21. Disjoint set operations

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Disjoint Set Operations :

We have a collection of disjoint sets of elements. Each set is identified by a representative element. We want to perform union operations, and tell which set something is in. This is useful in a minimum spanning tree algorithm and many other applications. Formally, we have the following operations.

- MAKE-SET(x): Create new set {x} with representative x.
- UNION(x,y): x and y are elements of two sets. Remove these sets and add their union. Choose a representative for it.
- FIND-SET(x): return the representative of the set containing x.

Example :

```
MAKE-SET(1)
MAKE-SET(2)
MAKE-SET(3)
MAKE-SET(4)
FIND(3)
                       (returns 3)
FIND(2)
                       (returns 2)
UNION(1,2)
                       (representative 1, say)
                                                 {1,2}
FIND(2)
                       (returns 1)
FIND(1)
                       (returns 1)
UNION(3,4)
                       (representative 4, say)
                                                 {3,4}
FIND(4)
                       (returns 4)
FIND(3)
                       (returns 4)
                       (representative 4, say)
UNION(1,3)
                                                 {1,2,3,4}
FIND(2)
                       (returns 4)
FIND(1)
                       (returns 4)
FIND(4)
                       (returns 4)
FIND(3)
                       (returns 4)
```

An application of disjoint-set data structure :

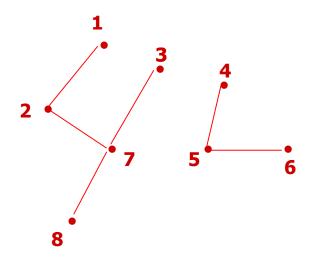
```
Connected-Components(G)
for each vertex v \in G.V
Make-Set(v)
for each edge (u,v) \in G.E
if Find-Set(u) \neq Find-Set(v)
Union(u,v)
```

Same-Component(u,v)

if Find-Set(u) == Find-Set(v)

return TRUE

else return FALSE



Linked List Implementation with Concatenation :

Each cell has an element, a pointer to the next member, and a pointer to the first element in the list, which is the set representative.

For union, we append one list to another, changing the pointers to the set representative on the list at the end.

Thus we get quadratic worst case performance.

Make-Set(
$$x_1$$
),....., Make-Set(x_n)

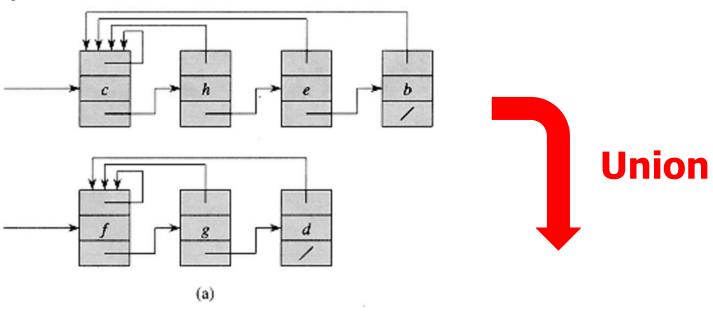
1 2 3 q-1

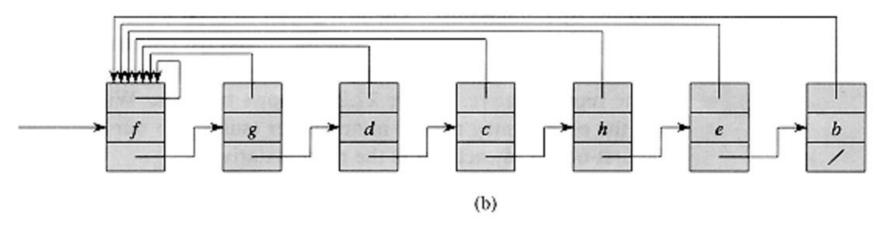
Union(x_1,x_2), Union(x_2,x_3), Union(x_3,x_4)....., Union(x_{q-1},x_q)

$$n = \lceil m/2 \rceil + 1, q = m - n$$

$$\sum_{j=1}^{q-1} i = O(q^2) = O(m^2)$$



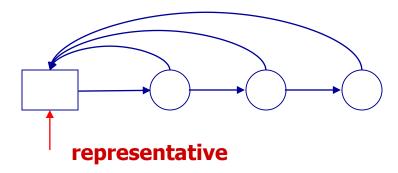






Disjoint-set union:

Use a linked list to represent a set



- Make-Set : O(1)
- Find-Set : O(1)
- Union(A,B): Copy elements of A into B, O(A).

