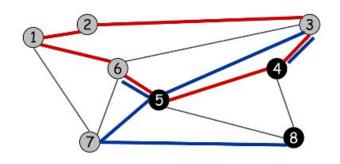
Greedy algorithms

CMPSC 465 - Yana Safonova

The minimum spanning tree

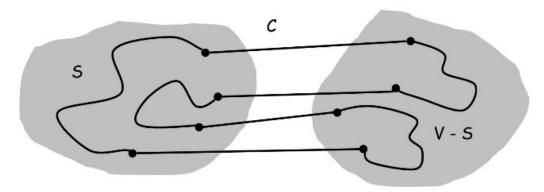
Kruskal's algorithm - proof of correctness

Claim. A cycle and a cutset intersect in an even number of edges.



Cycle C = 1-2, 2-3, 3-4, 4-5, 5-6, 6-1 Cutset D = 3-4, 3-5, 5-6, 5-7, 7-8 Intersection = 3-4, 5-6

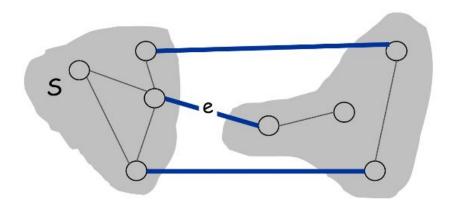
Pf. (by picture)



A cycle has to enter and leave the cut the same amount of times

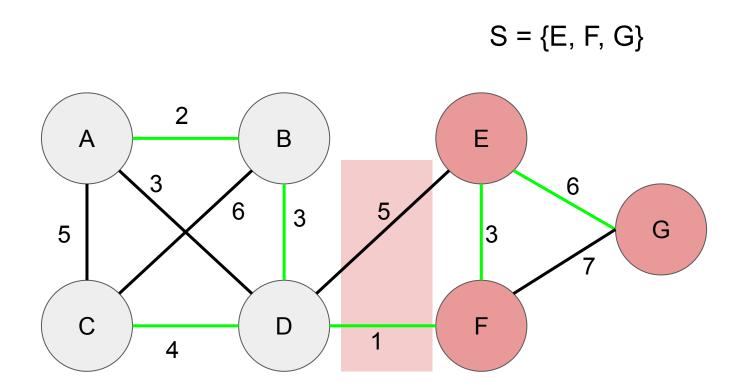
Kruskal's algorithm - proof of correctness

Cut property. Let S be any subset of nodes, and let e be the min cost edge with exactly one endpoint in S. Then the MST contains e.



e is in the MST

The cut property



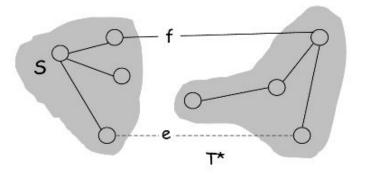
Cut property. Let S be any subset of nodes, and let e be the min cost edge with exactly one endpoint in S. Then the MST contains e.

Kruskal's algorithm - proof of correctness

Cut property. Let S be any subset of nodes, and let e be the min cost edge with exactly one endpoint in S. Then the MST T* contains e.

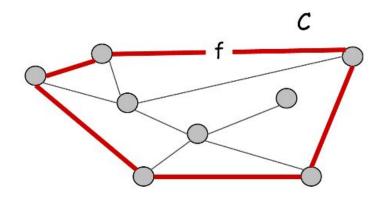
Pf. (exchange argument)

- Suppose e does not belong to T*, and let's see what happens.
- Adding e to T* creates a cycle C in T*.
- Edge e is both in the cycle C and in the cutset D corresponding to S ⇒ there exists another edge, say f, that is in both C and D.
- $T' = T^* \cup \{e\} \{f\}$ is also a spanning tree.
- Since c_e < c_f, cost(T') < cost(T*).
- This is a contradiction.



Kruskal's algorithm - proof of correctness

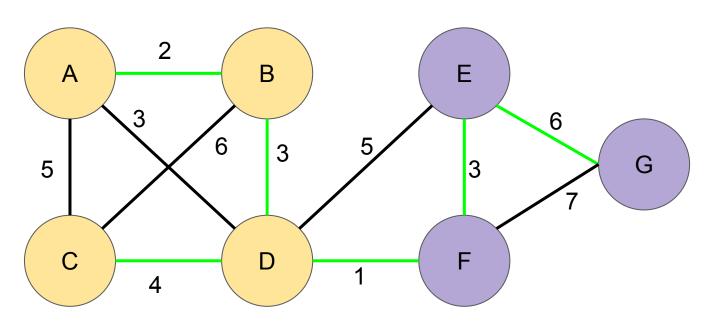
Cycle property. Let C be any cycle, and let f be the max cost edge belonging to C. Then the MST does not contain f.



f is not in the MST

The cycle property

Cycle property. Let C be any cycle, and let f be the max cost edge belonging to C. Then the MST does not contain f.

