1971. Find if Path Exists in Graph

vertex and check if the destination vertex can be reached.

Breadth-First Search Union Find Graph

Problem Description

Easy

Depth-First Search

the graph is bi-directional. It's also specified that each pair of vertices can be connected by at most one edge and that no vertex is connected to itself with an edge. Our goal is to find out if there exists a path from a given source vertex to a destination vertex. If such a path exists, the function should return true; otherwise, it should return false.

In this problem, we are given a bi-directional graph consisting of n vertices, labeled from 0 to n - 1. The connections or edges

between these vertices are provided in the form of a 2D array called edges, where each element of the array represents an edge

connecting two vertices. For example, if edges [i] = [u, v], there is an edge connecting vertex u to vertex v and vice versa since

Intuition

such as Depth-First Search (DFS) or Breadth-First Search (BFS). These methods can explore the graph starting from the source

The intuitive approach to determine if a valid path exists between two vertices in a graph is by using graph traversal methods

However, the solution provided uses a different approach known as Disjoint Set Union (DSU) or Union-Find algorithm, which is an efficient algorithm to check whether two elements belong to the same set or not. In the context of graphs, it helps determine if two vertices are in the same connected component, i.e., if there is a path between them.

The Union-Find algorithm maintains an array p where p[x] represents the parent or the representative of the set to which x belongs. In the given solution, the find function is a recursive function that finds the representative of a given vertex x. If x is the representative of its set, it returns x itself; otherwise, it recursively calls itself to find x's representative and performs path

compression along the way. Path compression is an optimization that flattens the structure of the tree by making every node directly point to the representative of the set, which speeds up future operations.

Then, in the solution, every edge [u, v] is considered, and the vertices u and v are united by setting their representatives to be

By using Union-Find, we can avoid the need to explicitly traverse the graph while still being able to determine connectivity between vertices efficiently.

The solution provided employs the Union-Find (Disjoint Set Union) algorithm. The implementation of this algorithm consists of

The algorithm uses an array named p where p[i], initially, is set to i for all vertices 0 to n-1. This step constitutes the make-set

representative (parent) of the set that x belongs to. This is done by checking if p[x] is equal to x. If p[x] != x, then x is not the

the same. In the end, it checks whether the source and the destination vertices have the same representative. If they do, it

means they are connected, and thus, a valid path exists between them, returning true; else, it returns false.

operation where each vertex is initially the parent of itself, meaning each vertex starts in its own set.

two main operations: find and union. In the context of the problem, Union-Find helps to efficiently check if there is a path between two vertices in the graph.

The find function is then defined. This function takes an integer x, which represents a vertex, and it recursively finds the

Solution Approach

representative of its set, so the function is called recursively with p[x]. During recursion, path compression is performed by setting p[x] = find(p[x]). Path compression is a crucial optimization as it helps to reduce the time complexity significantly by flattening the structure of the tree.

Next, a loop iterates over each edge in the edges list. For each edge [u, v], the union operation is performed implicitly by setting p[find(u)] = find(v). This essentially connects the two vertices u and v by ensuring they have the same representative. By performing this operation for all edges in the graph, all connected components in the graph are merged.

Finally, the solution checks whether there is a valid path between source and destination. This is done by comparing their representatives: if find(source) == find(destination), then source and destination are in the same connected component, signifying that there is a path between them, and hence true is returned. If the representatives are different, the function returns false, implying there is no valid path. It is worth noting that the Union-Find algorithm is particularly effective for problems involving connectivity queries in a static

graph, where the graph does not change over time. This algorithm allows such queries to be performed in nearly constant time,

Let's consider a small graph with 5 vertices (n = 5) labeled from 0 to 4. The edges array is provided as follows: edges = [[0, 1], [1, 2], [3, 4]]

can visualize the graph: 0 --- 1 --- 2 3 --- 4

1. Union operation on edge [0, 1]: set the parent of the representative of 1 (p[1]) to be the representative of 0 (p[0]). Hence, p[1] = 0.

