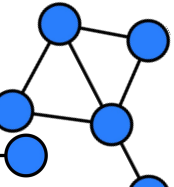
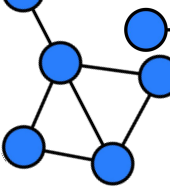
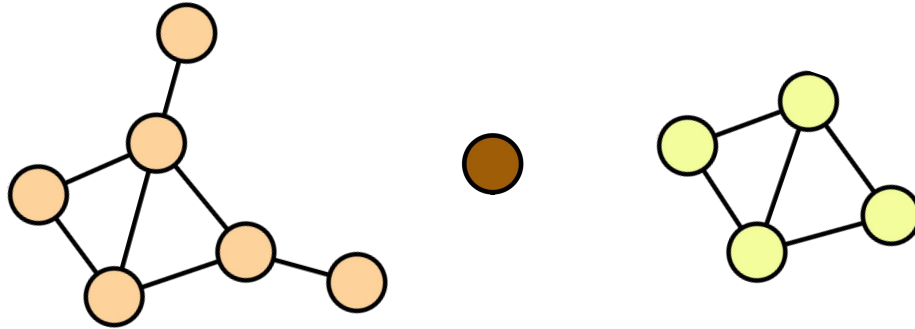
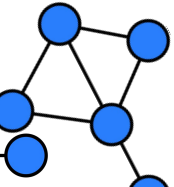
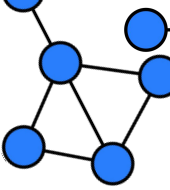


Union Find (Disjoint Set Union)



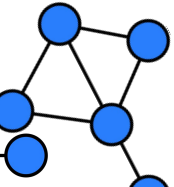
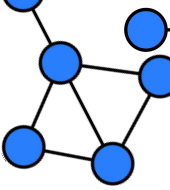
Objectives

- Understand Union Find basics
- Learn union and find operations
- Understand path compression
- Apply in problem-solving
- Analyze time and space complexities



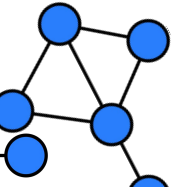
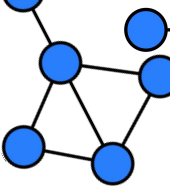
Lecture Flow

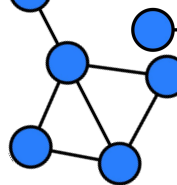
- 1) Pre-requisites
- 2) Problem definition
- 3) Brute force approach
- 4) Improved brute force approach
- 5) Union Find
- 6) Union Find optimization
- 7) Solve problem
- 8) Things to pay attention
- 9) Practice problems
- 10) Quote of the day



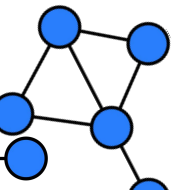
Pre-requisites

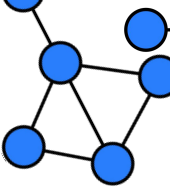
- Graph
- Connected components
- Cycles in graphs
- Set (intersection, union)





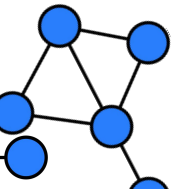
Problem Definition

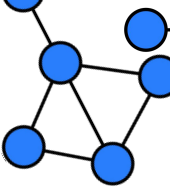




You have a graph of **cities and roads** connecting the cities.

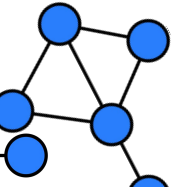
You are given **multiple queries** and in one query, you are asked to **determine if two cities are connected**.





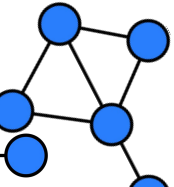
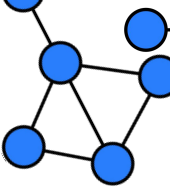
How would you approach this?

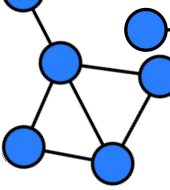
We need a quick way to determine if two nodes are connected.



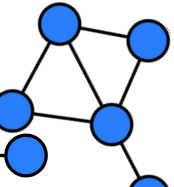
Brute Force Approach

- Build a graph
- For every query,
 - make one node a source and
 - the other a destination.
 - do graph traversal to check if path exists
- Time complexity: $O(\text{queries} * (V + E))$



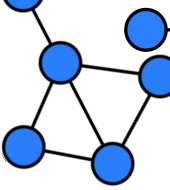
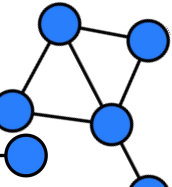


Can we do better?

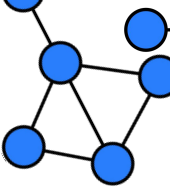


Improved Brute Force Approach

- What if we **group connected cities together?**
- For quicker look up we **use Set** to group
- We use dictionary to track which group each city belongs to
 - **City : Group**



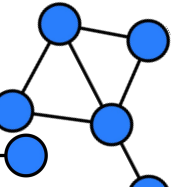
Improved Brute Force Approach (cont...)



- To check whether there is a connection between City A and City B

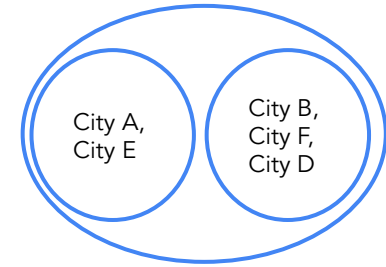
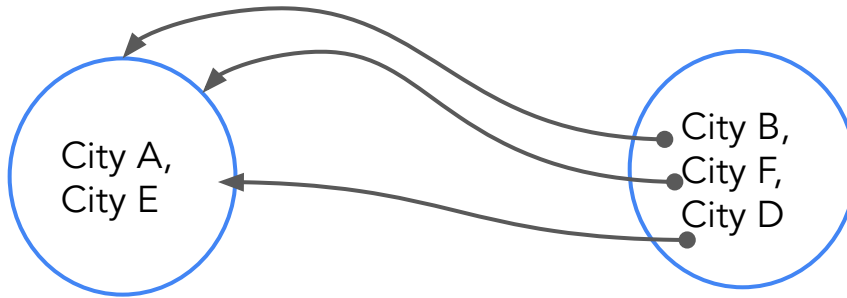
```
return dictionary[city_a] is dictionary[city_b]
```

- If both have the same Set reference, then they must belong to the same group

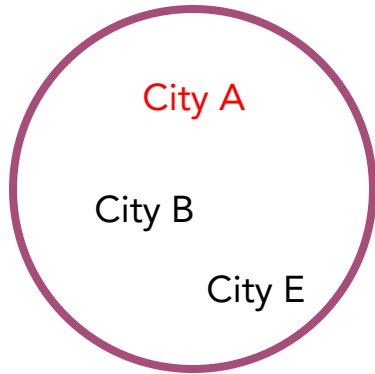
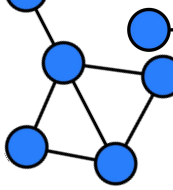


Improved Brute Force Approach (cont...)

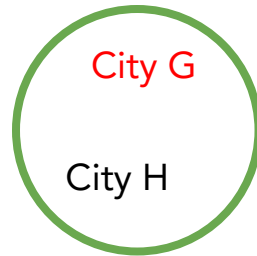
- For merging City A and City B, we'd need to **change the dictionary reference for every city merged** with one of the cities which **adds extra time complexity**.



