

## CHAPTER 8

### THE DISJOINT SET ADT

#### § 1 Equivalence Relations

**【Definition】** A *relation*  $R$  is defined on a set  $S$  if for every pair of elements  $(a, b)$ ,  $a, b \in S$ ,  $a R b$  is either true or false. If  $a R b$  is true, then we say that  $a$  is related to  $b$ .

**【Definition】** A relation,  $\sim$ , over a set,  $S$ , is said to be an *equivalence relation* over  $S$  iff it is *symmetric*, *reflexive*, and *transitive* over  $S$ .

**【Definition】** Two members  $x$  and  $y$  of a set  $S$  are said to be in the same *equivalence class* iff  $x \sim y$ .

## § 2 The Dynamic Equivalence Problem



Given an equivalence relation  $\sim$ , decide for any  $a$  and  $b$  if  $a \sim b$ .

**[[Example]]** Given  $S = \{ 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12 \}$  and 9 relations: **12=4**, **3=1**, **6=10**, **8=9**, **7=4**, **6=8**, **3=5**, **2=11**, **11=12**.

The equivalence classes are  $\{ 2, 4, 7, 11, 12 \}$ ,  $\{ 1, 3, 5 \}$ ,  $\{ 6, 8, 9, 10 \}$

### Algorithm: (Union / Find)

```
{ /* step 1: read the relations in */
  Initialize N disjoint sets;
  while ( read in a ~ b ) {
    if ( ! (Find(a) == Find(b)) )
      Union the two sets;
  } /* end-while */
  /* step 2: decide if a ~ b */
  while ( read in a and b )
    if ( Find(a) == Find(b) ) output( true );
    else output( false );
}
```

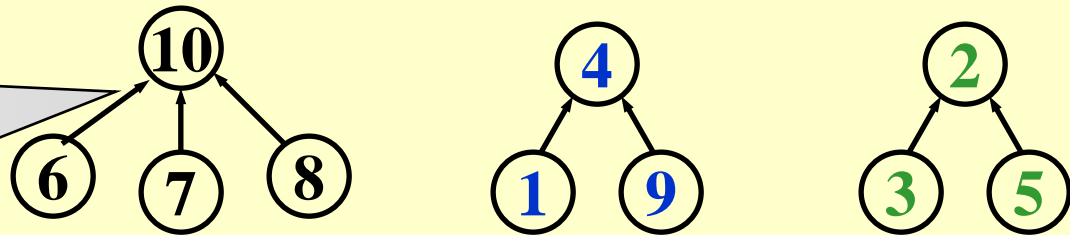
Dynamic (on-line)

✍ **Elements** of the sets:  $1, 2, 3, \dots, N$

✍ **Sets** :  $S_1, S_2, \dots$  and  $S_i \cap S_j = \emptyset$  ( if  $i \neq j$  ) ——— disjoint

[[**Example**]]  $S_1 = \{ 6, 7, 8, 10 \}$ ,  $S_2 = \{ 1, 4, 9 \}$ ,  $S_3 = \{ 2, 3, 5 \}$

**Note:**  
Pointers are  
from children  
to parents



A possible forest representation of these sets

✍ **Operations** :

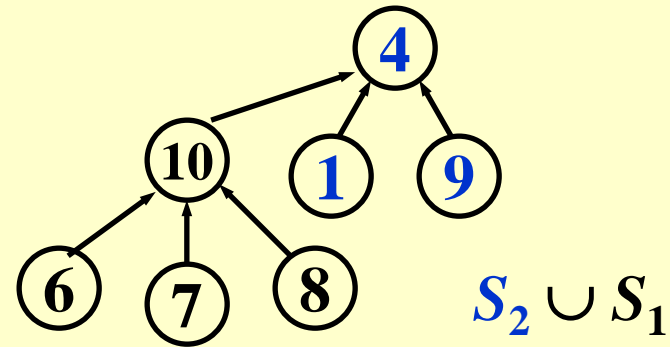
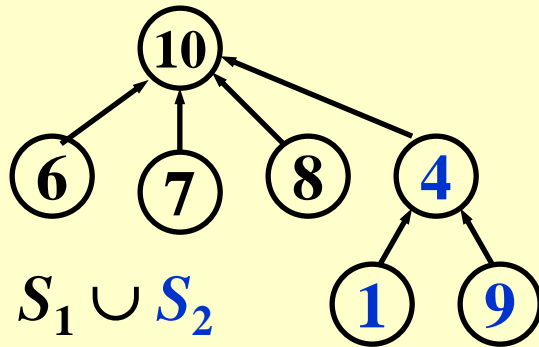
(1) **Union**(  $i, j$  ) ::= Replace  $S_i$  and  $S_j$  by  $S = S_i \cup S_j$

(2) **Find**(  $i$  ) ::= Find the set  $S_k$  which contains the element  $i$ .

