed: got {result3}"
```

```
print(" Test 3 passed")
```

Example Walkthrough

```
We'll trace Test 1:
image = [[1,1,1],[1,1,0],[1,0,1]], sr=1, sc=1, color=2
Initial state:
-original_color = image[1][1] = 1
- color = 2 \rightarrow \text{different} \rightarrow \text{proceed}
- rows = 3, cols = 3
Call dfs(1, 1):
- (1,1) is in bounds and image[1][1] == 1 \rightarrow valid
- Set image [1] [1] = 2 \rightarrow image now:
[[1,1,1],
                   [1,2,0],
                                    [1,0,1]] - Recurse to 4 neighbors:
   1. dfs(0,1) (up):
          • Valid (value = 1) \rightarrow set to 2
          • Image: row 0 becomes [1,2,1]
          • Recurse from (0,1):
               - Up: (-1,1) \rightarrow \text{invalid}
               - Down: (1,1) \rightarrow \text{now } 2 \quad 1 \rightarrow \text{skip}
               - Left: (0,0) \rightarrow \text{value}=1 \rightarrow \text{set to } 2
                    * From (0,0): up/down invalid; right=(0,1)=2 (skip); left invalid
               - Right: (0,2) \rightarrow \text{value}=1 \rightarrow \text{set to } 2
                    * From (0,2): neighbors checked \rightarrow no new changes
          • Row 0 now: [2,2,2]
   2. dfs(2,1) (down):
```

- image[2][1] = 0 $1 \rightarrow \text{return immediately}$
- 3. dfs(1,0) (left):
 - image[1][0] = 1 \rightarrow set to 2
 - Image row 1: [2,2,0]
 - From (1,0):

- Up:
$$(0,0) = 2 \to \text{skip}$$

- Down:
$$(2,0) = 1 \rightarrow \text{valid} \rightarrow \text{set to } 2$$

* From $(2,0)$: neighbors $\rightarrow (1,0)=2$, $(2,1)=0 \rightarrow \text{stop}$

- Row 2 becomes: [2,0,1]
- 4. dfs(1,2) (right):
 - $image[1][2] = 0 \rightarrow skip$

Final image:

[[2,2,2],

[2,2,0],

[2,0,1]]

Matches expected output

Complexity Analysis

• Time Complexity: O(m * n)

In worst case, we visit every pixel once (e.g., entire grid same color). Each pixel processed once via DFS.

• Space Complexity: O(m * n)

Recursion stack depth can be up to $\mathtt{m} * \mathtt{n}$ in worst case (e.g., snake-like path). No extra data structures beyond input.

3. 01 Matrix

Pattern: Multi-source BFS (Breadth-First Search)

Problem Statement

Given an $m \times n$ binary matrix mat, return the distance of the nearest 0 for each cell.

The distance between two adjacent cells is 1.

Clarification:

- "Adjacent" means up/down/left/right (4-directional).
- If a cell is 0, its distance is 0.
- All other cells must compute shortest distance to any 0.

```
Input: [[0,0,0],[0,1,0],[0,0,0]]
Output: [[0,0,0],[0,1,0],[0,0,0]]
Explanation: All 1s are already adjacent to 0s; distance = 1.

Input: [[0,0,0],[0,1,0],[1,1,1]]
Output: [[0,0,0],[0,1,0],[1,2,1]]
Explanation: Bottom row: middle cell is 2 steps from nearest 0.

Input: [[1]]
Output: [[-1]] → Wait! Actually: [[0]] if input were [[0]],
but [[1]] has no 0!
Correction: Input must contain at least one 0 per problem constraints.
So edge case: [[1,0]] → Output: [[1,0]]
```

```
Final test cases: - Normal: [[0,1,1],[1,1,1],[1,1,0]] \rightarrow [[0,1,2],[1,2,1],[2,1,0]] - Edge: [[0]] \rightarrow [[0]] - Tricky: [[1,0,1],[1,1,1],[1,1,1]] \rightarrow [[1,0,1],[2,1,2],[3,2,3]]
```

```
from collections import deque
from typing import List
class Solution:
    def updateMatrix(self, mat: List[List[int]]) -> List[List[int]]:
        # STEP 1: Initialize structures
          - Use multi-source BFS: start from all Os simultaneously.
            - dist matrix tracks shortest distance; init with 0s at 0-cells,
              and -1 (or large) for unvisited 1s.
        m, n = len(mat), len(mat[0])
        dist = [[-1] * n for _ in range(m)]
        q = deque()
        # Enqueue all Os as starting points (distance = 0)
        for i in range(m):
            for j in range(n):
                if mat[i][j] == 0:
                    dist[i][j] = 0
                    q.append((i, j))
        # STEP 2: Main loop / recursion
            - BFS guarantees shortest path in unweighted grid.
          - Process neighbors level-by-level (distance increases by 1).
        directions = [(1, 0), (-1, 0), (0, 1), (0, -1)]
        while q:
            x, y = q.popleft()
            # STEP 3: Update state / bookkeeping
            # - For each neighbor, if unvisited, set distance = current + 1
            for dx, dy in directions:
                nx, ny = x + dx, y + dy
                if (0 \le nx \le m \text{ and } 0 \le ny \le n \text{ and}
                    dist[nx][ny] == -1): # not visited
                    dist[nx][ny] = dist[x][y] + 1
                    q.append((nx, ny))
        # STEP 4: Return result
        # - All cells guaranteed reachable (problem ensures 1 zero)
        return dist
```

```
# ------ INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   mat1 = [[0,1,1],[1,1,1],[1,1,0]]
   out1 = sol.updateMatrix(mat1)
   expected1 = [[0,1,2],[1,2,1],[2,1,0]]
   assert out1 == expected1, f"Test 1 failed: got {out1}"
   print(" Test 1 passed")
      Test 2: Edge case - single zero
   mat2 = [[0]]
   out2 = sol.updateMatrix(mat2)
   expected2 = [[0]]
   assert out2 == expected2, f"Test 2 failed: got {out2}"
   print(" Test 2 passed")
   # Test 3: Tricky/negative - zero not at corner
   mat3 = [[1,0,1],[1,1,1],[1,1,1]]
   out3 = sol.updateMatrix(mat3)
   expected3 = [[1,0,1],[2,1,2],[3,2,3]]
   assert out3 == expected3, f"Test 3 failed: got {out3}"
   print(" Test 3 passed")
```

Example Walkthrough

```
We'll walk through Test 3: Input: [[1,0,1],[1,1,1],[1,1,1]] - dist initialized as [[-1,0,-1], [-1,-1,-1], [-1,-1,-1]] - Queue q = deque([(0,1)]) \leftarrow only the 0 at (0,1)

Step 1: Pop (0,1) from queue.

Check 4 neighbors: - (1,1): valid, unvisited \rightarrow set dist[1][1] = 0+1 = 1, enqueue (1,1) - (0,0): valid, unvisited \rightarrow dist[0][0] = 1, enqueue (0,0) - (0,2): valid, unvisited \rightarrow dist[0][2] = 1, enqueue (0,2) - (-1,1): invalid \rightarrow skip
```

Now:

dist =
$$[[1,0,1], [-1,1,-1], [-1,-1,-1]]$$

q = $[(1,1), (0,0), (0,2)]$

Step 2: Pop (1,1)

Neighbors: - (2,1): unvisited \rightarrow dist[2][1] = 1+1 = 2, enqueue - (1,0): unvisited \rightarrow dist[1][0] = 2, enqueue - (1,2): unvisited \rightarrow dist[1][2] = 2, enqueue - (0,1): already visited \rightarrow skip

dist =
$$[[1,0,1], [2,1,2], [-1,2,-1]]$$

q = $[(0,0), (0,2), (2,1), (1,0), (1,2)]$

Step 3: Pop (0,0)

Neighbors: (1,0) already set to $2 \to \text{skip}$; (-1,0) invalid; (0,-1) invalid; (0,1) visited. \to No new updates.

Step 4: Pop (0,2)

Neighbors: (1,2) already $2 \to \text{skip}$; others invalid/visited.

Step 5: Pop (2,1)

Neighbors: - (2,0): unvisited \rightarrow dist[2][0] = 2+1 = 3, enqueue - (2,2): unvisited \rightarrow dist[2][2] = 3, enqueue

Now dist = [[1,0,1], [2,1,2], [3,2,3]]

Step 6: Pop remaining (1,0), (1,2), (2,0), (2,2)

 \rightarrow All their neighbors already visited or out of bounds.

Final dist: [[1,0,1],[2,1,2],[3,2,3]]

Key insight: BFS from all zeros at once ensures first time we reach a 1, it's via shortest path.

Complexity Analysis

• Time Complexity: O(m * n)

Each cell is enqueued and dequeued at most once. We visit every cell a constant number of times (once for initialization, once in BFS). Total operations scale linearly with number of cells.

• Space Complexity: O(m * n)

The dist matrix uses O(mn) space. The queue in worst case (e.g., half the grid is 0) may store up to O(mn) cells. Thus total space is O(mn).

4. Rotting Oranges

Pattern: Multi-source BFS (Breadth-First Search)

Problem Statement

You are given an $m \times n$ grid where each cell can have one of three values:

- 0 representing an empty cell,
- 1 representing a fresh orange,
- 2 representing a rotten orange.

Every minute, any fresh orange that is **4-directionally adjacent** to a rotten orange becomes rotten.

Return the **minimum number of minutes** that must elapse until no cell has a fresh orange. If this is impossible, return -1.

```
Input: [[2,1,1],[1,1,0],[0,1,1]]
Output: 4
Explanation: All oranges rot in 4 minutes.
```

```
Input: [[2,1,1],[0,1,1],[1,0,1]]
Output: -1
Explanation: The orange at bottom-left (2,0) never rots.
```

```
Input: [[0,2]]
Output: 0
Explanation: No fresh oranges exist initially.
```

```
from typing import List
from collections import deque
class Solution:
    def orangesRotting(self, grid: List[List[int]]) -> int:
        # STEP 1: Initialize structures
        # Use a queue to store all initially rotten oranges (multi-source BFS)
        # Track fresh count to detect if all oranges rotted
        m, n = len(grid), len(grid[0])
        queue = deque()
        fresh = 0
        # Scan grid to find rotten oranges and count fresh ones
        for i in range(m):
            for j in range(n):
                if grid[i][j] == 2:
                    queue.append((i, j))
                elif grid[i][j] == 1:
                    fresh += 1
        # If no fresh oranges, time = 0
        if fresh == 0:
            return 0
        # STEP 2: Main loop / recursion
        # Process all currently rotten oranges level-by-level(minute-by-minute)
        # Each BFS level = 1 minute
        minutes = 0
        directions = [(1, 0), (-1, 0), (0, 1), (0, -1)]
        while queue and fresh > 0:
            # Process all oranges that rot at current minute
            for _ in range(len(queue)):
                x, y = queue.popleft()
                # Check 4 neighbors
                for dx, dy in directions:
                    nx, ny = x + dx, y + dy
                    # STEP 3: Update state / bookkeeping
                    # - Only rot fresh oranges; reduce fresh count
```

```
if (0 \le nx \le m \text{ and } 0 \le ny \le n \text{ and } 0
                        grid[nx][ny] == 1):
                        grid[nx][ny] = 2
                        fresh -= 1
                        queue.append((nx, ny))
           minutes += 1
       # STEP 4: Return result
       # - If any fresh oranges remain, return -1
       return minutes if fresh == 0 else -1
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   grid1 = [[2,1,1],[1,1,0],[0,1,1]]
   print(sol.orangesRotting(grid1)) # Expected: 4
   # Test 2: Edge case - impossible to rot all
   grid2 = [[2,1,1],[0,1,1],[1,0,1]]
   print(sol.orangesRotting(grid2)) # Expected: -1
   # Test 3: Tricky/negative - no fresh oranges
   grid3 = [[0,2]]
   print(sol.orangesRotting(grid3)) # Expected: 0
```

Example Walkthrough

We'll walk through Test 1: grid = [[2,1,1],[1,1,0],[0,1,1]].

Initial state:

- Grid: [2,1,1] [1,1,0] [0,1,1] - fresh = 6 (six 1s) - Queue starts with (0,0) (only rotten orange)

Minute $0 \rightarrow \text{Minute } 1$:

- Process (0,0): check neighbors $\rightarrow (0,1)$ and (1,0) are fresh.

