

Introduction to Algorithms

Meng-Tsung Tsai

11/19/2019

Data Structures for Disjoint Sets

What is a data structure for disjoint sets?

It is a data structure that maintains a collection of disjoint dynamic sets $C = \{S_1, S_2, \dots, S_t\}$.

- (1) disjoint: $S_i \cap S_j = \emptyset$ for every $i \neq j$
- (2) dynamic: S_i changes over time for every i

We need the data structure to support the following operations:

- (1) Make-Set(x)
- (2) Union(x, y)
- (3) Find-Set(x)

Make-Set(x) operation

Make-Set(x) operation creates a singleton set $\{x\}$.

Example.

Let $C = \{S_1 = \{a, b\}, S_2 = \{c, d, e\}\}$.

After Make-Set(**f**), we have

$C = \{S_1 = \{a, b\}, S_2 = \{c, d, e\}, S_3 = \{\mathbf{f}\}\}$.

Union(x, y) operation

Union(x, y) unites the sets that contain x and y, say S_x and S_y , into a new set that is a union of these two sets.

Example.

Let $C = \{S_1 = \{a, \textcolor{red}{b}\}, S_2 = \{c, \textcolor{blue}{d}, e\}, S_3 = \{f\}\}$.

After Union($\textcolor{red}{b}$, $\textcolor{blue}{d}$), we have

$C = \{S_1 = \{a, \textcolor{red}{b}, c, \textcolor{blue}{d}, e\}, S_2 = \{f\}\}$.

Find-Set(x) operation

Find-Set(x) returns the representative element of the set that contains x.

Example.

Let $C = \{S_1 = \{a, b\}, S_2 = \{c, d, e\}, S_3 = \{f\}\}$. Our algorithm picks an arbitrary element as the representative for each set S_i .

Find-Set(a) returns b.

Find-Set(b) returns b.

Find-Set(e) returns d.

Find-Set(f) returns f.

Find-Set(x) operation

Find-Set(x) returns the representative element of the set that contains x.

Example.

Let $C = \{S_1 = \{a, b\}, S_2 = \{c, d, e\}, S_3 = \{f\}\}$. Our algorithm picks an arbitrary element as the representative for each set S_i .

Find-Set(a) returns b.

Find-Set(b) returns b.

Find-Set(e) returns d.

Find-Set(f) returns f.

If S_i doesn't change, the representative element of S_i should be fixed.

Applications of Disjoint-Set Data Structures

Computing the connected components

Let $G = (V, E)$ be an undirected graph.

```
Connected-Component(G) {  
  foreach node  $x$  in  $V$  {  
    Make-Set( $x$ );  
  }  
  foreach edge  $(u, v)$  in  $E$  {  
    if(Find-Set( $u$ )  $\neq$  Find-Set( $v$ )) {  
      Union( $u, v$ );  
    }  
  }  
}
```

Computing the connected components

Example.

$G = (V = \{1, 2, 3, 4, 5, 6, 7\}, E = \{\{1, 2\}, \{1, 3\}, \{2, 3\}, \{5, 6\}\})$.

Initially, $C = \{\{1\}, \{2\}, \{3\}, \{4\}, \{5\}, \{6\}, \{7\}\}$.

If $(\text{Find-Set}(1) \neq \text{Find-Set}(2))$ Union(1, 2);

After which, $C = \{\{1, 2\}, \{3\}, \{4\}, \{5\}, \{6\}, \{7\}\}$.

Computing the connected components

Example.

$G = (V = \{1, 2, 3, 4, 5, 6, 7\}, E = \{\{1, 2\}, \{\textcolor{red}{1}, \textcolor{blue}{3}\}, \{2, 3\}, \{5, 6\}\})$.

$C = \{\{\textcolor{red}{1}, 2\}, \{\textcolor{blue}{3}\}, \{4\}, \{5\}, \{6\}, \{7\}\}$.

If (Find-Set($\textcolor{red}{1}$) \neq Find-Set($\textcolor{blue}{3}$)) Union($\textcolor{red}{1}$, $\textcolor{blue}{3}$);

After which, $C = \{\{\textcolor{red}{1}, 2, \textcolor{blue}{3}\}, \{4\}, \{5\}, \{6\}, \{7\}\}$.

Computing the connected components

Example.

$G = (V = \{1, 2, 3, 4, 5, 6, 7\}, E = \{\{1, 2\}, \{1, 3\}, \{2, 3\}, \{5, 6\}\})$.

$C = \{\{1, 2, 3\}, \{4\}, \{5\}, \{6\}, \{7\}\}$.

If (Find-Set(2) = Find-Set(3)) do nothing;

After which, $C = \{\{1, 2, 3\}, \{4\}, \{5\}, \{6\}, \{7\}\}$.

Computing the connected components

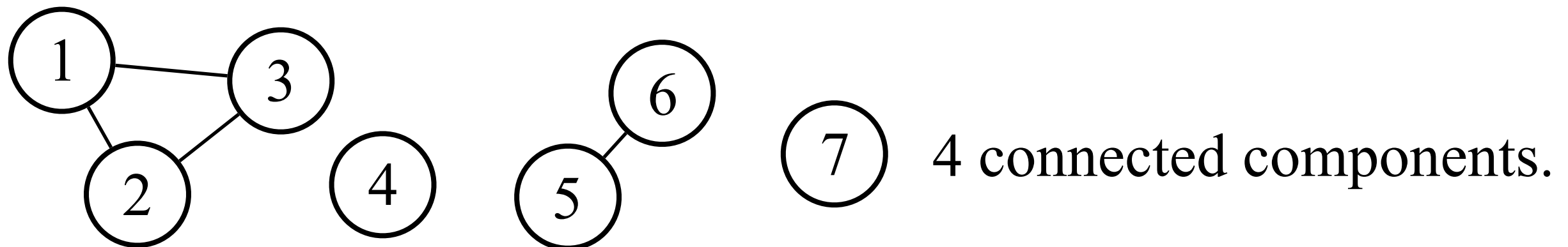
Example.

$G = (V = \{1, 2, 3, 4, 5, 6, 7\}, E = \{\{1, 2\}, \{1, 3\}, \{2, 3\}, \{5, 6\}\})$.

$C = \{\{1, 2, 3\}, \{4\}, \{5\}, \{6\}, \{7\}\}$.

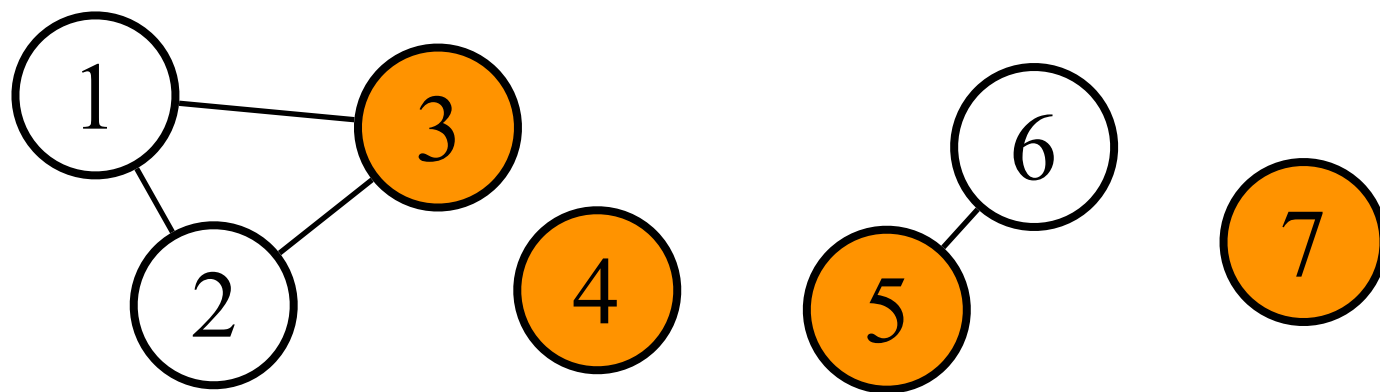
If ($\text{Find-Set}(5) \neq \text{Find-Set}(6)$) $\text{Union}(5, 6)$;

After which, $C = \{\{1, 2, 3\}, \{4\}, \{5, 6\}, \{7\}\}$.



