

# Graphs

## Chunk 1: Graph Traversal & BFS/DFS Patterns

### Pattern: DFS / BFS on Grid (2D Array)

*Used for connected components, flood fill, shortest path in unweighted grids.*

#### How to Recognize:

- Problem involves a 2D grid (**matrix**, **grid**, **board**) with cells that can be visited or blocked.
- You're asked to:
  - Count islands
  - Find shortest path
  - Flood-fill a region
  - Traverse all reachable cells from a starting point
- Movement is typically up/down/left/right (sometimes diagonals)

#### Step-by-Step Thinking Process (The Recipe):

1. **Check bounds:** Ensure row/col indices are within grid dimensions.
2. **Check visited state:** Use a **visited** set or mark grid in-place (e.g., change value).
3. **Base case:** If cell is invalid, out of bounds, or already visited → return.
4. **Process current cell:** Do work (e.g., increment count, update distance).
5. **Recursive/Queue-based traversal:** Call DFS/BFS on neighbors.
6. **Return result:** Based on problem (count, path length, etc.).

#### Common Pitfalls & Edge Cases:

- Forgetting to mark visited nodes → infinite loop.
  - Not handling edge cases like empty grid, single cell, or no valid path.
  - Using recursion for large grids → stack overflow (use iterative BFS instead).
  - Misinterpreting movement rules (e.g., only 4-directional vs 8-directional).
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## Problem 1: Flood Fill

### Summary:

Given a 2D image represented as a matrix of integers, start at a pixel (**sr**, **sc**) and replace all adjacent pixels with the same color as the starting pixel with a new color. Adjacency is defined as 4-directional.

### Pattern(s):

- DFS / BFS on Grid
- Connected Components

### Solution with Inline Comments:

```
def floodFill(image, sr, sc, newColor):
    # Get dimensions of the image
    rows, cols = len(image), len(image[0])

    # Original color at the starting pixel
    original_color = image[sr][sc]

    # If the new color is same as old, no need to do anything
    if original_color == newColor:
        return image

    # Use DFS to fill the region
    def dfs(r, c):
        # Base case: check bounds and if pixel has correct original color
        if (r < 0 or r >= rows or
            c < 0 or c >= cols or
            image[r][c] != original_color):
            return

        # Change the color of current pixel
        image[r][c] = newColor

        # Recursively apply DFS to 4 neighbors
        dfs(r + 1, c) # down
        dfs(r - 1, c) # up
```

```

        dfs(r, c + 1) # right
        dfs(r, c - 1) # left

# Start DFS from the initial pixel
dfs(sr, sc)

return image

```

### Official Example Walkthrough:

#### Input:

image = [[1,1,1],[1,1,0],[1,0,1]], sr = 1, sc = 1, newColor = 2

**Step-by-step:** 1. Start at (1,1) — original color = 1, new color = 2. 2. Mark (1,1) → 2. 3. Visit neighbors: - (2,1) → color 0 1 → skip - (0,1) → color 1 → mark as 2 - From (0,1): visit (0,0) → 1 → mark as 2 - From (0,0): neighbors (0,1) already done; (1,0) → 1 → mark as 2 - From (1,0): neighbors (2,0) → 1 → mark as 2 - From (2,0): neighbor (2,1) → 0 → skip - (1,2) → 0 → skip - (1,0) → already processed via chain

#### Final Image:

```

[[2,2,2],
 [2,2,0],
 [2,0,1]]

```

Output matches expected.

#### Complexity:

- **Time:**  $O(m \times n)$  — each cell visited once
- **Space:**  $O(m \times n)$  — worst-case recursion depth (stack) if entire grid is filled

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### Problem 2: [01 Matrix](#)

#### Summary:

Given a binary matrix, find the distance from each cell to the nearest 0. Distance is Manhattan distance.

### Pattern(s):

- Multi-source BFS
- Shortest Path in Unweighted Grid

### Solution with Inline Comments:

```
from collections import deque

def updateMatrix(mat):
    rows, cols = len(mat), len(mat[0])

    # Initialize result matrix with infinity (or large number)
    dist = [[float('inf')] * cols for _ in range(rows)]

    # Queue for BFS: store (row, col, distance)
    queue = deque()

    # Add all '0' cells as sources with distance 0
    for i in range(rows):
        for j in range(cols):
            if mat[i][j] == 0:
                dist[i][j] = 0
                queue.append((i, j))

    # Directions: up, down, left, right
    directions = [(0, 1), (0, -1), (1, 0), (-1, 0)]

    # Multi-source BFS
    while queue:
        r, c = queue.popleft()
        current_dist = dist[r][c]

        # Explore neighbors
        for dr, dc in directions:
            nr, nc = r + dr, c + dc

            # Check bounds and if we can update distance
            if (0 <= nr < rows and 0 <= nc < cols and
                dist[nr][nc] > current_dist + 1):
```

```

        # Update distance and add to queue
        dist[nr][nc] = current_dist + 1
        queue.append((nr, nc))

    return dist

```

### Official Example Walkthrough:

#### Input:

```
mat = [[0,0,0],[0,1,0],[1,1,1]]
```

**Step-by-step:** 1. All 0s go into queue with distance 0. 2. Process (0,0), (0,1), (0,2), (1,0), (1,2) → update their neighbors: - From (0,0): (1,0) already in queue → skip - From (0,1): (1,1) → set to 1, add to queue - From (1,0): (2,0) → set to 1, add to queue - From (1,2): (2,2) → set to 1, add to queue 3. Now process (1,1) → update (2,1) → set to 2 4. Process (2,0) → (2,1) already updated 5. Process (2,2) → (2,1) already updated 6. Final distances:

```

[[0,0,0],
 [0,1,0],
 [1,2,1]]

```

Matches expected output.

#### Complexity:

- **Time:**  $O(m \times n)$  — each cell processed at most once
- **Space:**  $O(m \times n)$  — for `dist` array and queue (max size  $\sim m \times n$ )

### Problem 3: Number of Islands

#### Summary:

Given a 2D grid of '1's (land) and '0's (water), count the number of distinct islands (connected groups of '1's).

### Pattern(s):

- DFS/BFS on Grid
- Connected Components

### Solution with Inline Comments:

```
def numIslands(grid):
    if not grid or not grid[0]:
        return 0

    rows, cols = len(grid), len(grid[0])
    visited = [[False] * cols for _ in range(rows)]
    islands = 0

    # Directions: up, down, left, right
    directions = [(0, 1), (0, -1), (1, 0), (-1, 0)]

    def dfs(r, c):
        # Base case: out of bounds or water or already visited
        if (r < 0 or r >= rows or
            c < 0 or c >= cols or
            grid[r][c] == '0' or
            visited[r][c]):
            return

        # Mark as visited
        visited[r][c] = True

        # Explore all 4 neighbors
        for dr, dc in directions:
            dfs(r + dr, c + dc)

    # Iterate through every cell
    for i in range(rows):
        for j in range(cols):
            # If land not visited, start DFS → new island
            if grid[i][j] == '1' and not visited[i][j]:
                dfs(i, j)
                islands += 1
```

```
return islands
```

### Official Example Walkthrough:

#### Input:

```
[
  ["1","1","1","1","0"],
  ["1","1","0","1","0"],
  ["1","1","0","0","0"],
  ["0","0","0","0","0"]
]
```

**Step-by-step:** - Start at (0,0) → land → DFS covers entire top-left island → count = 1 - Move to (0,4) → water → skip - Continue until (1,1) → already visited - At (1,3) → land → but not visited? Wait: (1,3) is part of first island? - Actually, (1,3) connects to (0,3) → which connects to (0,0) → so it's one island! - Only one connected component exists.