```
\rightarrow Rot them: set to 2, fresh = 4, add to queue.
- Queue now: [(0,1), (1,0)]
- minutes = 1
Minute 1 \rightarrow Minute 2:
- Process (0,1): neighbors \rightarrow (0,2) and (1,1) are fresh.
\rightarrow Rot them: fresh = 2, queue adds (0,2), (1,1)
- Process (1,0): neighbor (2,0) is 0 (empty); (1,1) already handled.
- Queue now: [(0,2), (1,1)]
-minutes = 2
Minute 2 \rightarrow Minute 3:
- Process (0,2): neighbor (1,2) is 0 \to \text{skip}.
- Process (1,1): neighbor (2,1) is fresh \rightarrow rot it.
\rightarrow fresh = 1, add (2,1) to queue.
- Queue now: [(2,1)]
-minutes = 3
Minute 3 \rightarrow Minute 4:
- Process (2,1): neighbor (2,2) is fresh \rightarrow rot it.
\rightarrow fresh = 0, add (2,2)
- Queue now: [(2,2)]
-minutes = 4
Loop ends (fresh == 0). Return 4.
 Final output: 4
```

Complexity Analysis

• Time Complexity: O(m * n)

We visit each cell at most once during initialization and once during BFS. Total operations scale linearly with grid size.

• Space Complexity: O(m * n)

In worst case (all oranges rotten), the queue holds all cells. Also, we modify grid in-place (no extra space beyond queue).

5. Shortest Path to Get Food

Pattern: BFS (Breadth-First Search)

Problem Statement

You are given an m x n character grid grid representing a kitchen layout.

- 'O' denotes an empty cell you can walk through.
- '#' denotes a food cell (your target).
- 'X' denotes an obstacle (blocked).
- '*' denotes your starting position.

Return the **length of the shortest path** from your starting position to any food cell.

If no path exists, return -1.

You can move up, down, left, or right. Each move counts as 1 step.

```
Input: grid = [["*","#"]]
Output: 1
Explanation: Start adjacent to food - 1 step.
```

```
from collections import deque
from typing import List
class Solution:
    def getFood(self, grid: List[List[str]]) -> int:
        # STEP 1: Initialize structures
        # - Find start position ('*')
        # - Use queue for BFS: stores (row, col, steps)
        # - Track visited cells to avoid cycles
        m, n = len(grid), len(grid[0])
        start = None
        for i in range(m):
            for j in range(n):
                if grid[i][j] == '*':
                    start = (i, j)
                    break
            if start:
                break
        if not start:
            return -1 # Should not happen per problem constraints
        queue = deque([(start[0], start[1], 0)])
        visited = [[False] * n for _ in range(m)]
        visited[start[0]][start[1]] = True
        # Directions: up, down, left, right
        directions = [(-1, 0), (1, 0), (0, -1), (0, 1)]
        # STEP 2: Main loop / recursion
        # - BFS explores level-by-level → guarantees shortest path
            - Stop when we hit a food cell ('#')
        while queue:
            r, c, steps = queue.popleft()
            # STEP 3: Update state / bookkeeping
            # - Check if current cell is food
```

```
if grid[r][c] == '#':
                return steps
           # Explore neighbors
           for dr, dc in directions:
               nr, nc = r + dr, c + dc
                # Validate bounds and accessibility
                if (0 \le nr \le m \text{ and } 0 \le nc \le n \text{ and } 0
                    not visited[nr][nc] and
                    grid[nr][nc] != 'X'):
                    visited[nr][nc] = True
                    queue.append((nr, nc, steps + 1))
       # STEP 4: Return result
       # - If queue empties without finding food → unreachable
       return -1
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   grid1 = [
        ["X","X","X","X","X","X"],
        ["X","*","0","0","0","X"],
        ["X","O","O","#","O","X"],
        ["X","X","X","X","X","X"]
   ]
   assert sol.getFood(grid1) == 3, f"Expected 3, got {sol.getFood(grid1)}"
   # Test 2: Edge case - unreachable food
   grid2 = [
        ["X","X","X","X","X"],
        ["X","*","X","O","X"],
        ["X","O","X","#","X"],
        ["X","X","X","X","X"]
   ]
   assert sol.getFood(grid2) == -1, f"Expected -1, got {sol.getFood(grid2)}"
   # Test 3: Tricky/negative - adjacent food
```

```
grid3 = [["*","#"]]
assert sol.getFood(grid3) == 1, f"Expected 1, got {sol.getFood(grid3)}"
print(" All tests passed!")
```

Example Walkthrough

We'll trace **Test 1** step-by-step:

Initial grid:

```
Row O: X X X X X X
Row 1: X * 0 0 0 X
Row 2: X 0 0 # 0 X
Row 3: X X X X X X
Step 1: Find start \rightarrow at (1,1).
- Initialize queue = [(1,1,0)]
- visited[1][1] = True
Step 2: Dequeue (1,1,0)
- Not food ('*' '#')
- Check 4 neighbors: - (0,1): 'X' \rightarrow skip
- (2,1): 'O' \rightarrow valid \rightarrow mark visited, enqueue (2,1,1)
- (1,0): 'X' \to \text{skip}
- (1,2): 'O' \rightarrow valid \rightarrow enqueue (1,2,1)
- Queue now: [(2,1,1), (1,2,1)]
Step 3: Dequeue (2,1,1)
- Cell = 'O' \rightarrow not food
- Neighbors: - (1,1): visited \rightarrow skip
- (3,1): 'X' \to \text{skip}
- (2,0): 'X' \to \text{skip}
- (2,2): 'O' \rightarrow enqueue (2,2,2)
- Queue: [(1,2,1), (2,2,2)]
Step 4: Dequeue (1,2,1)
- Cell = 'O'
```

- Neighbors: (0,2): 'X' \rightarrow skip
- (2,2): already enqueued (but not visited yet? Actually, it will be marked when enqueued so skip if already visited)
- \rightarrow In our code, we mark when enqueuing, so (2,2) is already visited \rightarrow skip
- (1,1): visited
- (1,3): 'O' \to enqueue (1,3,2)
- Queue: [(2,2,2), (1,3,2)]

Step 5: Dequeue (2,2,2)

- Cell = 'O'
- Neighbors: (1,2): visited
- (3,2): 'X'
- (2,1): visited
- (2,3): '#' \rightarrow FOOD! \rightarrow return steps + 1 = 2 + 1 = 3

Final output: 3

Key insight: BFS guarantees the first time we reach food is via the shortest path.

Complexity Analysis

• Time Complexity: O(m * n)

In worst case, we visit every cell once. Each cell is enqueued and dequeued at most once. Grid has m * n cells.

• Space Complexity: O(m * n)

The visited matrix uses O(m * n). The BFS queue can hold up to O(m * n) cells in worst case (e.g., all cells are open).

6. Word Search

Pattern: Backtracking (DFS on Grid)

Problem Statement

Given an $m \times n$ grid of characters board and a string word, return true if word exists in the grid.

The word can be constructed from letters of sequentially adjacent cells, where adjacent cells are horizontally or vertically neighboring. The same letter cell may not be used more than once.

Sample Input & Output

```
Input: board = [["A","B","C","E"],["S","F","C","S"],["A","D","E","E"]],
word = "ABCCED"
Output: true
Explanation: Path: A → B → C → C → E → D (valid adjacent path without reuse)

Input: board = [["A","B","C","E"],["S","F","C","S"],["A","D","E","E"]],
word = "SEE"
Output: true
Explanation: S → E → E (down then right)

Input: board = [["A","B","C","E"],["S","F","C","S"],["A","D","E","E"]],
word = "ABCB"
Output: false
Explanation: After A→B→C, next B is not reachable without reusing 'B'
```

```
from typing import List

class Solution:
    def exist(self, board: List[List[str]], word: str) -> bool:
        # STEP 1: Initialize structures
        # - Store board dimensions and visited state
```

```
rows, cols = len(board), len(board[0])
   visited = [[False] * cols for _ in range(rows)]
   # STEP 2: Main loop / recursion
   # - Try starting DFS from every cell
   # - If any path matches word, return True
   for r in range(rows):
       for c in range(cols):
            if self._dfs(board, word, 0, r, c, visited):
                return True
   return False
def _dfs(
   self,
   board: List[List[str]],
   word: str,
   idx: int,
   r: int,
   c: int,
   visited: List[List[bool]]
) -> bool:
   # Base case: full word matched
   if idx == len(word):
       return True
   # Boundary & validity checks
   if (
       r < 0 or r >= len(board) or
       c < 0 or c \ge len(board[0]) or
       visited[r][c] or
       board[r][c] != word[idx]
   ):
       return False
   # STEP 3: Update state / bookkeeping
   # - Mark current cell as used
   visited[r][c] = True
   # Explore 4 directions
   found = (
       self._dfs(board, word, idx + 1, r + 1, c, visited) or
        self._dfs(board, word, idx + 1, r - 1, c, visited) or
```

```
self._dfs(board, word, idx + 1, r, c + 1, visited) or
           self._dfs(board, word, idx + 1, r, c - 1, visited)
       )
       # Backtrack: unmark current cell
       visited[r][c] = False
       # STEP 4: Return result
       # - Propagate match status up recursion stack
       return found
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   board1 = [
       ["A", "B", "C", "E"],
       ["S", "F", "C", "S"],
       ["A","D","E","E"]
   assert sol.exist(board1, "ABCCED") == True
   # Test 2: Edge case - single letter
   board2 = [["A"]]
   assert sol.exist(board2, "A") == True
   # Test 3: Tricky/negative - reuse not allowed
   board3 = [
       ["A", "B", "C", "E"],
       ["S", "F", "C", "S"],
       ["A","D","E","E"]
   assert sol.exist(board3, "ABCB") == False
   print(" All tests passed!")
```

Example Walkthrough

We'll trace exist(board, "SEE") on the standard board:

```
Initial State:
```

- board = 3×4 grid as above
- word = "SEE" \rightarrow length = 3
- visited = 3×4 grid of False

Step 1: Outer loops try every cell as start.

- At (0,0): 'A' 'S' \rightarrow skip

- ...

- At (1,0): 'S' == 'S'
$$\rightarrow$$
 call _dfs(..., idx=0, r=1, c=0)

Step 2: Inside _dfs at (1,0), idx=0

- Not out of bounds
- visited[1][0] is False
- board[1][0] == 'S' == word[0] \rightarrow OK
- Mark visited[1][0] = True
- Now try 4 neighbors for next char 'E' (idx=1)

Step 3: Try down \rightarrow (2,0) = 'A' 'E' \rightarrow fail

Try up
$$\rightarrow$$
 (0,0) = 'A' 'E' \rightarrow fail

Try right
$$\rightarrow$$
 (1,1) = 'F' 'E' \rightarrow fail

Try left
$$\rightarrow$$
 invalid (c=-1) \rightarrow fail

$$ightarrow$$
 All fail $ightarrow$ backtrack: set visited[1][0] = False $ightarrow$ return False

Step 4: Continue outer loop...

At (1,3):
$$S' == S' \rightarrow \text{call dfs}(\ldots, \text{idx=0, r=1, c=3})$$

Step 5: In $_{dfs(1,3)}$, mark visited \rightarrow try neighbors:

- Down: (2,3) = 'E' == word[1]
$$\rightarrow$$
 recurse to idx=1

Step 6: In $_{dfs(2,3)}$, mark visited \rightarrow now look for 'E' (idx=2)

- Up: (1,3) \rightarrow visited \rightarrow skip
- Down: invalid
- Left: $(2,2) = E' = \text{word}[2] \rightarrow \text{recurse to idx=2}$

Step 7: In $_{dfs(2,2)}$, mark visited \rightarrow now idx=2, next call with idx=3

- Base case: idx == len(word) (3) \rightarrow return True

Step 8: True propagates up \rightarrow outer function returns True

Final output: True

Key takeaway: Backtracking explores all paths but **undoes choices** (via **visited**[r][c] = False) so other paths can reuse cells.

Complexity Analysis

• Time Complexity: $O(m * n * 4^L)$

For each of the m*n starting cells, we may explore up to 4 directions per character, and the word has length L. Worst-case exponential due to backtracking.

• Space Complexity: O(L)

The recursion stack depth is at most L (length of word). The visited matrix is O(m*n), but since we reuse it and it's input-sized, some consider auxiliary space as O(L). However, strictly: O(m*n) due to visited array. Clarification: LeetCode typically counts extra space beyond input. Since visited is extra, space = O(m*n). But in interviews, clarify assumptions.

7. Number of Connected Components in an Undirected Graph

Pattern: Graph Traversal (DFS/BFS) + Union-Find (Disjoint Set Union)

Problem Statement

You are given an undirected graph with n nodes labeled from 0 to n-1. The graph is represented as an integer n and a list of edges edges, where each edges[i] = [a, b] indicates an undirected edge between nodes a and b.

Return the number of **connected components** in the graph.

```
Input: n = 5, edges = [[0,1],[1,2],[3,4]]
Output: 2
Explanation: Nodes 0-1-2 form one component; nodes 3-4 form another.
```

```
Input: n = 5, edges = []
Output: 5
Explanation: No edges → each node is its own component.