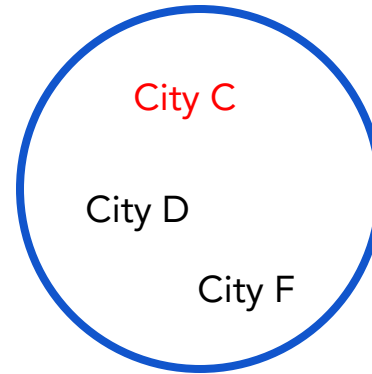
Improved Brute Force Approach (cont...)



Set 1



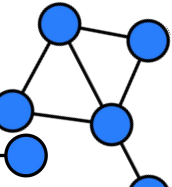
Set 2



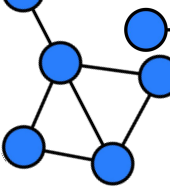
Set 3

Dictionary:

City A: Set_1
City B: Set_1
City E: Set_1
City G: Set_2
City H: Set_2
City C: Set_3
City D: Set_3
City F: Set_3



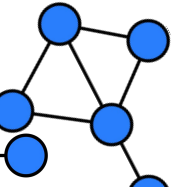
Improved Brute Force Approach (cont...)



- We don't need the entire set to check if two cities belong to the same connected component; we are only checking reference.
- In that case, instead of using sets as an identifier of a group, we can select a representative element for each set.

Dictionary:

City A: Set_1
City B: Set_1
City E: Set_1
City G: Set_2
City H: Set_2
City C: Set_3
City D: Set_3
City F: Set_3



Improved Brute Force Approach (cont...)

City A belongs to Set_1
City B belongs to Set_1

City A represented by City A
City B represented by City A

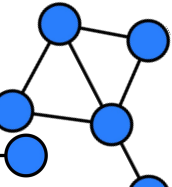
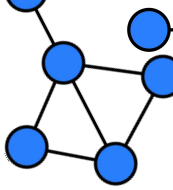
Dictionary:

City A: Set_1
City B: Set_1
City E: Set_1
City G: Set_2
City H: Set_2
City C: Set_3
City D: Set_3
City F: Set_3

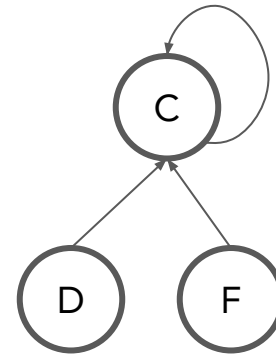
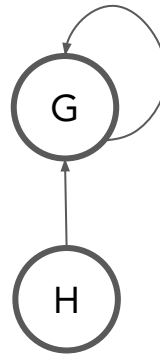
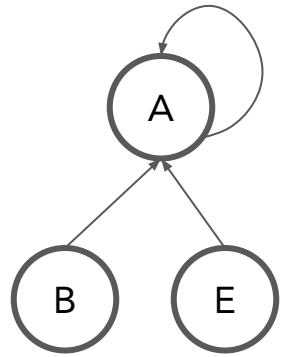


Dictionary:

City A: City A
City B: City A
City E: City A
City G: City G
City H: City G
City C: City C
City D: City C
City F: City C

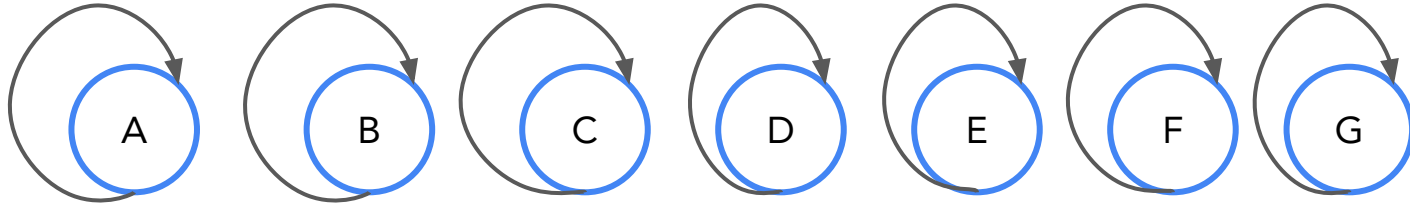


What does it mean to connect two cities?



Arrows pointing to their representative

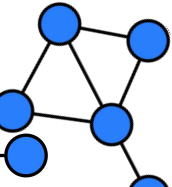
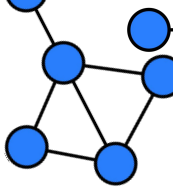
What does it mean to connect two cities?



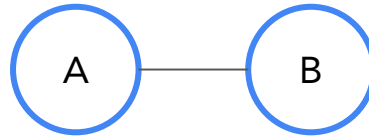
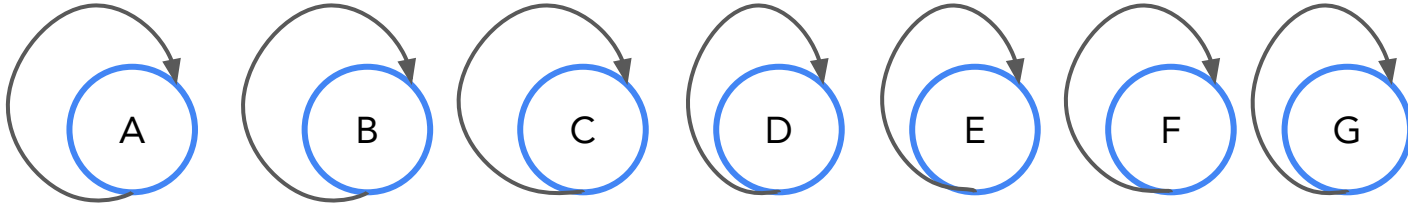
First, all nodes represent themselves

Dictionary:

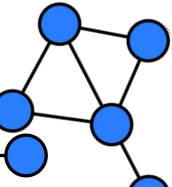
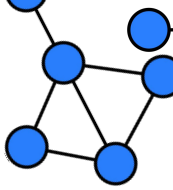
City A: City A
City B: City B
City E: City E
City G: City G
City H: City H
City C: City C
City D: City D
City F: City F



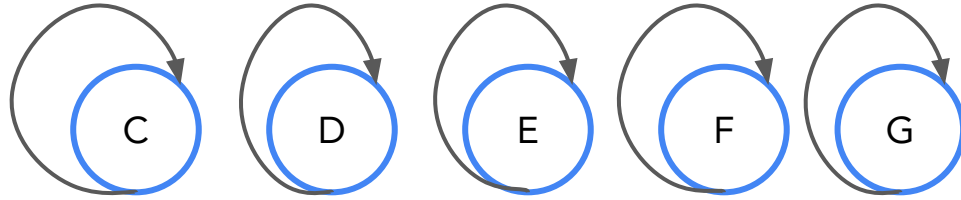
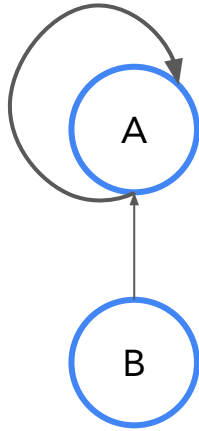
What does it mean to connect two cities?