Wait! But input shows two separate islands?

Let me recheck:

```
Row 0: 1 1 1 1 0
Row 1: 1 1 0 1 0
Row 2: 1 1 0 0 0
Row 3: 0 0 0 0 0
```

So: - Top-left block: all ones connected → one island - (1,3) is isolated from others? No — (0,3) → (1,3) → yes, connected!

But (1,3) is connected to (0,3) → which is part of main island → so still one island.

Wait — this is actually **one island**!

But expected output is 1.

Yes — only **one** connected group of '1'.

If we had:

```
["1","1","0"],
["0","0","1"]
```

→ Two islands.

So this example gives 1.

**Complexity:**

- **Time:**  $O(m \times n)$  — each cell visited once
  - **Space:**  $O(m \times n)$  — `visited` array + recursion stack (worst-case depth  $m \times n$ )
- 

**Problem 4: Rotting Oranges**

**Summary:**

In a grid with 0 (empty), 1 (fresh orange), 2 (rotten orange), every minute rotten oranges infect adjacent fresh oranges. Return minutes until all oranges rot, or -1 if impossible.

**Pattern(s):**

- Multi-source BFS
- Level-order traversal (time progression)

**Solution with Inline Comments:**

```
from collections import deque

def orangesRotting(grid):
    rows, cols = len(grid), len(grid[0])

    # Count fresh oranges and collect all rotten ones
    fresh_count = 0
    queue = deque()

    for i in range(rows):
        for j in range(cols):
            if grid[i][j] == 1:
                fresh_count += 1
            elif grid[i][j] == 2:
```



```

        queue.append((i, j))

# If no fresh oranges, return 0
if fresh_count == 0:
    return 0

# Directions: up, down, left, right
directions = [(0, 1), (0, -1), (1, 0), (-1, 0)]
minutes = 0

# Multi-source BFS
while queue and fresh_count > 0:
    # Process all rotten oranges at current time level
    level_size = len(queue)

    for _ in range(level_size):
        r, c = queue.popleft()

        for dr, dc in directions:
            nr, nc = r + dr, c + dc

            # Check bounds and if fresh orange
            if (0 <= nr < rows and 0 <= nc < cols and
                grid[nr][nc] == 1):

                # Rot the orange
                grid[nr][nc] = 2
                fresh_count -= 1
                queue.append((nr, nc))

    minutes += 1 # One minute passed

# If any fresh orange remains, return -1
return minutes if fresh_count == 0 else -1

```

### Official Example Walkthrough:

#### Input:

```

[
  [2,1,1],

```

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

**Step-by-step:** - Rotten oranges: (0,0) - Fresh: 5 - Minute 0: process (0,0) → infect (0,1) and (1,0) - Minute 1: process (0,1), (1,0) → infect (0,2), (1,1) - Minute 2: process (0,2), (1,1) → infect (2,1) - Minute 3: process (2,1) → no new infections - All fresh oranges now rotten → return 3

Output: 3

#### Complexity:

- **Time:**  $O(m \times n)$  — each cell processed once
  - **Space:**  $O(m \times n)$  — queue holds up to all rotten oranges
- 

### Problem 5: [Word Search](#)

#### Summary:

Given a 2D board and a word, determine if the word exists in the grid by moving up/down/left/right. Each letter can be used only once per path.

#### Pattern(s):

- DFS with Backtracking
- Grid Traversal

#### Solution with Inline Comments:

```
def exist(board, word):  
    if not board or not board[0]:  
        return False  
  
    rows, cols = len(board), len(board[0])  
  
    # Directions: up, down, left, right
```

```

directions = [(0, 1), (0, -1), (1, 0), (-1, 0)]

def dfs(r, c, index):
    # Base case: if we've matched all characters
    if index == len(word):
        return True

    # Check bounds and character match
    if (r < 0 or r >= rows or
        c < 0 or c >= cols or
        board[r][c] != word[index]):
        return False

    # Temporarily mark current cell as visited (by changing char)
    temp = board[r][c]
    board[r][c] = '#' # mark as visited

    # Try all 4 directions
    for dr, dc in directions:
        if dfs(r + dr, c + dc, index + 1):
            board[r][c] = temp # backtrack
            return True

    # Backtrack: restore original character
    board[r][c] = temp
    return False

# Try starting from every cell
for i in range(rows):
    for j in range(cols):
        if dfs(i, j, 0):
            return True

return False

```

### Official Example Walkthrough:

#### Input:

board = [["A","B","C","E"],["S","F","C","S"],["A","D","E","E"]], word = "ABCCED"

**Step-by-step:** - Start at (0,0) → 'A' → good - Go to (0,1) → 'B' → good - Go to (0,2) →

'C' → good - Go to (0,3) → 'E' → good - Go to (1,3) → 'S' → bad - Backtrack → try (1,2) → 'C' → good - Then (2,2) → 'E' → good - Then (2,3) → 'E' → good → word found!

Returns True

### Complexity:

- **Time:**  $O(m \times n \times 4^L)$  where  $L$  = length of word (each step has up to 4 choices)
  - **Space:**  $O(L)$  — recursion depth (path length)
- 

## Problem 6: Pacific Atlantic Water Flow

### Summary:

Given a height map, return all cells from which water can flow to both Pacific (top/left edges) and Atlantic (bottom/right edges). Water flows to adjacent cells with equal or lower height.

### Pattern(s):

- BFS/DFS from boundaries
- Reverse flow simulation

### Solution with Inline Comments:

```
from collections import deque

def pacificAtlantic(heights):
    if not heights or not heights[0]:
        return []

    rows, cols = len(heights), len(heights[0])

    # Sets to track cells reachable from each ocean
    pacific_reachable = set()
    atlantic_reachable = set()

    # Directions: up, down, left, right
```

```

directions = [(0, 1), (0, -1), (1, 0), (-1, 0)]

def bfs(queue, reachable_set):
    while queue:
        r, c = queue.popleft()
        reachable_set.add((r, c))

        for dr, dc in directions:
            nr, nc = r + dr, c + dc

            # Check bounds
            if (0 <= nr < rows and 0 <= nc < cols and
                (nr, nc) not in reachable_set and
                heights[nr][nc] >= heights[r][c]): # reverse flow

                queue.append((nr, nc))

# Initialize queues: start from top row (pacific) and leftmost column
pacific_queue = deque()
atlantic_queue = deque()

for c in range(cols):
    pacific_queue.append((0, c)) # top row
    atlantic_queue.append((rows - 1, c)) # bottom row

for r in range(rows):
    pacific_queue.append((r, 0)) # leftmost col
    atlantic_queue.append((r, cols - 1)) # rightmost col

# Run BFS from both oceans
bfs(pacific_queue, pacific_reachable)
bfs(atlantic_queue, atlantic_reachable)

# Return intersection of both sets
return list(pacific_reachable & atlantic_reachable)

```

### Official Example Walkthrough:

#### Input:

```
heights = [[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]]
```

**Step-by-step:** - Pacific starts from top and left edges → propagate inward where height previous - Atlantic starts from bottom and right edges → same - Cells reachable from both → returned

Example output: `[[0,4],[1,3],[1,4],[2,3],[2,4],[3,0],[4,0],[4,1],[4,2],[4,3],[4,4]]`

Correct.

### Complexity:

- **Time:**  $O(m \times n)$  — each cell visited at most twice
  - **Space:**  $O(m \times n)$  — sets and queues
- 

## Problem 7: **Shortest Path to Get Food**

### Summary:

In a grid with 'X' (walls), '.' (empty), '\*' (food), find minimum steps from robot (start) to food. Can move in 4 directions.