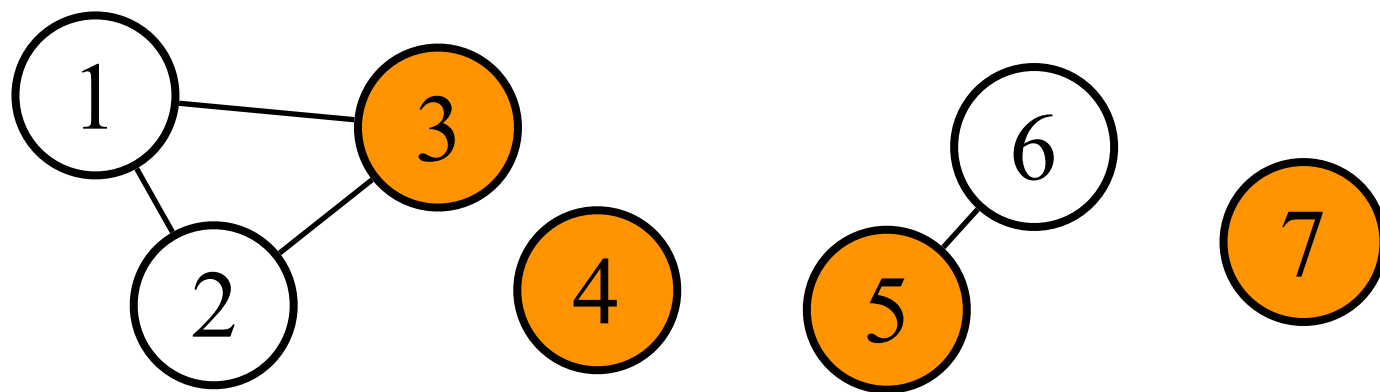
Deciding whether two nodes are in the same connected component

```
Same-Component(u, v) {  
    if(Find-Set(u) = Find-Set(v)) {  
        return True;  
    } else {  
        return False;  
    }  
}
```



Deciding whether two nodes are in the same connected component

```
Same-Component(u, v) {  
  if(Find-Set(u) = Find-Set(v)) {  
    return True;  
  } else {  
    return False;  
  }  
}
```

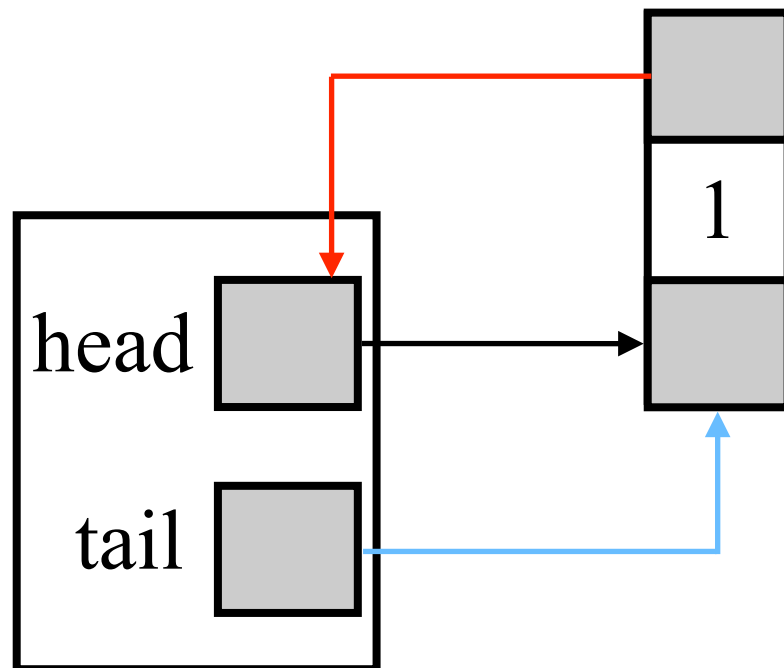


Same-Component(1, 2) returns True because
Find-Set(1) = 3 and Find-Set(2) = 3.