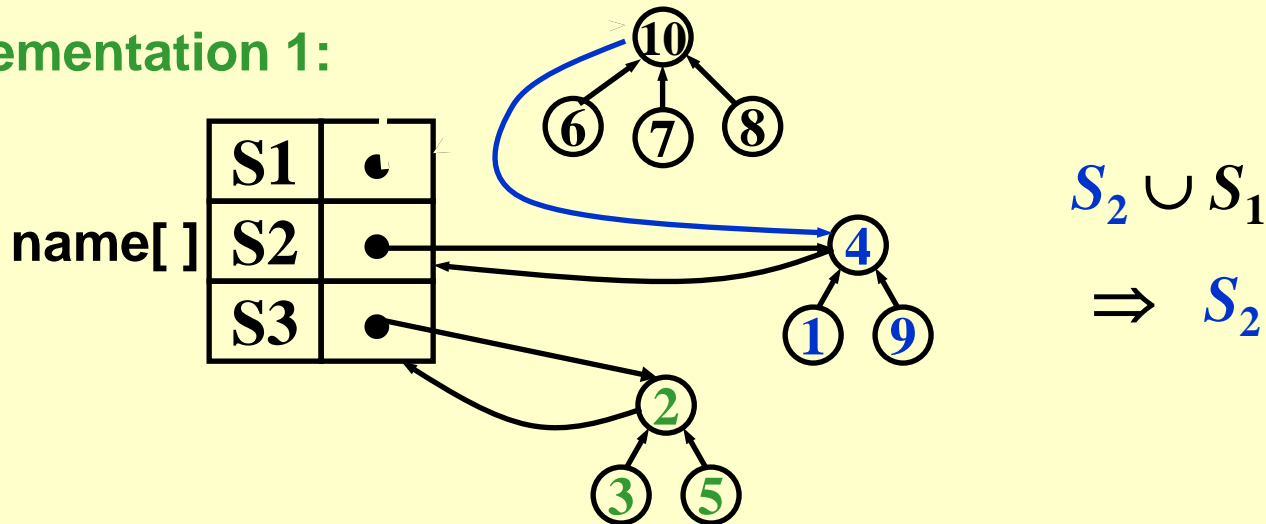
## § 3 Basic Data Structure

### ❖ Union ( $i, j$ )

**Idea:** Make  $S_i$  a subtree of  $S_j$ , or vice versa. That is, we can set the parent pointer of one of the roots to the other root.



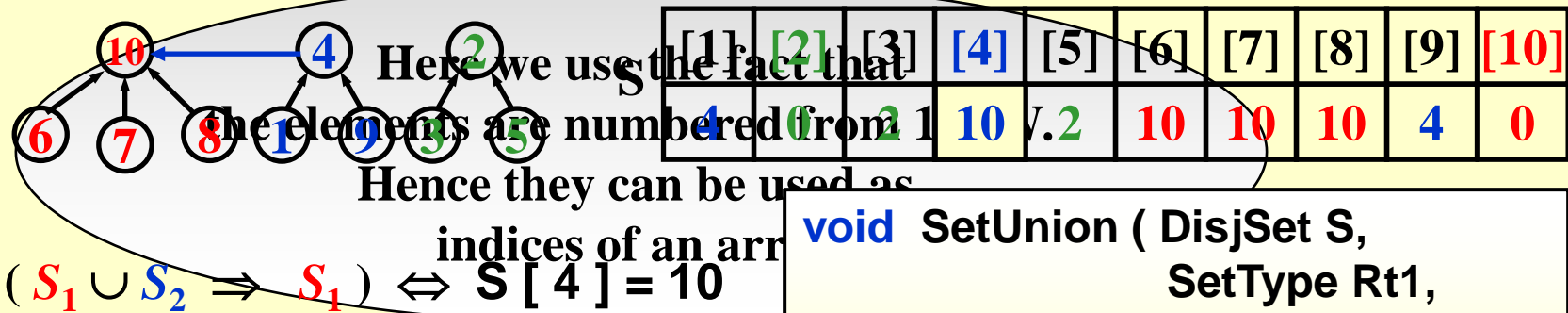
**Implementation 1:**



**Implementation 2:**  $S[\text{element}] = \text{the element's parent.}$

**Note:**  $S[\text{root}] = 0$  and set name = root index.

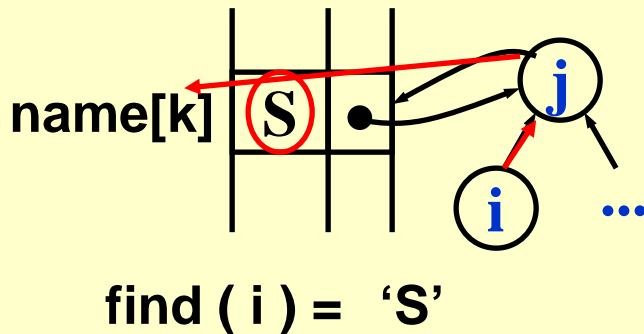
**[Example]** The array representation of the three sets is



```
void SetUnion ( DisjSet S,
                SetType Rt1,
                SetType Rt2 )
{   S [ Rt2 ] = Rt1 ;   }
```