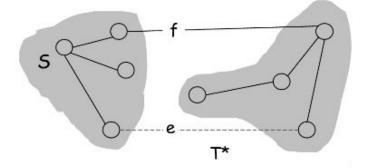


Kruskal's algorithm - proof of correctness

Cycle property. Let C be any cycle in G, and let f be the max cost edge belonging to C. Then the MST T* does not contain f.

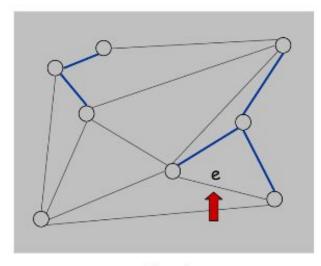
Pf. (exchange argument)

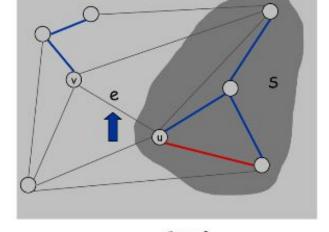
- Suppose f belongs to T*, and let's see what happens.
- Deleting f from T* creates a cut S in T*.
- Edge f is both in the cycle C and in the cutset D corresponding to S ⇒ there exists another edge, say e, that is in both C and D.
- $T' = T^* \cup \{e\} \{f\}$ is also a spanning tree.
- Since c_e < c_f, cost(T') < cost(T*).
- This is a contradiction.



Kruskal's algorithm - proof of correctness

- Consider edges in ascending order of weight.
- Case 1: If adding e to T creates a cycle, discard e according to cycle property.
- Case 2: Otherwise, insert e = (u, v) into T according to cut property where S = set of nodes in u's connected component.





Case 1 Case 2

Kruskal's algorithm - implementation

Implementation. Use the union-find data structure.

- Build set T of edges in the MST.
- Maintain set for each connected component.

Disjoint sets or union-find sets or merge-find sets

What is a disjoint set?

A data structure that is also known as the **disjoint set** or the **merge-find set**

Example:

There are 5 people: A, B, C, D, and E.

A is B's friend, B is C's friend, and D is E's friend, therefore, the following is true:

- A, B, and C are connected to each other
- D and E are connected to each other

{A, B, C} and {D, E} are two disjoint sets

How can we quickly compute if two people are in the same friend group?

Disjoint set - description

Operations:

Union(A, B): Connect two elements A and B

Find(A, B): Find whether the two elements A and B are connected

Example:

A set of elements $S = \{0, 1, 2, 3, 4, 5, 6, 7, 8, 9\}.$

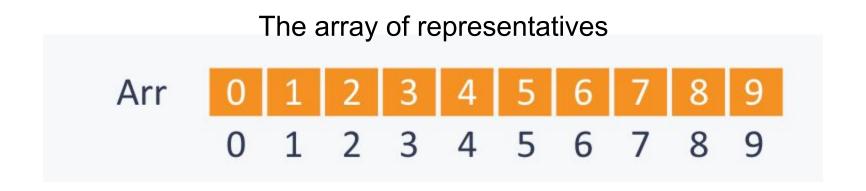
Arr[] that is indexed by elements of sets, which are of size N (N = 10) is used for the operations of *union* and *find*.

Assumption:

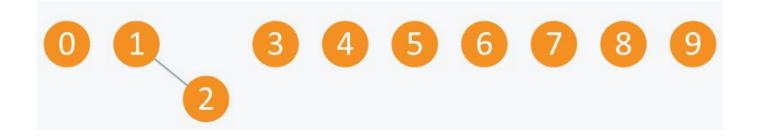
A and B objects are connected only if arr[A] = arr[B].