Linked-list Representation of Disjoint Sets

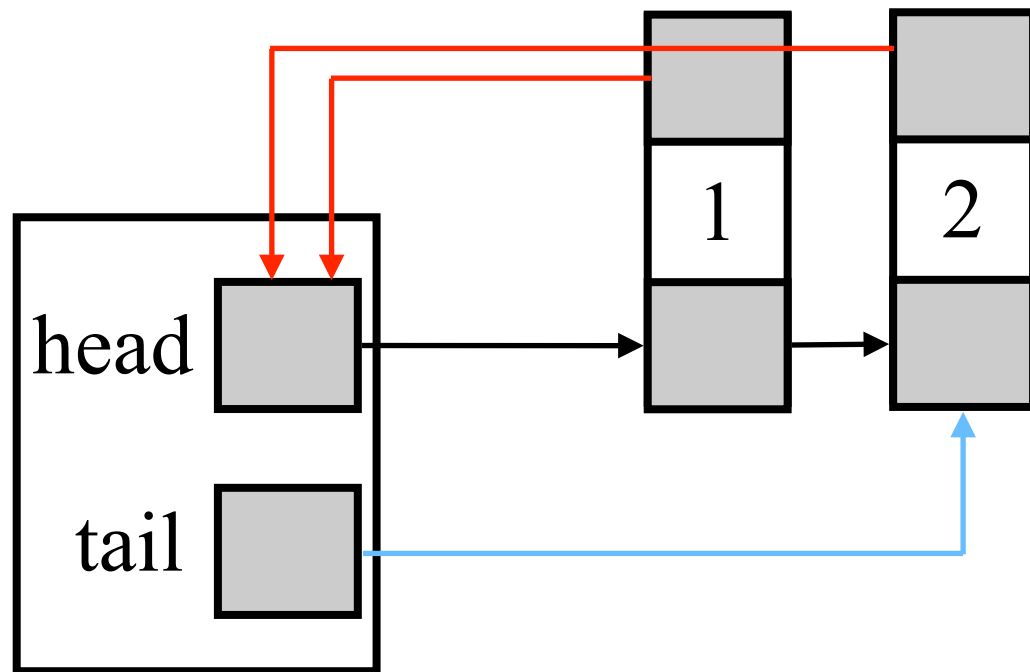
Implementation by linked-lists

Make-Set(1): // $O(1)$ time



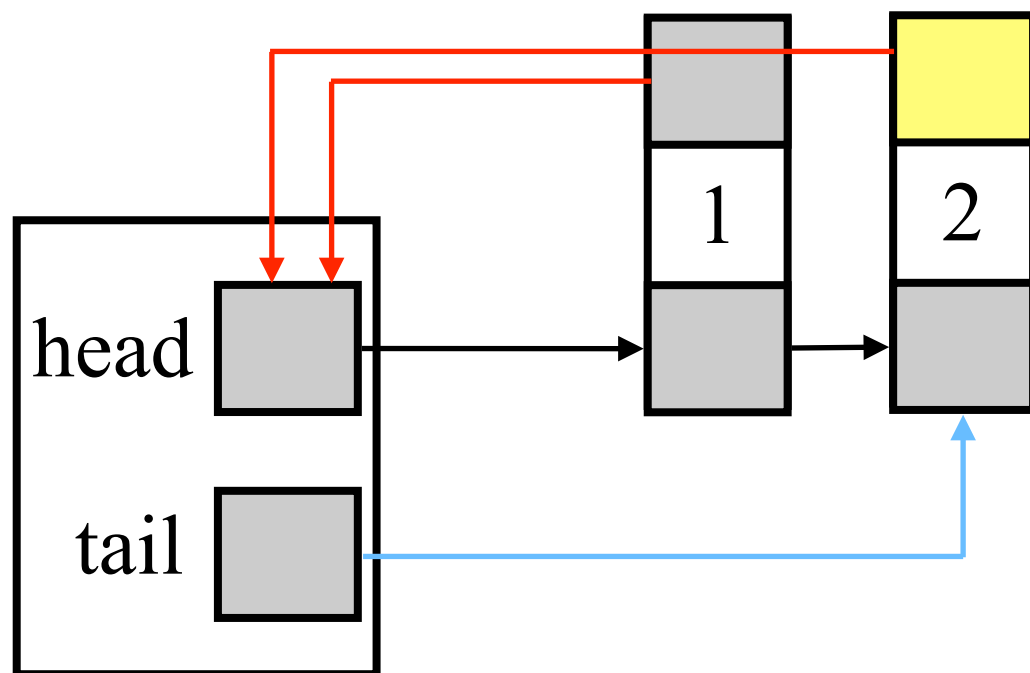
Implementation by linked-lists

Find-Set(2): // $O(1)$ time. Let the first element on the linked list be the representative.



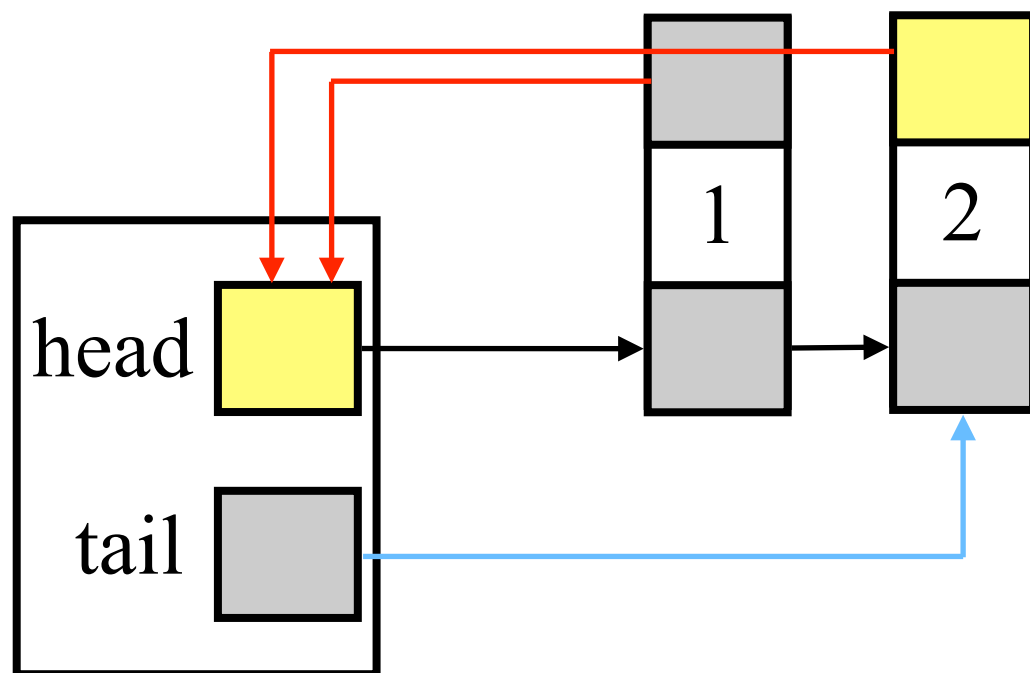
Implementation by linked-lists

Find-Set(2): // $O(1)$ time. Let the first element on the linked list be the representative.



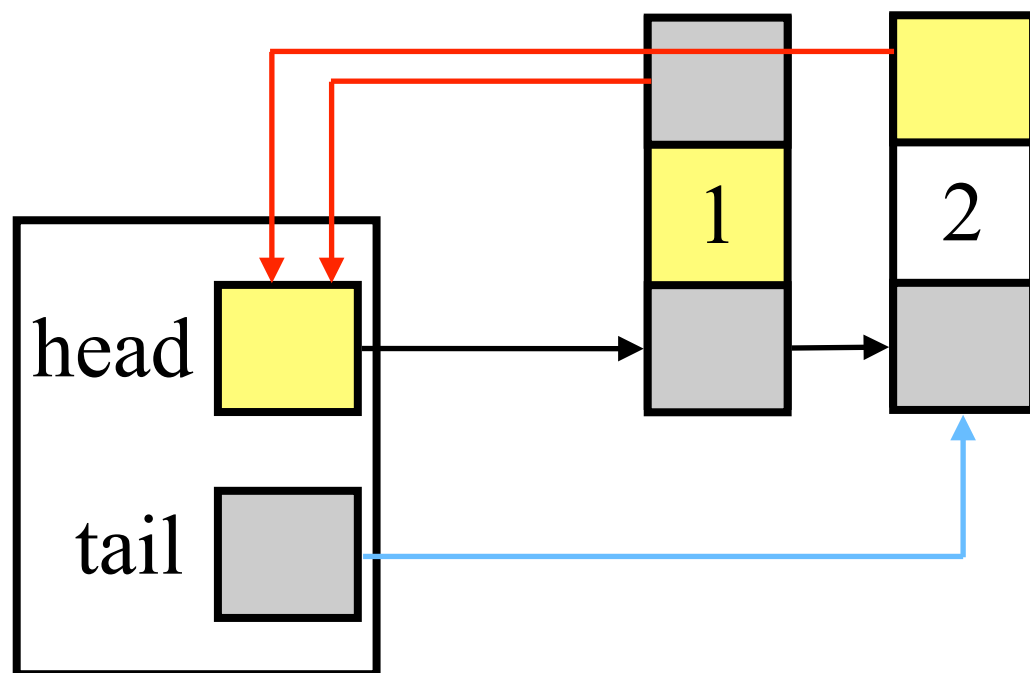
Implementation by linked-lists

Find-Set(2): // $O(1)$ time. Let the first element on the linked list be the representative.



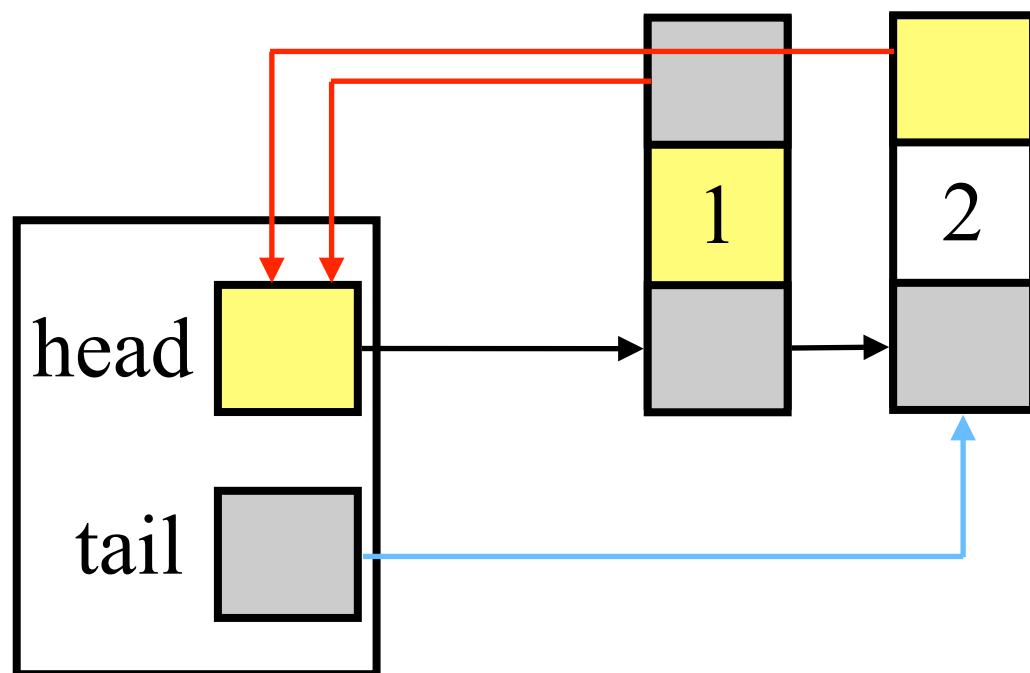
Implementation by linked-lists

Find-Set(2): // $O(1)$ time. Let the first element on the linked list be the representative.