### Pattern(s):

- BFS (Shortest Path in Unweighted Grid)

### Solution with Inline Comments:

```
from collections import deque

def getFood(grid):
    rows, cols = len(grid), len(grid[0])

    # Find robot position
    start_r, start_c = None, None
    for i in range(rows):
        for j in range(cols):
            if grid[i][j] == '*':
                start_r, start_c = i, j
```

```

        break
    if start_r is not None:
        break

# BFS setup
queue = deque([(start_r, start_c, 0)]) # (row, col, steps)
visited = {(start_r, start_c)}

# Directions: up, down, left, right
directions = [(0, 1), (0, -1), (1, 0), (-1, 0)]

while queue:
    r, c, steps = queue.popleft()

    # Explore neighbors
    for dr, dc in directions:
        nr, nc = r + dr, c + dc

        # Check bounds and wall
        if (0 <= nr < rows and 0 <= nc < cols and
            grid[nr][nc] != 'X' and
            (nr, nc) not in visited):

            if grid[nr][nc] == '#': # food found!
                return steps + 1

            visited.add((nr, nc))
            queue.append((nr, nc, steps + 1))

return -1 # no path to food

```

### Official Example Walkthrough:

#### Input:

```

grid = [
    ["X","X","X","X","X","X"],
    ["X","*","0","0","0","X"],
    ["X","0","0","#","0","X"],
    ["X","X","X","X","X","X"]
]

```

- Robot at (1,1), food at (2,3)
- BFS:
  - Step 0: (1,1)
  - Step 1: (1,2), (2,1)
  - Step 2: (1,3), (2,2)
  - Step 3: (2,3) → food → return 3

Output: 3

### Complexity:

- **Time:**  $O(m \times n)$
- **Space:**  $O(m \times n)$

## Chunk 2: Topological Sort, Union-Find, Shortest Path, Backtracking + DFS

We'll now cover the following problems:

- **Topological Sort**
  - Course Schedule → Detect cycle (DFS/Kahn's)
  - Course Schedule II → Return ordering (DFS/Kahn's)
  - Alien Dictionary → Build graph from dictionary order + topo sort
- **Union-Find**
  - Accounts Merge → DSU with emails
  - Graph Valid Tree → DSU cycle detection
  - Number of Connected Components in Undirected Graph → DSU
- **Shortest Path**
  - Word Ladder → BFS
  - Minimum Knight Moves → BFS on infinite grid (symmetry trick)
  - Bus Routes → BFS with routes as graph
  - Cheapest Flights Within K Stops → BFS/Dijkstra with (node, stops) state
- **Backtracking + DFS**
  - Word Search → DFS with visited (already covered in Chunk 1 — skip)
  - Word Search II → DFS + Trie for pruning

We'll reprocess **Word Search II** here since it's a more advanced variant.



## Pattern: Topological Sort

*Used when tasks have dependencies; output is an ordering where each task comes after its prerequisites.*

### How to Recognize:

- Problem involves dependencies: “A must happen before B”
- You’re asked to find an order of execution or detect if impossible
- Graph is **Directed Acyclic Graph (DAG)** — no cycles allowed

### Step-by-Step Thinking Process (The Recipe):

1. **Build adjacency list and indegree array** from edges.
2. **Find all nodes with indegree 0** → they can start first.
3. **Use queue (Kahn’s Algorithm)**: repeatedly remove indegree 0 nodes.
4. For each removed node, decrease indegree of its neighbors.
5. If all nodes are processed → valid topological order exists.
6. Else → cycle → return empty list.

### Common Pitfalls & Edge Cases:

- Not checking for cycles → return invalid order.
  - Using DFS without proper backtracking → missing nodes.
  - Forgetting that multiple valid orders exist → any valid one is acceptable.
  - Handling self-loops or duplicate edges.
- 

## Problem 1: [Course Schedule](#)

### Summary:

Given  $n$  courses labeled 0 to  $n-1$  and a list of prerequisite pairs, determine if it’s possible to finish all courses.

### Pattern(s):

- **Topological Sort (Cycle Detection)**

### Solution with Inline Comments:

```
from collections import deque

def canFinish(numCourses, prerequisites):
    # Build adjacency list and indegree array
    graph = [[] for _ in range(numCourses)]
    indegree = [0] * numCourses

    for course, prereq in prerequisites:
        graph[prereq].append(course) # prereq -> course
        indegree[course] += 1

    # Queue for nodes with indegree 0
    queue = deque()
    for i in range(numCourses):
        if indegree[i] == 0:
            queue.append(i)

    # Count processed nodes
    processed = 0

    while queue:
        node = queue.popleft()
        processed += 1

        # Reduce indegree of neighbors
        for neighbor in graph[node]:
            indegree[neighbor] -= 1
            if indegree[neighbor] == 0:
                queue.append(neighbor)

    # If all courses processed → no cycle
    return processed == numCourses
```

### Official Example Walkthrough:

#### Input:

numCourses = 2, prerequisites = [[1,0]]

- Course 1 depends on course 0.

- Graph:  $0 \rightarrow 1$
- Indegree:  $[0,1]$
- Start: `queue = [0]`
- Process 0  $\rightarrow$  reduce indegree of 1  $\rightarrow$  becomes 0  $\rightarrow$  add to queue
- Process 1  $\rightarrow$  done
- `processed = 2`  $\rightarrow$  equals `numCourses`  $\rightarrow$  return `True`

Output: `true`

### Complexity:

- **Time:**  $O(V + E)$  — visit each node and edge once
- **Space:**  $O(V + E)$  — graph + indegree + queue

## Problem 2: [Course Schedule II](#)

### Summary:

Return **any valid order** of courses to complete all courses, or return an empty array if impossible.

### Pattern(s):

- Topological Sort (Ordering)

### Solution with Inline Comments:

```
from collections import deque

def findOrder(numCourses, prerequisites):
    # Build graph and indegree
    graph = [[] for _ in range(numCourses)]
    indegree = [0] * numCourses

    for course, prereq in prerequisites:
        graph[prereq].append(course)
        indegree[course] += 1
```

```

# Queue for indegree 0 nodes
queue = deque()
for i in range(numCourses):
    if indegree[i] == 0:
        queue.append(i)

result = []

while queue:
    node = queue.popleft()
    result.append(node)

    for neighbor in graph[node]:
        indegree[neighbor] -= 1
        if indegree[neighbor] == 0:
            queue.append(neighbor)

# Check if all courses are included
return result if len(result) == numCourses else []

```

### Official Example Walkthrough:

#### Input:

numCourses = 4, prerequisites = [[1,0],[2,0],[3,1],[3,2]]

- Dependencies:  $0 \rightarrow 1$ ,  $0 \rightarrow 2$ ,  $1 \rightarrow 3$ ,  $2 \rightarrow 3$
- Graph:  $0 \rightarrow 1 \rightarrow 3$ ,  $0 \rightarrow 2 \rightarrow 3$
- Indegree: [0,1,1,2]
- Start: queue = [0]
- Process 0  $\rightarrow$  update 1 and 2  $\rightarrow$  both indegree=0  $\rightarrow$  add to queue
- Process 1  $\rightarrow$  update 3  $\rightarrow$  indegree[3]=1
- Process 2  $\rightarrow$  update 3  $\rightarrow$  indegree[3]=0  $\rightarrow$  add to queue
- Process 3  $\rightarrow$  done
- Result: [0,1,2,3]

Output: [0,1,2,3] (or any valid order like [0,2,1,3])

#### Complexity:

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

---

### Problem 3: Alien Dictionary

#### Summary:

Given a list of words sorted lexicographically in an alien language, reconstruct the order of characters.