Input: n = 1, edges = []
Output: 1
Explanation: Single node with no edges → one component.
```

```
from typing import List
class Solution:
    def countComponents(self, n: int, edges: List[List[int]]) -> int:
        # STEP 1: Build adjacency list
        # - Why? To enable efficient graph traversal.
        # - Undirected → add both directions.
        graph = [[] for _ in range(n)]
        for a, b in edges:
            graph[a].append(b)
            graph[b].append(a)
        # STEP 2: Track visited nodes
        # - Prevent revisiting and infinite loops.
        visited = [False] * n
        components = 0
        # STEP 3: DFS helper to mark all nodes in a component
        def dfs(node):
            visited[node] = True
            for neighbor in graph[node]:
                if not visited[neighbor]:
                    dfs(neighbor)
        # STEP 4: Iterate through all nodes
        # - Each unvisited node starts a new component.
        for i in range(n):
```

Example Walkthrough

```
3. **Loop over nodes**:
   - `i = O`: not visited → start DFS
     - `dfs(0)`:
       - Mark `visited[0] = True`
       - Visit neighbor `1` → not visited → `dfs(1)`
         - Mark `visited[1] = True`
         - Neighbors: `0` (visited), `2` → `dfs(2)`
           - Mark `visited[2] = True`
           - Neighbor `1` already visited → return
       - Backtrack → DFS ends
     - `components = 1`
   - `i = 1`: already visited → skip
   - `i = 2`: already visited → skip
   - `i = 3`: not visited → start DFS
     - idfs(3)i:
       - Mark `visited[3] = True`
       - Visit `4` → not visited → `dfs(4)`
         - Mark `visited[4] = True`
         - Neighbor `3` visited → return
     - `components = 2`
   - `i = 4`: visited → skip
4. **Return** `2`
Final `visited = [True, True, True, True, True]`
Output: `2`
___
```

Complexity Analysis

• Time Complexity: O(n + e)

We visit each node once (n) and each edge twice (once per direction, but still O(e) total). DFS visits every reachable node/edge exactly once.

• Space Complexity: O(n + e)

Adjacency list uses O(n + e) space. Recursion stack in worst case (e.g., a line graph) uses O(n) space. So total is O(n + e).

8. Graph Valid Tree

Pattern: Graph — Union-Find / DFS Cycle Detection

Problem Statement

You are given n nodes labeled from 0 to n-1 and a list of undirected edges (each edge is a pair of nodes). Write a function to check whether these edges make up a valid tree.

A valid tree must satisfy two conditions:

- 1. There are exactly n 1 edges.
- 2. The graph is **fully connected and acyclic** (i.e., one connected component with no cycles).

```
Input: n = 5, edges = [[0,1],[0,2],[0,3],[1,4]]
Output: True
Explanation: 5 nodes, 4 edges, connected and no cycles → valid tree.
```

```
Input: n = 5, edges = [[0,1],[1,2],[2,3],[1,3],[1,4]]
Output: False
Explanation: Contains a cycle (1-2-3-1), so not a tree.
```

```
Input: n = 1, edges = []
Output: True
Explanation: Single node with no edges is a valid tree.
```

We'll use Union-Find (Disjoint Set Union) — a classic pattern for cycle detection in undirected graphs.

- If we ever try to union two nodes already in the same set \rightarrow cycle detected.
- Also verify edge count = n 1.

```
from typing import List
class Solution:
    def validTree(self, n: int, edges: List[List[int]]) -> bool:
       # STEP 1: Quick edge count check
       # - A tree must have exactly n - 1 edges
       if len(edges) != n - 1:
           return False
       # STEP 2: Initialize Union-Find parent array
       # - Each node starts as its own parent
       parent = list(range(n))
       # Helper: Find root with path compression
       def find(x):
           if parent[x] != x:
               parent[x] = find(parent[x]) # Path compression
           return parent[x]
       # STEP 3: Process each edge
       # - If two nodes share root → cycle → invalid
       for a, b in edges:
           root_a = find(a)
           root_b = find(b)
           if root_a == root_b:
               return False # Cycle detected!
           parent[root_a] = root_b # Union
       # STEP 4: Return True
       # - Passed edge count + no cycles → valid tree
       return True
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
```

```
# Test 1: Normal case
assert sol.validTree(5, [[0,1],[0,2],[0,3],[1,4]]) == True

# Test 2: Edge case - single node
assert sol.validTree(1, []) == True

# Test 3: Tricky/negative - cycle present
assert sol.validTree(5, [[0,1],[1,2],[2,3],[1,3],[1,4]]) == False
print(" All tests passed!")
```

Example Walkthrough

```
Let's trace Test 1: n = 5, edges = [[0,1],[0,2],[0,3],[1,4]]
Initial state:
- parent = [0, 1, 2, 3, 4]
- Edge count = 4 \rightarrow \text{equals 5} - 1 \rightarrow \text{proceed}.
Edge [0,1]:
- find(0) \rightarrow 0, find(1) \rightarrow 1 \rightarrow different roots
- Union: set parent [0] = 1 \rightarrow parent = [1, 1, 2, 3, 4]
Edge [0,2]:
- find(0) \rightarrow find(1) \rightarrow 1; find(2) \rightarrow 2
- Union: parent[1] = 2 \rightarrow parent = [1, 2, 2, 3, 4]
Edge [0,3]:
- find(0) \rightarrow find(1) \rightarrow find(2) \rightarrow 2; find(3) \rightarrow 3
- Union: parent[2] = 3 \rightarrow parent = [1, 2, 3, 3, 4]
Edge [1,4]:
- find(1) \rightarrow find(2) \rightarrow find(3) \rightarrow 3; find(4) \rightarrow 4
- Union: parent[3] = 4 \rightarrow parent = [1, 2, 3, 4, 4]
  No cycles found \rightarrow return True.
Now Test 3: edges = [[0,1],[1,2],[2,3],[1,3],[1,4]]
- First 3 edges connect 0-1-2-3 into one component.
- When processing [1,3]:
```

- find(1) \rightarrow root = 3 (after unions)
- find(3) \rightarrow root = 3
- Same root \rightarrow cycle detected \rightarrow return False immediately.

Complexity Analysis

• Time Complexity: $O(n \cdot (n)) O(n)$

We process n-1 edges. Each find uses path compression, making amortized cost nearly constant (= inverse Ackermann function).

• Space Complexity: O(n)

The parent array stores one entry per node. No recursion stack (iterative union).

9. Course Schedule

Pattern: Topological Sort (Graph — Directed Acyclic Graph / Cycle Detection)

Problem Statement

There are a total of numCourses courses you have to take, labeled from 0 to numCourses - 1. You are given an array prerequisites where prerequisites[i] = [a_i, b_i] indicates that you must take course b_i first if you want to take course a_i.

Return true if you can finish all courses. Otherwise, return false.

Sample Input & Output

```
Input: numCourses = 2, prerequisites = [[1,0]]
Output: true
Explanation: To take course 1, you must first take course 0. This is valid.

Input: numCourses = 2, prerequisites = [[1,0],[0,1]]
Output: false
Explanation: Course 0 requires course 1,
and course 1 requires course 0 → cycle.

Input: numCourses = 1, prerequisites = []
Output: true
Explanation: Only one course with no prerequisites → always finishable.
```

```
from typing import List
class Solution:
    def canFinish(self, numCourses: int, prerequisites: List[List[int]]):
        # STEP 1: Build adjacency list and in-degree array
        # - adj[i] = list of courses that depend on course i
        # - indegree[i] = number of prerequisites for course i
        adj = [[] for _ in range(numCourses)]
        indegree = [0] * numCourses
        for course, prereq in prerequisites:
            adj[prereq].append(course)
            indegree[course] += 1
        # STEP 2: Initialize queue with all courses having no prerequisites
        # - These can be taken immediately (indegree == 0)
        from collections import deque
        queue = deque()
        for i in range(numCourses):
```

```
if indegree[i] == 0:
               queue.append(i)
       # STEP 3: Process courses in topological order
       # - For each course taken, reduce indegree of its dependents
       # - If a dependent's indegree becomes 0, add to queue
       taken = 0
       while queue:
           curr = queue.popleft()
           taken += 1
           for neighbor in adj[curr]:
               indegree[neighbor] -= 1
               if indegree[neighbor] == 0:
                   queue.append(neighbor)
       # STEP 4: Return result
       # - If we took all courses, no cycle exists → return True
       return taken == numCourses
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case - linear dependency
   assert sol.canFinish(2, [[1,0]]) == True
   # Test 2: Edge case - single course, no prereqs
   assert sol.canFinish(1, []) == True
   # Test 3: Tricky/negative - cycle in graph
   assert sol.canFinish(2, [[1,0],[0,1]]) == False
   print(" All tests passed!")
```

Example Walkthrough

We'll walk through Test 3: numCourses = 2, prerequisites = [[1,0],[0,1]].

1. Initialize structures

- adj = [[], []] \rightarrow two empty lists for courses 0 and 1
- indegree = [0, 0]

2. Build graph from prerequisites

- Process [1,0]: course 1 depends on 0
 → adj[0].append(1) → adj = [[1], []]
 → indegree[1] += 1 → indegree = [0, 1]
- Process [0,1]: course 0 depends on 1
 → adj[1].append(0) → adj = [[1], [0]]
 → indegree[0] += 1 → indegree = [1, 1]

3. Initialize queue

- Check indegree [0] = $1 \rightarrow \text{skip}$
- Check indegree[1] = $1 \rightarrow \text{skip}$ $\rightarrow \text{queue} = \text{deque}() \text{ (empty)}$

4. Process queue

• Queue is empty \rightarrow while loop never runs \rightarrow taken = 0

5. Return result

• taken == numCourses \rightarrow 0 == 2 \rightarrow False

Final output: False — correctly detects cycle.

Complexity Analysis

• Time Complexity: O(V + E)

We visit each course (V = numCourses) once and each prerequisite edge (E = len(prerequisites)) once during graph building and BFS traversal.

• Space Complexity: O(V + E)

The adjacency list stores E edges, and indegree + queue use O(V) space.

10. Course Schedule II

Pattern: Topological Sort (Graph – DAG Detection + Ordering)

Problem Statement

There are a total of numCourses courses you have to take, labeled from 0 to numCourses - 1. You are given an array prerequisites where prerequisites[i] = [ai, bi] indicates that you must take course bi first if you want to take course ai.

Return the ordering of courses you should take to finish all courses. If there are many valid answers, return **any** of them. If it is impossible to finish all courses, return an **empty array**.

Sample Input & Output

```
Input: numCourses = 2, prerequisites = [[1,0]]
Output: [0,1]
Explanation: To take course 1, you must first take course 0.
```

```
Input: numCourses = 4, prerequisites = [[1,0],[2,0],[3,1],[3,2]]
Output: [0,1,2,3] or [0,2,1,3]
Explanation: Courses 1 and 2 depend on 0; course 3 depends on both 1 and 2.
```

```
Input: numCourses = 2, prerequisites = [[1,0],[0,1]]
Output: []
Explanation: Circular dependency → impossible to finish.
```

```
from typing import List
from collections import deque, defaultdict
class Solution:
    def findOrder(self, numCourses: int, prerequisites: List[List[int]]):
        # STEP 1: Initialize structures
        # - Build adjacency list (graph) and in-degree array
        graph = defaultdict(list)
        in_degree = [0] * numCourses
        for course, prereq in prerequisites:
            graph[prereq].append(course)
            in_degree[course] += 1
        # STEP 2: Main loop / recursion
        # - Use Kahn's algorithm: start with zero in-degree nodes
        queue = deque()
        for i in range(numCourses):
            if in_degree[i] == 0:
                queue.append(i)
        topo_order = []
        # STEP 3: Update state / bookkeeping
        # - Process nodes level by level; reduce in-degree of neighbors
        while queue:
            current = queue.popleft()
           topo_order.append(current)
           for neighbor in graph[current]:
                in_degree[neighbor] -= 1
                if in_degree[neighbor] == 0:
                   queue.append(neighbor)
        # STEP 4: Return result
        # - If topo_order doesn't include all courses → cycle exists
        return topo_order if len(topo_order) == numCourses else []
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
```

```
# Test 1: Normal case
result1 = sol.findOrder(2, [[1,0]])
print("Test 1:", result1) # Expected: [0, 1]

# Test 2: Edge case - no prerequisites
result2 = sol.findOrder(3, [])
print("Test 2:", result2) # Expected: [0, 1, 2] (any order)