Connect Node A with Node B?



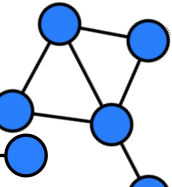
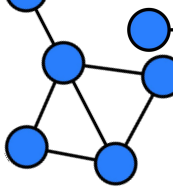
What does it mean to connect two cities?



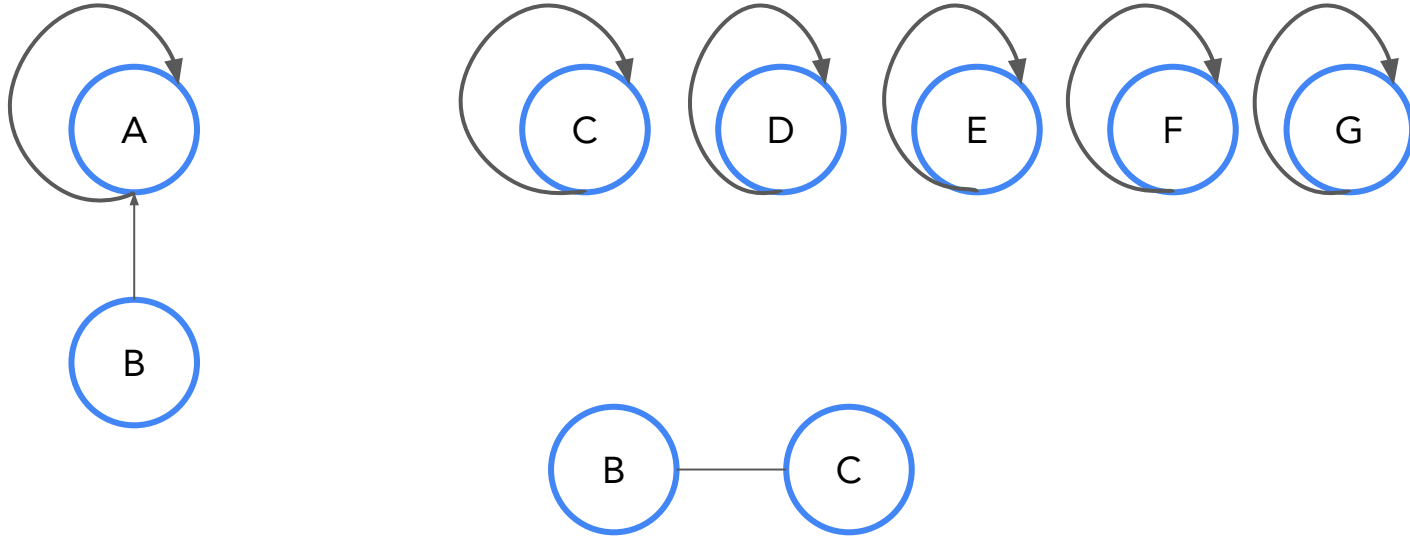
1. Pick representative
2. Then, assign both of them the same representative

Dictionary:

City A: City A
City B: City A
City E: City E
City G: City G
City H: City H
City C: City C
City D: City D
City F: City F

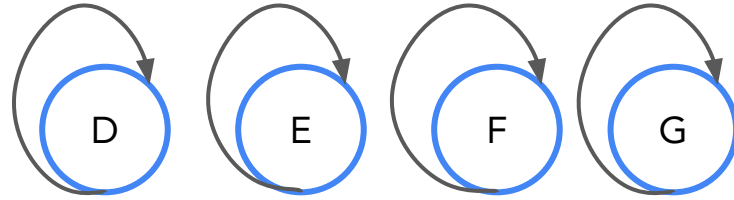
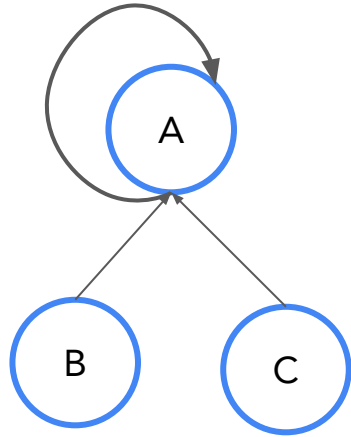


What does it mean to connect two cities?



Connect Node C with Node B?

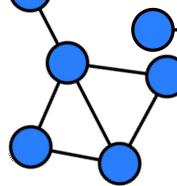
What does it mean to connect two cities?



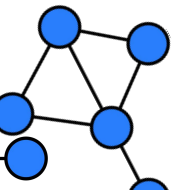
1. Find the representative of the group where B belongs
2. Assign C's representative to B's representative

Dictionary:

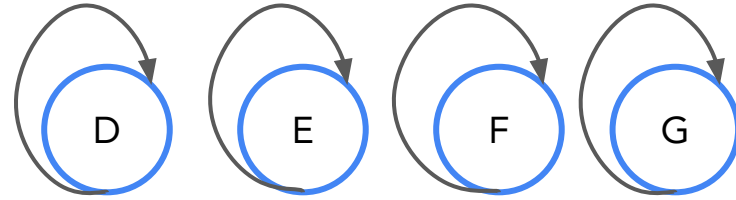
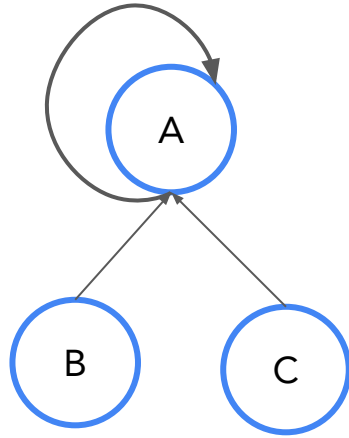
City A: City A
City B: City A
City E: City E
City G: City G
City H: City H
City C: City A
City D: City D
City F: City F

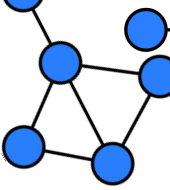


In summary,
assign the same representative to both



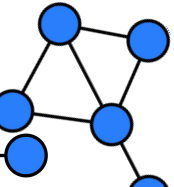
How do we check if they belong to same group?





Check if they have the same
representative

```
dictionary[city_a] == dictionary[city_b]
```





Implement Here

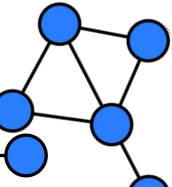
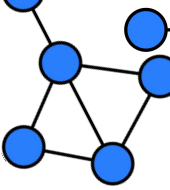


Time & Space Complexity Analysis

Time complexity:

- Union: $O(V)$
- Find: $O(1)$
- Connected: $O(1)$

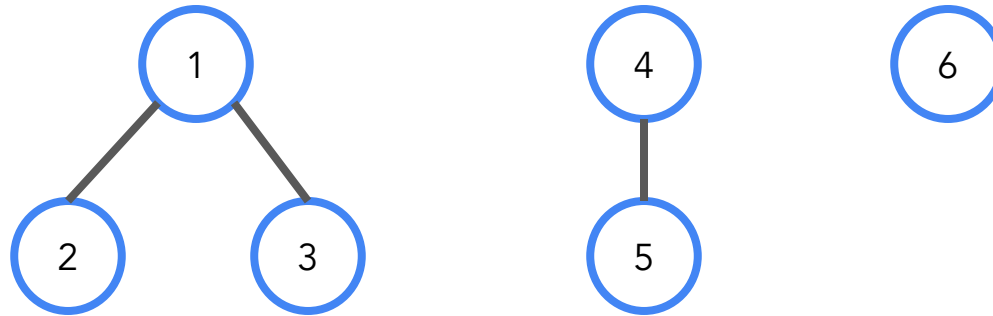
Space Complexity: $O(V)$

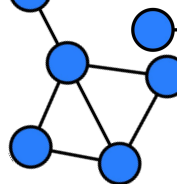


Topic Introduction

Union-find is a way to **group objects** and efficiently **determine if two objects belong in the same group**, as well as **join two groups**.

