

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 .

对称，自反，传递

【 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)

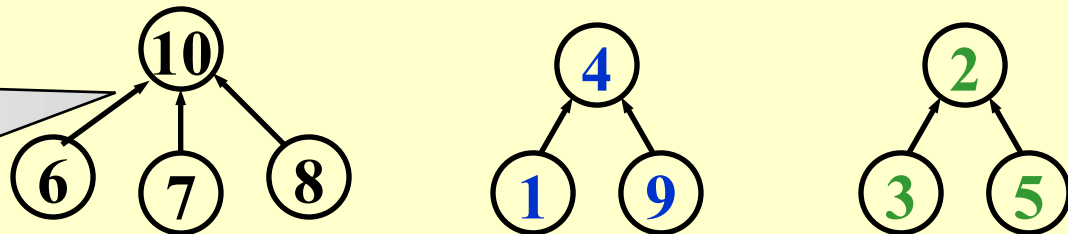
✍ **Elements** of the sets: $1, 2, 3, \dots, N$

✍ **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** :

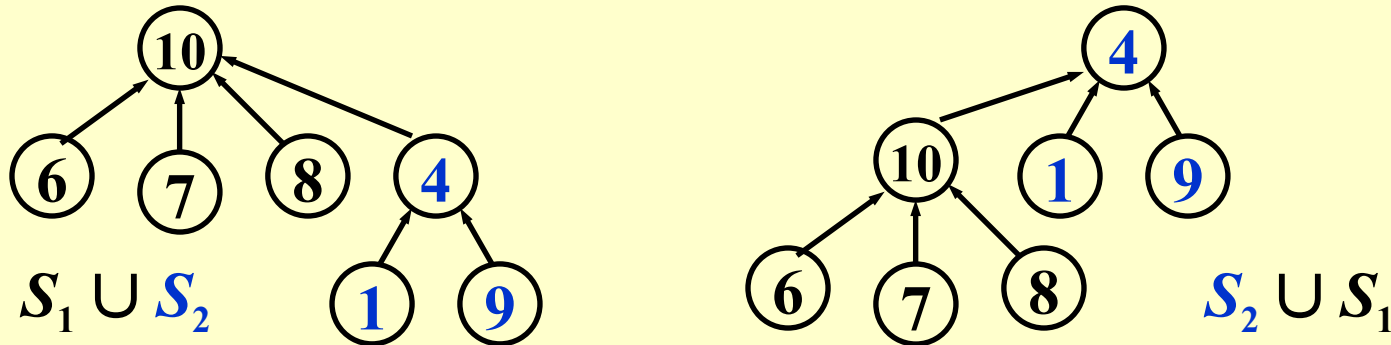
(1) **Union**(i, j) ::= Replace S_i and S_j by $S = S_i \cup S_j$