Disjoint set



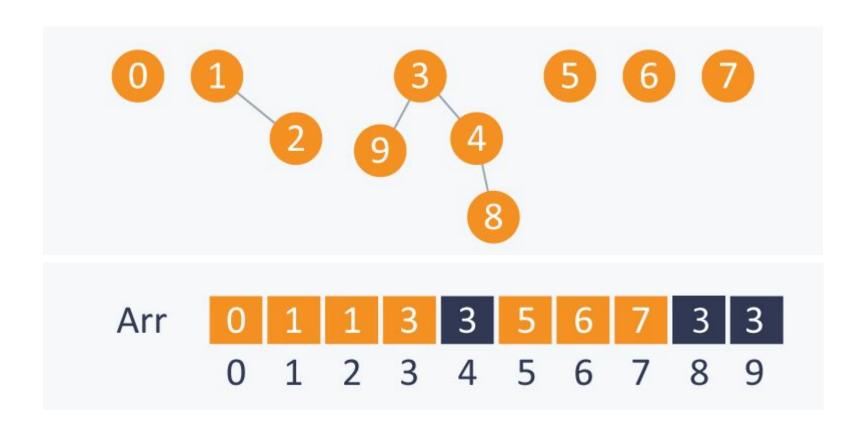


Disjoint set - Union(2, 1)

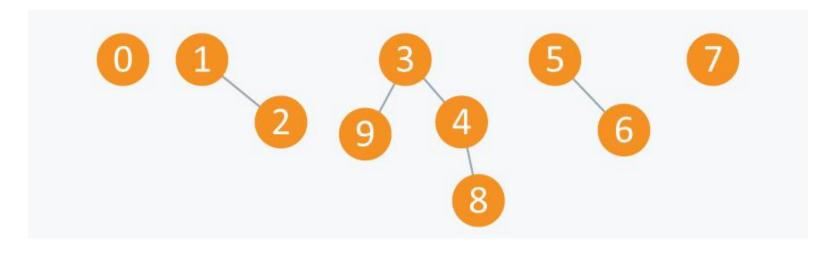


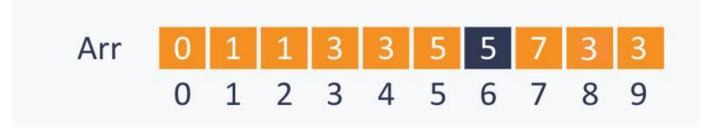


Disjoint set - Union(4, 3), Union(8, 4), Union(9, 3)

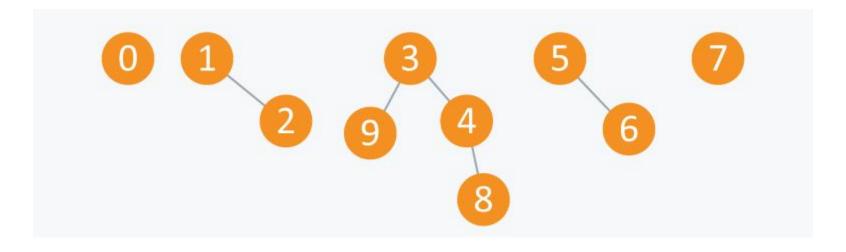


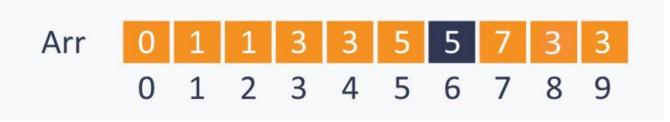
Disjoint set - Union(6, 5)