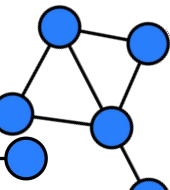
It is also known as **Disjoint Set** and **Merge Find Set**.

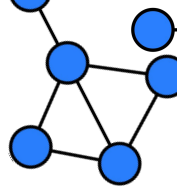




Problems solved using Union-Find

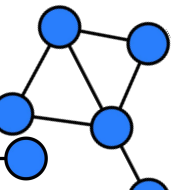
- Check if **path between two nodes exists** in undirected graph in **$< O(N)$ time**
- **Count connected components** in undirected graph
- Detecting **cycles** in an undirected graph
- **Merging** sets efficiently
- Kruskal's **Minimum Spanning Tree** algorithm



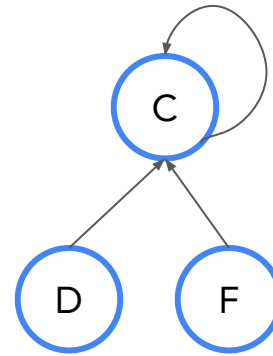
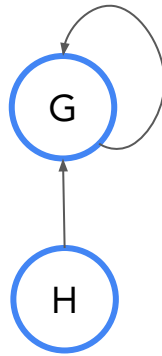
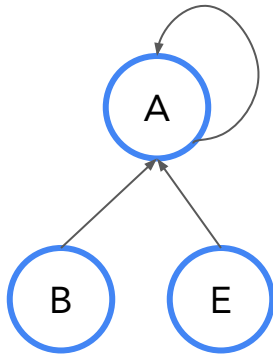


Union Find implementation
can be **optimized**.

Let's brainstorm



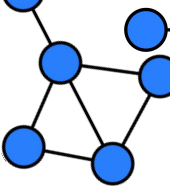
Quick Find: Make all nodes point to their representative



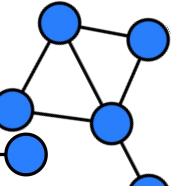
Dictionary:

City A: City A
City B: City A
City E: City A
City G: City G
City H: City G
City C: City C
City D: City C
City F: City C

Note: What happens during UNION
for Quick Find?

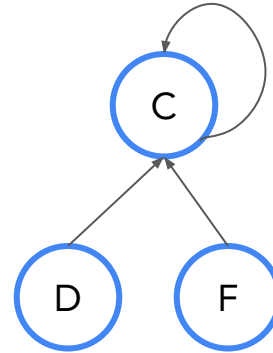
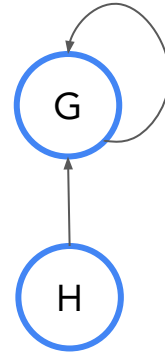
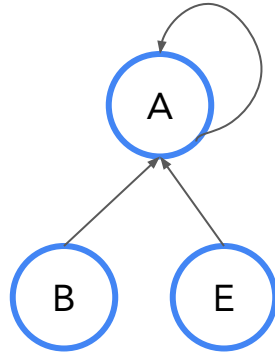


Let's explore Quick Union



Quick Union Steps

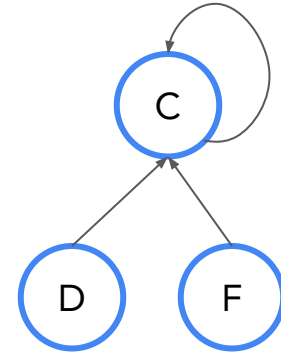
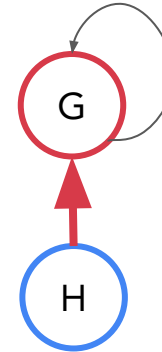
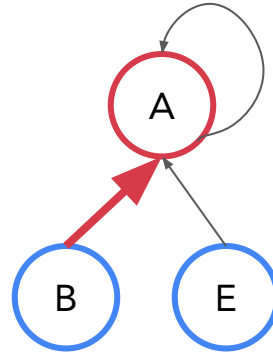
Union H <> B



Quick Union Steps

Union H <> B

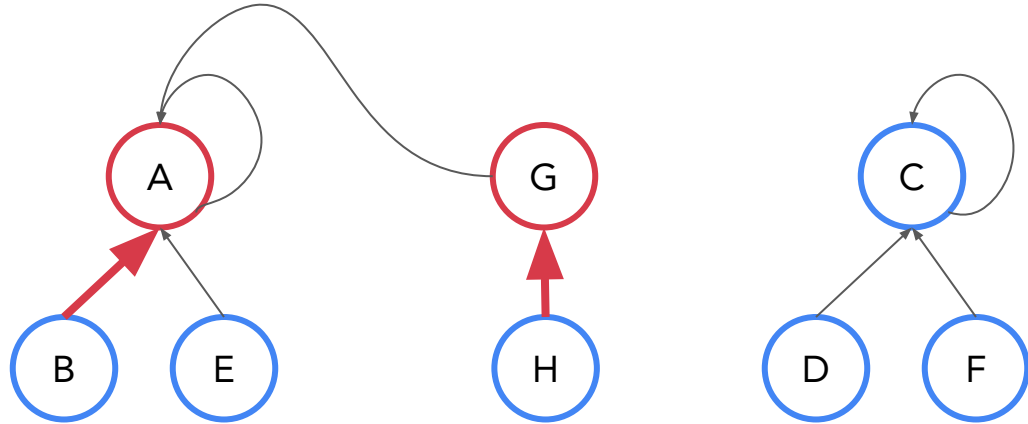
1. Find representative of H
2. Find representative of B
3. ?



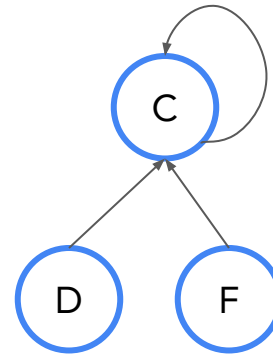
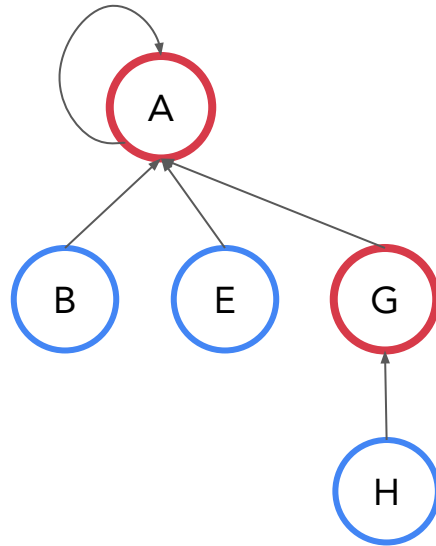
Quick Union Steps

Union H <> B

1. Find representative of H
2. Find representative of B
3. Make G's representative B's Representative



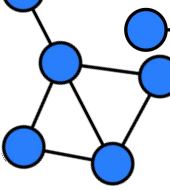
Quick Union H <> B



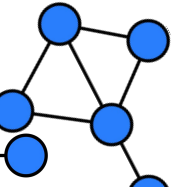
Dictionary:

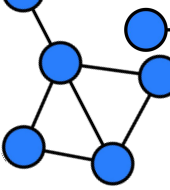
City A: City A
City B: City A
City E: City A
City G: City A
City H: City G
City C: City C
City D: City C
City F: City C

Note: Parent of H is not updated immediately as in Quick Find.