Implementation by linked-lists

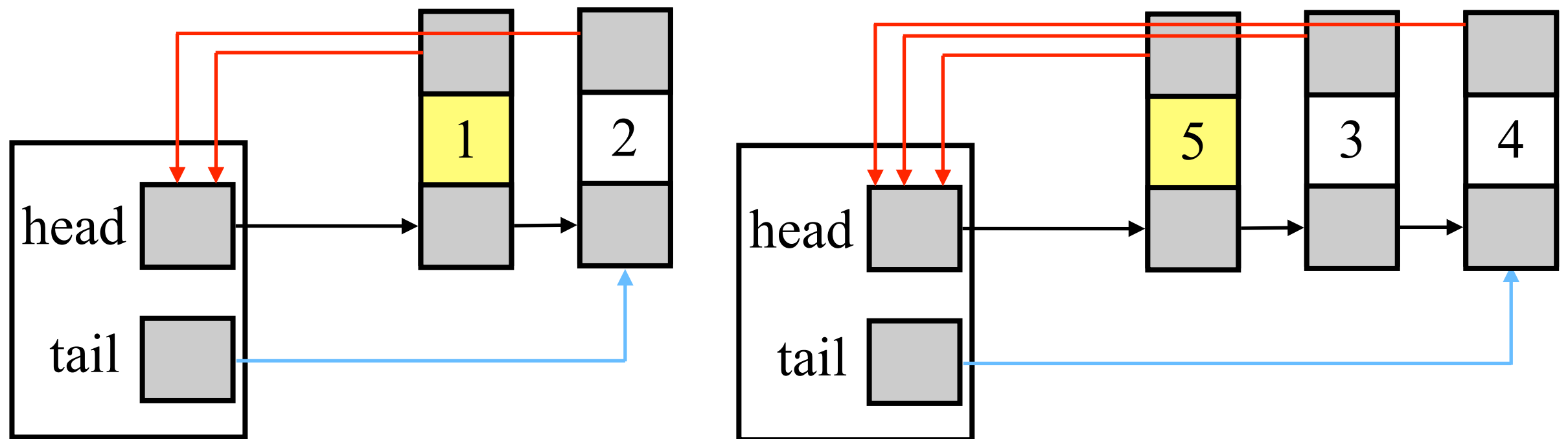
Find-Set(2): // $O(1)$ time. Let the first element on the linked list be the representative.



Find-Set(2) returns 1. It takes $O(1)$ steps.

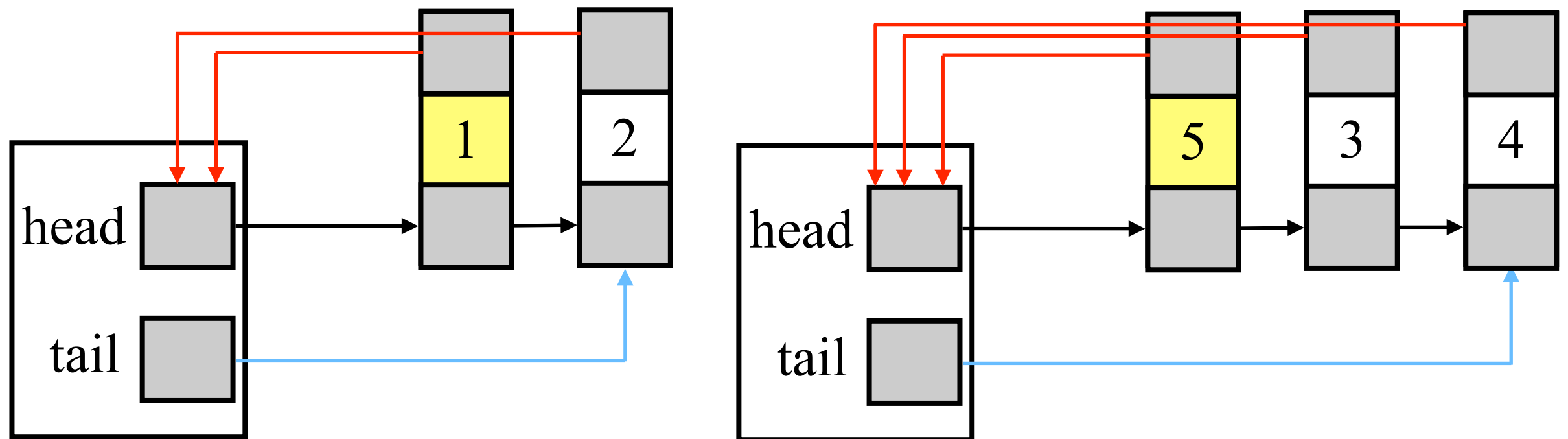
Implementation by linked-lists

Union(2, 4): // $O(\text{length of one list})$



Implementation by linked-lists

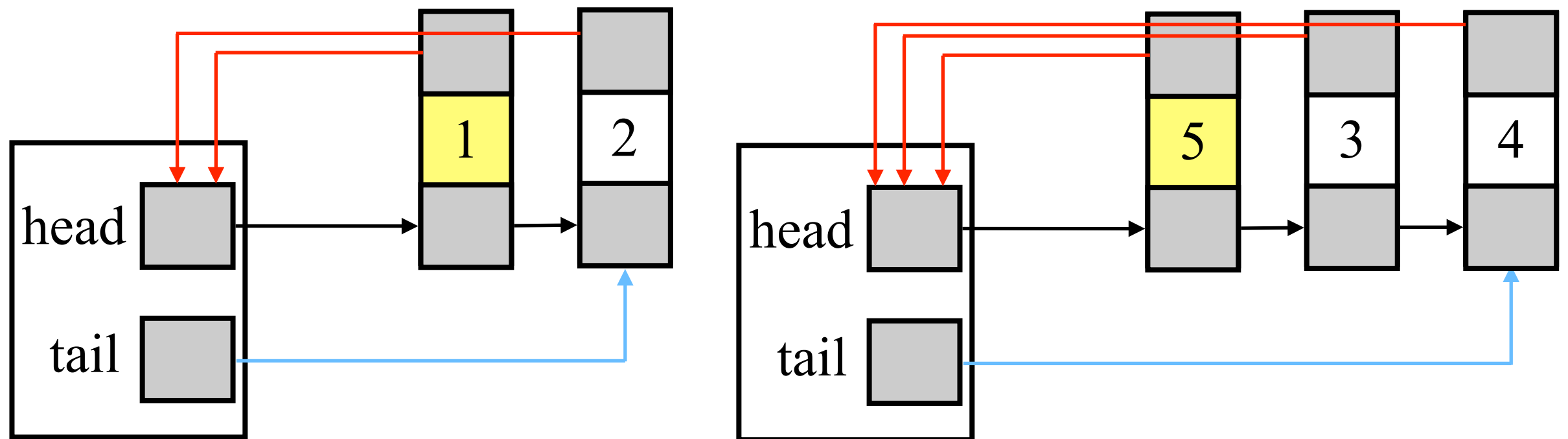
Union(2, 4): // $O(\text{length of one list})$