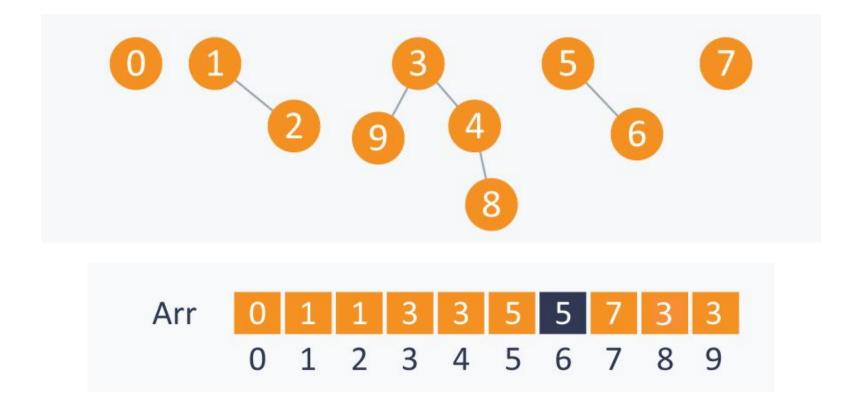
Disjoint set - 5 disjoint sets





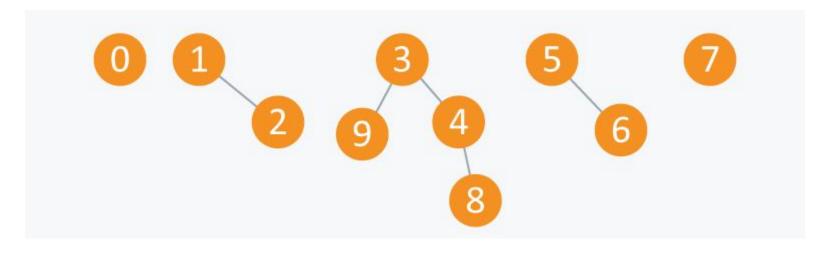
- First subset comprises the elements {3, 4, 8, 9}
- Second subset comprises the elements {1, 2}
- Third subset comprises the elements {5, 6}
- Fourth subset comprises the elements {0}
- Fifth subset comprises the elements {7}

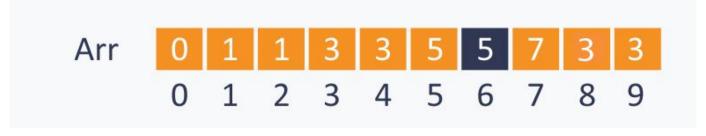
Disjoint set - find operations



- Find(0, 7): 0 and 7 are disconnected: FALSE
- Find(8, 9): Although 8 and 9 are not connected directly, there is a path that connects them: TRUE

Disjoint set - final representatives





Disjoint set - naive implementation

```
def initialize(Arr, N):
        for i in range(N):
            Arr[ i ] = i
    # returns true if A and B are connected, else returns
false
   def find(Arr, A, B):
        return Arr[ A ] == Arr[ B ]
        #if Arr[ A ] == Arr[ B ]:
        # return True
        #else:
        # return False
    #change all entries from Arr[ A ] to Arr[ B ]
    def union(Arr, N, A, B):
        TEMP = Arr[A]
        for i in range(N):
            if Arr[ i ] == TEMP:
                Arr[ i ] = Arr[ B ]
```

Disjoint set - naive implementation

```
def initialize(Arr, N):
           for i in range(N):
                Arr[ i ] = i
     # returns true if A and B are connected, else returns
false
     def find(Arr, A, B):
           return Arr[ A ] == Arr[ B ]
           #if Arr[ A ] == Arr[ B ]:
                 return True
           #else
           #
             return False
    #change all entries from Arr[ A ] to Arr[ B ]
def union(Arr, N, A, B):
   TEMP = Arr[ A ]
   for i in range(N):
        if Arr[ i ] == TEMP:
        Arr[ i ] = Arr[ B ]
                     Arr[i] = Arr[B]
```

Kruskal's algorithm + naive disjoint sets

- M is the number of edges
- N is the number of vertices
- The maximum number of M is N(N 1) / 2 or O(N²)
- $O(M) * O(N) = O(N^3)$ too slow!

```
Kruskal(G, c) { Sort edges weights so that c_1 \le c_2 \le \ldots \le c_m. T \leftarrow \phi foreach (u \in V) make a set containing singleton u for i = 1 to m are u and v in different connected components? (u,v) = e_i if (u and v are in different sets) { T \leftarrow T \cup \{e_i\} merge the sets containing u and v } return v merge two components
```

Disjoint sets

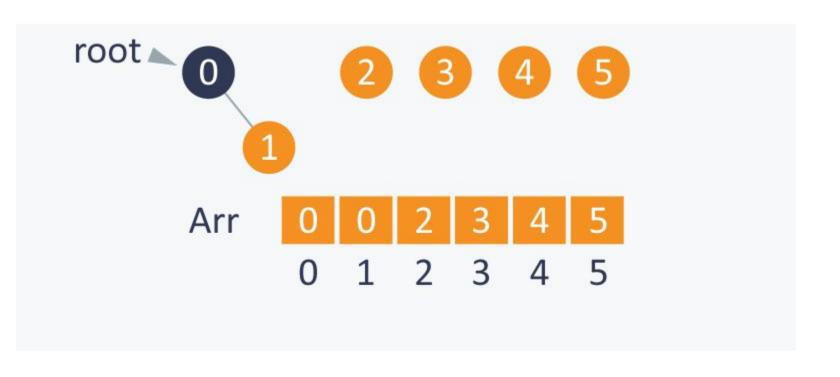
- Stores non-overlapping sets
- Union(a, b) = merges sets where a and b are located
- Find(a, b) = tells whether or not a and b are located in the same set