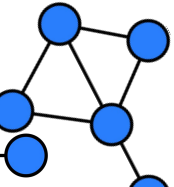


Implement Quick Union Here

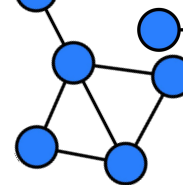




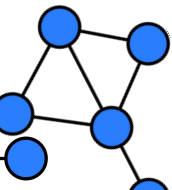
Which one is faster,
Quick Union or Quick Find?

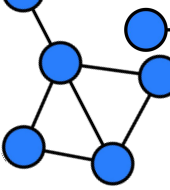


Quick Union is faster with small few tweaks

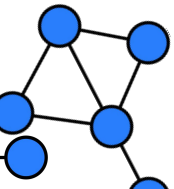


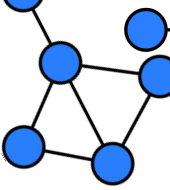
- Quick-union is generally considered better than quick-find because:
 - The union operation can be **optimized using weighting or height**
 - The find operation can be **optimized using path compression**



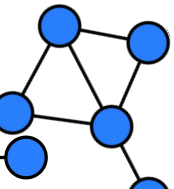


How can we improve Quick Union?



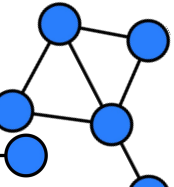
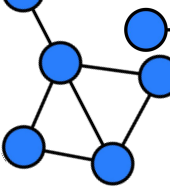


When we merge two sets,
what **relevant information** should be considered
when deciding **who to make a representative**?



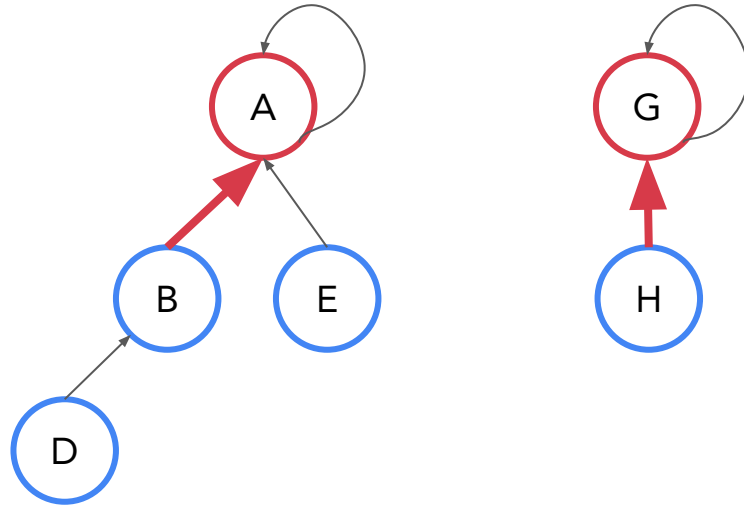
When merging

- Optimize by considering the size of the group
- Merging the smaller group into the larger one prevents an increase in tree depth
- A shallower tree improves the time complexity of the find operation

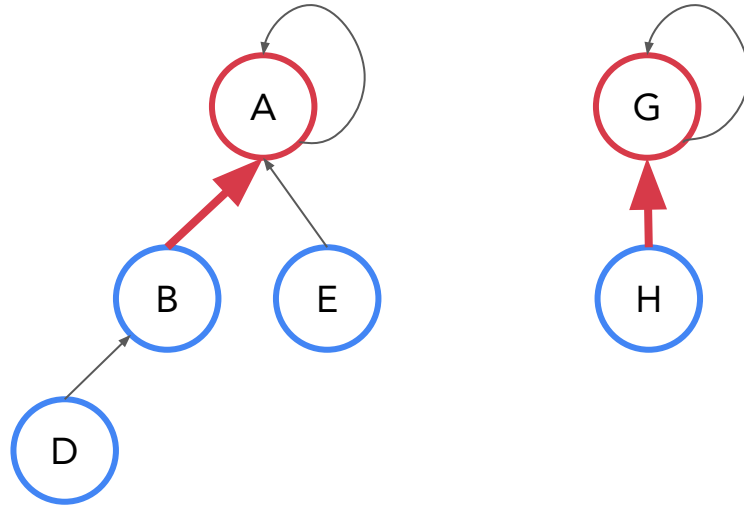


Let's use the **size**

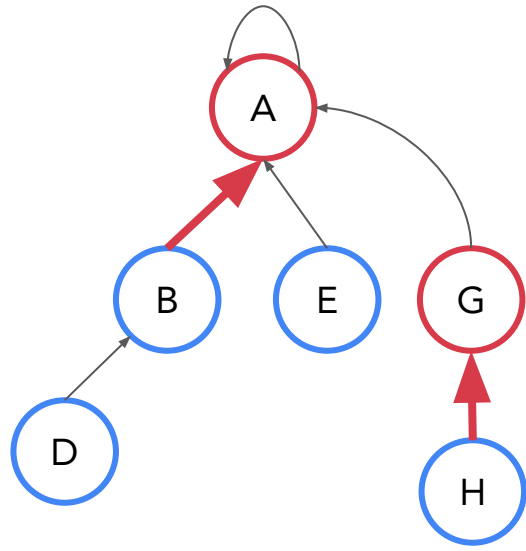
- Size is the number of members in the group



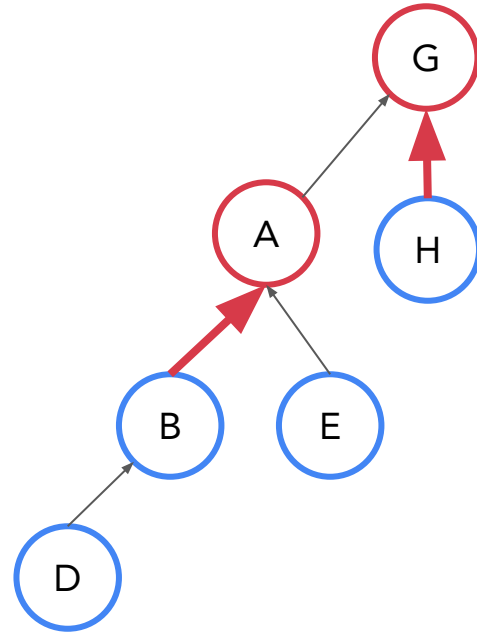
Union H <> B



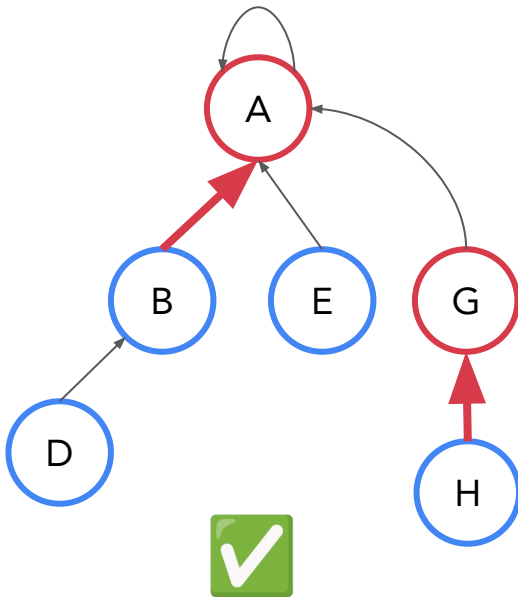
Union $H \leftrightarrow B$, which merging is better?