Worst-case analysis :

$$|S_i| = 1$$
 initially.

Union(S₁, S₂) 1
Union(S₂, S₃) 2

Union(S_{n-1}, S_n) n-1
$$\Theta(n^2)$$

Improvement : Copy smaller set into larger.

A single Union can still take $\Theta(n)$, but n unions take only $O(n \log n)$ time.

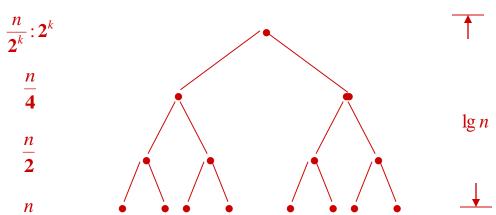
Refined Linked List Implementation :

(Append shorter list onto longer one)

MAKE-SET O(1) time
FIND-SET O(1) time
UNION O(log n) time (rough bound)

Proof of the bound for UNION:

A set has to double in size each time it is concatenated onto the end. Thus we get O(n log n) worst case performance for n operations.





Amortized analysis :

- In a set of size n, any element can be copied at most Ign times.
 Each time copied, it was in smaller set.
- n elements each copied O(lgn) times → n Unions take O(nlgn) time.
- Each UNION takes O(lgn).

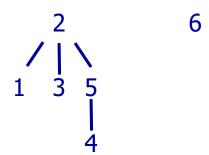
Forest representation :

Here we represent each set as a tree, and the representative is the root . For example, the following forest represents the set $\{1,2,3\}$, $\{4,5\}$, $\{6\}$:

Implementation

```
MAKE-SET(x) Create a tree
FIND-SET(x) Go to the root
UNION(x,y) Add a pointer
```

Thus we would get the following form UNION(1,4)



This representation does not improve the running time in the worst case over the linked list representation.



Path compression and ranks :

These are refinements of the forest representation which make it significantly faster.

FIND-SET: Do path compression

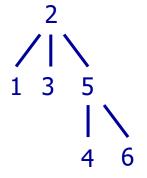
UNION: Use ranks

"Path compression" means that when we do FIND-SET(X), we make all nodes encountered point directly to the representative element for x. Initially, all elements have rank 0. The ranks of representative elements are updated so that if two sets with representatives of the same rank are unioned, then the new representative is incremented by one.

Example :

We show a sequence of operations and what the forest would look like.

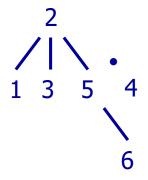
UNION(4,3)



$$RANK(2)=2$$

 $RANK(5)=1$

FIND(4)



FIND(3) no change

FIND(6) (path compression)

Make_Set(x):

Algorithm :

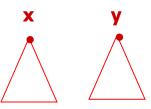
```
Make_Set(x)
\{ x.p = x
x.rank = 0 \}
```

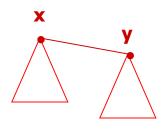
- Union(x,y):
 - Algorithm :

```
Union(x,y)
{ Link(Find-Set(x),Find-Set(y))
```

- Link(x,y):
 - Algorithm :

```
Link(x,y)
{
    if x.rank > y.rank
       y.p = x
    else x.p = y
       if x.rank == y.rank  y.rank++ }
```

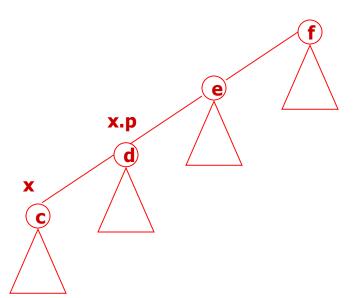


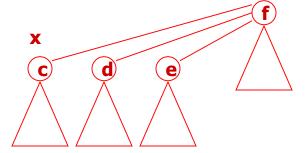


Find_Set(x) :

Algorithm :

```
Find_Set(x)
\{ if x \neq x.p
x.p = Find-Set(x.p)
return x.p \}
```





Analysis of MST with disjoint set union:

• Sort : $\theta(E \lg E) = \theta(E \lg V)$

O(V) : Make-Set's

O(E) : Find-Set's

■ O(V): Union's

- Above disjoint-set operations together take $O(E \cdot \alpha(E,V))$, using best known algorithm for disjoint-set union.
- Total : $O(E \lg V)$
- m operations on n sets $O(m \cdot \alpha(m,n))$, where $\alpha(m,n)$ is a functional inverse" of Ackermann's function.