	Naive implementation
Find	O(1)
Union	O(N)
Union for all elements	$O(N^2)$

Optimized disjoint set

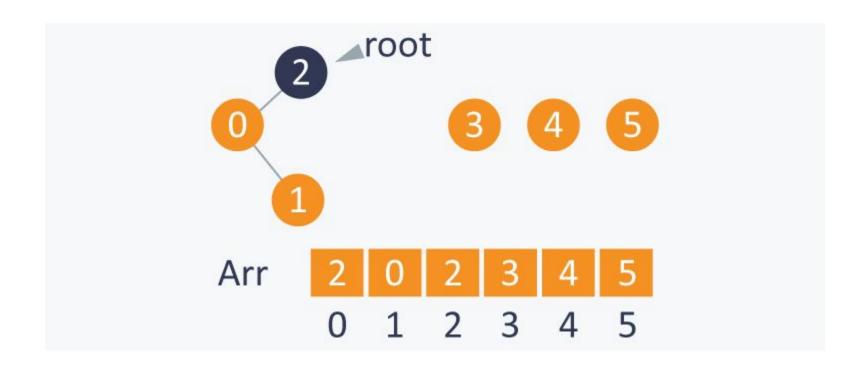


Optimized disjoint set - union(0, 1)



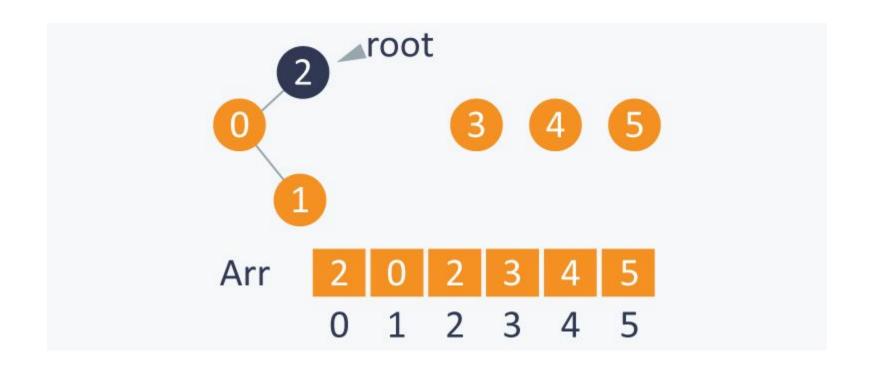
Optimized disjoint set - union(0, 2)

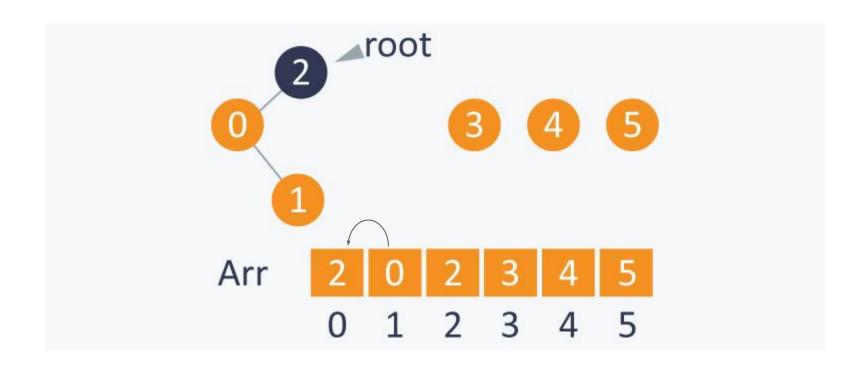
We only change the representative for the old root!