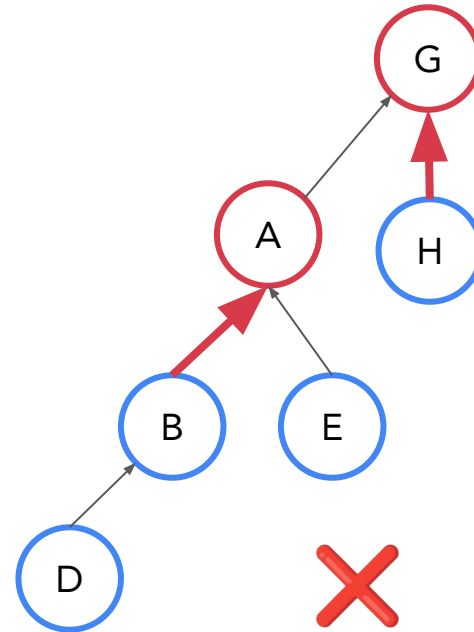
OR

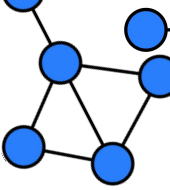


Which one would take longer to find the parent for node 'D'?

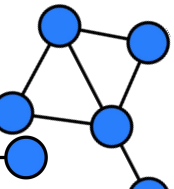


OR



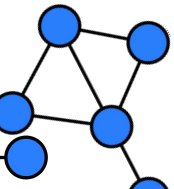
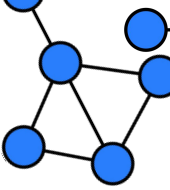


Implement Union by Size Here

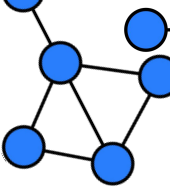


Union by Rank

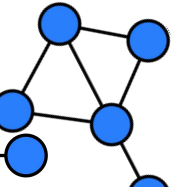
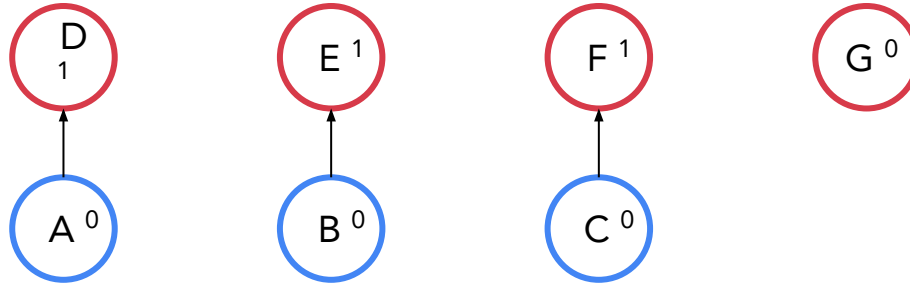
- Rank is an **upper bound** for its height.
- Initially, its rank is **set to zero**.
- When merging, first compare their ranks:
 - If the ranks are **different**, then **the larger rank tree becomes the parent**, and the ranks **do not change**.
 - If the ranks are the **same**, then either one can become the parent, but **the new parent's rank is incremented by one**.



Union by Rank



- After **union**(A, D), **union**(B, E), and **union**(C, F):

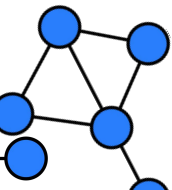
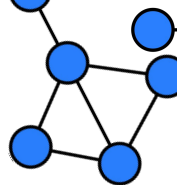
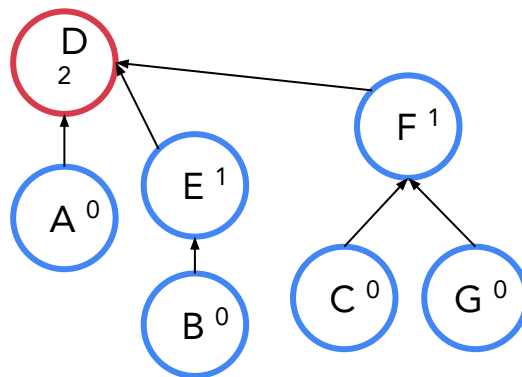


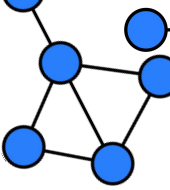
Union by Rank

- After `union(C, G)`, and `union(E, A)`:

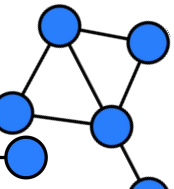


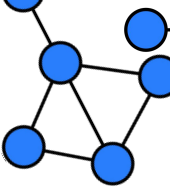
- After `union(B, G)`:



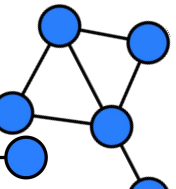


Implement Union by Rank Here





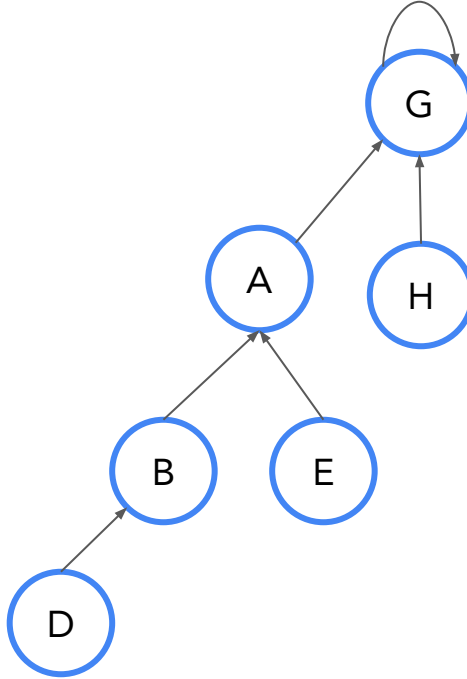
Can we improve Find?



