323. Number of Connected Components in an Undirected Graph

**Depth-First Search Breadth-First Search Union Find** 

### **Problem Description**

edges[i] = [a\_i, b\_i] represents an undirected edge between nodes a\_i and b\_i in the graph. The goal is to determine the number of connected components in the graph. A connected component is a set of nodes in a graph that are connected to each other by paths, and those nodes are not connected to any other nodes outside of the component. The task is to return the total count of such connected components.

Intuition

In this problem, we have a graph that consists of n nodes. We are given an integer n and an array edges where each element

#### The objective is to count the number of separate groups of interconnected nodes where there are no connections between the

Medium

groups into one larger group. Therefore, we are looking to find the total number of distinct sets that cannot be merged any further.

To address this problem, we use the Union-Find algorithm (also known as Disjoint Set Union or DSU). The Union-Find algorithm is efficient in keeping track of elements which are split into one or more disjoint sets. It has two primary operations: find, which determines the identity of the set containing a particular element, and union, which merges two sets together.

groups. We can think of each node as starting in its own group, and each edge as a connection that, potentially, merges two

own set (group).

2. Union Operation: We loop through the list of edges. For each edge, we apply the union operation. This means we find the parents (representatives) of the two nodes connected by the edge. If they have different parents, we set the parent of one to

Initialization: We start by initializing each node to be its own parent, representing that each node is the representative of its

- be the parent of the other, effectively merging the two sets.

  The function find(x) is a recursive function that finds the root (or parent) of the set to which x belongs. It includes path
- compression, which means that during the find operation, we make each looked-up node point directly to the root. Path compression improves efficiency, making subsequent find operations faster.

Counting Components: After processing all the edges, we iterate through the nodes and count how many nodes are their

own parent. This count is equal to the number of connected components since nodes that are their own parent represent the

This solution's beauty is that it is relatively simple but powerful, allowing us to solve the connectivity problem in nearly linear time complexity, which is very efficient.

Solution Approach

understanding this approach is recognizing how these functions work together to determine the number of connected components.

• Parent Array: An array p is used to keep track of the representative (or parent) for each node in the graph. Initially, each node's representative is

The implementation of the solution follows the Union-Find algorithm, using two primary functions: find and union. The key to

#### 1. **The Find Function**: The find function is used to find the root (or parent) of the node in the graph. When calling find(x), the

def find(x):

**Data Structure Used:** 

itself, meaning p[i] = i.

**Algorithms and Patterns Used:** 

The solution has three main parts:

root of a connected component.

function checks if p[x] != x. If this condition is true, it means that x is not its own parent, and there is a path to follow to find the root. This is done recursively until the root is found. The line p[x] = find(p[x]) applies path compression by directly connecting x to the root of its set, which speeds up future find operations.

Union Operation: The union operation is not explicitly defined but is performed within the loop that iterates over the edges

Counting Components: Finally, the number of connected components is determined by counting the number of nodes that

are their own parents (i.e., the roots). This is done by iterating through each node i in the range 0 to n-1 and checking if i ==

find(i). If this condition is true, it means that i is the representative of its component, contributing to the total count.

The solution approach effectively groups nodes into sets based on the connections (edges) between them, and then it identifies

the unique sets to arrive at the total number of connected components. Each connected component is represented by a unique

root node, which is why counting these root nodes gives the correct count for the connected components.

**Initialization**: We start with each node being its own parent, hence p = [0, 1, 2, 3, 4].

## array. For each edge, the roots of the nodes a and b are found using the find operation. If they have different roots, it means they belong to different sets and should be connected. The line p[find(a)] = find(b) effectively merges the two sets by

for a, b in edges:

p[find(a)] = find(b)

**if** p[x] != x:

return p[x]

p[x] = find(p[x])

setting the parent of a's root to be b's root.

return sum(i == find(i) for i in range(n))

Union Operation: We process the edges one by one:

p[find(0)] = find(1) resulting in p = [1, 1, 2, 3, 4].

- Example Walkthrough

  Let's use a small example to illustrate the solution approach. Imagine we have n=5 nodes and the following edges: edges = [[0,1], [1,2], [3,4]].
  - find(2), leading to p = [1, 1, 1, 3, 4].
    Lastly, for the edge [3,4], we find the parents of 3 and 4, which are 3 and 4. Again, they are different, so we unite them, resulting in p = [1, 1, 1, 4, 4].
    Counting Components: Now we count the number of nodes that are their own parents, which corresponds to the root nodes:

In this example, we have two nodes (1 and 4) that are their own parents, so the total number of connected components is 2.

o For the edge [0,1], we find the parents of 0 and 1, which are 0 and 1, respectively. Since they are different, we connect them by setting

Next, for the edge [1,2], we find the parents of 1 and 2, which are 1 and 2. They are different, so we connect them by setting p[find(1)] =

Python

**Solution Implementation** 

def find(node):

parent = list(range(n))

for a, b in edges:

parent = new int[n];

int count = 0;

parent[i] = i;

for (int[] edge : edges)

for (int i = 0; i < n; ++i) {

union(vertex1, vertex2);

for (int i = 0; i < n; ++i) {

// Find function with path compression

private void union(int node1, int node2) {

if (parent[node] != node) {

count++;

private int find(int node) {

/\*\*

class Solution:

0 is not its own parent (its parent is 1).