# Test 3: Tricky/negative - cycle
result3 = sol.findOrder(2, [[1,0],[0,1]])
print("Test 3:", result3) # Expected: []
```

Example Walkthrough

We'll walk through Test 1: numCourses = 2, prerequisites = [[1,0]].

- 1. Initialize graph and in_degree:
 - graph starts empty (defaultdict of lists).
 - in_degree = [0, 0] (for courses 0 and 1).
- 2. Process prerequisites:
 - For [1,0]: course 1 depends on 0.
 - Add 1 to graph [0] \rightarrow graph = {0: [1]}
 - Increment in_degree[1] → in_degree = [0, 1]
- 3. Find zero in-degree nodes:
 - Course 0: $in_{degree}[0] == 0 \rightarrow add$ to queue.
 - Course 1: $in_{degree}[1] == 1 \rightarrow skip$.
 - queue = deque([0])
- 4. Begin BFS (Kahn's algorithm):
 - Pop 0 from queue → add to topo_order = [0]
 - Look at neighbors of 0: [1]
 - Reduce in_degree[1] from 1 to 0
 - Now in_degree[1] == $0 \rightarrow \text{add 1}$ to queue

- 5. Next iteration:
 - Pop 1 \rightarrow topo_order = [0, 1]
 - graph[1] has no neighbors \rightarrow nothing to update
- 6. Queue empty \rightarrow exit loop.
 - len(topo_order) = 2 == numCourses \rightarrow return [0,1]

Final Output: [0, 1] — correct topological order.

Key takeaway: **Topological sort only works on DAGs**. If a cycle exists, some nodes never reach in-degree 0, so the result list will be shorter than numCourses.

Complexity Analysis

• Time Complexity: O(V + E)

We visit each course (V = numCourses) once and each prerequisite edge (E = len(prerequisites)) once during graph building and BFS traversal.

• Space Complexity: O(V + E)

The adjacency list (graph) stores E edges. The in_degree array and queue use O(V) space. Total: O(V + E).

11. Alien Dictionary

 ${\bf Pattern:}\ {\bf Topological\ Sort\ (Graph\,+\,DFS/Kahn's\ Algorithm)}$

Problem Statement

There is a new alien language that uses the English alphabet. However, the order among letters is unknown to you.

You are given a list of strings words from the alien dictionary, where the strings are sorted lexicographically by the rules of this new language.

Return a string of the unique letters in the new alien language sorted in **lexico-graphically increasing order** by the new rules. If there is no solution, return "". If there are multiple solutions, return **any** of them.

A string **s** is lexicographically smaller than a string **t** if at the first letter where they differ, the letter in **s** comes before the letter in **t** in the alien language.

Constraints:

```
- 1 <= words.length <= 100
- 1 <= words[i].length <= 100
```

- words[i] consists of only lowercase English letters.

Sample Input & Output

```
Input: words = ["wrt", "wrf", "er", "ett", "rftt"]
Output: "wertf"
Explanation: From word pairs, we infer: t < f, w < e, r < t, e < r
→ valid topological order exists.</pre>
```

```
Input: words = ["z", "x", "z"]

Output: ""

Explanation: Contradiction: z < x and x < z \rightarrow cycle \rightarrow invalid.
```

```
Input: words = ["abc", "ab"]
Output: ""
Explanation: Prefix "ab" comes after "abc" - invalid lexicographic order.
```

```
from typing import List
from collections import defaultdict, deque
class Solution:
    def alienOrder(self, words: List[str]) -> str:
        # STEP 1: Initialize structures
            - Build graph: char -> set of chars that come after it
           - Track in-degree for each unique char
        graph = defaultdict(set)
        in_degree = {}
        # Add all unique chars with in-degree 0 initially
        for word in words:
            for char in word:
                in_degree[char] = 0
        # STEP 2: Main loop / recursion
        # - Compare adjacent words to infer ordering
        for i in range(len(words) - 1):
            word1, word2 = words[i], words[i + 1]
            # Edge case: prefix violation (e.g., ["abc", "ab"])
            if len(word1) > len(word2) and word1.startswith(word2):
                return ""
            # Find first differing character
            min_len = min(len(word1), len(word2))
            for j in range(min_len):
                c1, c2 = word1[j], word2[j]
                if c1 != c2:
                    # c1 must come before c2
                    if c2 not in graph[c1]:
                        graph[c1].add(c2)
                        in_degree[c2] += 1
                    break # Only first difference matters
        # STEP 3: Update state / bookkeeping
        # - Use Kahn's algorithm (BFS) for topological sort
        queue = deque([char for char in in_degree if in_degree[char] == 0])
        result = []
        while queue:
```

```
char = queue.popleft()
           result.append(char)
           for neighbor in graph[char]:
               in_degree[neighbor] -= 1
               if in_degree[neighbor] == 0:
                   queue.append(neighbor)
       # STEP 4: Return result
       # - If result doesn't include all chars → cycle exists
       if len(result) != len(in_degree):
           return ""
       return "".join(result)
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   assert sol.alienOrder(["wrt", "wrf", "er", "ett", "rftt"]) == "wertf"
   # Test 2: Edge case - cycle
   assert sol.alienOrder(["z", "x", "z"]) == ""
   # Test 3: Tricky/negative - invalid prefix order
   assert sol.alienOrder(["abc", "ab"]) == ""
   print(" All tests passed!")
```

Example Walkthrough

We'll walk through Test 1: ["wrt", "wrf", "er", "ett", "rftt"].

- 1. Initialize in_degree with all unique chars: Chars: w, r, t, f, $e \rightarrow in_degree = \{'w':0, 'r':0, 't':0, 'f':0, 'e':0\}$
- 2. Compare adjacent words:

- "wrt" vs "wrf" → first diff at index 2: t vs f → add edge t → f
 → graph['t'] = {'f'}, in_degree['f'] = 1
- "wrf" vs "er" \rightarrow first diff: w vs e \rightarrow edge w \rightarrow e
 - \rightarrow graph['w'] = {'e'}, in_degree['e'] = 1
- "er" vs "ett" \rightarrow first diff: r vs t \rightarrow edge r \rightarrow t
 - \rightarrow graph['r'] = {'t'}, in_degree['t'] = 1
- "ett" vs "rftt" → first diff: e vs r → edge e → r
 → graph['e'] = {'r'}, in_degree['r'] = 1
- 3. Final in_degree:

```
{'w':0, 'r':1, 't':1, 'f':1, 'e':1}
```

- 4. Kahn's BFS:
 - Start with queue = ['w'] (only char with in-degree 0)
 - Pop 'w' → add to result → process neighbors: 'e' → decrement in_degree['e'] to 0 → enqueue 'e'
 - Pop 'e' \rightarrow result = ['w','e'] \rightarrow process 'r' \rightarrow in_degree['r'] = 0 \rightarrow enqueue 'r'
 - Pop 'r' \rightarrow result = ['w','e','r'] \rightarrow process 't' \rightarrow in_degree['t'] = 0 \rightarrow enqueue 't'
 - Pop 't' \rightarrow result = ['w','e','r','t'] \rightarrow process 'f' \rightarrow in_degree['f'] = 0 \rightarrow enqueue 'f'
 - Pop 'f' \rightarrow result = ['w','e','r','t','f']
- 5. Result length = 5 = total chars → return "wertf"

Final Output: "wertf"

Key Insight: Lexicographic order gives pairwise constraints \rightarrow model as DAG \rightarrow topological sort.

Complexity Analysis

• Time Complexity: O(C)

Where C = total number of characters in all words.

We scan each character once to build the graph and once during BFS.

Each edge is processed once. Number of edges number of unique char pairs $26 \times 26 = O(1)$, but dominated by input size C.

• Space Complexity: O(1) (or O(U + E))

```
U = \text{number of unique characters (26)}, E = \text{edges (26^2)}. So technically O(1) since alphabet is fixed, but more precisely O(U + E).
```

12. Clone Graph

Pattern: Graph Traversal (BFS/DFS) + Hash Map (Node Mapping)

Problem Statement

Given a reference of a node in a **connected undirected graph**, return a **deep copy (clone)** of the graph.

Each node in the graph contains a value (int) and a list (List[Node]) of its neighbors.

```
class Node:
    def __init__(self, val = 0, neighbors = None):
        self.val = val
        self.neighbors = neighbors if neighbors is not None else []
```

Note: The graph is represented using an adjacency list.

You must return the copy of the given node as a reference to the cloned graph.

Sample Input & Output

```
Input: adjList = [[2,4],[1,3],[2,4],[1,3]]
Output: [[2,4],[1,3],[2,4],[1,3]]
Explanation: There are 4 nodes. Node 1 connects to 2 and 4, etc.
The cloned graph has same structure but new node objects.
```

```
Input: adjList = [[]]
Output: [[]]
Explanation: Single node with no neighbors.
```

```
Input: adjList = [[2],[1]]
Output: [[2],[1]]
Explanation: Two nodes connected bidirectionally.
```

```
from typing import Optional
# Definition for a Node (as provided by LeetCode)
class Node:
    def __init__(self, val = 0, neighbors = None):
        self.val = val
        self.neighbors = neighbors if neighbors is not None else []
class Solution:
    def cloneGraph(self, node: Optional['Node']) -> Optional['Node']:
        # STEP 1: Initialize structures
        # - Use a hash map to track original -> cloned node mapping
            - Prevents cycles and duplicate cloning
        if not node:
            return None
        visited = {}
        # STEP 2: Main loop / recursion
            - DFS via helper function
            - Clone current node, then recursively clone neighbors
        def dfs(n: 'Node') -> 'Node':
            if n in visited:
                return visited[n] # Return existing clone
            # Create clone of current node (without neighbors yet)
            clone = Node(n.val)
            visited[n] = clone # Record mapping immediately
            # STEP 3: Update state / bookkeeping
            # - Recursively clone all neighbors and link them
            for neighbor in n.neighbors:
```

```
clone.neighbors.append(dfs(neighbor))
           return clone
       # STEP 4: Return result
       # - Start DFS from input node
       return dfs(node)
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case - 4-node cycle
   n1 = Node(1)
   n2 = Node(2)
   n3 = Node(3)
   n4 = Node(4)
   n1.neighbors = [n2, n4]
   n2.neighbors = [n1, n3]
   n3.neighbors = [n2, n4]
   n4.neighbors = [n1, n3]
   cloned = sol.cloneGraph(n1)
   # Verify structure by values
   assert cloned.val == 1
   assert set(n.val for n in cloned.neighbors) == {2, 4}
   assert set(n.val for n in cloned.neighbors[0].neighbors) == {1, 3}
   print(" Test 1 passed: 4-node graph cloned correctly.")
   # Test 2: Edge case - single node with no neighbors
   single = Node(1)
   cloned_single = sol.cloneGraph(single)
   assert cloned_single.val == 1
   assert len(cloned_single.neighbors) == 0
   print(" Test 2 passed: Single node cloned correctly.")
   # Test 3: Tricky case - two nodes connected bidirectionally
   a = Node(1)
   b = Node(2)
   a.neighbors = [b]
   b.neighbors = [a]
   cloned_ab = sol.cloneGraph(a)
   assert cloned_ab.val == 1
```

```
assert len(cloned_ab.neighbors) == 1
assert cloned_ab.neighbors[0].val == 2
assert cloned_ab.neighbors[0].neighbors[0] is cloned_ab # cycle preserved
print(" Test 3 passed: 2-node cycle cloned correctly.")
```

Example Walkthrough

We'll walk through Test 3: two nodes a(1) and b(2) connected to each other.

- 1. Call cloneGraph(a)
 - node = a (not None), so proceed.
 - visited = {} (empty dict).
- 2. Enter dfs(a)
 - a not in visited.
 - Create clone_a = Node(1).
 - visited = {a: clone_a}.
- 3. Loop over a.neighbors \rightarrow [b]
 - Call dfs(b).
- 4. Enter dfs(b)
 - b not in visited.
 - Create clone_b = Node(2).
 - visited = {a: clone_a, b: clone_b}.
- 5. Loop over b.neighbors \rightarrow [a]
 - Call dfs(a) again.
- 6. Re-enter dfs(a)
 - Now a is in visited \rightarrow return clone_a.

7. Back in dfs(b):

- Append clone_a to clone_b.neighbors.
- Return clone_b.

8. Back in dfs(a):

- Append clone_b to clone_a.neighbors.
- Return clone_a.

9. Final state:

- clone_a.val = 1, clone_a.neighbors = [clone_b]
- clone_b.val = 2, clone_b.neighbors = [clone_a]
- The cycle is perfectly replicated with new objects.

Key insight: The **visited** map ensures we never clone the same node twice and correctly rebuilds cycles.

Complexity Analysis

• Time Complexity: O(N + M)

Where $\mathbb{N}=$ number of nodes, $\mathbb{M}=$ number of edges. Each node is visited once, and each edge is traversed once during neighbor iteration.

• Space Complexity: O(N)

The visited hash map stores one entry per node. Recursion stack depth is O(N) in worst case (e.g., linear chain).

13. Accounts Merge

Pattern: Union-Find (Disjoint Set Union) + Hashing

Problem Statement

Given a list of accounts where each element accounts[i] is a list of strings, where the first element accounts[i][0] is a name, and the rest of the elements are emails representing emails of the account.

Now, we would like to merge these accounts. Two accounts definitely belong to the same person if there is some common email to both accounts. Note that even if two accounts have the same name, they may belong to different people as people could have the same name. A person can have any number of accounts initially, but all of their accounts definitely have the same name.

After merging the accounts, return the accounts in the following format: the first element of each account is the name, and the rest of the elements are emails in **sorted order**. The accounts themselves can be returned in any order.

Sample Input & Output

```
Input: accounts = [
    ["John", "johnsmith@mail.com", "john_newyork@mail.com"],
    ["John", "johnsmith@mail.com", "john00@mail.com"],
    ["Mary", "mary@mail.com"],
    ["John", "johnnybravo@mail.com"]
]
Output: [
    ["John", "john00@mail.com", "john_newyork@mail.com", "johnsmith@mail.com"],
    ["John", "johnnybravo@mail.com"],
    ["Mary", "mary@mail.com"]
]
Explanation: The first two accounts share "johnsmith@mail.com",
so they are merged.
```

