

Union-Find Problem

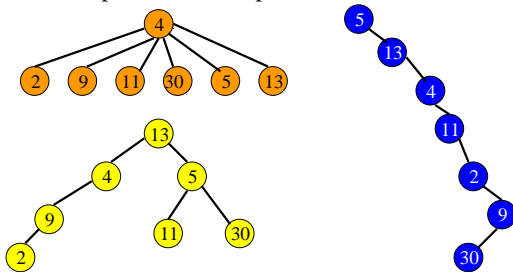
- Given a set $\{1, 2, \dots, n\}$ of n elements.
- Initially each element is in a different set.
 - $\{1\}, \{2\}, \dots, \{n\}$
- An intermixed sequence of union and find operations is performed.
- A union operation combines two sets into one.
 - Each of the n elements is in exactly one set at any time.
- A find operation identifies the set that contains a particular element.

Using Arrays And Chains

- See Section 7.7 for applications as well as for solutions that use arrays and chains.
- Best time complexity obtained in Section 7.7 is $O(n + u \log u + f)$, where u and f are, respectively, the number of union and find operations that are done.
- Using a tree (not a binary tree) to represent a set, the time complexity becomes almost $O(n + f)$ (assuming at least $n/2$ union operations).

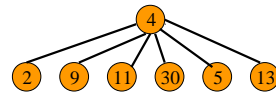
A Set As A Tree

- $S = \{2, 4, 5, 9, 11, 13, 30\}$
- Some possible tree representations:



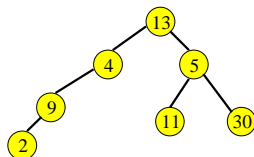
Result Of A Find Operation

- $\text{find}(i)$ is to identify the set that contains element i .
- In most applications of the union-find problem, the user does not provide set identifiers.
- The requirement is that $\text{find}(i)$ and $\text{find}(j)$ return the same value iff elements i and j are in the same set.



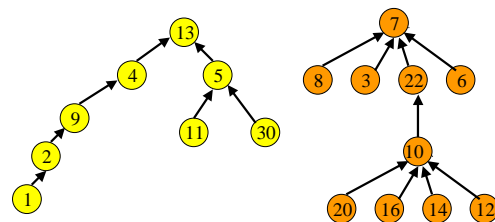
$\text{find}(i)$ will return the element that is in the tree root.

Strategy For $\text{find}(i)$



- Start at the node that represents element i and climb up the tree until the root is reached.
- Return the element in the root.
- To climb the tree, each node must have a parent pointer.

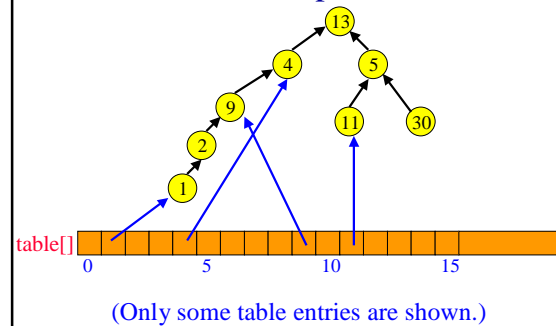
Trees With Parent Pointers



Possible Node Structure

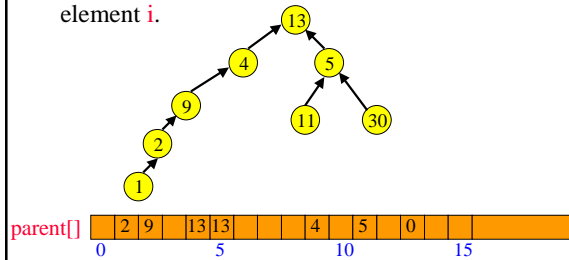
- Use nodes that have two fields: **element** and **parent**.
 - Use an array **table[]** such that **table[i]** is a pointer to the node whose element is **i**.
 - To do a **find(i)** operation, start at the node given by **table[i]** and follow parent fields until a node whose parent field is null is reached.
 - Return element in this root node.