#### Pattern(s):

- Topological Sort
- Graph Construction from String Comparison

#### Solution with Inline Comments:

```
from collections import defaultdict, deque

def alienOrder(words):
    # Step 1: Initialize graph and indegree
    graph = defaultdict(set)
    indegree = {char: 0 for word in words for char in word}

    # Step 2: Compare adjacent words to build edges
    for i in range(len(words) - 1):
        w1, w2 = words[i], words[i + 1]
        min_len = min(len(w1), len(w2))

        found_diff = False
        for j in range(min_len):
            c1, c2 = w1[j], w2[j]
            if c1 != c2:
                # Add edge: c1 -> c2
                if c2 not in graph[c1]:
                    graph[c1].add(c2)
                    indegree[c2] += 1
                found_diff = True
                break

    # Topological Sort using Kahn's algorithm
    queue = deque([char for char, degree in indegree.items() if degree == 0])
    order = []
    while queue:
        char = queue.popleft()
        order.append(char)
        for neighbor in graph[char]:
            indegree[neighbor] -= 1
            if indegree[neighbor] == 0:
                queue.append(neighbor)

    return ''.join(order) if order else ''
```

```

        # If shorter word is prefix of longer → invalid (e.g., "abc" before "ab")
        if not found_diff and len(w1) > len(w2):
            return ""

# Step 3: Kahn's algorithm
queue = deque([c for c in indegree if indegree[c] == 0])
result = []

while queue:
    node = queue.popleft()
    result.append(node)

    for neighbor in graph[node]:
        indegree[neighbor] -= 1
        if indegree[neighbor] == 0:
            queue.append(neighbor)

# Check if all characters are used
return ''.join(result) if len(result) == len(indegree) else ""

```

### Official Example Walkthrough:

**Input:** words = ["wrt","wrf","er","ett","rft"]

- Compare "wrt" and "wrf" → 't' vs 'f' → add t → f
- Compare "wrf" and "er" → 'w' vs 'e' → add w → e
- Compare "er" and "ett" → 'e' vs 'e', then 'r' vs 't' → add r → t
- Compare "ett" and "rft" → 'e' vs 'r' → add e → r

Edges: t→f, w→e, r→t, e→r

Graph: - w → e - e → r - r → t - t → f

Indegree: w:0, e:1, r:1, t:1, f:1

Start: queue = [w] - Process w → e now indegree 0 - Process e → r → 0 - Process r → t → 0 - Process t → f → 0 - Process f

Result: "wertf"

Output: "wertf" (or valid permutation)

### Complexity:

- **Time:**  $O(C)$  where  $C$  = total characters across all words
  - **Space:**  $O(1)$  — at most 26 letters, so constant
- 

### Pattern: Union-Find (Disjoint Set Union)

*Used to manage disjoint sets, detect cycles, merge groups, count components.*

### How to Recognize:

- Problems about connected components, merging groups, detecting cycles in undirected graphs.
- You're told to "merge" or "union" things, or check if two items are connected.

### Step-by-Step Thinking Process (The Recipe):

1. **Initialize parent and rank arrays** (each node points to itself).
2. **Find root with path compression** (`find(x)`).
3. **Union by rank** (`union(x,y)`): attach smaller tree under larger.
4. Use `union` to connect components.
5. Use `find` to check connectivity or detect cycles.

### Common Pitfalls & Edge Cases:

- Not using path compression/rank  $\rightarrow$  slow performance.
  - Incorrect union logic (e.g., always attaching  $x$  to  $y$ ).
  - Not handling self-loops or duplicate edges.
  - Returning wrong result (e.g., returning `True` when union fails due to cycle).
- 

### Problem 4: [Accounts Merge](#)

#### Summary:

Given accounts (name + emails), merge accounts with common emails into one.

### Pattern(s):

- Union-Find (DSU)
- Graph of Emails

### Solution with Inline Comments:

```
class UnionFind:
    def __init__(self):
        self.parent = {}

    def find(self, x):
        if x not in self.parent:
            self.parent[x] = x
        if self.parent[x] != x:
            self.parent[x] = self.find(self.parent[x]) # path compression
        return self.parent[x]

    def union(self, x, y):
        px, py = self.find(x), self.find(y)
        if px != py:
            self.parent[px] = py

def accountsMerge(accounts):
    uf = UnionFind()
    email_to_name = {}

    # Step 1: Union emails within same account
    for account in accounts:
        name = account[0]
        emails = account[1:]

        # Link all emails in this account
        for email in emails:
            email_to_name[email] = name
            uf.union(emails[0], email) # link all to first email

    # Step 2: Group emails by their root parent
    groups = defaultdict(list)
    for email in email_to_name:
        root = uf.find(email)
```



```

        groups[root].append(email)

# Step 3: Sort and format output
result = []
for group in groups.values():
    result.append([email_to_name[group[0]] + sorted(group)])

return result

```

### Official Example Walkthrough:

#### Input:

```

accounts = [
    ["John", "johnsmith@mail.com", "john_newyork@mail.com"],
    ["John", "johnsmith@mail.com", "john00@mail.com"],
    ["Mary", "mary@mail.com"]
]

```

- John's emails: johnsmith@mail.com, john\_newyork@mail.com, john00@mail.com
- First account: union johnsmith john\_newyork
- Second account: union johnsmith john00 → so all three linked
- Mary: standalone

Groups: - Root of johnsmith → all three emails grouped - Mary → her own

#### Output:

```

[
    ["John", "john00@mail.com", "john_newyork@mail.com", "johnsmith@mail.com"],
    ["Mary", "mary@mail.com"]
]

```

Correct.

#### Complexity:

- **Time:**  $O(A \times M \times (M))$  where  $A$  = accounts,  $M$  = avg emails per account,  $=$  inverse Ackermann
- **Space:**  $O(E)$  — number of emails

## Problem 5: Graph Valid Tree

### Summary:

Given  $n$  nodes and `edges`, determine if the graph forms a valid tree (connected, acyclic,  $n-1$  edges).

### Pattern(s):

- Union-Find (Cycle Detection)
- Connectivity Check

### Solution with Inline Comments:

```
class UnionFind:
    def __init__(self, n):
        self.parent = list(range(n))
        self.rank = [0] * n

    def find(self, x):
        if self.parent[x] != x:
            self.parent[x] = self.find(self.parent[x])
        return self.parent[x]

    def union(self, x, y):
        px, py = self.find(x), self.find(y)
        if px == py:
            return False # cycle detected
        if self.rank[px] < self.rank[py]:
            px, py = py, px
        self.parent[py] = px
        if self.rank[px] == self.rank[py]:
            self.rank[px] += 1
        return True

def validTree(n, edges):
    # Must have exactly n-1 edges
    if len(edges) != n - 1:
        return False
```

```
uf = UnionFind(n)

# Try to union all edges
for u, v in edges:
    if not uf.union(u, v):
        return False # cycle found

return True
```

### Official Example Walkthrough:

**Input:**  $n = 5$ ,  $\text{edges} = [[0,1], [0,2], [0,3], [1,4]]$

- Edges = 4  $\rightarrow$   $n-1 = 4 \rightarrow$  ok
- Union: 0-1, 0-2, 0-3, 1-4  $\rightarrow$  no cycles
- All nodes connected? Yes
- Return True

Output: true

### Complexity:

- **Time:**  $O(E \times (N))$
  - **Space:**  $O(N)$
- 

## Problem 6: Number of Connected Components in Undirected Graph

### Summary:

Count how many connected components exist in an undirected graph.