If(Find-Set(2) \neq Find-Set(4))
Link(Find-Set(2), Find-Set(4));

Implementation by linked-lists

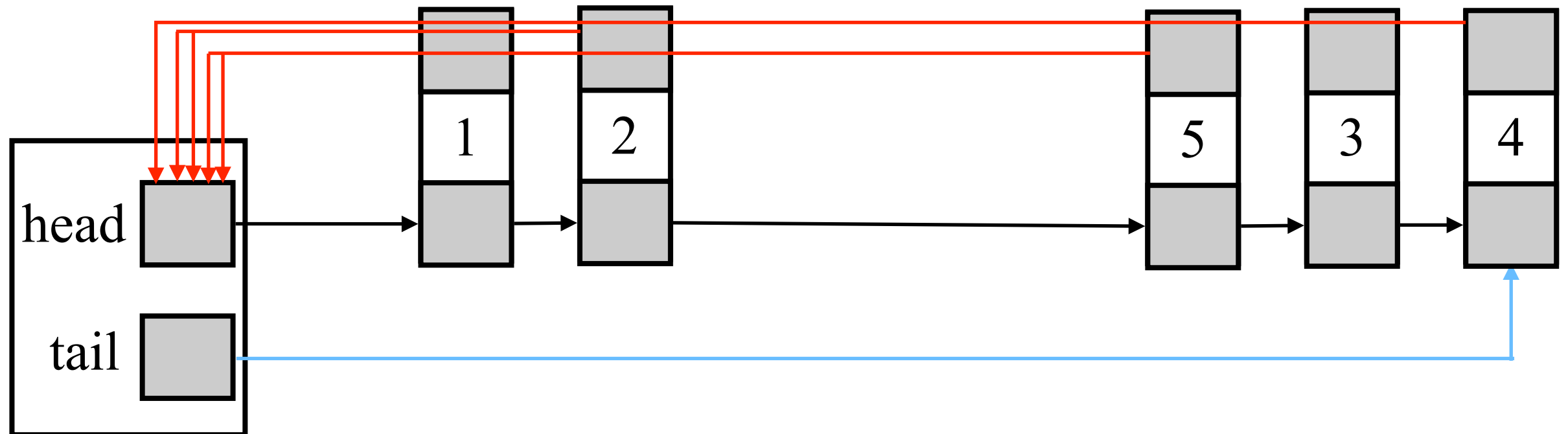
Union(2, 4): // $O(\text{length of one list})$



It needs to modify all the head-pointers of one list to the new head.

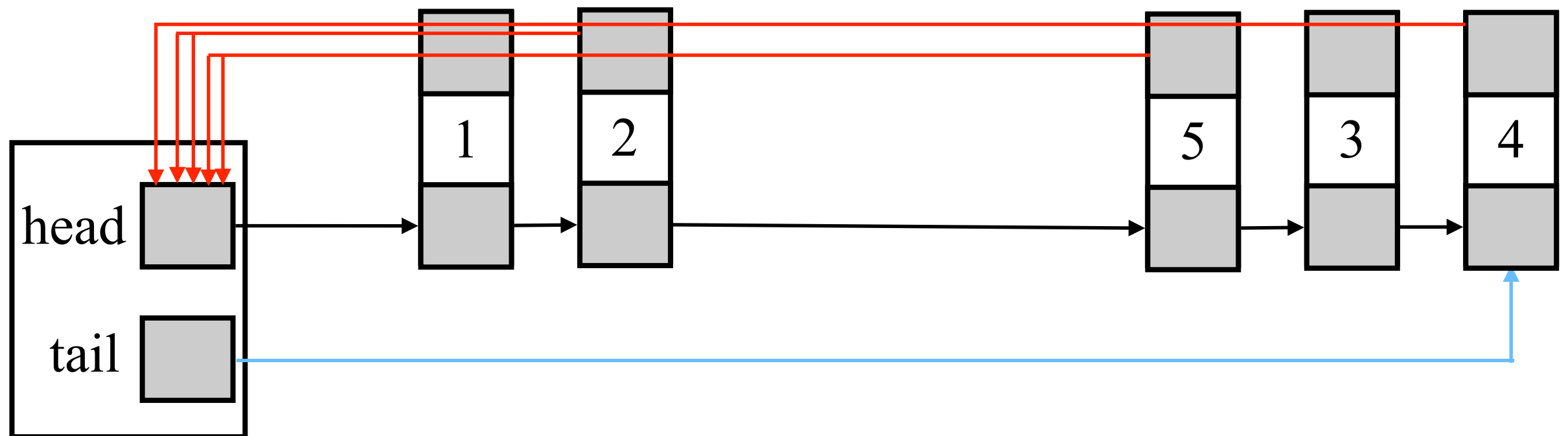
Implementation by linked-lists

Union(2, 4): // $O(\text{length of one list})$



Implementation by linked-lists

Union(2, 4): // $O(\text{length of one list})$



Alternatively, we could append the shorter list to the longer one, which we call **weighted-union heuristic**.

Running time of using linked-list representation and weighted-union heuristic

Claim. It takes $O(n \log n + m)$ running time to perform any sequence of m Make-Set(x), Find-Set(x), Union(x, y) operations, in which there are n Make-Set(x) operations.

Proof. Every operation runs in $O(1)$ time, except that a Union(x, y) operation may update $\omega(1)$ head-pointers.

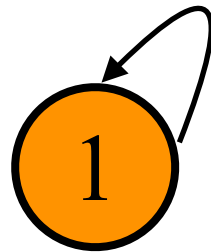
For each element x , every time we link its head-pointer to the new head, the length of x 's list is doubled (because x is on the shorter list). Hence, in all $(n-1)$ Union operations, x 's head-pointer is updated at most $O(\log n)$ times.

In total, the head-pointers of the n elements are updated at most $O(n \log n)$ times. We thus get the $O(m+n \log n)$ bound.

Forest Representation of Disjoint Sets

Implementation by forests

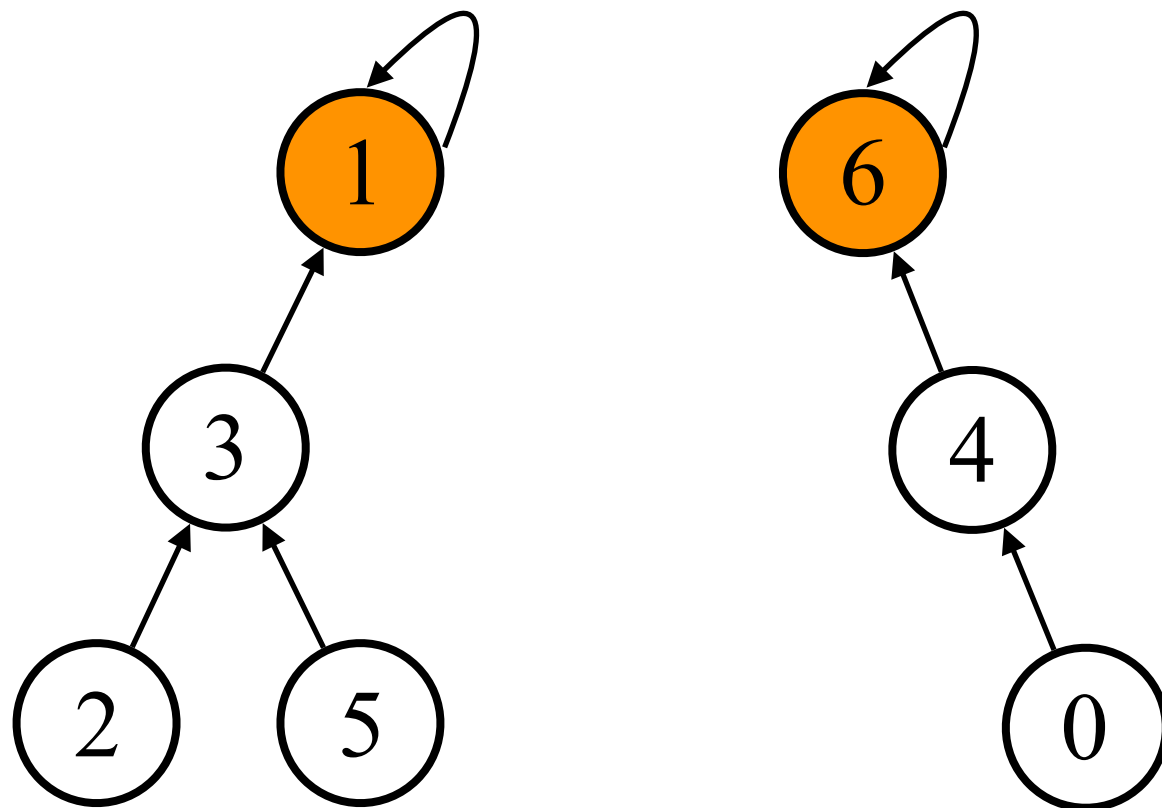
Make-Set(1): // $O(1)$ time



```
Make-Set(x) {  
    x.parent = x;  
    x.rank = 0; // x.rank is an  
                inaccurate upper bound of  
                the height of x (# edges in  
                the longest simple path  
                between x and any  
                descendant leaf)  
}
```