Path Compression

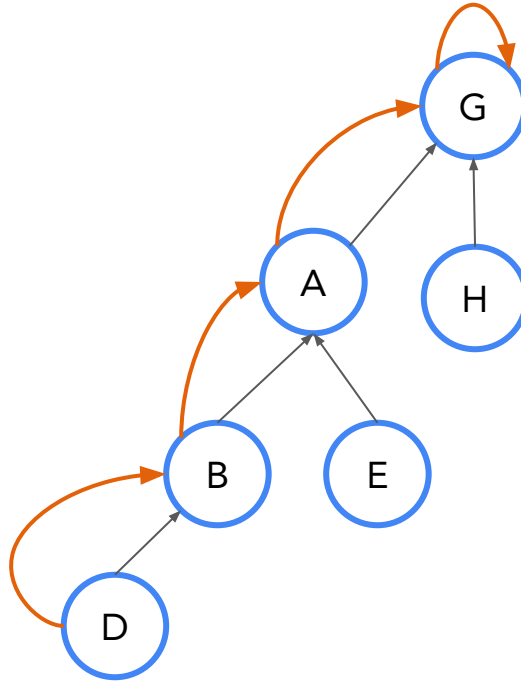
Who is the
representative of **D**?

How would we find it?

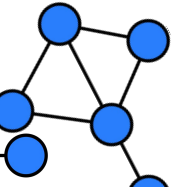
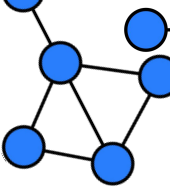


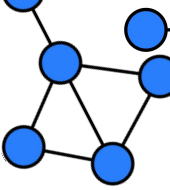
Path Compression

Who is the representative of **D**?

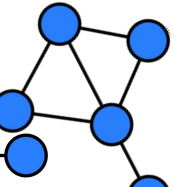


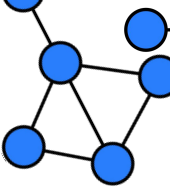
We would traverse up the tree until the root node





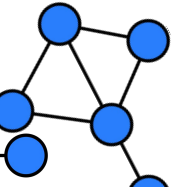
What information have we gained by going up the tree?



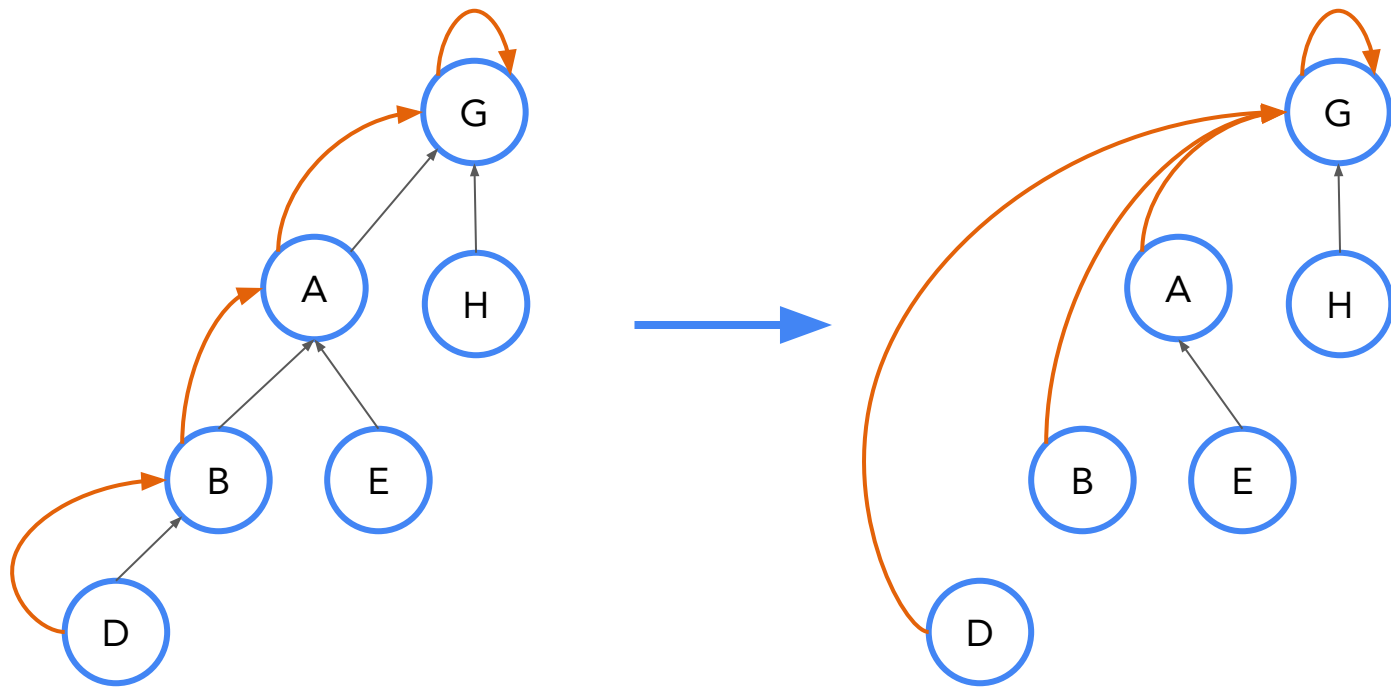


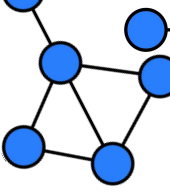
All the children we touched going up have
the root as their representative

So, why not update them accordingly?

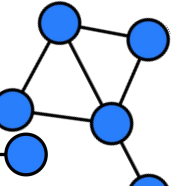


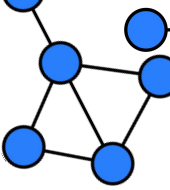
Path Compression



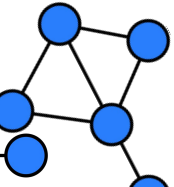


Implement Path Compression for Find

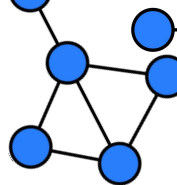




What if we use **path compression** and **union by size (rank)** together?



Union by Size with Path Compression



```
class UnionFind:
    def __init__(self, size):
        self.root = [i for i in range(size)]
        self.size = [1] * size

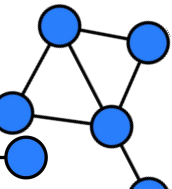
    def find(self, x):
        if x == self.root[x]:
            return x
        self.root[x] = self.find(self.root[x])
        return self.root[x]

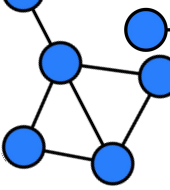
    def union(self, x, y):
        rootX = self.find(x)
        rootY = self.find(y)
        if rootX != rootY:
            if self.size[rootX] > self.size[rootY]:
                self.root[rootY] = rootX
                self.size[rootX] += self.size[rootY]
            else:
                self.root[rootX] = rootY
                self.size[rootY] += self.size[rootX]
```

Path Compression

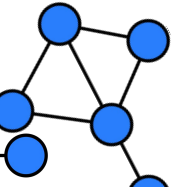


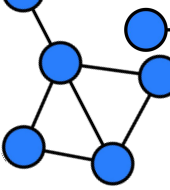
- $O(\text{find}) = \text{amortized } 0(?)$
- $O(\text{union}) = \text{amortized } 0(?)$
- $O(\text{constructor}) = 0(?)$





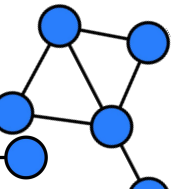
Is there an **iterative** way of **path**
compression?





Yes, there are.