### Pattern(s):

- Union-Find
- Component Counting

### Solution with Inline Comments:

```
class UnionFind:
    def __init__(self, n):
        self.parent = list(range(n))
        self.rank = [0] * n
        self.components = n # count of distinct components

    def find(self, x):
        if self.parent[x] != x:
            self.parent[x] = self.find(self.parent[x])
        return self.parent[x]

    def union(self, x, y):
        px, py = self.find(x), self.find(y)
        if px == py:
            return
        if self.rank[px] < self.rank[py]:
            px, py = py, px
        self.parent[py] = px
        if self.rank[px] == self.rank[py]:
            self.rank[px] += 1
        self.components -= 1

def countComponents(n, edges):
    uf = UnionFind(n)

    for u, v in edges:
        uf.union(u, v)

    return uf.components
```

### Official Example Walkthrough:

**Input:** n = 5, edges = [[0,1],[1,2],[3,4]]

- Initially 5 components
- Union 0-1  $\rightarrow$  4
- Union 1-2  $\rightarrow$  3
- Union 3-4  $\rightarrow$  2
- Final count: 2

Output: 2

**Complexity:**

- **Time:**  $O(E \times (N))$
  - **Space:**  $O(N)$
- 

**Pattern: Shortest Path in Weighted Graphs**

*BFS for unweighted, Dijkstra for weighted, with constraints (e.g., max stops).*

**How to Recognize:**

- Find shortest path in graph with weights or steps.
  - Constraints: max stops, limited moves, etc.
  - Often use **priority queue (Dijkstra)** or **BFS with state tracking**.
- 

**Problem 7: [Word Ladder](#)**

**Summary:**

Transform `beginWord` to `endWord` by changing one letter at a time, with each intermediate word in `wordList`.

**Pattern(s):**

- **BFS (Unweighted Shortest Path)**
- **Graph via Word Transformation**

**Solution with Inline Comments:**

```

from collections import deque

def ladderLength(beginWord, endWord, wordList):
    wordSet = set(wordList)

    # If endWord not in wordList, impossible
    if endWord not in wordSet:
        return 0

    # BFS queue: (word, steps)
    queue = deque([(beginWord, 1)])
    visited = {beginWord}

    # Letters a-z
    alphabet = 'abcdefghijklmnopqrstuvwxyz'

    while queue:
        word, steps = queue.popleft()

        # Try changing each character
        for i in range(len(word)):
            for c in alphabet:
                new_word = word[:i] + c + word[i+1:]

                if new_word == endWord:
                    return steps + 1

                if new_word in wordSet and new_word not in visited:
                    visited.add(new_word)
                    queue.append((new_word, steps + 1))

    return 0

```

### Official Example Walkthrough:

```

**Input:** `beginWord = "hit"`,
          `endWord = "cog"`,
          `wordList = ["hot","dot","dog","lot","log","cog"]`

```

- hit → hot → dot → dog → cog → 5 steps
- But output counts number of words in sequence → 5

Output: 5

**Complexity:**

- **Time:**  $O(L \times M \times 26)$  where  $L$  = word length,  $M$  = number of words
  - **Space:**  $O(M)$
- 

**Problem 8: Minimum Knight Moves**

**Summary:**

Find minimum moves for knight to go from (0,0) to (x,y) on infinite chessboard.

**Pattern(s):**

- BFS on Infinite Grid
- Symmetry Optimization

**Solution with Inline Comments:**

```
from collections import deque

def minKnightMoves(x, y):
    # Use symmetry: only consider positive quadrant
    x, y = abs(x), abs(y)

    # Directions for knight
    directions = [(2,1),(2,-1),(-2,1),(-2,-1),
                  (1,2),(1,-2),(-1,2),(-1,-2)]

    # BFS
    queue = deque([(0, 0, 0)]) # (row, col, steps)
    visited = {(0, 0)}

    while queue:
        r, c, steps = queue.popleft()
```

```

    if r == x and c == y:
        return steps

    for dr, dc in directions:
        nr, nc = r + dr, c + dc

        # Use symmetry: mirror negative coordinates
        nr, nc = abs(nr), abs(nc)

        if (nr, nc) not in visited and nr <= x + 2 and nc <= y + 2:
            visited.add((nr, nc))
            queue.append((nr, nc, steps + 1))

    return -1

```

### Official Example Walkthrough:

**Input:**  $x = 2, y = 1$

- Knight at (0,0)  $\rightarrow$  (2,1) in 1 move? No  $\rightarrow$  try (1,2), (2,1)  $\rightarrow$  yes!
- From (0,0)  $\rightarrow$  (2,1)  $\rightarrow$  valid move  $\rightarrow$  1 step

**Output:** 1

### Complexity:

- **Time:**  $O(\max(|x|, |y|)^2)$  — bounded by symmetry
- **Space:**  $O(\max(|x|, |y|)^2)$

## Problem 9: [Bus Routes](#)

### Summary:

Given bus routes, find minimum number of buses to take from source to target stop.



**Pattern(s):**

- BFS with Route-to-Stop Mapping
- Graph: Stop → Bus

**Solution with Inline Comments:**

```
from collections import deque, defaultdict

def numBusesToDestination(routes, source, target):
    if source == target:
        return 0

    # Map each stop to list of buses (routes)
    stop_to_buses = defaultdict(list)
    for bus_id, route in enumerate(routes):
        for stop in route:
            stop_to_buses[stop].append(bus_id)

    # BFS: (bus_id, stops_taken)
    queue = deque()
    visited_buses = set()

    # Start from all buses that serve source
    for bus_id in stop_to_buses[source]:
        queue.append((bus_id, 1))
        visited_buses.add(bus_id)

    while queue:
        bus_id, count = queue.popleft()

        # Visit all stops on this bus
        for stop in routes[bus_id]:
            if stop == target:
                return count

        # Take other buses from this stop
        for next_bus_id in stop_to_buses[stop]:
            if next_bus_id not in visited_buses:
                visited_buses.add(next_bus_id)
                queue.append((next_bus_id, count + 1))
```

```
return -1
```

### Official Example Walkthrough:

**Input:** routes = [[1,2,7],[3,6,7]], source = 1, target = 6

- Bus 0 serves stops 1,2,7
- Bus 1 serves stops 3,6,7
- Source = 1 → take bus 0 → reach stop 7
- Stop 7 → take bus 1 → reach stop 6 → target!

So: 2 buses

Output: 2

### Complexity:

- **Time:**  $O(B \times S)$  where  $B$  = buses,  $S$  = stops
  - **Space:**  $O(B \times S)$
- 

## Problem 10: Cheapest Flights Within K Stops

### Summary:

Find cheapest price from `src` to `dst` with at most  $k$  stops.

### Pattern(s):

- BFS / Dijkstra with (node, stops, cost) state
- State-space search

### Solution with Inline Comments:

```

from collections import deque
import heapq

def findCheapestPrice(n, flights, src, dst, k):
    # Build adjacency list
    graph = [[] for _ in range(n)]
    for u, v, w in flights:
        graph[u].append((v, w))

    # Min-heap: (cost, stops, node)
    heap = [(0, 0, src)]
    min_cost = [float('inf')] * n

    while heap:
        cost, stops, node = heapq.heappop(heap)

        if node == dst:
            return cost

        if stops > k:
            continue

        if cost > min_cost[node]:
            continue

        for neighbor, flight_cost in graph[node]:
            new_cost = cost + flight_cost
            new_stops = stops + 1

            if new_cost < min_cost[neighbor]:
                min_cost[neighbor] = new_cost
                heapq.heappush(heap, (new_cost, new_stops, neighbor))

    return -1

```

### Official Example Walkthrough:

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

- $0 \rightarrow 1 \rightarrow 2$ :  $cost = 200$ ,  $stops = 1 \rightarrow$  valid
- $0 \rightarrow 2$ :  $cost = 500$ ,  $stops = 0 \rightarrow$  but only 1 stop allowed  $\rightarrow$  valid
- $Min = 200$

Output: 200

**Complexity:**

- **Time:**  $O(E \times \log V \times (K+1))$  — each edge may be pushed multiple times
  - **Space:**  $O(V + E)$
- 

**Problem 11: [Word Search II](#)**

**Summary:**

Given a board and list of words, return all words found in the board.