Example



Better Representation

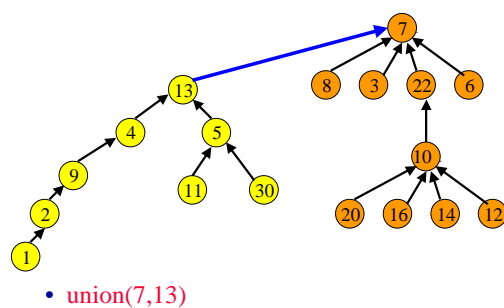
- Use an integer array **parent[]** such that **parent[i]** is the element that is the parent of element **i**.



Union Operation

- union(i,j)**
 - i** and **j** are the roots of two different trees, **i != j**.
- To unite the trees, make one tree a subtree of the other.
 - parent[j] = i**

Union Example



The Find Method

```
public int find(int theElement)
{
    while (parent[theElement] != 0)
        theElement = parent[theElement]; // move up
    return theElement;
}
```


The Union Method

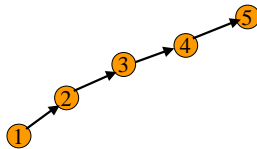
```
public void union(int rootA, int rootB)
{ parent[rootB] = rootA; }
```

Time Complexity Of union()

- $O(1)$


Time Complexity of find()

- Tree height may equal number of elements in tree.
 - $\text{union}(2,1), \text{union}(3,2), \text{union}(4,3), \text{union}(5,4) \dots$

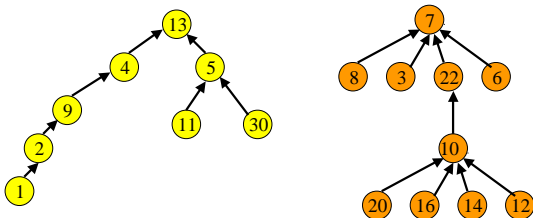


So complexity is $O(u)$.

u Unions and f Find Operations

- $O(u + uf) = O(uf)$
- Time to initialize $\text{parent}[i] = 0$ for all i is $O(n)$.
- Total time is $O(n + uf)$.
- Worse than solution of Section 7.7!
- Back to the drawing board. 

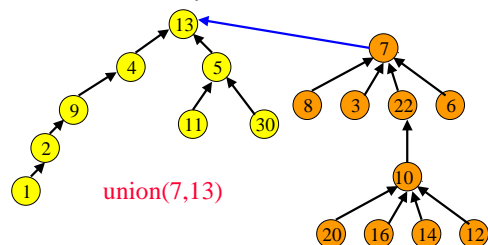
Smart Union Strategies



- $\text{union}(7,13)$
- Which tree should become a subtree of the other?

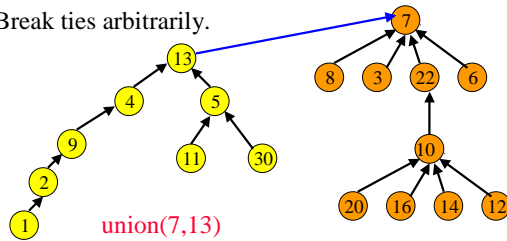
Height Rule

- Make tree with smaller height a subtree of the other tree.
- Break ties arbitrarily.



Weight Rule

- Make tree with fewer number of elements a subtree of the other tree.
- Break ties arbitrarily.



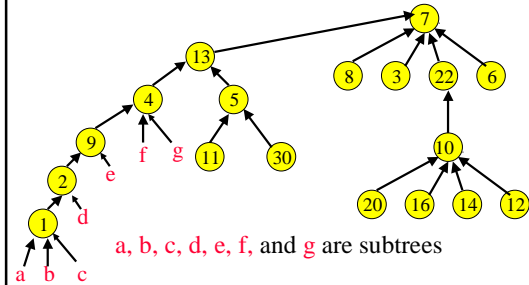
Implementation

- Root of each tree must record either its height or the number of elements in the tree.
- When a union is done using the height rule, the height increases only when two trees of equal height are united.
- When the weight rule is used, the weight of the new tree is the sum of the weights of the trees that are united.