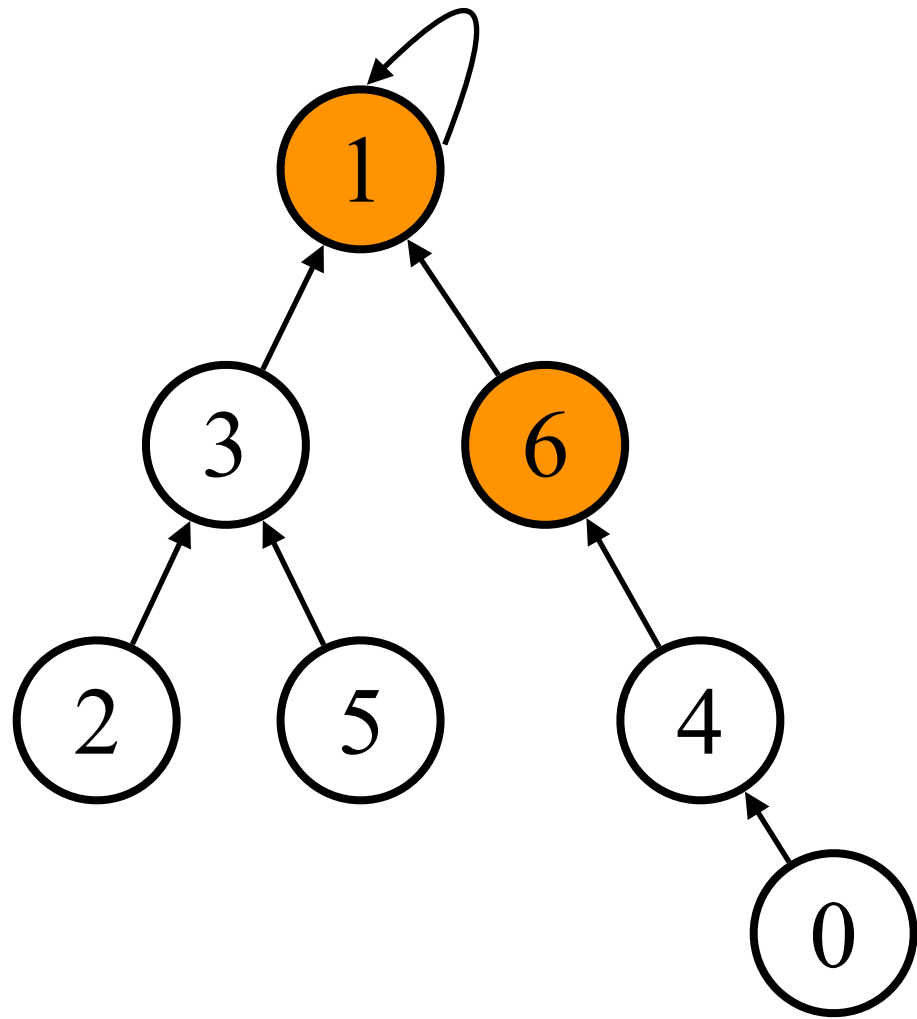
Implementation by forests

Union(2, 4): // $O(1)$



Implementation by forests

Union(2, 4): // $O(1)$

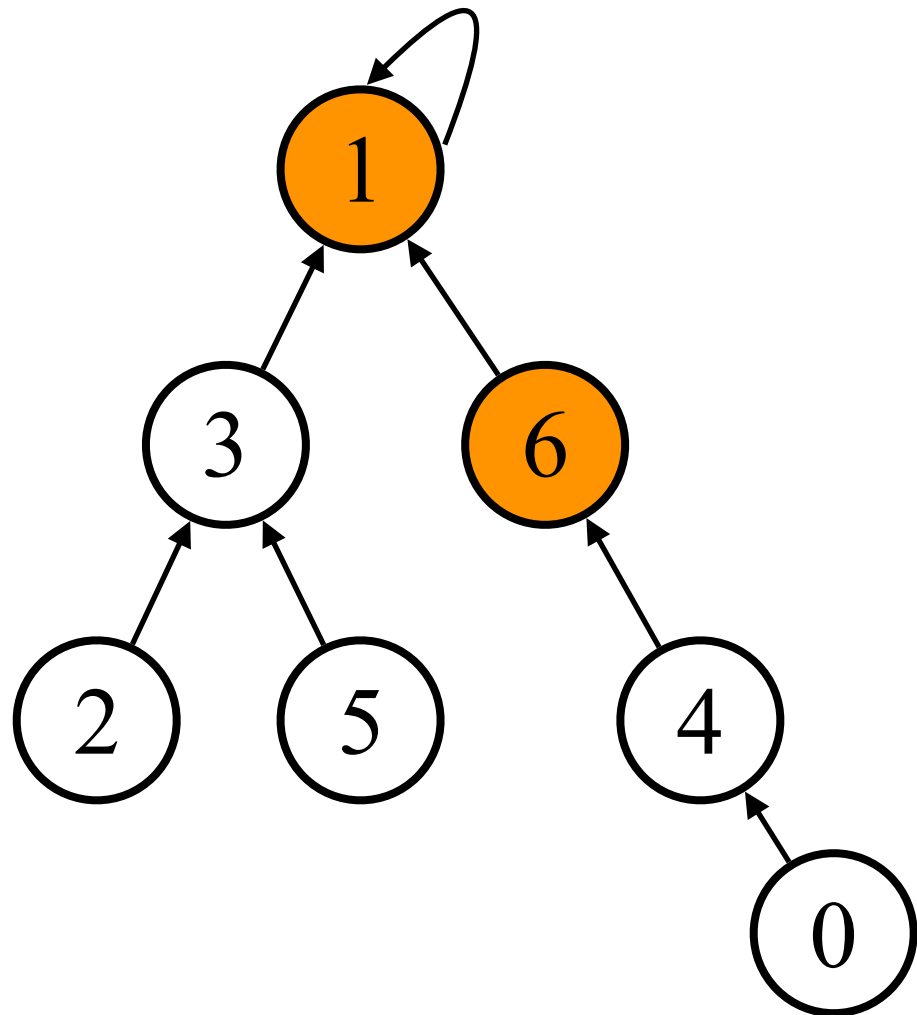


```
Union(x, y){
    if(Find-Set(x) ≠ Find-Set(y)){
        Link(Find-Set(x), Find-Set(y));
    }
}

Link(x, y){
    if(x.rank > y.rank){
        y.parent = x;
    }else{
        x.parent = y;
        if(x.rank equals y.rank)
            y.rank ← y.rank + 1;
    }
}
```