**Pattern(s):**

- Trie + DFS with Backtracking
- Pruning via Prefix Matching

**Solution with Inline Comments:**

```
class TrieNode:
    def __init__(self):
        self.children = {}
        self.word = None # store full word at leaf

class Trie:
    def __init__(self):
        self.root = TrieNode()

    def insert(self, word):
        node = self.root
        for c in word:
            if c not in node.children:
                node.children[c] = TrieNode()
            node = node.children[c]
        node.word = word
```

```

def findWords(board, words):
    if not board or not board[0]:
        return []

    rows, cols = len(board), len(board[0])
    trie = Trie()
    for word in words:
        trie.insert(word)

    result = set()
    directions = [(0,1),(0,-1),(1,0),(-1,0)]

    def dfs(r, c, node):
        char = board[r][c]
        current_node = node.children.get(char)

        if not current_node:
            return

        # Found a word
        if current_node.word:
            result.add(current_node.word)

        # Mark visited
        board[r][c] = '#'

        for dr, dc in directions:
            nr, nc = r + dr, c + dc
            if (0 <= nr < rows and 0 <= nc < cols and
                board[nr][nc] != '#'):
                dfs(nr, nc, current_node)

        # Backtrack
        board[r][c] = char

    # Try starting from every cell
    for i in range(rows):
        for j in range(cols):
            dfs(i, j, trie.root)

    return list(result)

```

### Official Example Walkthrough:

```
**Input:** `board = [ ["o","a","a","n"],
                      ["e","t","a","e"], ["i","h","k","r"], ["i","f","l","v"] ], `
          words = ["oath","pea","eat","rain"]`
```

- Words “oath”, “eat” found → return ["oath","eat"]

Output: ["oath","eat"]

### Complexity:

- **Time:**  $O(M \times N \times 4^L)$  where  $L = \text{max word length}$
  - **Space:**  $O(W \times L)$  — trie size
- 

## Chunk 3: Dynamic Programming on Graphs

We'll now cover the following two advanced problems that combine **Graph Traversal** with **Dynamic Programming**:

- [Longest Increasing Path in a Matrix](#)
- [Minimum Height Trees](#)

These are excellent examples of **DP on graphs**, where we use memoization or iterative pruning (like topological sort) to solve path optimization problems.

---

### Pattern: Dynamic Programming on Graphs

*Used when you need to compute optimal paths, longest/shortest sequences, or values that depend on neighbors — often with constraints like monotonicity.*

### How to Recognize:

- Problem involves a grid or graph where each node has a value.
- You're asked to find the **longest increasing path**, **minimum height tree**, or similar.
- The solution depends on values of neighboring nodes (e.g., `current > neighbor`).
- Recursion with memoization is natural → avoid recomputation.

### Step-by-Step Thinking Process (The Recipe):

1. **Define state:** What does `dp[i][j]` represent? (e.g., longest path starting at `(i,j)`)
2. **Base case:** If no valid next step  $\rightarrow$  return 1.
3. **Transition:** For each neighbor, if condition holds (e.g., `matrix[i][j] < neighbor`), take max of `1 + dp[neighbor]`.
4. **Memoize:** Cache results to avoid recalculating.
5. **Iterate over all starting points** and return maximum.

### Common Pitfalls & Edge Cases:

- Forgetting to **memoize**  $\rightarrow$  exponential time.
  - Misunderstanding directionality (e.g., increasing vs decreasing).
  - Not handling edge cases like single cell, or all equal values.
  - Using recursion without stack limit awareness (use iterative DP or increase recursion limit if needed).
- 

### Problem 1: Longest Increasing Path in a Matrix

#### Summary:

Given an `m x n` matrix of integers, find the length of the longest increasing path where you can only move up/down/left/right, and each step must go to a strictly larger number.

#### Pattern(s):

- DFS + Memoization
- DP on Grid

#### Solution with Inline Comments:

```
def longestIncreasingPath(matrix):  
    if not matrix or not matrix[0]:  
        return 0  
  
    rows, cols = len(matrix), len(matrix[0])
```

```

# Memoization cache: dp[r][c] = longest path starting at (r,c)
memo = {}

# Directions: up, down, left, right
directions = [(0, 1), (0, -1), (1, 0), (-1, 0)]

def dfs(r, c):
    # If already computed, return cached result
    if (r, c) in memo:
        return memo[(r, c)]

    # Base case: start with path length 1 (just current cell)
    max_length = 1

    # Try all 4 neighbors
    for dr, dc in directions:
        nr, nc = r + dr, c + dc

        # Check bounds and increasing condition
        if (0 <= nr < rows and 0 <= nc < cols and
            matrix[nr][nc] > matrix[r][c]):

            # Recursively compute path from neighbor
            length = 1 + dfs(nr, nc)
            max_length = max(max_length, length)

    # Cache result
    memo[(r, c)] = max_length
    return max_length

# Try starting from every cell
result = 0
for i in range(rows):
    for j in range(cols):
        result = max(result, dfs(i, j))

return result

```

### Official Example Walkthrough:

Input:



```
matrix = [
    [9,9,4],
    [6,6,8],
    [2,1,1]
]
```

**Step-by-step:** - Start at  $(0,0) = 9 \rightarrow$  can't go anywhere  $\rightarrow$  path = 1 - Start at  $(0,1) = 9 \rightarrow$  same - Start at  $(0,2) = 4 \rightarrow$  go to  $(1,2) = 8 \rightarrow$  then to  $(1,1) = 6$ ? No  $\rightarrow 8 > 6 \rightarrow$  invalid - But wait:  $4 \rightarrow 8 \rightarrow 8 \rightarrow ? \rightarrow$  nothing bigger  $\rightarrow$  path = 2 - Start at  $(1,0) = 6 \rightarrow$  go to  $(1,1) = 6 \rightarrow$  not greater  $\rightarrow$  skip - Start at  $(1,1) = 6 \rightarrow$  no outgoing - Start at  $(1,2) = 8 \rightarrow$  go to  $(0,2) = 4 \rightarrow$  smaller  $\rightarrow$  invalid - Start at  $(2,0) = 2 \rightarrow$  go to  $(1,0) = 6 \rightarrow$  good  $\rightarrow$  then to  $(0,0) = 9 \rightarrow$  good  $\rightarrow$  then stop - Path:  $2 \rightarrow 6 \rightarrow 9 \rightarrow$  length = 3 - Also:  $2 \rightarrow 6 \rightarrow 8 \rightarrow$  length = 3 - So max = 3

Output: 3

### Complexity:

- **Time:**  $O(m \times n)$  — each cell computed once due to memoization
- **Space:**  $O(m \times n)$  — memoization table + recursion stack (depth  $m \times n$ )

---

## Problem 2: Minimum Height Trees

### Summary:

Given an undirected tree (graph with  $n$  nodes and  $n-1$  edges), find all **root nodes** such that the tree's height is minimized.

A “height” of a tree is the longest path from root to any leaf.