```
from typing import List
from collections import defaultdict
class Solution:
    def accountsMerge(self, accounts: List[List[str]]) -> List[List[str]]:
        # STEP 1: Initialize Union-Find structures
        # - email_to_id: map each email to a unique integer ID
            - email to name: map email to its owner's name
        # - parent: list for DSU (indexed by email ID)
        email to id = {}
        email_to_name = {}
        id_counter = 0
        # Assign unique IDs to all emails and record names
        for account in accounts:
            name = account[0]
            for email in account[1:]:
                if email not in email_to_id:
                    email_to_id[email] = id_counter
                    email_to_name[email] = name
                    id_counter += 1
        # Initialize parent array for DSU
        parent = list(range(id_counter))
        # Helper: find root with path compression
        def find(x: int) -> int:
            if parent[x] != x:
                parent[x] = find(parent[x])
```

```
return parent[x]
       # Helper: union two email IDs
       def union(x: int, y: int):
           rx, ry = find(x), find(y)
           if rx != ry:
               parent[ry] = rx
       # STEP 2: Union emails within the same account
           - All emails in one account belong to same person
           - Union first email with every other in the list
       for account in accounts:
           first_email = account[1]
           first_id = email_to_id[first_email]
           for email in account[2:]:
               union(first_id, email_to_id[email])
       # STEP 3: Group emails by root parent
       # - Use root ID as key to collect all connected emails
       root_to_emails = defaultdict(list)
       for email in email_to_id:
           root = find(email_to_id[email])
           root_to_emails[root].append(email)
       # STEP 4: Build final result
       # - For each group: get name from any email, sort emails
       result = []
       for emails in root_to_emails.values():
           name = email_to_name[emails[0]]
           result.append([name] + sorted(emails))
       return result
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case - two accounts share an email
   accounts1 = [
        ["John", "johnsmith@mail.com", "john_newyork@mail.com"],
        ["John", "johnsmith@mail.com", "john00@mail.com"],
        ["Mary", "mary@mail.com"],
```

```
["John", "johnnybravo@mail.com"]
1
expected1 = [
  ["John", "john00@mail.com", "john_newyork@mail.com", "johnsmith@mail.com"],
  ["John", "johnnybravo@mail.com"],
  ["Mary", "mary@mail.com"]
output1 = sol.accountsMerge(accounts1)
assert sorted(
  [sorted(x) for x in output1]) == sorted([sorted(x) for x in expected1])
print(" Test 1 passed")
# Test 2: Edge case - no shared emails
accounts2 = [
    ["Gabe", "Gabe0@m.co", "Gabe3@m.co", "Gabe1@m.co"],
    ["Kevin", "Kevin3@m.co", "Kevin5@m.co", "Kevin0@m.co"]
output2 = sol.accountsMerge(accounts2)
# Should return same structure (emails sorted per account)
expected2 = [
    ["Gabe", "Gabe0@m.co", "Gabe1@m.co", "Gabe3@m.co"],
    ["Kevin", "Kevin0@m.co", "Kevin3@m.co", "Kevin5@m.co"]
1
assert sorted(
    [sorted(x) for x in output2]) == sorted([sorted(x) for x in expected2])
print(" Test 2 passed")
# Test 3: Tricky case - chain of shared emails
accounts3 = [
    ["David", "David0@m.co", "David1@m.co"],
    ["David", "David3@m.co", "David4@m.co"],
    ["David", "David4@m.co", "David5@m.co"],
    ["David", "David2@m.co", "David3@m.co"],
    ["David", "David1@m.co", "David2@m.co"]
output3 = sol.accountsMerge(accounts3)
expected3 = [["David","David0@m.co","David1@m.co","David2@m.co",
              "David3@m.co", "David4@m.co", "David5@m.co"]]
assert len(output3) == 1
assert sorted(output3[0]) == sorted(expected3[0])
print(" Test 3 passed")
```

Example Walkthrough

We'll trace **Test 1** step-by-step:

1. Email ID Assignment

- Process each account:
 - Account 0: "John" \to emails: johnsmith@mail.com \to ID 0, john_newyork@mail.com \to ID 1
 - Account 1: "John" \to johnsmith@mail.com (already ID 0), john00@mail.com \to ID 2
 - Account 2: "Mary" \rightarrow mary@mail.com \rightarrow ID 3
 - Account 3: "John" o johnnybravo@mail.com o ID 4
- email_to_id = {johnsmith:0, john_newyork:1, john00:2, mary:3, johnnybravo:4}
- parent = [0,1,2,3,4] (each email its own root)

2. Union Within Accounts

- Account 0: union(0,1) \rightarrow parent[1] = 0 \rightarrow parent = [0,0,2,3,4]
- Account 1: union(0,2) \rightarrow find(0)=0, find(2)=2 \rightarrow parent[2]=0 \rightarrow parent = [0,0,0,3,4]
- Account 2: only one email \rightarrow no union
- Account 3: only one email \rightarrow no union

3. Path Compression on Find

- When we later call find(1), it becomes 0 and updates parent[1]=0 (already done)
- find(2) \rightarrow parent[2]=0, so returns 0 and sets parent[2]=0

4. Group by Root

- For each email:
 - johnsmith (ID 0): root = $0 \rightarrow \text{group}[0] += [\text{johnsmith}]$
 - john_newyork (ID 1): $find(1)=0 \rightarrow group[0] += [john_newyork]$
 - john00 (ID 2): $find(2)=0 \rightarrow group[0] += [john00]$
 - mary (ID 3): root= $3 \rightarrow \text{group}[3] = [\text{mary}]$
 - johnnybravo (ID 4): $root=4 \rightarrow group[4] = [johnnybravo]$

5. Build Result

- Group 0: name = "John", emails sorted \rightarrow ["John", "john00...", "john_newyork...", "johnsmith..."]
- Group 3: ["Mary", "mary@mail.com"]
- Group 4: ["John", "johnnybravo@mail.com"]

Final output matches expected.

Complexity Analysis

- Time Complexity: O(N * (N))
 - > Where N is total number of emails. Each find/union is nearly O(1) due to path compression and union by rank (implicit here). Sorting emails per group adds O(K log K) per group, but total over all groups is O(N log N) in worst case. However, since DSU dominates the merging logic and sorting is unavoidable for output, overall is often cited as O(N log N) due to sorting. But core union-find is O(N (N)), and is inverse Ackermann (~constant).
- Space Complexity: O(N)
 - > We store email_to_id, email_to_name, parent, and root_to_emails all proportional to number of unique emails N.

14. Word Ladder

Pattern: BFS (Breadth-First Search) on Implicit Graph

Problem Statement

A transformation sequence from word beginWord to word endWord using a dictionary wordList is a sequence of words such that:

- Every adjacent pair of words differs by exactly one letter.
- Every word in the sequence is in wordList. Note that beginWord does not need to be in wordList.
- endWord must be in wordList.

Given two words, beginWord and endWord, and a dictionary wordList, return the number of words in the shortest transformation sequence from beginWord to endWord, or 0 if no such sequence exists.

Sample Input & Output

```
Input: beginWord = "hit", endWord = "cog",
wordList = ["hot","dot","dog","lot","log","cog"]
Output: 5
Explanation: One shortest transformation is "hit" →
"hot" → "dot" → "dog" → "cog" (5 words).
```

```
Input: beginWord = "hit", endWord = "cog",
wordList = ["hot","dot","dog","lot","log"]
Output: 0
Explanation: "cog" is not in wordList, so no valid path exists.
```

```
Input: beginWord = "a", endWord = "c", wordList = ["a","b","c"]
Output: 2
Explanation: "a" → "c" (differs by one letter), so sequence length = 2.
```

```
from collections import deque
from typing import List
class Solution:
    def ladderLength(
        self, beginWord: str, endWord: str, wordList: List[str]
    ) -> int:
        # STEP 1: Initialize structures
        # - Convert wordList to a set for O(1) lookups
        # - Use a queue for BFS: stores (current_word, level)
        word_set = set(wordList)
        if endWord not in word_set:
            return 0 # Early exit if endWord missing
        queue = deque([(beginWord, 1)])
        visited = {beginWord}
        # STEP 2: Main loop / recursion
        # - BFS explores all words at current level before moving deeper
          - For each word, try changing every char to 'a'-'z'
        while queue:
            word, level = queue.popleft()
            # STEP 3: Update state / bookkeeping
            # - If we reach endWord, return level (number of words)
            if word == endWord:
                return level
            # Generate all possible one-letter mutations
            for i in range(len(word)):
                for c in 'abcdefghijklmnopqrstuvwxyz':
                    next_word = word[:i] + c + word[i+1:]
                    if next_word in word_set and next_word not in visited:
                        visited.add(next_word)
                        queue.append((next_word, level + 1))
        # STEP 4: Return result
        # - If BFS completes without finding endWord, return 0
        return 0
```

Example Walkthrough

```
Let's trace Test 1:
beginWord = "hit", endWord = "cog",
wordList = ["hot","dot","dog","lot","log","cog"]

Initial State:
- word_set = {"hot","dot","dog","lot","log","cog"}
- queue = deque([("hit", 1)])
- visited = {"hit"}
```

Step 1: Dequeue ("hit", 1)
- Not "cog" → continue
- Try all 1-letter changes of "hit":

```
- Change index 0: "ait", "bit", ..., "hot" \rightarrow only "hot" in word_set
```

```
- Add "hot" to visited, enqueue ("hot", 2)
```

State:

```
- queue = [("hot", 2)]
```

Step 2: Dequeue ("hot", 2)

- Not "cog"
- Mutations:
- Index 1: "hat", "hbt", ..., "dot", "lot" \rightarrow both in set
- Enqueue ("dot", 3) and ("lot", 3), mark visited

State:

Step 3: Dequeue ("dot", 3)

- Mutations:
- Index 2: "doa", ..., "dog" \rightarrow "dog" valid
- Enqueue ("dog", 4)

State:

- visited adds "dog"

Step 4: Dequeue ("lot", 3)

- Mutations \rightarrow "log" valid
- Enqueue ("log", 4)

State:

Step 5: Dequeue ("dog", 4)

- Mutations \rightarrow "cog" found!
- Enqueue ("cog", 5)

Step 6: Dequeue ("log", 4)

- Also finds "cog", but already visited \rightarrow skip

Step 7: Dequeue ("cog", 5)

- Matches endWord \rightarrow return 5

Final Output: 5

Key Takeaway: BFS guarantees shortest path in unweighted graphs. Each level = one more word in sequence.

Complexity Analysis

- Time Complexity: O(M² × N)
 - N = number of words in wordList
 - -M = length of each word
 - For each word (up to N), we generate M positions \times 26 letters = O(26M) O(M)
 - But slicing word[:i] + c + word[i+1:] takes O(M), so total per word: O(M²)
 - Overall: $O(N \times M^2)$
- Space Complexity: O(N × M)
 - word_set stores N words of length M \rightarrow O(NM)
 - visited and queue may store up to N words \rightarrow O(NM)
 - Total: O(NM)

15. Minimum Knight Moves

 ${\bf Pattern} \colon {\rm BFS} \mbox{ (Breadth-First Search)}$ on Implicit Graph

Problem Statement

In an infinite chessboard, a knight starts at [0, 0] and wants to reach a target position [x, y].

Return the **minimum number of moves** required for the knight to reach the target.

A knight moves in an L-shape: 2 squares in one direction and 1 square perpendicular, or vice versa.

Note: Due to symmetry, you can assume x = 0, y = 0. The answer is the same for all quadrants.

Sample Input & Output

```
Input: x = 2, y = 1
Output: 1
Explanation: Knight moves from (0,0) \rightarrow (2,1) in one move.
```

```
Input: x = 5, y = 5
Output: 4
Explanation: One optimal path: (0,0) \rightarrow (2,1) \rightarrow (3,3) \rightarrow (4,5) \rightarrow (5,5)
```

```
Input: x = 1, y = 1
Output: 2
Explanation: Cannot reach (1,1) in 1 move. Minimum: (0,0) \rightarrow (2,-1) \rightarrow (1,1)
```

```
from collections import deque
from typing import Tuple
class Solution:
    def minKnightMoves(self, x: int, y: int) -> int:
        # Normalize to first quadrant using symmetry
        x, y = abs(x), abs(y)
        # STEP 1: Initialize BFS structures
            - Queue holds (current_x, current_y, move_count)
        # - Visited set prevents reprocessing same cell
        queue = deque([(0, 0, 0)])
        visited = \{(0, 0)\}
        # All 8 possible knight moves
        directions = [
            (2, 1), (2, -1), (-2, 1), (-2, -1),
            (1, 2), (1, -2), (-1, 2), (-1, -2)
        ]
        # STEP 2: Main BFS loop
            - Process level by level (guarantees minimum moves)
            - Stop when target is reached
        while queue:
            curr_x, curr_y, moves = queue.popleft()
            # STEP 3: Check if target reached
            if curr_x == x and curr_y == y:
                return moves
            # STEP 4: Explore all valid next positions
            for dx, dy in directions:
                nx, ny = curr_x + dx, curr_y + dy
                # Optimization: only explore near target
                # Since knight can't go too far negatively,
                # we bound search to x+2, y+2 (empirically safe)
                if (nx, ny) not in visited and nx >= -2 and ny >= -2:
                    visited.add((nx, ny))
                    queue.append((nx, ny, moves + 1))
        # Should never reach here for valid input
```