Implementation by forests

Union(2, 4): // $O(1)$

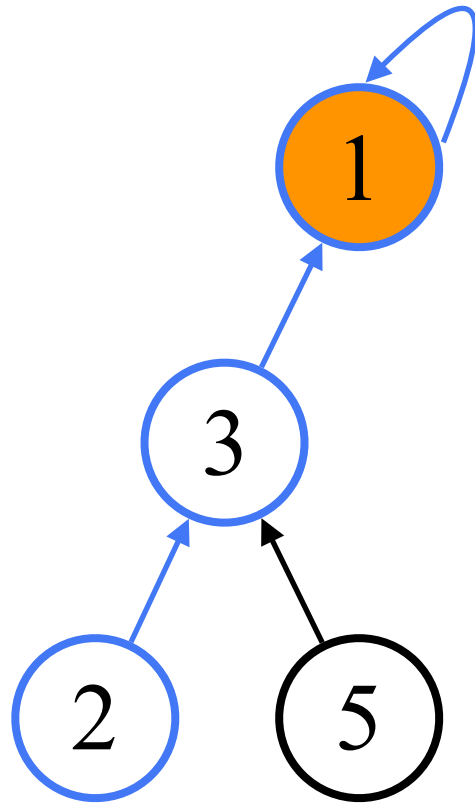


Similarly, we could append the shorter tree to the taller one, which we call **union by rank**.

```
Union(x, y) {  
    if (Find-Set(x) ≠ Find-Set(y)) {  
        Link(Find-Set(x), Find-Set(y));  
    }  
}  
  
Link(x, y) {  
    if (x.rank > y.rank) {  
        y.parent = x;  
    } else {  
        x.parent = y;  
        if (x.rank equals y.rank)  
            y.rank ← y.rank + 1;  
    }  
}
```

Implementation by forests

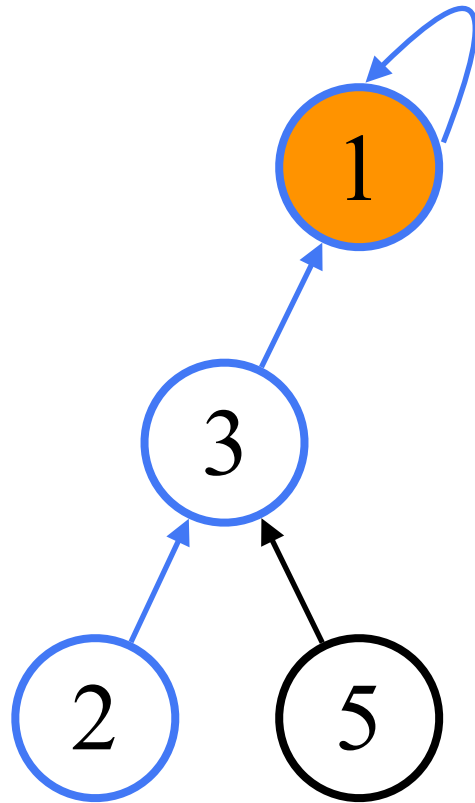
Find-Set(2): // $O(\text{height of the tree})$ time



```
Find-Set(x) {  
    while(x.parent  $\neq$  x) {  
        x = x.parent;  
    }  
    return x; // the representative  
}
```

Implementation by forests

Find-Set(2): // $O(\text{height of the tree})$ time



```
Find-Set(x) {  
    while(x.parent  $\neq$  x) {  
        x = x.parent;  
    }  
    return x; // the representative  
}
```

Using forest representation and union by rank, the height of tree is at most $O(\log n)$.

Exercise

Prove that the height of trees is $O(\log n)$ in the forest representation that uses union by rank to merge two trees.

Running time of using forest representation and union by rank heuristic

Claim. It takes $O(m \log n)$ running time to perform any sequence of m Make-Set(x), Find-Set(x), Union(x, y) operations, in which there are n Make-Set(x) operations.

Proof. Make-Set(x) and Union(x, y) operations need $O(1)$ time each. Find-Set(x) can be done in $O(\log n)$ time. We thus get the $O(m \log n)$ bound.

Running time of using forest representation and union by rank heuristic

Claim. It takes $O(m \log n)$ running time to perform any sequence of m Make-Set(x), Find-Set(x), Union(x, y) operations, in which there are n Make-Set(x) operations.

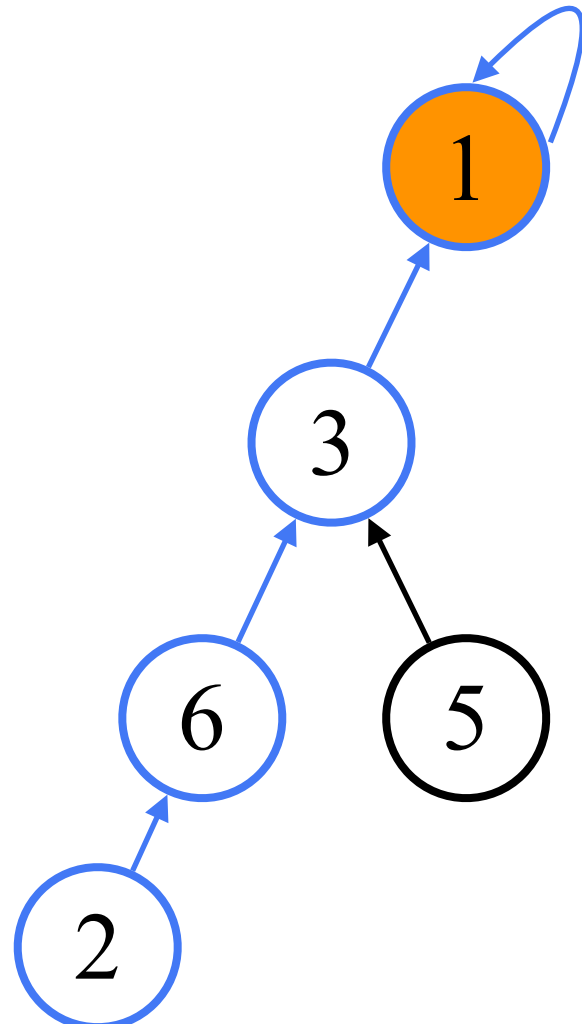
Proof. Make-Set(x) and Union(x, y) operations need $O(1)$ time each. Find-Set(x) can be done in $O(\log n)$ time. We thus get the $O(m \log n)$ bound.

Note that n must be $\leq m$. Hence, $O(m \log n)$ is slower than $O(n \log n + m)$.

Implementation by forests (faster)

Find-Set(2): // $O(\text{height of the tree})$ time. In the meanwhile, we flatten the tree so that Find-Set(x) runs faster next time.

