

Analysis & Design of Algorithms (CSCE 321)



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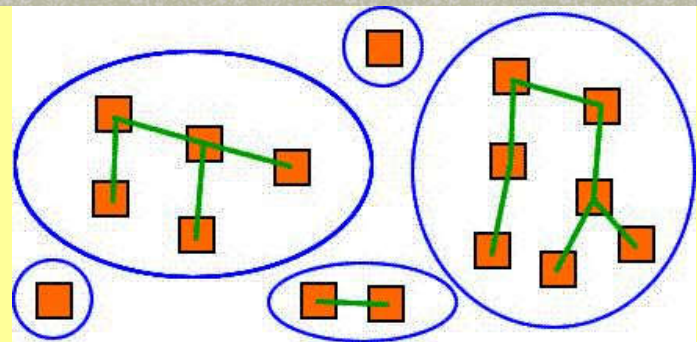
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Part R4. Disjoint Sets



Disjoint Sets

- What are Disjoint Sets?
- Tree Representation
- Basic Operations
- Parent Array Representation
- Simple Find and Simple Union
- Disjoint Sets Class
- Some Applications





What are Disjoint Sets?

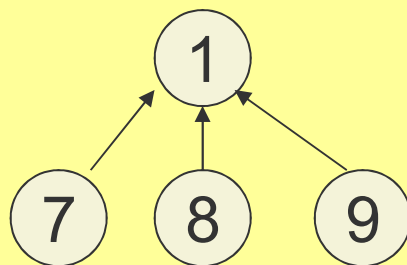
- A set S is a collection of elements of the same type
- We assume that the elements are the numbers $1, 2, \dots, n$. In practice, these numbers may be indices to a symbol table containing the names of the sets.
- Disjoint sets are collections of elements with no common elements between the sets. If S_i and S_j , $i \neq j$ are two sets, then $S_i \cap S_j = \emptyset$
- Examples:
 $S1 = \{1, 7, 8, 9\}, \quad S2 = \{5, 2, 10\}, \quad S3 = \{3, 4, 6\}$



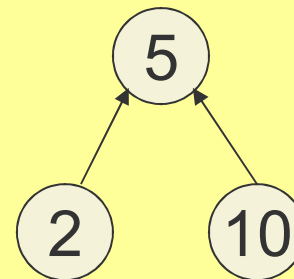
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Union-Find Data Structure: Tree Representation

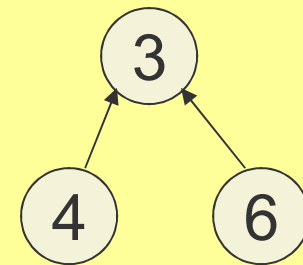
- One possible representation is a tree where a set can be identified by a **parent** node and **children** nodes.
- In this representation, children point to their parents, rather than the reverse:



$S1 = \{1,7,8,9\}$



$S2 = \{5,2,10\}$



$S3 = \{3,4,6\}$



Basic Operations

- Find(i):

Given the element (i), find the set containing (i), i.e. find the root of the tree.

- Union(i,j):

Given two disjoint sets S_i and S_j , obtain a set containing the elements in both sets, i.e.,

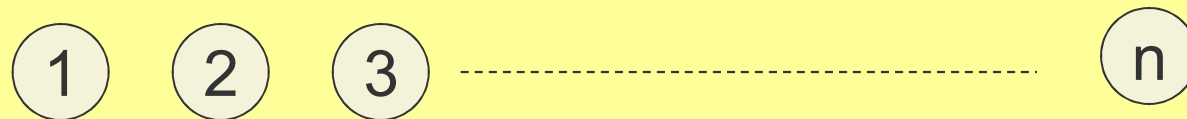
$$S_i \cup S_j$$



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Parent Array Representation

- n disjoint nodes can be represented as n disjoint sets where each node is its own parent, i.e. $p[i] = -1$:



i	1	2	3				n
p[i]	-1	-1	-1				-1



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Basic Operations

- Find(i):

Finding out which subset we are in is simple, for we keep traversing up the parent pointers until we hit the root.



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Simple Find

- The value in $p[i]$ represents the parent of node (i). For the sets $S1, S2, S3$ shown before, we have:

i	1	2	3	4	5	6	7	8	9	10
p[i]	-1	5	-1	3	-1	3	1	1	1	5

- Algorithm for Simple find:

Find the set containing node (i) = find the parent of (i):

```
int find(int i)
{
    while (p[i] >= 0) i = p[i];
    return i;
}
```

find (1) \rightarrow 1
find (4) \rightarrow 3
find (10) \rightarrow 5



Basic Operations

- Union(i,j):

Making a union between two subsets is also easy. Just make the root of one of two trees point to the other, so now all elements have the same root and thus the same subset name.