```
return -1

# ------- INLINE TESTS ------

if __name__ == "__main__":
    sol = Solution()

# Test 1: Normal case
    result1 = sol.minKnightMoves(2, 1)
    print(f"Test 1 - Input: (2,1) → Output: {result1}") # Expected: 1

# Test 2: Edge case (diagonal close)
    result2 = sol.minKnightMoves(1, 1)
    print(f"Test 2 - Input: (1,1) → Output: {result2}") # Expected: 2

# Test 3: Tricky/negative (symmetry test)
    result3 = sol.minKnightMoves(-5, -5)
    print(f"Test 3 - Input: (-5,-5) → Output: {result3}") # Expected: 4
```

Example Walkthrough

We'll walk through minKnightMoves(1, 1) step by step.

Initial state:

```
- x = 1, y = 1 \rightarrow normalized to (1, 1)

- queue = deque([(0, 0, 0)])

- visited = {(0, 0)}
```

Step 1: Dequeue (0, 0, 0)

- Not target \rightarrow explore neighbors.
- Generate 8 moves:

```
(2,1), (2,-1), (-2,1), (-2,-1),
(1,2), (1,-2), (-1,2), (-1,-2)
```

- Filter by nx >= -2 and ny >= -2 \rightarrow all pass.
- Add all 8 to visited and queue with moves = 1.

State after Step 1:

- queue has 8 entries, e.g., (2,1,1), (2,-1,1), ..., (-1,-2,1)
- visited has 9 points.

Step 2: Process each of the 8 positions (BFS level 1).

None equal (1,1). For each, generate next moves.

Many are duplicates \rightarrow skipped by visited.

But from (2,-1), one move is (2-1, -1+2) = (1,1) \to target!

Step 3: When processing (2, -1, 1):

- Generate $(1, 1) \rightarrow \text{not visited} \rightarrow \text{add to queue as } (1,1,2).$

Step 4: Later, dequeue $(1,1,2) \rightarrow \text{matches target} \rightarrow \text{return 2}$.

Final output: 2

Complexity Analysis

• Time Complexity: $O(\max(|x|, |y|)^2)$

BFS explores a grid roughly proportional to the distance from origin to target. The pruning $(nx \ge -2, ny \ge -2)$ bounds the search space to a constant factor around the target.

• Space Complexity: O(max(|x|, |y|)²)

The visited set and queue store positions in the explored region, which scales quadratically with distance.

16. Bus Routes

Pattern: Graph BFS (Shortest Path in Unweighted Graph)

Problem Statement

You are given an array routes representing bus routes where routes[i] is a bus route that the ith bus repeats forever.

For example, if routes [0] = [1, 5, 7], this means the 0th bus travels in a circular route: $1 \rightarrow 5 \rightarrow 7 \rightarrow 1 \rightarrow 5 \rightarrow 7 \rightarrow \dots$

You start at bus stop source and want to go to bus stop target. Return the least number of buses you must take to travel from source to target. Return -1 if it is not possible.

Constraints:

```
- 1 <= routes.length <= 500

- 1 <= routes[i].length <= 10

- All values in routes[i] are unique.

- sum(routes[i].length) <= 10

- 0 <= source, target < 10
```

Sample Input & Output

```
from collections import defaultdict, deque
from typing import List
class Solution:
    def numBusesToDestination(
        self, routes: List[List[int]], source: int, target: int
    ) -> int:
        # STEP 1: Initialize structures
        # - stop_to_buses: map each stop to list of bus indices
        # - queue: BFS queue of (stop, num_buses)
        # - visited_stops: avoid revisiting stops
          - visited_buses: avoid re-boarding same bus
        if source == target:
            return 0
        stop_to_buses = defaultdict(list)
        for bus_id, stops in enumerate(routes):
            for stop in stops:
                stop_to_buses[stop].append(bus_id)
        queue = deque([(source, 0)])
        visited_stops = set([source])
        visited_buses = set()
        # STEP 2: Main loop / recursion
            - BFS over stops; when we board a bus, we mark all its
              stops as reachable with +1 bus count
        while queue:
            current_stop, num_buses = queue.popleft()
            # STEP 3: Update state / bookkeeping
            # - For each bus that serves current_stop,
                  if not already taken, add all its stops to queue
            for bus id in stop to buses[current stop]:
                if bus_id in visited_buses:
                    continue
                visited_buses.add(bus_id)
                for next_stop in routes[bus_id]:
                    if next_stop == target:
                        return num_buses + 1
```

```
if next_stop not in visited_stops:
                       visited_stops.add(next_stop)
                       queue.append((next_stop, num_buses + 1))
       # STEP 4: Return result
       # - If BFS ends without finding target, impossible
       return -1
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   assert sol.numBusesToDestination(
       [[1,2,7],[3,6,7]], 1, 6
   ) == 2, "Test 1 Failed"
   # Test 2: Edge case - already at target
   assert sol.numBusesToDestination(
       [[1,2,7],[3,6,7]], 7, 7
   ) == 0, "Test 2 Failed"
   # Test 3: Tricky/negative - unreachable
   assert sol.numBusesToDestination(
       [[7,12],[4,5,15],[6],[15,19],[9,12,13]], 15, 12
   ) == -1, "Test 3 Failed"
   print(" All tests passed!")
```

Example Walkthrough

```
We'll walk through Test 1:

routes = [[1,2,7],[3,6,7]], source = 1, target = 6.

Initial Setup: - stop_to_buses: - 1 \rightarrow [0] - 2 \rightarrow [0] - 7 \rightarrow [0,1] - 3 \rightarrow [1] - 6 \rightarrow [1] - queue = deque([(1, 0)]) - visited_stops = {1} - visited_buses = set()
```

```
Step 1: Dequeue (1, 0)
- Current stop = 1, buses taken = 0
- Buses at stop 1: [0]
- Bus 0 not visited → mark visited_buses = {0}
- Explore all stops on bus 0: [1, 2, 7]
- 1: already visited \rightarrow skip
- 2: not visited \rightarrow add (2, 1) to queue, visited_stops = {1,2}
- 7: not visited \rightarrow add (7, 1) to queue, visited_stops = {1,2,7}
Queue now: [(2,1), (7,1)]
Step 2: Dequeue (2, 1)
- Buses at stop 2: [0] \rightarrow \text{already visited} \rightarrow \text{skip}
- No new stops added.
Queue now: [(7,1)]
Step 3: Dequeue (7, 1)
- Buses at stop 7: [0,1]
- Bus 0: already visited \rightarrow skip
- Bus 1: not visited \rightarrow add to visited_buses = {0,1}
- Explore stops on bus 1: [3,6,7]
- 3: not visited \rightarrow add (3,2), visited_stops = {1,2,7,3}
- 6: this is target! \rightarrow return 1 + 1 = 2
```

Key Insight:

Final Output: 2

We don't BFS over individual stops one-by-one. Instead, when we reach any stop on a bus route, we "take" that entire bus, and all its stops become reachable in one additional step. This avoids $O(N^2)$ edge explosion.

Complexity Analysis

• Time Complexity: O(S), where S = sum(len(route) for route in routes)

We visit each stop at most once, and for each stop, we iterate over all buses that serve it. But each bus is processed only once (via visited_buses), and processing a bus touches all its stops. Total work is proportional to total number of stop-bus memberships, which is S 10.

• Space Complexity: O(S)

stop_to_buses stores S entries. visited_stops and queue store at most
O(unique stops) S. visited_buses stores at most O(number of buses)
500. Dominated by O(S).

17. Cheapest Flights Within K Stops

Pattern: Graph — Shortest Path with Limited Steps (BFS / Modified Dijkstra / DP)

Problem Statement

There are n cities connected by some number of flights. You are given an array flights where flights[i] = [from_i, to_i, price_i] indicates that there is a flight from city from_i to city to_i with cost price_i.

You are also given three integers src, dst, and k, and you need to find the cheapest price from src to dst with at most k stops. If there is no such route, return -1.

Sample Input & Output

```
Input: n = 4, flights = [[0,1,100],[1,2,100],[2,0,100],[1,3,600],[2,3,200]], src = 0, dst = 3, k = 1
Output: 700
Explanation: The path 0 \rightarrow 1 \rightarrow 3 uses 1 stop ( k) and costs 100 + 600 = 700.
```

```
Input: n = 3, flights = [[0,1,100],[1,2,100],[0,2,500]], src = 0, dst = 2, k = 0

Output: 500

Explanation: Direct flight 0 \rightarrow 2 is the only option with 0 stops.
```

```
Input: n = 3, flights = [[0,1,100]], src = 0, dst = 2, k = 1
Output: -1
Explanation: No route from 0 to 2 within 1 stop.
```

```
from typing import List
from collections import deque
class Solution:
    def findCheapestPrice(
        self, n: int, flights: List[List[int]], src: int, dst: int, k: int
    ) -> int:
        # STEP 1: Build adjacency list for graph
        # - Each entry: {city: [(neighbor, price), ...]}
        graph = [[] for _ in range(n)]
        for u, v, w in flights:
            graph[u].append((v, w))
        # STEP 2: BFS with stops tracking
        # - Queue stores (cost_so_far, city, stops_used)
        # - We allow up to k stops → max k+1 edges
        queue = deque()
        queue.append((0, src, 0)) # (cost, city, stops)
        # Track min cost to reach each city (pruning)
        min_cost = [float('inf')] * n
        min_cost[src] = 0
        # STEP 3: Process queue level by level
        while queue:
            cost, city, stops = queue.popleft()
```

```
# Skip if we've exceeded stop limit
           if stops > k:
               continue
           # Explore neighbors
           for neighbor, price in graph[city]:
               new cost = cost + price
               # Only proceed if we found a cheaper way
               # to reach 'neighbor'
               if new_cost < min_cost[neighbor]:</pre>
                   min_cost[neighbor] = new_cost
                   queue.append((new_cost, neighbor, stops + 1))
       # STEP 4: Return result or -1 if unreachable
       return min_cost[dst] if min_cost[dst] != float('inf') else -1
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   n1 = 4
   flights1 = [[0,1,100],[1,2,100],[2,0,100],[1,3,600],[2,3,200]]
   src1, dst1, k1 = 0, 3, 1
   print(f"Test 1: {sol.findCheapestPrice(n1, flights1, src1, dst1, k1)}")
   # Expected: 700
   # Test 2: Edge case - direct flight only
   flights2 = [[0,1,100],[1,2,100],[0,2,500]]
   src2, dst2, k2 = 0, 2, 0
   print(f"Test 2: {sol.findCheapestPrice(n2, flights2, src2, dst2, k2)}")
   # Expected: 500
   # Test 3: Tricky/negative - unreachable
   n3 = 3
   flights3 = [[0,1,100]]
   src3, dst3, k3 = 0, 2, 1
   print(f"Test 3: {sol.findCheapestPrice(n3, flights3, src3, dst3, k3)}")
   # Expected: -1
```

Example Walkthrough

```
We'll walk through Test 1:

n = 4, flights = [[0,1,100],[1,2,100],[2,0,100],[1,3,600],[2,3,200]],

src = 0, dst = 3, k = 1
```

Goal: Find cheapest price from city 0 to city 3 with 1 stop.

Step 0: Build Graph

We create an adjacency list:

```
-graph[0] = [(1, 100)]
```

$$-graph[1] = [(2, 100), (3, 600)]$$

$$-graph[2] = [(0, 100), (3, 200)]$$

-graph[3] = []

 $min_cost = [0, \omega, \omega, \omega]$ (only src=0 has cost 0)

Queue starts as: $deque([(0, 0, 0)]) \rightarrow (cost=0, city=0, stops=0)$

Step 1: Pop (0, 0, 0)

- stops = 0 $k=1 \rightarrow OK$
- From city 0, go to city 1 with price 100

$$- \text{ new_cost} = 0 + 100 = 100$$

- 100 < $\omega \rightarrow \mathrm{update}\, \mathtt{min_cost[1]}$ = 100
- Push (100, 1, 1) into queue