❖ **Find (i)**

**Implementation 1:**



**Implementation 2:**

```
SetType Find ( ElementType X,
              DisjSet S )
{   for ( ; S[X] > 0; X = S[X] ) ;
    return X ;
}
```

## ❖ Analysis

Realistically speaking, union and find performance

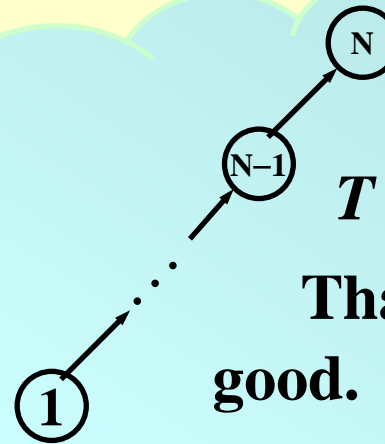
Sure. Try this one:  
 union(2, 1), find(1);  
 union(3, 2), find(1);  
 ..., ...;  
 union(N, N - 1), find(1).

1, 2, 3, 5, ..., all

to

the cl

2, { 1,



$$T = \Theta(N^2) !$$

That's not good.

## Algorithm using

```

{ Initialize S
for ( k = 1
  if ( Find(
    SetUnion
}
}
  
```



$\equiv j * /$

## § 4 Smart Union Algorithms

❖ **Union-by-Size** -- Always change the smaller tree

$S[\text{Root}] = -\text{size};$  /\* initialized to be -1 \*/

**【Lemma】** Let  $T$  be a tree created by union-by-size with  $N$  nodes, then

$$\text{height}(T) \leq \lceil \log_2 N \rceil + 1$$

for the worst case

**Proof:** By induction. (Each element can have its set name changed at most  $\log_2 N$  times.)

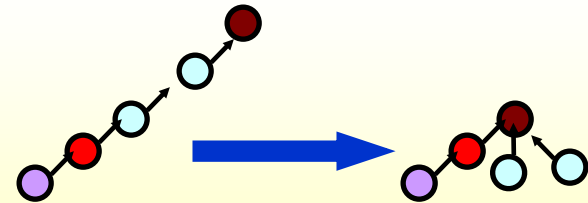
**Time complexity** of  $N$  Union and  $M$  Find operations is now  $O(N + M \log_2 N)$ .

❖ **Union-by-Height** -- Always change the shallow tree

Please read Figure 8.13 on p.273 for detailed implementation.

## § 5 Path Compression

```
SetType Find ( ElementType X, DisjSet S )
{
    if ( S[ X ] <= 0 ) return X;
    else return S[ X ] = Find( S[ X ], S );
}
```



```
SetType Find ( ElementType X, DisjSet S )
{ ElementType root, trail, lead;
  for ( root = X; S[ root ] > 0; root = S[ root ] )
    ; /* find the root */
  for ( trail = X; trail != root; trail = lead ) {
    lead = S[ trail ];
    S[ trail ] = root ;
  } /* collapsing */
  return root ;
}
```

Slower for  
a single find, but  
faster for a sequence of  
find operations.

**Note:** Not compatible with union-by-height since it changes the heights. Just take “height” as an estimated *rank*.



## § 6 Worst Case for Union-by-Rank and Path Compression

**【Lemma (Tarjan)】** Let  $T(M, N)$  be the maximum time required to process an intermixed sequence of  $M \geq N$  finds and  $N - 1$  unions. Then:

$$k_1 M \alpha(M, N) \leq T(M, N) \leq k_2 M \alpha(M, N)$$

for some positive constants  $k_1$  and  $k_2$ .

☞ Ackermann's Function and  $\alpha(M, N)$

$$A(i, j) = \begin{cases} 2^j & i = 1 \text{ and } j \geq 1 \\ A(i-1, 2) & i \geq 2 \text{ and } j = 1 \\ A(i-1, A(i, j-1)) & i \geq 2 \text{ and } j \geq 2 \end{cases}$$

$\log^* 2^{65536} = 5$   
since  
 $\log \log \log \log \log (2^{65536}) = 1$

<http://mathworld.wolfram.com/AckermannFunction.html>

$$\alpha(M, N) = \min\{ i \geq 1 \mid A(i, \lfloor M/N \rfloor) > \log N \} \leq O(\log^* N) \leq 4$$

$\log^* N$  (inverse Ackermann function)

= # of times the logarithm is applied to  $N$  until the result  $\leq 1$ .