2 is not its own parent (its parent is 1).

3 is not its own parent (its parent is 4).

1 is the parent of itself, so it counts as a component.

4 is the parent of itself, so it counts as a component.

if parent[node] != node:
 parent[node] = find(parent[node])
 return parent[node]

return sum(i == find(i) for i in range(n))

public int countComponents(int n, int[][] edges) {

// For each edge, perform a union of the two vertices

return count; // Return the total count of connected components

return parent[node]; // Return the root parent of the node

int root1 = find(node1); // Find the root parent of the first node

int root2 = find(node2); // Find the root parent of the second node

parent[root1] = root2; /\* Make one root parent the parent of the other \*/

// Union function to join two subsets into a single subset

\* Counts the number of connected components in an undirected graph

function countComponents(n: number, edges: number[][]): number {

// Function to find the root of the set that 'x' belongs to

// by counting the number of nodes that are their own parent

def countComponents(self, n: int, edges: List[List[int]]) -> int:

# Function to find the root of a node using path compression

// Perform union by setting the parent of the representative of 'a'

// Parent array to represent the disjoint set forest

// Path compression for efficiency

parent[x] = find(parent[x]);

// Union the sets that the edges connect

// to the representative of 'b'

// Count the number of connected components

parent[find(a)] = find(b);

 $* @param {number} n - The number of nodes in the graph$ 

\* @return {number} - The number of connected components

\* @param {number[][]} edges - The edges of the graph

// Initialize each node to be its own parent

let parent: number[] = new Array(n);

function find(x: number): number {

**if** (parent[x] !== x) {

return parent[x];

let count = 0;

return count;

def find(node):

class Solution:

**Time Complexity** 

for (const [a, b] of edges) -

for (let i = 0; i < n; ++i) {

if (i === find(i)) {

count++;

for (let i = 0; i < n; ++i) {

parent[i] = i;

int vertex1 = edge[0], vertex2 = edge[1];

parent[find(a)] = find(b)

def countComponents(self, n: int, edges: List[List[int]]) -> int:

# Function to find the root of a node using path compression

# Initialization of parent list where each node is its own parent initially

// Initialize parent array, where initially each node is its own parent

# Count the number of components by checking how many nodes are their own parents

// Count the number of components by counting the nodes that are their own parents

parent[node] = find(parent[node]); // Path compression for efficiency

if  $(i == find(i)) \{ // If the node's parent is itself, it's the root of a component$ 

```
Java
class Solution {
    private int[] parent; // This array will hold the parent for each node representing the components
```

# Union operation: join two components by pointing the parent of one's root to the other's root

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C++
#include <vector>
#include <numeric>
#include <functional>
class Solution {
public:
    // Function to count the number of connected components in an undirected graph
    int countComponents(int n, vector<vector<int>>& edges) {
       // Parent array to represent disjoint sets
       vector<int> parent(n);
       // Initialize each node to be its own parent, forming n separate sets
        iota(parent.begin(), parent.end(), 0);
       // Lambda function to find the representative (leader) of a set
        function<int(int)> find = [&](int x) -> int {
            if (parent[x] != x) {
                // Path compression: update the parent of x to be the parent of its current parent
                parent[x] = find(parent[x]);
            return parent[x];
       // Iterate through all edges to merge sets
        for (auto& edge : edges) {
            int nodeA = edge[0], nodeB = edge[1];
            // Union by rank not used, simply attach the tree of `nodeA` to `nodeB`
            parent[find(nodeA)] = find(nodeB);
       // Count the number of sets by counting the number of nodes that are self-parented
        int componentCount = 0;
        for (int i = 0; i < n; ++i) {
            if (i == find(i)) { // If the node is the leader of a set
                componentCount++;
       return componentCount;
};
TypeScript
```

compression in the find function and the union operation in the loop iterating through the edges.

2. **Union Operation (The for loop)** - The time to traverse all edges and perform the union operation for each edge. There are 'e' edges, with e being the length of the edges list. As the path compression makes the union operation effectively near-constant

The time complexity of the given code is primarily derived from two parts: the union-find operation that involves both path

Path Compression (find function) - The recursive find function includes path compression which optimizes the time

complexity of finding the root representative of a node to  $0(\log n)$  on average and  $0(\alpha(n))$  in the amortized case, where  $\alpha$  is

the inverse Ackermann function, which grows very slowly. For most practical purposes, it can be considered almost constant.

3. **Summation Operation** - The final sum iterates over all nodes to count the number of components, taking <code>0(n)</code> time.

# Space Complexity

Parent Array (p) - It's a list of length n, so it uses O(n) space.

Recursive Stack Space - Because the find function is recurs

The space complexity consists of:

time, this also takes O(e) operations.

- 2. **Recursive Stack Space** Because the find function is recursive (though with path compression), in the worst case, it could go as deep as O(n) before path compression is applied. However, with path compression, this is drastically reduced.
- Given these considerations, the overall space complexity is 0(n).

Considering all parts, the overall time complexity is 0(e + n).