```
Queue: [(100, 1, 1)]
min_cost = [0, 100, \omega, \omega]
```

Step 2: Pop (100, 1, 1)

- stops = 1 $k=1 \rightarrow OK$
- From city 1:
 - To city 2: new_cost = 100 + 100 = 200 \rightarrow update min_cost[2] = 200, push (200, 2, 2)
 - To city 3: new_cost = 100 + 600 = 700 \rightarrow update min_cost[3] = 700, push (700, 3, 2)

Queue: [(200, 2, 2), (700, 3, 2)] min_cost = [0, 100, 200, 700]

Step 3: Pop (200, 2, 2)

• stops = 2 > k=1 \rightarrow skip (too many stops)

Step 4: Pop (700, 3, 2)

• stops = 2 > k=1 \rightarrow skip

Queue empty \rightarrow done.

Final min_cost[3] = $700 \rightarrow \text{return } 700$

Matches expected output!

Complexity Analysis

- Time Complexity: O((k + 1) * E)
 - > In worst case, we explore each edge up to k+1 times (once per allowed stop level).
 - > E = len(flights) \rightarrow each edge may be relaxed once per stop layer.
- Space Complexity: O(n + E)
 - > O(E) for adjacency list, O(n) for min_cost array and queue (queue holds at most O(n) per level, but total bounded by graph size).

18. Pacific Atlantic Water Flow

Pattern: Graph Traversal (Multi-source BFS/DFS)

Problem Statement

There is an $m \times n$ rectangular island that borders both the **Pacific Ocean** (top and left edges) and **Atlantic Ocean** (bottom and right edges).

The island is partitioned into a grid of square cells. You are given an $m \times n$ integer matrix heights where heights[r][c] represents the height above sea level of the cell at coordinate (r, c).

The island receives a lot of rain, and the rainwater can flow to neighboring cells directly north, south, east, or west if the neighboring cell's height is less than or equal to the current cell's height.

Water can flow from any cell adjacent to an ocean into the ocean.

Return a **2D** list of grid coordinates result where result[i] = [r, c] denotes that rainwater can flow from cell (r, c) to both the Pacific and Atlantic oceans.

Sample Input & Output

```
Input: heights = [[1]]
Output: [[0,0]]
Explanation: Single cell touches both oceans
(top-left = Pacific, bottom-right = Atlantic).
```

```
from typing import List
class Solution:
    def pacificAtlantic(self, heights: List[List[int]]) -> List[List[int]]:
        # STEP 1: Initialize structures
            - Use two sets to track cells reachable from each ocean.
            - Start DFS from ocean edges (multi-source).
        if not heights or not heights[0]:
            return []
        m, n = len(heights), len(heights[0])
        pacific_reachable = set()
        atlantic_reachable = set()
        # STEP 2: Main loop / recursion
        # - Perform DFS from all Pacific-border cells (top row, left col).
        # - Perform DFS from all Atlantic-border cells (bottom row, right col).
        # - DFS moves to neighbors with >= height (reverse flow logic).
        def dfs(r: int, c: int, reachable: set):
            reachable.add((r, c))
            for dr, dc in [(1, 0), (-1, 0), (0, 1), (0, -1)]:
                nr, nc = r + dr, c + dc
                if (0 \le nr \le m \text{ and } 0 \le nc \le n \text{ and } 0
                     (nr, nc) not in reachable and
                    heights[nr][nc] >= heights[r][c]):
                    dfs(nr, nc, reachable)
        # Pacific: top row and left column
        for i in range(m):
```

```
dfs(i, 0, pacific_reachable)
       for j in range(n):
           dfs(0, j, pacific_reachable)
       # Atlantic: bottom row and right column
       for i in range(m):
           dfs(i, n - 1, atlantic_reachable)
       for j in range(n):
           dfs(m - 1, j, atlantic_reachable)
       # STEP 3: Update state / bookkeeping
       # - Intersection of both sets = cells reaching both oceans.
       # - Convert to list of lists for output.
       both = pacific_reachable & atlantic_reachable
       # STEP 4: Return result
       # - Order doesn't matter per problem statement.
       return [[r, c] for r, c in both]
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   heights1 = [
       [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]
   expected1 = [[0,4],[1,3],[1,4],[2,2],[3,0],[3,1],[4,0]]
   result1 = sol.pacificAtlantic(heights1)
   assert sorted(result1) == sorted(expected1), f"Test 1 failed: {result1}"
   # Test 2: Edge case - single cell
   heights2 = [[1]]
   expected2 = [[0,0]]
   result2 = sol.pacificAtlantic(heights2)
   assert result2 == expected2, f"Test 2 failed: {result2}"
   # Test 3: Tricky/negative - spiral with high center
```

```
heights3 = [
        [1,2,3],
        [8,9,4],
        [7,6,5]
]
expected3 = [[0,2],[1,1],[1,2],[2,0],[2,1],[2,2]]
result3 = sol.pacificAtlantic(heights3)
assert sorted(result3) == sorted(expected3), f"Test 3 failed: {result3}"
print(" All tests passed!")
```

Example Walkthrough

We'll trace **Test 1** (heights1) step by step:

- 1. Initialization:
 - m = 5, n = 5
 - pacific_reachable = set(), atlantic_reachable = set()
- 2. Pacific DFS starts:
 - From left column: (0,0), (1,0), (2,0), (3,0), (4,0)
 - From top row: (0,1), (0,2), (0,3), (0,4)
 - DFS from (0,0) (height=1): can go to (1,0) (3 1) \rightarrow then (2,0) (2 3? No! Wait—reverse logic: we allow **higher or equal** neighbors because we're flowing from ocean inward. So from (1,0)=3, we can go to (2,0)=2? No—2 < 3 \rightarrow not allowed. But from (3,0)=6, we can go to (2,0)=2? No. However, from (3,0)=6, we can go to (3,1)=7 (7 6) \rightarrow yes! So (3,1) gets added to Pacific set.
- 3. Atlantic DFS starts:
 - From right column: (0,4), (1,4), (2,4), (3,4), (4,4)
 - From bottom row: (4,0), (4,1), (4,2), (4,3)

• DFS from $(4,4)=4 \to \text{can go to } (4,3)=2$? No (2 < 4). But $(3,4)=5 \to 4 \to \text{yes}$. Then from (3,4)=5, go to (3,3)=4 $(4 \to 3)=4$ allowed in reverse), etc.

4. After both DFS runs:

- pacific_reachable includes (3,0), (3,1), (4,0), (0,4), etc.
- atlantic_reachable includes (0,4), (1,4), (2,2), (3,0), etc.

5. Intersection:

- Common cells: (0,4), (1,3), (1,4), (2,2), (3,0), (3,1), (4,0)
- Converted to list of lists \rightarrow matches expected output.

Key Insight: Instead of checking every cell (expensive), we **reverse the flow**—start from oceans and climb up (to higher/equal ground). This ensures we only traverse reachable regions.

Complexity Analysis

• Time Complexity: O(m × n)

Each cell is visited at most once by Pacific DFS and once by Atlantic DFS. Total work is linear in number of cells.

• Space Complexity: O(m × n)

In worst case (flat terrain), both pacific_reachable and atlantic_reachable store all m×n cells. Recursion stack depth also up to O(m×n) in worst case (though typically much less).

19. Longest Increasing Path in a Matrix

Pattern: DFS with Memoization (Graph Traversal + Dynamic Programming)

Problem Statement

Given an m x n integers matrix, return the length of the longest increasing path in the matrix.

From each cell, you can move in **four directions**: left, right, up, or down.

You **may not** move diagonally or move outside the boundary (i.e., wrap-around is not allowed).

A path is *increasing* if each subsequent value is **strictly greater** than the previous.

Sample Input & Output

```
Input: matrix = [[9,9,4],[6,6,8],[2,1,1]]
Output: 4
Explanation: The longest increasing path is [1 → 2 → 6 → 9].

Input: matrix = [[3,4,5],[3,2,6],[2,2,1]]
Output: 4
Explanation: One path is [3 → 4 → 5 → 6].

Input: matrix = [[1]]
Output: 1
Explanation: Single cell → path length = 1.
```

```
from typing import List

class Solution:
    def longestIncreasingPath(self, matrix: List[List[int]]) -> int:
        if not matrix or not matrix[0]:
            return 0

    rows, cols = len(matrix), len(matrix[0])
```

```
memo = [[0] * cols for _ in range(rows)]
       directions = [(0, 1), (1, 0), (0, -1), (-1, 0)]
       def dfs(r: int, c: int) -> int:
           # STEP 1: Return cached result if already computed
            if memo[r][c] != 0:
                return memo[r][c]
           max_path = 1  # At least the cell itself
           # STEP 2: Explore all 4 directions
           for dr, dc in directions:
               nr, nc = r + dr, c + dc
                # STEP 3: Only move to strictly greater neighbor
                if (0 \le nr \le rows \text{ and } 0 \le nc \le rols \text{ and } 0
                    matrix[nr][nc] > matrix[r][c]):
                    # Recursively get path length from neighbor
                    path_from_neighbor = dfs(nr, nc)
                    max_path = max(max_path, 1 + path_from_neighbor)
           # STEP 4: Cache and return result
           memo[r][c] = max_path
           return max_path
       result = 0
       for r in range(rows):
           for c in range(cols):
               result = max(result, dfs(r, c))
       return result
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   assert sol.longestIncreasingPath([[9,9,4],[6,6,8],[2,1,1]]) == 4
   # Test 2: Edge case - single cell
   assert sol.longestIncreasingPath([[1]]) == 1
```

```
# Test 3: Tricky/negative - all equal values
assert sol.longestIncreasingPath([[7,7,7],[7,7,7],[7,7,7]]) == 1
```

Example Walkthrough

We'll trace matrix = [[9,9,4],[6,6,8],[2,1,1]].

- 1. Initialization:
 - rows = 3, cols = 3
 - memo is a 3×3 grid of zeros.
 - Directions: right, down, left, up.
- 2. Outer loop starts at $(0,0) \rightarrow \text{value} = 9$.
 - All neighbors (9,6) $9 \rightarrow$ no valid moves.
 - dfs(0,0) returns $1 \rightarrow \text{memo}[0][0] = 1$.
- 3. Continue to $(0,1) \rightarrow also 9 \rightarrow same \rightarrow memo[0][1] = 1$.
- 4. At $(0,2) \to \text{value} = 4$.
 - Down to $(1,2) = 8 (>4) \rightarrow \text{call dfs(1,2)}.$
 - From (1,2)=8: down to (2,2)=1 (no), up=4 (no), left=(1,1)=6 ($<8 \rightarrow$ no), right=invalid.
 - So dfs(1,2) returns $1 \rightarrow \text{memo[1][2]} = 1$.
 - So dfs(0,2) = 1 + 1 = 2 \rightarrow memo[0][2] = 2.
- 5. At $(1,0) \to \text{value} = 6$.
 - Down to $(2,0)=2 \ (<6 \to no)$.
 - Up to (0,0)=9 (>6) \rightarrow call dfs(0,0) \rightarrow returns 1 (cached).
 - So path: $6 \rightarrow 9 \rightarrow \text{length} = 2$.
 - But also check left/right: (1,1)=6 (not >), so max = 2.
- 6. At $(2,0) \to \text{value} = 2$.
 - Up to (1,0)=6 (>2) \rightarrow call dfs(1,0) \rightarrow returns 2 (from above).
 - So path: $2 \to 6 \to 9 \to \text{length} = 3$.

- Also right to (2,1)=1 ($<2 \rightarrow no$).
- So memo[2][0] = 3.
- 7. At $(2,1) \to \text{value} = 1$.
 - Up to $(1,1)=6 (>1) \rightarrow \text{call dfs(1,1)}$.
 - From (1,1)=6: up= $(0,1)=9 \rightarrow dfs(0,1)=1 \rightarrow path = 2$.
 - Left=(1,0)=6 (not >), right= $(1,2)=8 \text{ (>6)} \rightarrow dfs(1,2)=1 \rightarrow path = 2$.
 - So dfs(1,1) = 2.
 - So from (2,1): $1 \rightarrow 6 \rightarrow 9 \rightarrow \text{length} = 3$.
 - Also left to $(2,0)=2 (>1) \to dfs(2,0)=3 \to path = 1 + 3 = 4!$
 - So memo[2][1] = 4.
- 8. Final result = max(..., 4) = 4.

Final Output: 4

Key Insight: Without memoization, we'd recompute paths like $6\rightarrow 9$ many times. Memoization turns exponential time into polynomial.

Complexity Analysis

• Time Complexity: O(m * n)

Each cell is visited **once** due to memoization. For each cell, we check 4 neighbors \rightarrow constant work per cell. Total = m * n * O(1) = O(mn).

• Space Complexity: O(m * n)

The memo table uses O(mn) space.

Recursion depth is at most O(mn) in worst-case (e.g., strictly increasing spiral), but in practice limited by path length. Still, worst-case stack space is O(mn). Thus, total space = O(mn).

20. Word Search II

Pattern: Trie + Backtracking (DFS)

Problem Statement

Given an $m \times n$ board of characters and a list of strings words, return all words on the board.

Each word must be constructed from letters of sequentially adjacent cells, where adjacent cells are horizontally or vertically neighboring. The same letter cell may not be used more than once in a word.

Sample Input & Output

```
Input: board = [["a","b"],["c","d"]], words = ["abcb"]
Output: []
Explanation: "abcb" requires reusing 'b', which is not allowed.
```

```
Input: board = [["a"]], words = ["a","aa"]
Output: ["a"]
Explanation: Only single-letter word "a" is possible.
```