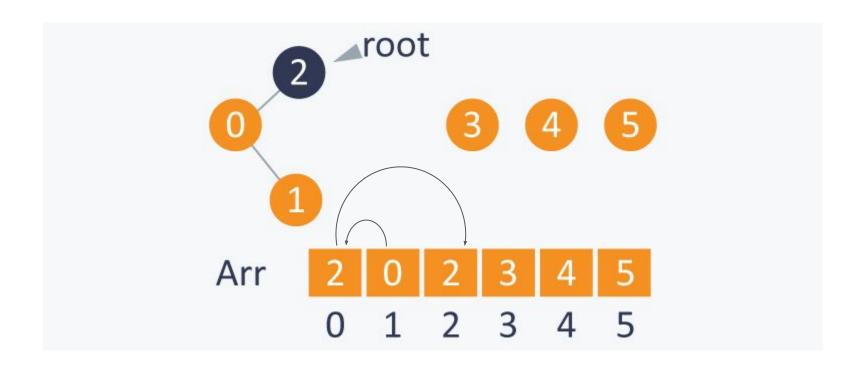


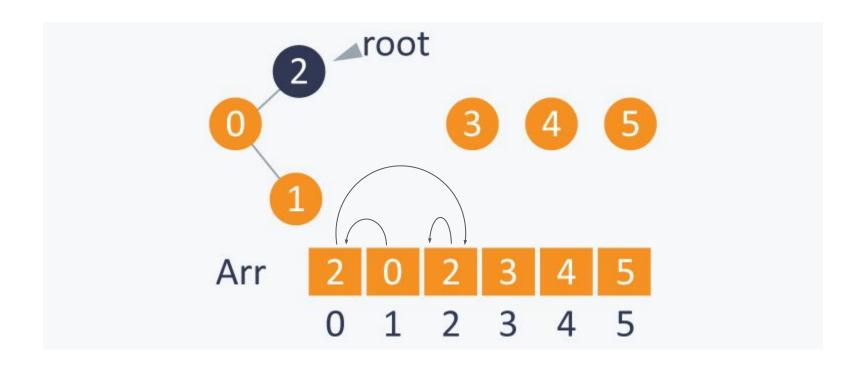
Optimized disjoint set - union(0, 2)

In the previous implementation, Arr would be [2, 2, 2, 3, 45]