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

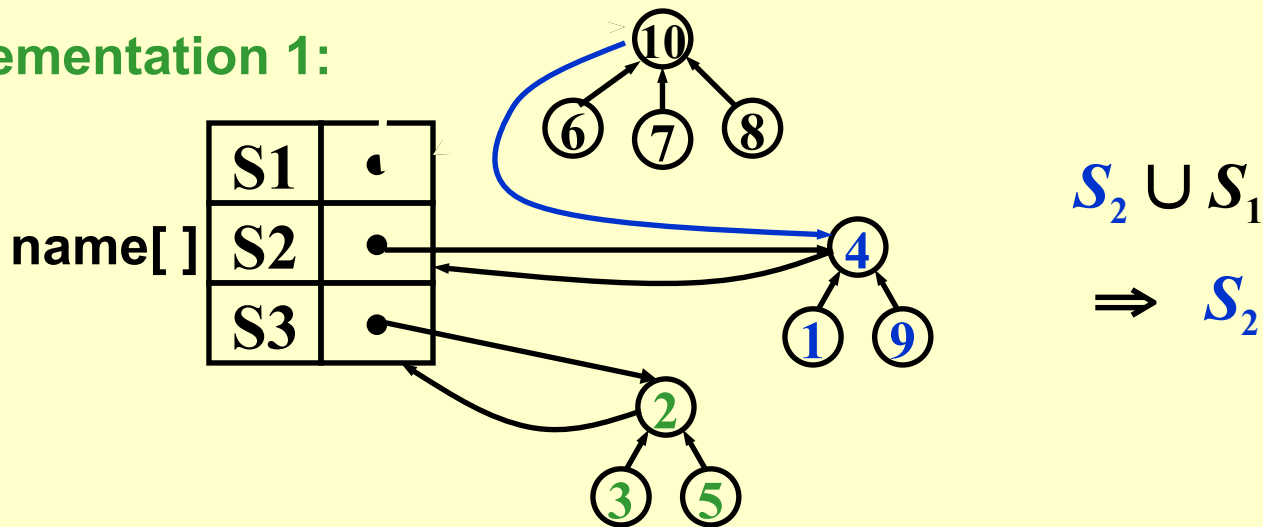
§3 Basic Data Structure

❖ 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.



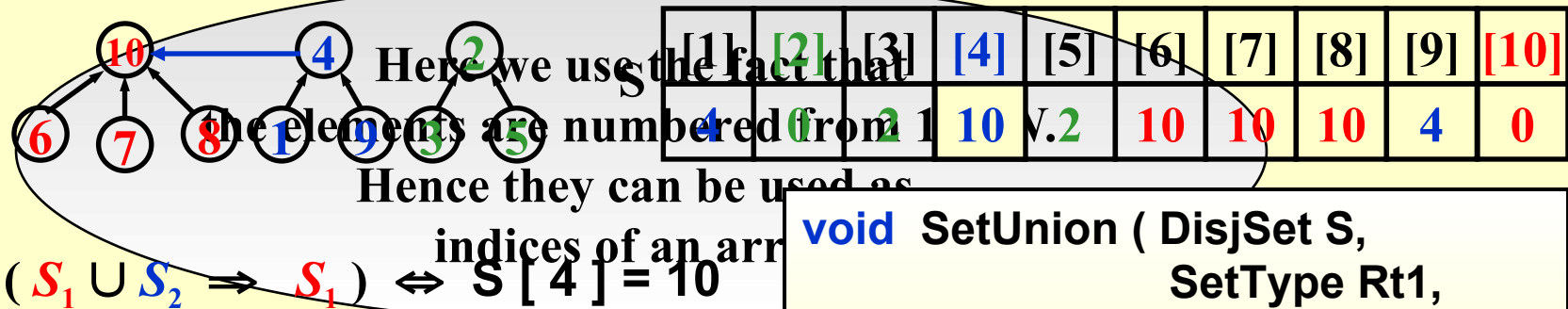
Implementation 1:



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

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

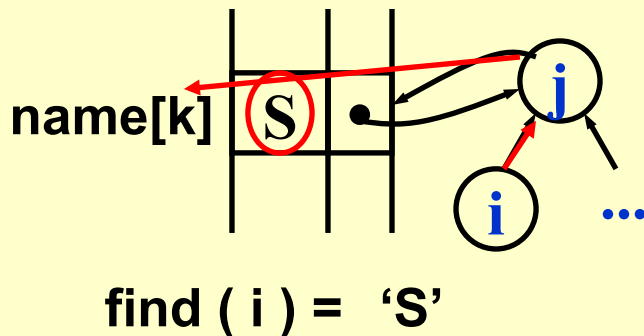
[[Example]] The array representation of the three sets is



```
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

... breaking, union and
... rmanc

Sure. Try this one:

union(2, 1), find(1);

union(3, 2), find(1);

..., ... ;

union(N, N - 1), find(1).

1

3, 5

}, all

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

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

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 $O(N + M \log_2 N)$.

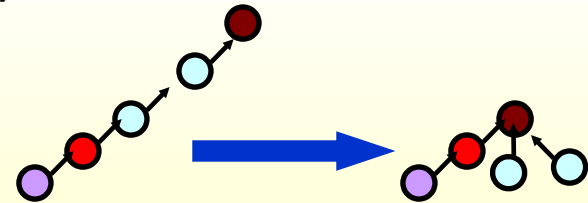
❖ **Union-by-Height** -- Always change the shallow tree

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 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 $\alpha(M, N)$

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

<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 ≤ 1 .

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

6536