```
from typing import List

class TrieNode:
    def __init__(self):
```

```
self.children = {}
       self.word = None # Stores full word at end node
class Solution:
   def findWords(self, board: List[List[str]], words: List[str]) -> List[str]:
       # STEP 1: Build Trie from words
           - Why Trie? Avoid recomputing shared prefixes;
              enables early pruning during DFS.
       root = TrieNode()
       for word in words:
           node = root
           for char in word:
                if char not in node.children:
                    node.children[char] = TrieNode()
               node = node.children[char]
           node.word = word # Mark end of word
       result = []
       rows, cols = len(board), len(board[0])
       # STEP 2: DFS from every cell
       # - Invariant: current path forms a prefix in Trie
       # - Signal: reached a node where node.word is set
       def dfs(r, c, node):
           char = board[r][c]
            if char not in node.children:
               return
           next_node = node.children[char]
            if next_node.word:
                result.append(next_node.word)
               next_node.word = None # Avoid duplicates
           # STEP 3: Mark visited & recurse
            # - Why mark? Prevent reuse in same path
           # - What breaks? Infinite loops or invalid paths
           board[r][c] = "#" # Temporary mark
            for dr, dc in [(0,1), (1,0), (0,-1), (-1,0)]:
               nr, nc = r + dr, c + dc
                if 0 \le nr \le rows and 0 \le nc \le cols:
                    dfs(nr, nc, next_node)
            board[r][c] = char # Backtrack
```

```
# Optional optimization: prune leaf nodes
           if not next_node.children:
               del node.children[char]
       # STEP 4: Launch DFS from every cell
       # - Handle edge: empty board or words
       for r in range(rows):
           for c in range(cols):
               dfs(r, c, root)
       return result
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   board1 = [["o", "a", "a", "n"],
             ["e","t","a","e"],
             ["i", "h", "k", "r"],
              ["i", "f", "l", "v"]]
   words1 = ["oath","pea","eat","rain"]
   print("Test 1:", sorted(sol.findWords(board1, words1)) == ["eat","oath"])
   # Test 2: Edge case - no matches
   board2 = [["a","b"],["c","d"]]
   words2 = ["abcb"]
   print("Test 2:", sol.findWords(board2, words2) == [])
   # Test 3: Tricky - single cell, duplicate prevention
   board3 = [["a"]]
   words3 = ["a", "aa"]
   print("Test 3:", sorted(sol.findWords(board3, words3)) == ["a"])
```

Example Walkthrough

We'll trace Test 1 with board1 and words1 = ["oath", "pea", "eat", "rain"].

1. Build Trie:

- Insert "oath": root \rightarrow 'o' \rightarrow 'a' \rightarrow 't' \rightarrow 'h' (word="oath")
- Insert "pea": root \rightarrow 'p' \rightarrow 'e' \rightarrow 'a' (word="pea")
- Insert "eat": root \rightarrow 'e' \rightarrow 'a' \rightarrow 't' (word="eat")
- Insert "rain": root \rightarrow 'r' \rightarrow 'a' \rightarrow 'i' \rightarrow 'n' (word="rain")

2. Start DFS at (0,0) = 'o':

- 'o' in root.children \rightarrow go to node 'o'
- From 'o', check neighbors: (0,1)='a' \rightarrow valid
- Path: "oa" \rightarrow continue to 't' at (1,1), then 'h' at (1,2)
- At 'h', node.word = "oath" \rightarrow add to result, set word=None

3. Later, start DFS at (1,0) = e:

- 'e' in root.children \rightarrow go to 'e'
- (1,1)='t' \rightarrow but need 'a' next! Wait—(0,0) is 'o', (1,0)='e'
- Actually: $(1,0)=\text{'e'} \to (0,0)=\text{'o'} \text{ (not 'a')}, (1,1)=\text{'t'}, (2,0)=\text{'i'}$
- But (0,1)='a' is neighbor of (1,0)? No—(1,0) neighbors: up=(0,0)='o', right=(1,1)='t', down=(2,0)='i'
- So how do we get "eat"?

Correction: "eat" starts at (1,2)='a'? No.

Actually:

- (1,0) = 'e'
- $(0,0) = \text{`o'} \rightarrow \text{skip}$
- $(1,1) = 't' \to \text{not 'a'}$
- $(2,0) = 'i' \to \text{not 'a'}$
- Wait—look again:

Row 0: o a a n

Row 1: e t a e \leftarrow (1,2) = 'a'

So "eat" = (1,0)='e' $\to (1,1)$ ='t'? No, that's "et"

Correct path for "eat":

- Start at (1,2) = 'a'? No.
- Actually: $(1,0)=\text{'e'} \to (0,0)=\text{'o'} \text{ (no)}, (1,1)=\text{'t'} \text{ (no)}, (2,0)=\text{'i'} \text{ (no)}$
- But (0,1)='a' is **not adjacent to** (1,0) it's diagonal! Not allowed.

Ah! Real path:

- Start at $(1,2) = 'a' \rightarrow no$, need 'e' first.
- Look at $(1,3) = \text{`e'} \to \text{up to } (0,3) = \text{`n'}, \text{ left to } (1,2) = \text{`a'}, \text{ down to } (2,3) = \text{`r'}$
- From (1,3)='e' \rightarrow left to (1,2)='a' \rightarrow up to (0,2)='a' \rightarrow no 't'

Correct path:

- (1,0) = 'e'
- $(0,0) = \text{`o'} \rightarrow \text{skip}$
- $(1,1) = \text{`t'} \rightarrow \text{skip}$
- $(2,0) = 'i' \rightarrow \text{skip}$
- Wait! (1,0)='e' \rightarrow right is (1,1)='t', but we need 'a' next for "ea"

Mistake in board reading:

Board row 1: ["e","t","a","e"] So:

- (1,0) = 'e'
- (1,1) = 't'
- (1,2) = 'a'
- (1,3) = 'e'

So "eat" = (1,0)='e' \rightarrow (0,0) is 'o', but (1,0) up is (0,0), right is (1,1)='t' — no 'a' adjacent?

Actually:

• (0,1) = 'a' is above (1,1), not (1,0).

• But (1,0) has no 'a' neighbor.

Then how is "eat" found?

Answer: Start at (1,3) ='e'

- (1,3) = 'e'
- Left \to (1,2) = 'a'
- Up \rightarrow (0,2) = 'a' \rightarrow not 't'
- Down \to (2,3) = 'r'
- Left again? No.

Correct path:

- Start at (1,0) = 'e'
- \mathbf{Up} : (0,0) = 'o'
- **Right**: (1,1) = 't'
- **Down**: (2,0) = 'i'
- Left: invalid

Still no 'a'.

Re-express board with coordinates:

$$(0,0)=o$$
 $(0,1)=a$ $(0,2)=a$ $(0,3)=n$

$$(1,0)=e$$
 $(1,1)=t$ $(1,2)=a$ $(1,3)=e$

$$(2,0)=i$$
 $(2,1)=h$ $(2,2)=k$ $(2,3)=r$

$$(3,0)=i$$
 $(3,1)=f$ $(3,2)=l$ $(3,3)=v$

"eat" path:

- Start at (1,3) = 'e'
- Move **left** to (1,2) = 'a'
- Move up to $(0,2) = 'a' \rightarrow \text{not 't'}$
- Move **down** to $(2,2) = k' \rightarrow no$

Alternative:

- Start at (1,0) = 'e'
- Is there an 'a' adjacent? No.

Wait! Look at (0,1) = 'a' — who is its neighbor?

- Down = (1,1) = 't'
- Left = (0,0) = 'o'
- Right = (0,2) = 'a'
- Up = invalid

So to get "eat", we need: $(e' \rightarrow (a' \rightarrow (t')))$

Where is an 'e' next to an 'a' that is next to a 't'?

Found it:

- (1,3) = 'e'
- (1,2) = 'a' (left of 'e')
- (1,1) = 't' (left of 'a') \rightarrow Path: $(1,3) \rightarrow (1,2) \rightarrow (1,1) =$ "e"-"a"-"t" = "eat" (but backwards!)

But word is "eat", so must start with 'e'.

So: start at (1,3)='e' $\rightarrow (1,2)=$ 'a' $\rightarrow (1,1)=$ 't' \rightarrow forms "eat"

DFS will find this when starting at (1,3).

4. **DFS** from (1,3):

- char = 'e' \rightarrow in root.children (from "eat" and "pea"? No, "pea" starts with 'p')
- 'e' is in root (from "eat") \rightarrow proceed
- next node = node after 'e'
- Check neighbors of (1,3): (0,3)='n', (1,2)='a', (2,3)='r'
- (1,2)='a' is in next_node.children \rightarrow recurse
- Now at (1,2), node after "ea"
- Neighbors: $(1,1)='t' \rightarrow \text{in children} \rightarrow \text{recurse}$
- At (1,1), node after "eat" \rightarrow word="eat" \rightarrow add to result

5. Backtracking:

• After exploring (1,1), mark restored to 't'

- Then (1,2) restored to 'a', then (1,3) to 'e'
- 6. **Result**: ["oath", "eat"] (order may vary)

Complexity Analysis

- Time Complexity: O(M × N × 4^L)
 - > Where M×N is board size, L is max word length.
 - > In worst case, DFS explores 4 directions up to L depth per cell.
 - > But Trie pruning reduces this significantly paths not in Trie are cut early.
- Space Complexity: O(K × L)
 - > K = number of words, L = average word length.
 - > Space used by Trie.
 - $> {\rm DFS}$ recursion stack: ${\tt O(L)}$ not dominant.

21. Minimum Height Trees

Pattern: Graphs — Topological Sorting (Leaf Pruning)

Problem Statement

A tree is an undirected graph in which any two vertices are connected by exactly one path. In other words, any connected graph without simple cycles is a tree.

Given a tree of n nodes labeled from 0 to n - 1, and an array edges where edges[i] = [ai, bi] indicates that there is an undirected edge between nodes ai and bi in the tree.

You are asked to find all the minimum height trees (MHTs) and return a list of their root labels. You can return the answer in any order.

The height of a rooted tree is the number of edges on the longest downward path between the root and a leaf.

Sample Input & Output

```
from typing import List
from collections import defaultdict, deque
class Solution:
    def findMinHeightTrees(self, n: int, edges: List[List[int]]) -> List[int]:
        # STEP 1: Handle trivial case
        # - Single node has no edges; it's the only MHT root.
        if n == 1:
           return [0]
        # Build adjacency list and degree count
        graph = defaultdict(set)
        degree = [0] * n
        for u, v in edges:
            graph[u].add(v)
            graph[v].add(u)
            degree[u] += 1
            degree[v] += 1
```

```
# STEP 2: Initialize queue with all leaves (degree == 1)
       # - Leaves cannot be optimal roots (they maximize height).
           - We iteratively prune leaves layer by layer.
       leaves = deque()
       for i in range(n):
           if degree[i] == 1:
               leaves.append(i)
       # STEP 3: Topological pruning (BFS from leaves inward)
          - Remove current leaves, reduce neighbor degrees.
           - New leaves emerge; repeat until 2 nodes remain.
           - Why 2? In a tree, the "center" is either 1 or 2 nodes.
       remaining_nodes = n
       while remaining_nodes > 2:
           leaves_count = len(leaves)
           remaining_nodes -= leaves_count
           for _ in range(leaves_count):
               leaf = leaves.popleft()
               for neighbor in graph[leaf]:
                   degree[neighbor] -= 1
                   graph[neighbor].discard(leaf)
                   if degree[neighbor] == 1:
                       leaves.append(neighbor)
       # STEP 4: Remaining nodes in queue are MHT roots
       return list(leaves)
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   assert sol.findMinHeightTrees(4, [[1,0],[1,2],[1,3]]) == [1]
   # Test 2: Edge case - single node
   assert sol.findMinHeightTrees(1, []) == [0]
   # Test 3: Tricky case - two centers
   result = sol.findMinHeightTrees(6, [[3,0],[3,1],[3,2],[3,4],[5,4]])
   assert sorted(result) == [3, 4] # order doesn't matter
```

```
print(" All tests passed!")
```

Example Walkthrough

Let's trace Test 3: n = 6, edges = [[3,0],[3,1],[3,2],[3,4],[5,4]].

Initial Graph (adjacency list):

- 0: {3}
- 1: {3}
- 2: {3}
- 3: {0,1,2,4}
- 4: {3,5}
- 5: {4}

Degrees: $[1, 1, 1, 4, 2, 1] \rightarrow \text{Leaves}$: nodes 0,1,2,5 (degree = 1).

Round 1 (prune leaves):

- Remove $0,1,2,5 \rightarrow remaining_nodes = 6 4 = 2$
- Update neighbors:
- Remove 0,1,2 from node 3 \rightarrow degree[3] = 4 3 = 1
- Remove 5 from node $4 \rightarrow \text{degree}[4] = 2 1 = 1$
- Now both 3 and 4 have degree 1 \rightarrow added to new leaves queue.

But wait! remaining_nodes = 2, so we stop before next pruning.

Final leaves queue: $[3, 4] \rightarrow \text{returned as answer.}$

This matches expected output [3,4].

Complexity Analysis

• Time Complexity: O(n)

Each edge is processed exactly twice (once per endpoint). Each node is enqueued and dequeued at most once. Total operations scale linearly with ${\tt n}.$

• Space Complexity: O(n)

Adjacency list stores 2*(n-1) edges $\to 0(n)$. Degree array and queue also use 0(n) space.