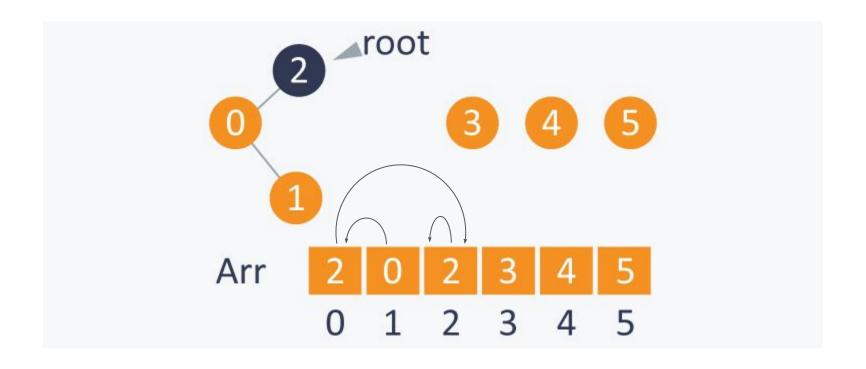




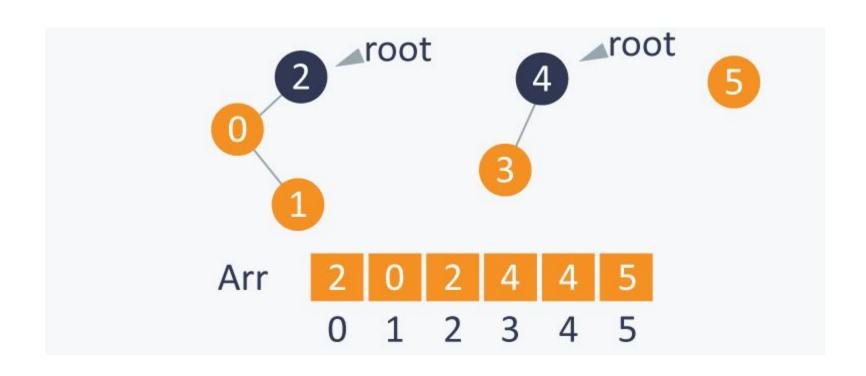




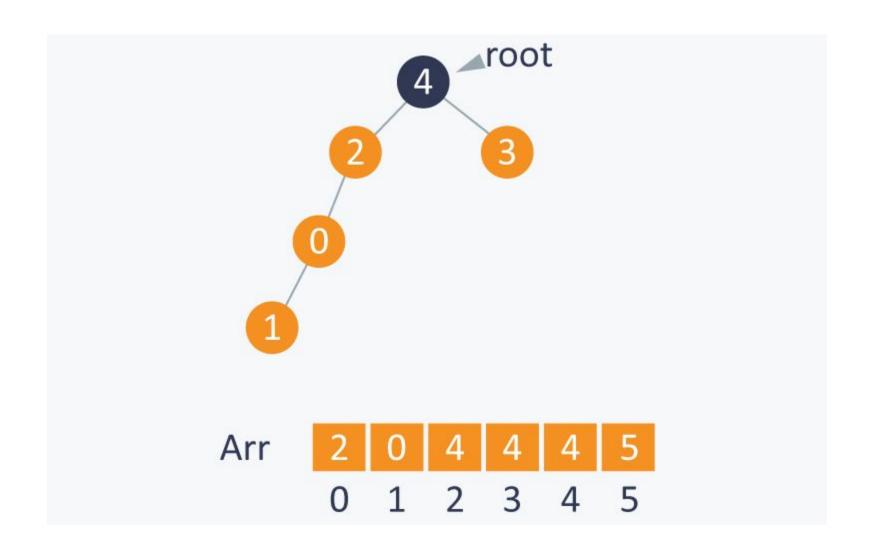
UNION becomes faster FIND becomes longer

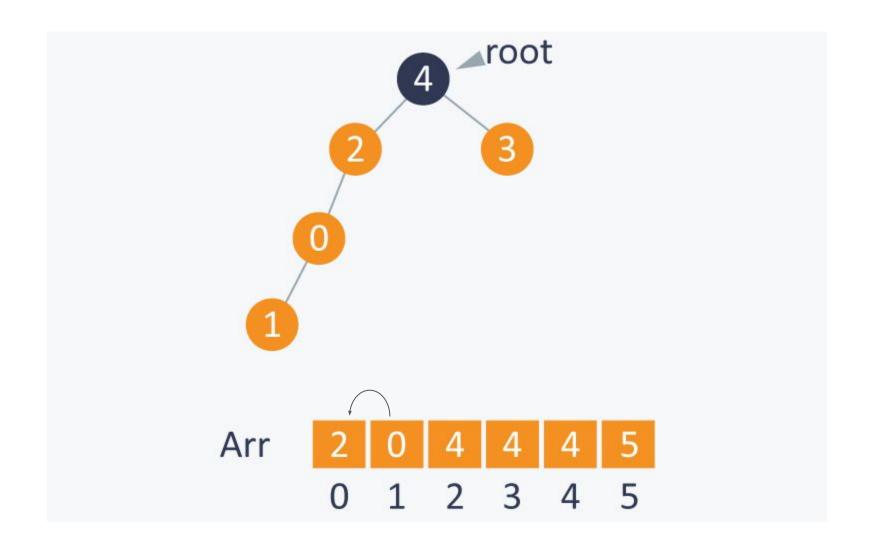


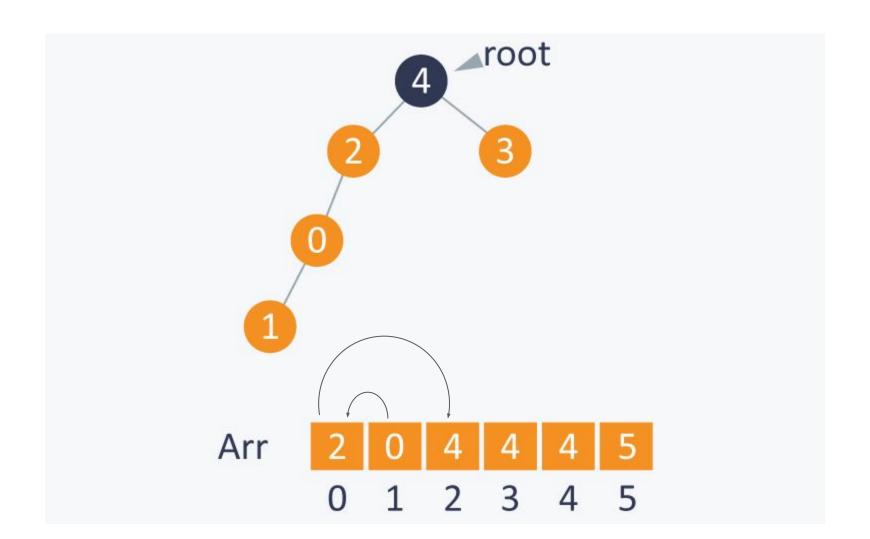
Optimized disjoint set - union(3, 4)