2. Union operation on edge [1, 2]: set the parent of the representative of 2 (p[2]) to be the representative of 1 (which was set to 0 previously).

Our goal is to determine if there is a path between the source vertex 0 and the destination vertex 3. From the edges given, we

p = [0, 1, 2, 3, 4]

Example Walkthrough

After step 1, our updated p array is: p = [0, 0, 2, 3, 4]

We would use the Union-Find algorithm for this.

Initially, all vertices are their own parents:

making it a powerful tool for solving such problems.

```
After step 2, our updated p array is:
```

After step 3, our updated p array is:

Find the representative of the source vertex 0:

Find the representative of the destination vertex 3:

Hence, p[2] = 0.

p = [0, 0, 0, 3, 4]

p = [0, 0, 0, 3, 3]

We start by uniting the vertices that have an edge between them using the union operation.

```
Now we check if there is a valid path between the source vertex 0 and the destination vertex 3 by comparing their
representatives.
```

Since the representatives of vertex 0 and vertex 3 are different (0 is not equal to 3), we conclude there is no path between them,

connected component, and vertices 3 and 4 form another component. There's no edge that connects these two components,

3. Union operation on edge [3, 4]: set the parent of the representative of 4 (p[4]) to be the representative of 3 (p[3]). Hence, p[4] = 3.

and the function should return false. It is evident from the p array and the disconnected components in the graph visualization that vertices 0, 1, and 2 are in one

confirming our conclusion.

find(0) will return 0 since p[0] is 0.

find(3) will return 3 since p[3] is 3.

```
Python
```

def find_root(node: int) -> int:

for start_node, end_node in edges:

from typing import List

class Solution:

Java

class Solution {

private int[] parent;

parent = new int[n];

// source - starting vertex

parent.resize(n);

// destination — ending vertex

parent[i] = i;

for (auto& edge : edges)

function initializeParent(n: number): void {

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

function unionSet(x: number, y: number): void {

// Function to perform the union operation on two subsets

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

function find(x: number): number {

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

const rootX = find(x);

const rootY = find(y);

if (rootX !== rootY) {

parent[rootX] = rootY;

return parent[x];

parent[i] = i;

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

parent[i] = i;

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

Solution Implementation

```
if parent[node] != node:
        parent[node] = find_root(parent[node])
    return parent[node]
# Initialize parent pointers for each node to point to itself.
parent = list(range(n))
```

def validPath(self, n: int, edges: List[List[int]], source: int, destination: int) -> bool:

Union the sets by updating the root of one to the root of the other.

// Parent array that stores the root of each node's tree in the disjoint set forest

public boolean validPath(int n, int[][] edges, int source, int destination) {

bool validPath(int n, vector<vector<int>>& edges, int source, int destination) {

// Iterate through all edges to perform the union operation

// Check if the source and destination have the same root parent

// Function to initialize the parent array with each vertex as its own parent

// Function to find the root parent of a vertex 'x' using path compression

unionSet(find(edge[0]), find(edge[1]));

return find(source) == find(destination);

// If they do, there is a valid path between them

// Helper function to find the root parent of a vertex x

// Initialize parent array so that each vertex is its own parent initially

Helper function to find the root of a node 'x' in the disjoint set.

It also applies path compression to optimize future lookups.