Height Of A Tree

- Suppose we start with single element trees and perform unions using either the height or the weight rule.
- The height of a tree with p elements is at most $\text{floor}(\log_2 p) + 1$.
- Proof is by induction on p . See text.

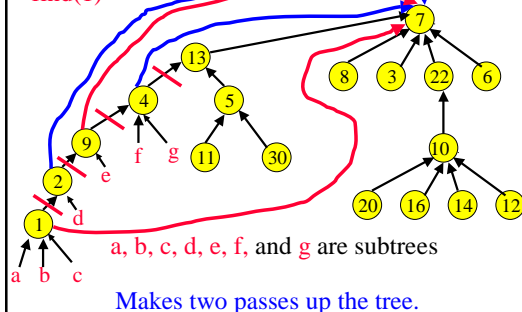
Sprucing Up The Find Method



- $\text{find}(1)$
- Do additional work to make future finds easier.

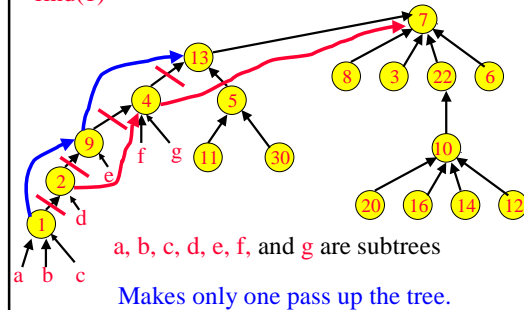
Path Compaction

- Make all nodes on find path point to root.
- $\text{find}(1)$



Path Splitting

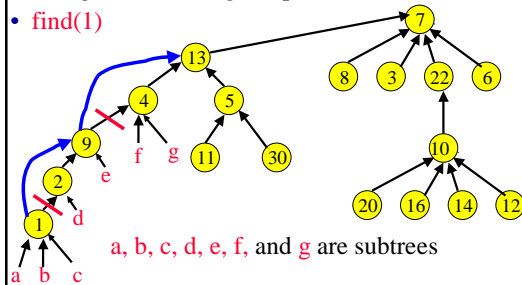
- Nodes on find path point to former grandparent.
- $\text{find}(1)$



Path Halving

- Parent pointer in every other node on find path is changed to former grandparent.

- $\text{find}(1)$



Changes half as many pointers.

Time Complexity



- Ackermann's function.
 - $A(i,j) = 2^i$, $i = 1$ and $j \geq 1$
 - $A(i,j) = A(i-1,2)$, $i \geq 2$ and $j = 1$
 - $A(i,j) = A(i-1, A(i,j-1))$, $i, j \geq 2$
- Inverse of Ackermann's function.
 - $\alpha(p,q) = \min\{z \geq 1 \mid A(z, p/q) > \log_2 q\}$, $p \geq q \geq 1$

Time Complexity



- Ackermann's function grows very rapidly as i and j are increased.
 - $A(2,4) = 2^{65,536}$
- The inverse function grows very slowly.
 - $\alpha(p,q) < 5$ until $q = 2^{A(4,1)}$
 - $A(4,1) = A(2,16) \gg \gg \gg A(2,4)$
- In the analysis of the union-find problem, q is the number, n , of elements; $p = n + f$; and $u \geq n/2$.
- For all practical purposes, $\alpha(p,q) < 5$.

Time Complexity



Theorem 12.2 [Tarjan and Van Leeuwen]

Let $T(f,u)$ be the maximum time required to process any intermixed sequence of f finds and u unions. Assume that $u \geq n/2$.

$$a^*(n + f^*\alpha(f+n, n)) \leq T(f,u) \leq b^*(n + f^*\alpha(f+n, n))$$

where a and b are constants.

These bounds apply when we start with singleton sets and use either the weight or height rule for unions and any one of the path compression methods for a find.