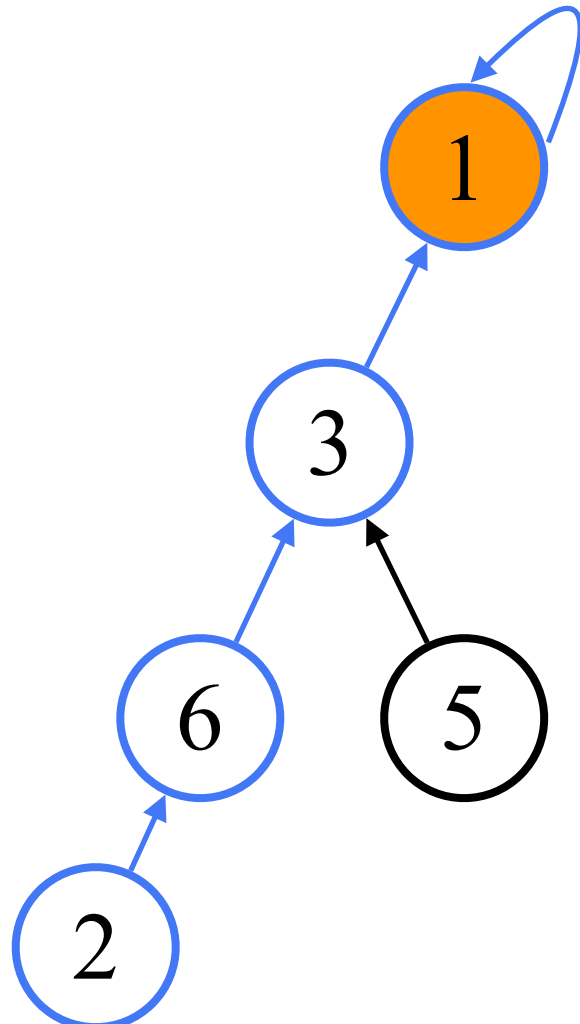
```
Find-Set(x) {  
    if(x.parent  $\neq$  x) {  
        x.parent = Find-Set(x.parent);  
    }  
    return x.parent; // the representative  
}
```



Implementation by forests (faster)

Find-Set(2): // $O(\text{height of the tree})$ time. In the meanwhile, we flatten the tree so that Find-Set(x) runs faster next time.

```
Find-Set(x) {  
    if(x.parent  $\neq$  x) {  
        x.parent = Find-Set(x.parent);  
    }  
    return x.parent; // the representative  
}
```

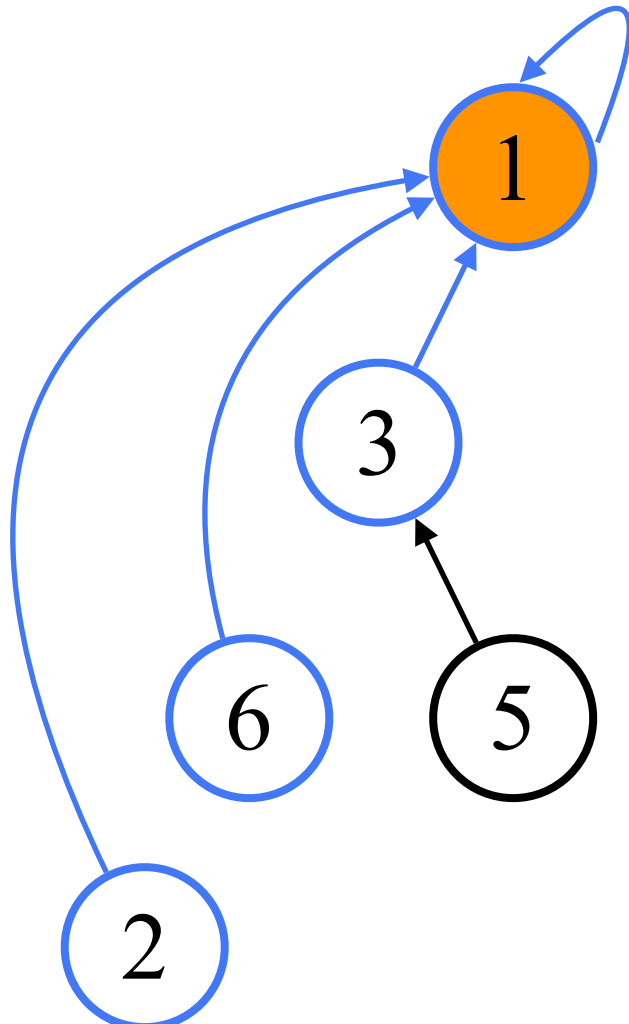


This technique is called
path compression.

Implementation by forests

Find-Set(2): // $O(\text{height of the tree})$ time. In the meanwhile, we flatten the tree so that Find-Set(x) runs faster next time.

```
Find-Set(x) {  
    if(x.parent  $\neq$  x) {  
        x.parent = Find-Set(x.parent);  
    }  
    return x.parent; // the representative  
}
```



This technique is called
path compression.

Running time of using forest representation and union by rank and path compression

Claim. It takes $O(m \alpha(n))$ running time to perform any sequence of m Make-Set(x), Find-Set(x), Union(x, y) operations, in which there are n Make-Set(x) operations.

Proof. The proof can be found in Chap 21.4.

Running time of using forest representation and union by rank and path compression

Claim. It takes $O(m \alpha(n))$ running time to perform any sequence of m Make-Set(x), Find-Set(x), Union(x, y) operations, in which there are n Make-Set(x) operations.

Proof. The proof can be found in Chap 21.4.

Note that $\alpha(n)$ is a very slowly growing function that $\alpha(n) \leq 4$ for any $n \leq 10^{80}$.

Hence, $O(m \alpha(n))$ is virtually linear in practice.

A very quickly growing function $A_k(j)$

$$A_k(j) = \begin{cases} j + 1 & \text{if } k = 0 \\ A_{k-1}^{(j+1)}(j + 1) & \text{if } k > 1 \end{cases}$$

where $A_{k-1}^{(0)}(j) = j$ and $A_{k-1}^{(i)}(j) = A_{k-1}(A_{k-1}^{(i-1)}(j))$

A very quickly growing function $A_k(j)$

$$A_k(j) = \begin{cases} j + 1 & \text{if } k = 0 \\ A_{k-1}^{(j+1)}(j + 1) & \text{if } k > 1 \end{cases}$$

where $A_{k-1}^{(0)}(j) = j$ and $A_{k-1}^{(i)}(j) = A_{k-1}(A_{k-1}^{(i-1)}(j))$

By simple calculations, we have:

- (1) $A_1(j) = 2j+1$ and $A_1(1) = 3$
- (2) $A_2(j) = 2^{j+1}(j+1) - 1$ and $A_2(1) = 7$
- (3) $A_3(1) = 2047$
- (4) $A_4(1) \gg 10^{80}$
- (5) $A_1(1) < A_2(1) < A_3(1) < \dots$

A very slowly growing function $\alpha(n)$

Let $\alpha(n) = \min \{k : A_k(1) \geq n\}$, the inverse of A_k .

A very slowly growing function $\alpha(n)$

Let $\alpha(n) = \min \{k : A_k(1) \geq n\}$, the inverse of A_k .

For every $n \leq 10^{80}$, $\alpha(n) \leq 4$.

In practice

Running time of using forest representation and ~~union by rank~~ and path compression

Claim. It takes $O(n + f(1 + \log_{2+f/n} n))$ running time to perform any sequence of n Make-Set(x), f Find-Set(x), and $<n$ Union(x, y) operations.

Running time of using forest representation and ~~union by rank~~ and path compression

Claim. It takes $O(n + f(1 + \log_{2+f/n} n))$ running time to perform any sequence of n Make-Set(x), f Find-Set(x), and $<n$ Union(x, y) operations.

Path compression is already efficient enough for the practical use. You **may** ignore union by rank.

Pseudocode of using forest representation and ~~union by rank~~ and path compression

```
int rep[1..n];
```

```
for(int i=1; i ≤ n; ++i){  
    rep[i] = i; // set i to be the representative of the i-th tree  
}
```

```
Find-Set(x, rep){  
    if(rep[x] equals x) return x;  
    return rep[x] = Find-Set(rep[x], rep); // path compression  
}
```

```
Union(x, y, rep){  
    if(Find-Set(x, rep) ≠ Find-Set(y, rep))  
        rep[rep[x]] = rep[y]; // unite arbitrarily  
}
```