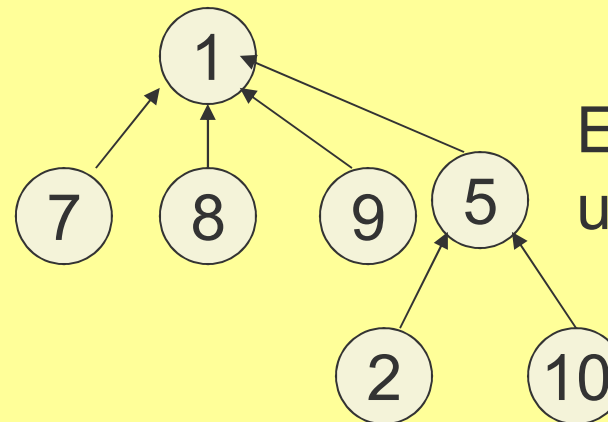
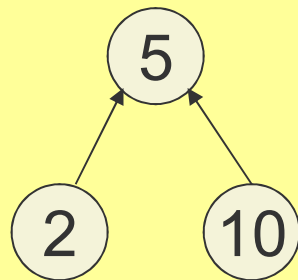
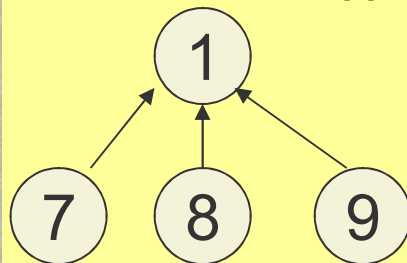


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Simple Union

- Simple (disjoint) Union:

Make set (i) the child of set (j) = union (i,j)



Example:
union(5,1)



Simple Union

- The parent array would change to:

i	1	2	3	4	5	6	7	8	9	10
p[i]	-1	5	-1	3	1	3	1	1	1	5

- Algorithm:

```
void union (int i, int j)
{ p [i] = j ; }
```



Disjoint sets class

// Make set(i) the child of set(j)

```
void simpleUnion(int i, int j)
{   p[i] = j;   }
```

// Find the parent set of subset(i)

```
int simpleFind(int i)
{   while (p[i]>=0) i = p[i];
    return i;
}
```




Analysis

- Union: takes constant time, i.e., $O(1)$
- Find: Suppose we do the following sequence of unions:

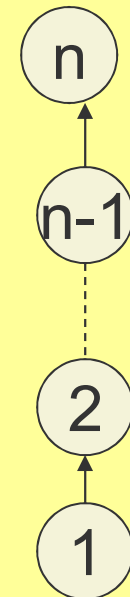
$\text{union}(1,2)$, $\text{union}(2,3)$,..., $\text{union}(n-1,n)$

The time of find parent of node (i) will be i.

To process all finds, we need

$(1+2+\dots+n)$ steps = $O(n^2)$

Hence, the average find cost is $O(n)$





Improvement

- We can improve the find cost by modifying the union algorithm to be:

*“make the root of the smaller tree point to the root of the bigger tree” (known as “**union by size**”)*

- In this case, the cost of find is **$O(\log n)$**

- **Implementation:**

Store the node count in each root. Modify the count when making a union.

Note: there is also “**union by rank**”



Disjoint sets class

// Make a union between set(i) and set(j)

```
void simpleUnion(int i, int j)
{
    int sum = c[i] + c[j];
    if (c[i] > c[j]) {p[j] = i; c[i] = sum;}
    else { p[i] = j; c[j] = sum;}
}
```

// Find the parent set of subset(i)

```
int simpleFind(int i)
{
    while (p[i] >= 0) i = p[i];
    return i;
}
```



Path compression

```
function Find(x)
    if x.parent != -1           // if not root
        x.parent := Find(x.parent)
    return x.parent
```

This is valid, since each element visited on the way to a root is part of the same set. The resulting flatter tree speeds up future operations.

Check the wiki page: both operations are semi-linear: $O(a \cdot n)$



Application example

MST with Kruskal

KRUSKAL(G):

$A = \emptyset$

foreach $v \in G.V$:

 MAKE-SET(v)

foreach (u, v) in $G.E$ ordered by $\text{weight}(u, v)$, increasing:

if FIND-SET(u) \neq FIND-SET(v):

$A = A \cup \{(u, v)\}$

 UNION(u, v)

return A