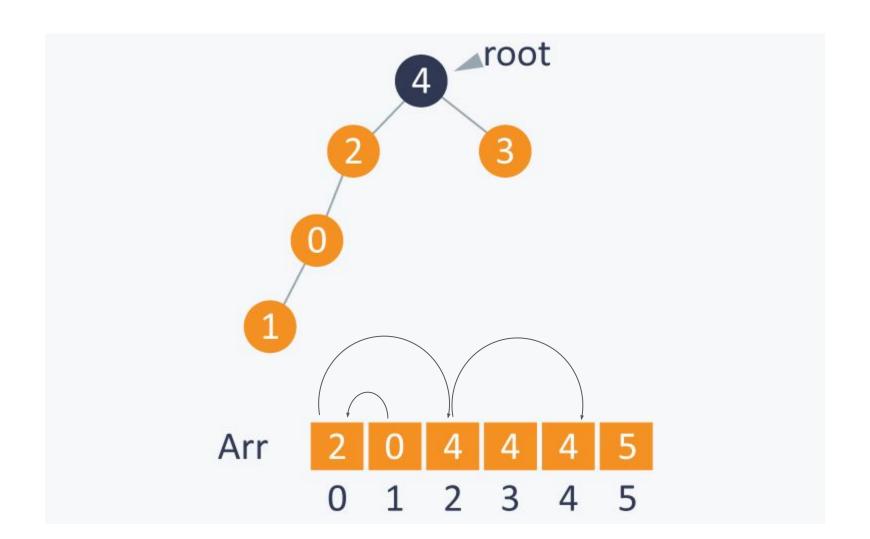


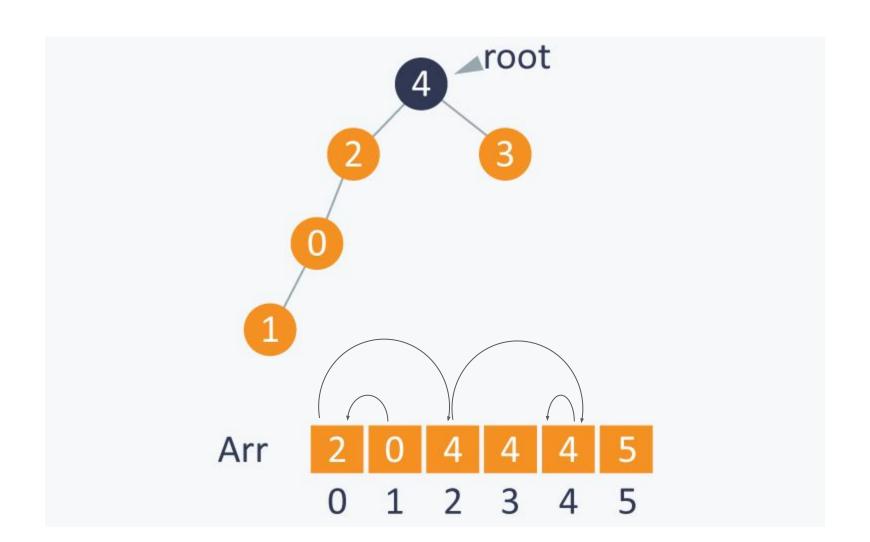
Optimized disjoint set - union(1, 4)











Optimized disjoint set - implementation

```
#finding root of an element
def root(Arr, i):
    while Arr[ i ] != i:
        i = Arr[ i ]
    return i
def union(Arr, A, B):
    root A = root(Arr, A)
    root B = root(Arr, B)
    Arr[ root_A ] = root_B
def find(Arr, A, B):
    return root(Arr, A)==root(Arr, B)
```

Optimized disjoint set - only a half-way through!

```
#finding root of an element
A loop is
         def root(Arr, i):
hidden
              while Arr[ i ] != i:
here
              i = Arr[ i ]
               return i
          def union(Arr, A, B):
              root_A = root(Arr, A)
We don't
have a
             root_B = root(Arr, B)
loop here
             Arr[ root_A ] = root_B
anymore
```

```
def find(Arr, A, B):
    return root(Arr, A)==root(Arr, B)
```