### Pattern(s):

- **BFS (Topological Sort-like)** — trimming leaves iteratively
- **Graph DP / Greedy Pruning**
- **Tree Center Finding**

### Solution with Inline Comments:

```
from collections import deque, defaultdict

def findMinHeightTrees(n, edges):
    # Handle base case
    if n == 1:
        return [0]

    # Build adjacency list and degree array
    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

    # Queue for leaf nodes (degree = 1)
    queue = deque()
    for i in range(n):
        if degree[i] == 1:
            queue.append(i)

    remaining_nodes = n
    # Trim leaves layer by layer until 1 or 2 nodes remain
    while remaining_nodes > 2:
        # Remove current level of leaves
        leaves_count = len(queue)
        remaining_nodes -= leaves_count

        for _ in range(leaves_count):
            leaf = queue.popleft()

            # Remove leaf from its neighbor
            for neighbor in graph[leaf]:
                degree[neighbor] -= 1
                graph[neighbor].remove(leaf)

            # If neighbor becomes leaf, add it
            if degree[neighbor] == 1:
```

```

queue.append(neighbor)

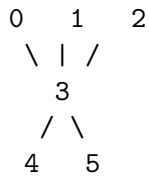
# Remaining nodes are centers (min height roots)
return list(queue)

```

### Official Example Walkthrough:

**Input:**  $n = 6$ ,  $\text{edges} = [[3,0], [3,1], [3,2], [3,4], [5,4]]$

Graph:



- Leaves: 0,1,2,5  $\rightarrow$  degree 1
- First round: remove 0,1,2,5  $\rightarrow$  update neighbors
  - Node 3 loses 3 connections  $\rightarrow$  degree becomes 1
  - Node 4 loses one connection (5)  $\rightarrow$  still degree 1
- Now leaves: 3,4  $\rightarrow$  both have degree 1
- Remaining nodes: 2  $\rightarrow$  stop
- Return [3,4]

But wait: `remaining_nodes > 2`  $\rightarrow$  continue?

Let's simulate: - Initial:  $\text{degree} = [1,1,1,4,1,1]$  - Queue: [0,1,2,5] - Remove them  $\rightarrow$  reduce degree of 3 and 4  $\rightarrow$   $\text{degree}[3]=1$ ,  $\text{degree}[4]=0$ ? Wait:

Wait: edge 5-4  $\rightarrow$  so removing 5  $\rightarrow$  reduces degree of 4 from 2 to 1

Edge 3-0, 3-1, 3-2  $\rightarrow$  remove 0,1,2  $\rightarrow$   $\text{degree}[3]$  goes from 4  $\rightarrow$  1

So after first round: - degree: [0,0,0,1,1,0] - Queue: [3,4]

Now `remaining_nodes = 2`  $\rightarrow$  break loop

Return [3,4]

Output: [3,4]

This makes sense — center nodes minimize height.

**Complexity:**

- **Time:**  $O(n)$  — each edge processed once
  - **Space:**  $O(n)$  — graph, degree, queue
- 

**Problem 3: Clone Graph****Summary:**

Given a reference to a node in a connected undirected graph, return a deep copy (clone) of the graph. Each node has a `val` and a list of its neighbors.

This is a classic **graph traversal + hashing** problem using **DFS or BFS** with a **visited map**.

---

**Pattern(s):**

- Graph Traversal (DFS/BFS)
  - Hash Map for Cloning (Node  $\rightarrow$  Clone Mapping)
  - Deep Copy
- 

**Why It Fits Your List:**

- It's a **core graph problem** like others you've done.
  - Requires **handling adjacency lists** and **avoiding cycles** via visited tracking.
  - Perfect fit for **DFS + memoization** or **BFS + queue**.
-

## Step-by-Step Thinking Process (The Recipe):

1. Use a **hash map** (`old_to_new`) to store mapping from original node to its clone.
2. Start from the given node.
3. If already cloned → return the clone.
4. Otherwise:
  - Create a new node with same `val`.
  - Add to map.
  - Recursively clone all neighbors.
5. Return the cloned node.

Key insight: **Don't re-create nodes you've already cloned** → avoid infinite loops.

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## Solution Template (DFS + Memoization)

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

def cloneGraph(node):
    # Hash map: original node -> cloned node
    old_to_new = {}

    def dfs(n):
        # Base case: if already cloned, return it
        if n in old_to_new:
            return old_to_new[n]

        # Create new node
        clone = Node(n.val)
        old_to_new[n] = clone # mark as cloned

        # Clone all neighbors
        for neighbor in n.neighbors:
            clone.neighbors.append(dfs(neighbor))
```

```
    return clone

return dfs(node) if node else None
```

**Time:**  $O(V + E)$  — each node and edge processed once

**Space:**  $O(V)$  — hash map + recursion stack

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### Official Example Walkthrough:

**Input:**

node = 1, neighbors: [2,4]

node 2: neighbors: [1,3]

node 3: neighbors: [2,4]

node 4: neighbors: [1,3]

This forms a cycle: 1-2-3-4-1

**Steps:** 1. Start at node 1 2. Create clone of 1 → new\_1 3. Clone 2 → new\_2, then clone 3 → new\_3, then clone 4 → new\_4 4. All connections are preserved in new graph 5. Return new\_1

Output: Deep copy of the graph

## LeetCode Interview Cheat Sheet: 10 Core Patterns + Templates

Designed for rapid review before interviews

Covers **Graphs, DP, BFS/DFS, Topo Sort, Union-Find**, and more

Each pattern includes: - How to recognize - Step-by-step thinking recipe - Common pitfalls - Code template (Python) - Example walkthrough (from real problems)

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### 1. DFS / BFS on Grid

**Recognize:**

- 2D grid (`matrix`, `board`)
- Flood fill, island counting, shortest path in unweighted grid

### Thinking Process:

1. Check bounds & visited state
2. Base case: invalid or visited → return
3. Mark current cell as visited
4. Recursively visit neighbors
5. Return result (count, path, etc.)

### Pitfalls:

- Stack overflow (use iterative BFS for large grids)
- Forgetting to mark visited → infinite loop
- Wrong movement directions (4 vs 8)

### Template (DFS):

```
def dfs(grid, r, c, visited):
    if (r < 0 or r >= len(grid) or
        c < 0 or c >= len(grid[0]) or
        grid[r][c] == '0' or (r,c) in visited):
        return

    visited.add((r,c))
    # Do work here
    for dr, dc in [(0,1),(0,-1),(1,0),(-1,0)]:
        dfs(grid, r+dr, c+dc, visited)
```

### Example: **Number of Islands**

→ Count connected '1' groups using DFS.

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## 2. Multi-source BFS

### Recognize:

- Shortest distance to nearest target (e.g., 0s)

- Multiple starting points (e.g., all rotten oranges, all sources)
- Unweighted graph/grid

### Thinking Process:

1. Add all sources to queue with distance 0
2. Process level by level (BFS)
3. Update neighbor distances only if better
4. Stop when queue empty

### Pitfalls:

- Not initializing distances properly
- Revisiting nodes without checking improvement

### Template:

```
queue = deque()
dist = [[float('inf')] * cols for _ in range(rows)]
for each source:
    queue.append((r, c))
    dist[r][c] = 0

while queue:
    r, c = queue.popleft()
    for dr, dc in directions:
        nr, nc = r+dr, c+dc
        if valid and dist[nr][nc] > dist[r][c] + 1:
            dist[nr][nc] = dist[r][c] + 1
            queue.append((nr, nc))
```

### Example: 01 Matrix

→ Find distance to nearest 0.

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### 3. Topological Sort (Kahn's Algorithm)

#### Recognize:

- Tasks with dependencies (“A must come before B”)
- Detect cycle or find valid order
- Graph is directed

#### Thinking Process:

1. Build graph + indegree array
2. Queue all nodes with indegree 0
3. Remove node → reduce indegree of neighbors
4. If all nodes processed → no cycle → valid order

#### Pitfalls:

- Not handling self-loops
- Returning invalid order when cycle exists
- Forgetting to check `len(result) == n`

#### Template:

```
graph = [[] for _ in range(n)]
indegree = [0] * n
for u, v in edges:
    graph[u].append(v)
    indegree[v] += 1

queue = deque([i for i in range(n) if indegree[i] == 0])
result = []

while queue:
    u = queue.popleft()
    result.append(u)
    for v in graph[u]:
        indegree[v] -= 1
        if indegree[v] == 0:
            queue.append(v)
```

```
return result if len(result) == n else []
```

#### Examples:

- [Course Schedule](#)
  - [Alien Dictionary](#)
- 

## 4. Union-Find (DSU)

#### Recognize:

- Connected components
- Cycle detection in undirected graphs
- Merge groups (e.g., accounts, islands)

#### Thinking Process:

1. Initialize parent and rank arrays
2. Use `find(x)` with path compression
3. Use `union(x,y)` by rank
4. Use `find` to check connectivity

#### Pitfalls:

- Not using path compression/rank  $\rightarrow$  slow
- Incorrect union logic (e.g., always attach x to y)

#### Template:

```
class UnionFind:
    def __init__(self, n):
        self.parent = list(range(n))
        self.rank = [0] * n

    def find(self, x):
```

```

    if self.parent[x] != x:
        self.parent[x] = self.find(self.parent[x])
    return self.parent[x]

def union(self, x, y):
    px, py = self.find(x), self.find(y)
    if px == py:
        return False
    if self.rank[px] < self.rank[py]:
        px, py = py, px
    self.parent[py] = px
    if self.rank[px] == self.rank[py]:
        self.rank[px] += 1
    return True

```

#### Examples:

- [Accounts Merge](#)
  - [Graph Valid Tree](#)
- 

## 5. Shortest Path (BFS/Dijkstra)

#### Recognize:

- Minimum steps/distance in weighted/unweighted graph
- Constraints: max stops, limited moves

#### Thinking Process:

- **Unweighted:** BFS (queue)
- **Weighted:** Dijkstra (min-heap)
- Track state: (node, steps, cost)
- Prune if worse than best known

#### Pitfalls:

- Using BFS for weighted graphs → wrong answer
- Not tracking state (e.g., stops, cost)

### Template (Dijkstra):

```
heapq.heappush(heap, (cost, node))
while heap:
    cost, node = heapq.heappop(heap)
    if cost > min_cost[node]: continue
    for neighbor, w in graph[node]:
        new_cost = cost + w
        if new_cost < min_cost[neighbor]:
            min_cost[neighbor] = new_cost
            heapq.heappush(heap, (new_cost, neighbor))
```

### Examples:

- [Word Ladder](#)
  - [Cheapest Flights Within K Stops](#)
- 

## 6. Backtracking + DFS

### Recognize:

- Search for combinations/subsets/permutations
- “Try all paths” — e.g., word search, Sudoku
- One cell used once per path

### Thinking Process:

1. Try placing item at current position
2. Recurse
3. Backtrack: restore state
4. Avoid revisiting same path

### Pitfalls:

- Not marking/unmarking visited → incorrect results
- Exponential time → optimize with pruning

### Template:

```
def backtrack(path, choices):
    if goal_reached:
        result.append(path[:])
        return

    for choice in choices:
        path.append(choice)
        backtrack(path, remaining_choices)
        path.pop() # backtracking
```

### Example: [Word Search II](#)

→ Use Trie + DFS + backtracking

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## 7. Dynamic Programming on Graphs

### Recognize:

- Longest increasing path
- Optimal path based on values
- Recursive dependency on neighbors

### Thinking Process:

1. Define  $dp[r][c]$  = longest path starting at  $(r, c)$
2. Base: 1 (just current cell)
3. Recursion:  $\max(1 + dp[\text{neighbor}])$  if condition holds
4. Memoize to avoid recomputation

### Pitfalls:

- No memoization → exponential time
- Wrong direction (increasing vs decreasing)

### Template:

```
memo = {}
def dfs(r, c):
    if (r,c) in memo: return memo[(r,c)]
    max_len = 1
    for dr,dc in directions:
        nr, nc = r+dr, c+dc
        if valid and matrix[nr][nc] > matrix[r][c]:
            max_len = max(max_len, 1 + dfs(nr, nc))
    memo[(r,c)] = max_len
    return max_len
```

### Example: Longest Increasing Path in Matrix

→ DFS + memoization

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## 8. Tree Center / Minimum Height Trees

### Recognize:

- Find root(s) that minimize tree height
- Only works on trees (acyclic undirected graphs)

### Thinking Process:

1. Trim leaves layer by layer
2. Until 1–2 nodes remain
3. These are the centers

### Pitfalls:

- Not handling base case (n=1)
- Misunderstanding “height” definition

### Template:

```
queue = deque([i for i in range(n) if degree[i] == 1])
while len(nodes) > 2:
    size = len(queue)
    for _ in range(size):
        leaf = queue.popleft()
        for neighbor in graph[leaf]:
            degree[neighbor] -= 1
            if degree[neighbor] == 1:
                queue.append(neighbor)
return list(queue)
```

### Example: Minimum Height Trees

→ Iterative leaf trimming

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## 9. Trie + DFS (Word Search II)

### Recognize:

- Multiple words to search in grid
- Prefix-based pruning needed

### Thinking Process:

1. Build Trie from all words
2. DFS from each cell
3. Prune if prefix not in Trie
4. Collect full words when found

### Pitfalls:

- Not storing full word at leaf
- Not removing duplicates (use set)

### Template:

```
class TrieNode:
    def __init__(self): self.children = {}; self.word = None

def insert(root, word):
    node = root
    for c in word:
        if c not in node.children:
            node.children[c] = TrieNode()
        node = node.children[c]
    node.word = word

def dfs(board, i, j, node, result):
    char = board[i][j]
    if char not in node.children: return
    node = node.children[char]
    if node.word:
        result.add(node.word)
    board[i][j] = '#'
    for di, dj in directions:
        ni, nj = i+di, j+dj
        if 0<=ni<m and 0<=nj<n and board[ni][nj]!='#':
            dfs(board, ni, nj, node, result)
    board[i][j] = char
```

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## 10. Symmetry Optimization (e.g., Knight Moves)

### Recognize:

- Infinite grid
- Symmetric solution space
- Can reduce search to one quadrant

### Thinking Process:

1. Use absolute coordinates
2. Mirror negative indices



3. Limit search space (e.g.,  $x+2$ ,  $y+2$ )
4. BFS with symmetry

### Template:

```
x, y = abs(x), abs(y)
queue = deque([(0,0,0)])
visited = {(0,0)}
while queue:
    r, c, steps = queue.popleft()
    if r == x and c == y: return steps
    for dr,dc in knight_moves:
        nr, nc = r+dr, c+dc
        nr, nc = abs(nr), abs(nc)
        if (nr,nc) not in visited and nr <= x+2 and nc <= y+2:
            visited.add((nr,nc))
            queue.append((nr,nc,steps+1))
```

### Example: Minimum Knight Moves

→ Use symmetry to reduce state space

### Final Tips for Interviews

| Pattern          | When to Use                                |
|------------------|--|
| DFS/BFS Grid     | Island count, flood fill, shortest path    |
| Multi-source BFS | Distance to nearest target                 |
| Topo Sort        | Task scheduling, dependency resolution     |
| Union-Find       | Connected components, cycle detection      |
| Dijkstra/BFS     | Shortest path with constraints             |
| Backtracking     | Word search, permutation generation        |
| DP on Graphs     | Longest increasing path, optimal sequences |
| Tree Center      | Minimize tree height                       |
| Trie + DFS       | Multiple word search                       |
| Symmetry         | Infinite grid problems                     |