Iterate through each edge and perform union operation.

parent[find_root(start_node)] = find_root(end_node)

Check if the source and destination are in the same set.

return find_root(source) == find_root(destination)

If the find_root of both is the same, they are connected.

```
// Union operation: merge the sets containing the two nodes of each edge
        for (int[] edge : edges) {
            parent[find(edge[0])] = find(edge[1]);
       // If the source and destination nodes have the same parent/root, they are connected; otherwise, they are not
        return find(source) == find(destination);
   // Method to find the root of the set that contains node x utilizing path compression for efficiency
   private int find(int x) {
       if (parent[x] != x) { // If x is not its own parent, it's not the representative of its set
            parent[x] = find(parent[x]); // Recurse to find the root of the set and apply path compression
       return parent[x]; // Return the root of the set that contains x
C++
class Solution {
public:
   // Parent vector to represent the disjoint set (union-find) structure
   vector<int> parent;
   // Main function to check if there is a valid path between source and destination
   // Parameters:
   // n - number of vertices
   // edges - list of edges represented as pairs of vertices
```

// Method to determine if there is a valid path between the source and the destination nodes within an undirected graph

// Initialize the parent array where each node is initially its own parent (representative of its own set)

```
int find(int x) {
       // Path Compression: Recursively makes the parents of the vertices
       // along the path from x to its root parent point directly to the root parent
       if (parent[x] != x)
            parent[x] = find(parent[x]);
        return parent[x];
    // Helper function to perform the union operation on two subsets
    void unionSet(int x, int y) {
       // Find the root parents of the vertices
       int rootX = find(x);
       int rootY = find(y);
       // Union by setting the parent of rootX to rootY
       if (rootX != rootY)
            parent[rootX] = rootY;
};
TypeScript
// Global parent array to represent the disjoint set (union-find) structure
const parent: number[] = [];
```

```
// Main function to check if there is a valid path between 'source' and 'destination'
  function validPath(n: number, edges: number[][], source: number, destination: number): boolean {
      initializeParent(n);
      // Perform the union operation for each edge to connect the vertices in the union-find structure
      for (const edge of edges) {
          unionSet(find(edge[0]), find(edge[1]));
      // If the source and destination have the same root parent, a valid path exists
      return find(source) === find(destination);
from typing import List
class Solution:
   def validPath(self, n: int, edges: List[List[int]], source: int, destination: int) -> bool:
       # Helper function to find the root of a node 'x' in the disjoint set.
       # It also applies path compression to optimize future lookups.
       def find_root(node: int) -> int:
            if parent[node] != node:
               parent[node] = find_root(parent[node])
            return parent[node]
       # Initialize parent pointers for each node to point to itself.
        parent = list(range(n))
       # Iterate through each edge and perform union operation.
        for start_node, end_node in edges:
            # Union the sets by updating the root of one to the root of the other.
            parent[find_root(start_node)] = find_root(end_node)
       # Check if the source and destination are in the same set.
       # If the find_root of both is the same, they are connected.
        return find_root(source) == find_root(destination)
```

Time Complexity:

Time and Space Complexity

The time complexity of this algorithm can be considered as $0(E * \alpha(N))$ for the Union-Find operations, where E is the number of edges and $\alpha(N)$ is the Inverse Ackermann function which grows very slowly and is nearly constant for all practical values of N. The reason for this time complexity is that each union operation, which combines the sets containing u and v, and each find operation,

The given Python code represents a solution to check if there's a valid path between the source and destination nodes in an

undirected graph. The code utilizes the Union-Find algorithm, also known as the Disjoint Set Union (DSU) data structure.

which finds the root of the set containing a particular element, takes $\alpha(N)$ time on average. The find function uses path compression which flattens the structure of the tree by making every node point to the root

which uses find, becomes nearly constant. Given that there are E iterations to process the edges array, the time complexity of the for loop would be $O(E * \alpha(N))$. Additionally, there are two more find operations after the for loop, but these do not significantly affect the overall time complexity, as they also work in $\alpha(N)$ time.

whenever find is used on it. Because of path compression, the average time complexity of find, and thus the union operation

Space Complexity: The space complexity of the code is O(N), where N is the number of nodes in the graph. This is because we maintain an array p

Therefore, the space complexity of maintaining this array is linear with respect to the number of nodes.

that holds the representative (parent) for each node, and it's of size equal to the number of nodes.