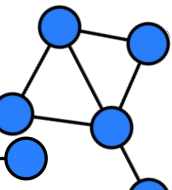
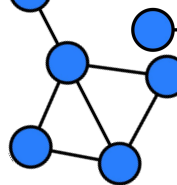
You can implement the previous code iteratively
or you can use Path Halving

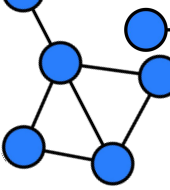


Path Halving

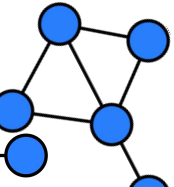
- Makes every other node on the find path link to its **grandparent**.
- Retains **the same worst-case complexity** but are **more efficient in practice**.
- Simple to implement:

```
def find(self, x):  
    while x != self.root[x]:  
        x = self.root[x] = self.root[self.root[x]]  
    return x
```





Let's solve a problem



Number of Provinces

Description

Editorial

Solutions (3.5K)

Submissions

547. Number of Provinces



Medium



7.5K



287



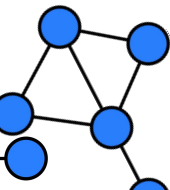
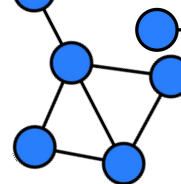
Companies

There are n cities. Some of them are connected, while some are not. If city a is connected directly with city b , and city b is connected directly with city c , then city a is connected indirectly with city c .

A **province** is a group of directly or indirectly connected cities and no other cities outside of the group.

You are given an $n \times n$ matrix `isConnected` where `isConnected[i][j] = 1` if the i^{th} city and the j^{th} city are directly connected, and `isConnected[i][j] = 0` otherwise.

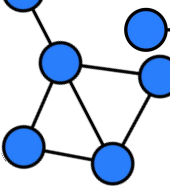
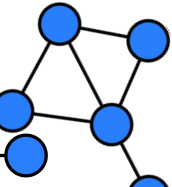
Return the total number of **provinces**.



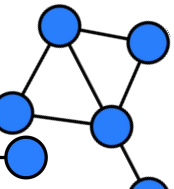
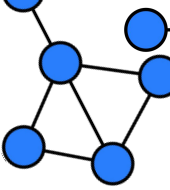
Time and Space Complexity

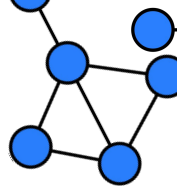
Time complexity: $O(n^2)$

Space Complexity: $O(n)$

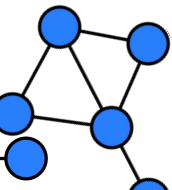


Checkpoint



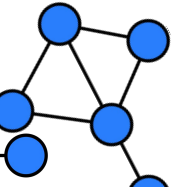
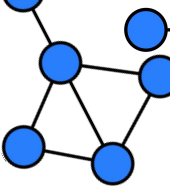


Things to pay attention to



Pitfall 1: Improper initialization

```
class UnionFind:
    def __init__(self, size):
        self.parent = [0] * size # Incorrect initialization
        self.rank = [0] * size
        self.count = size
```

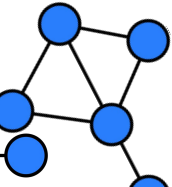
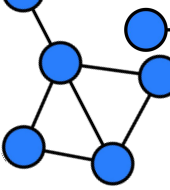


Pitfall 1: Improper initialization

```
class UnionFind:
    def __init__(self, size):
        # Incorrect initialization
        self.parent = [0] * size
        self.rank = [0] * size
        self.count = size
```



```
class UnionFind:
    def __init__(self, size):
        # Correct initialization
        self.parent = [i for i in range(size)]
        self.rank = [0] * size
        self.count = size
```

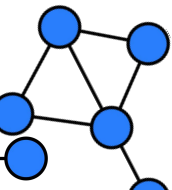
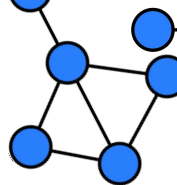


Pitfall 2: Forgetting to use path compression

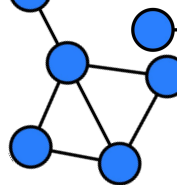
```
class UnionFind:
    def __init__(self, size):
        . . .

    def find(self, member):
        # No path compression implemented
        while member != self.parent[member]:
            member = self.parent[member]
        return member

    def union_set(self, member1, member2):
        . . .
```



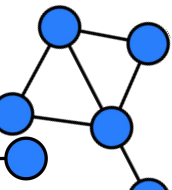
Pitfall 2: Forgetting to use path compression



```
class UnionFind:
    def __init__(self, size):
        . . .

    def find(self, member):
        # No path compression implemented
        while member != self.parent[member]:
            member = self.parent[member]
        return member

    def union_set(self, member1, member2):
        . . .
```



```
class UnionFind:
    def __init__(self, size):
        . . .

    def find(self, member):
        # Path compression implemented
        root = member
        while root != self.parent[root]:
            root = self.parent[root]

        while member != root:
            parent = self.parent[member]
            self.parent[member] = root
            member = parent

        return root
```

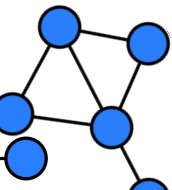
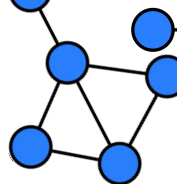


Pitfall 3: Forgetting to use rank-based union

```
class UnionFind:
    def __init__(self, size):
        . . .

    def find(self, member):
        . . .

    def union_set(self, member1, member2):
        root1 = self.find(member1)
        root2 = self.find(member2)
        # No rank-based union implemented
        self.parent[root1] = root2
```



Pitfall 3: Forgetting to use rank-based union

```
class UnionFind:
    def __init__(self, size):
        . . .

    def find(self, member):
        . . .

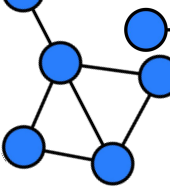
    def union_set(self, member1, member2):
        root1 = self.find(member1)
        root2 = self.find(member2)
        # No rank-based union implemented
        self.parent[root1] = root2
```



```
class UnionFind:
    . . .
    # rank-based union implemented
    def union(self, x, y):
        rootX = self.find(x)
        rootY = self.find(y)
        if rootX != rootY:
            if self.rank[rootX] > self.rank[rootY]:
                self.root[rootY] = rootX
            elif self.rank[rootX] < self.rank[rootY]:
                self.root[rootX] = rootY
            else:
                self.root[rootY] = rootX
                self.rank[rootX] += 1
        . . .
```



Practice Questions



Check if There is a Valid Path in a Grid

Redundant Connection

Regions Cut by Slashes

Accounts Merge

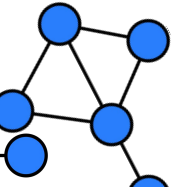
Most Stones Removed with same Row or Column

Satisfiability of Equality Equations

Checking Existence of Edge Length Limited Paths

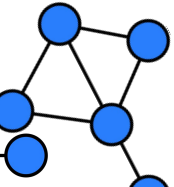
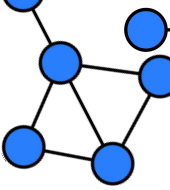
Remove Max Number of Edges to Keep Graph Fully Traversable

Minimize Malware Spread



Resources and more readables

- [Codeforce EDU Course](#)
- [Leetcode Explore Card](#)



In union there is strength.

Aesop

ASV
Africa To Silicon Valley