

# Lecture #13

## Disjoint Set

---

Algorithm

JBNU

Jinhong Jung

# In This Lecture

---

## □ Advanced data structure - disjoint set

- What is the disjoint set?
- How to represent and implement disjoint sets?
  - Basic version of disjoint set
- How to improve efficiency?
  - Union by rank
  - Path compression

# Outline

---

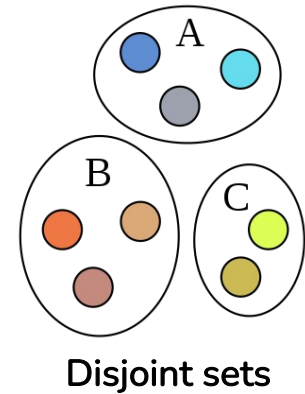
- ❑ Definition of disjoint set
- ❑ Disjoint set using non-binary tree
- ❑ How to improve efficiency?
- ❑ Analysis of disjoint set

# Disjoint Set

---

## □ What is disjoint set?

- Disjoint set is a data structure managing **non-overlapping** sets
  - A **set** is used to contain unique objects.
  - Each intersection of two sets are empty.



## □ Applications

- Used when we need to manage multiple partitions or groups in a problem
  - Connected components in a graph
  - Minimum spanning tree in a graph (Kruskal's algorithm)

# Main Operations

---

## □ make-set(u)

- Create a new set containing only given element u

## □ find-set(u)

- Return the set containing given element u

## □ union(u, v)

- Merge (or union) the set having u and the set having v

## □ Notes

- No need to consider intersect operation in disjoint set
- Due to the operations, it's also known as **union-find**.

# Outline

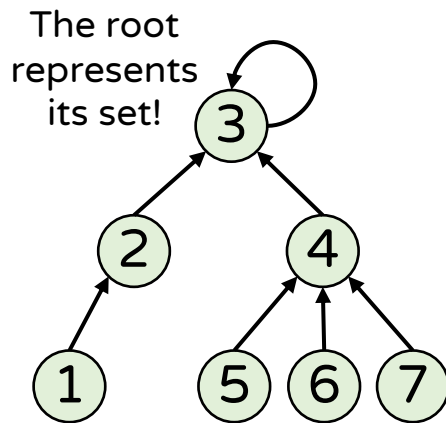
---

- ❑ Definition of disjoint set
- ❑ Disjoint set using non-binary tree
- ❑ How to improve efficiency?
- ❑ Analysis of disjoint set

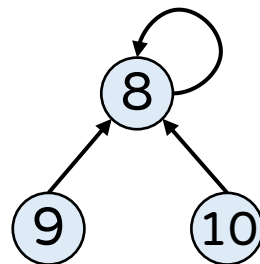
# How To Represent Disjoint Set

□ A disjoint set is represented by a non-binary tree.

- Unlike normal trees, we use **parent pointer tree**.
  - A child points to its parent, and the root points to itself (self-looped).
- Each tree represents a set (i.e., forest = multiple sets)
- This tree is implemented by an 1D array called  $p$ .
  - Assume an element in the set is a positive integer.



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



$\{8, 9, 10\}$

	Node's key									
Index	1	2	3	4	5	6	7	8	9	10
Value	2	3	3	3	4	4	4	8	8	8
	Parent									

$p[u]$ : parent of node  $u$

# Main Operations of Basic Version

## □ make-set(u)

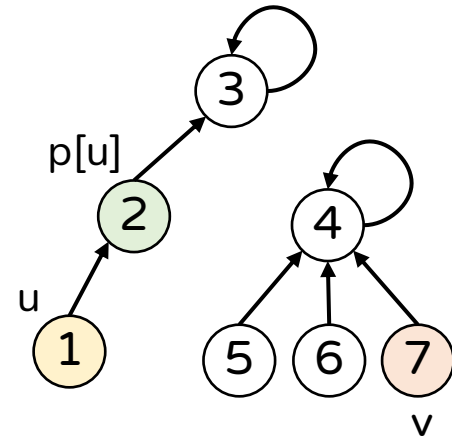
- It's implemented as u's parent points to u

## □ find-set(u)

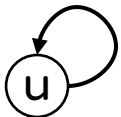
- Return the root of the set containing given u
- Recursively walk up from u to the root

## □ union(u, v)

- Merge the set having u and the set having v
  - Let the root of one set point to the root of other set



```
def make-set(u):
    p[u] ← u
```



```
def find-set(u):
    if u is p[u]: # if self-looped,
                  # it's root
        return u
    else:
        return find-set(p[u]) # go up one level
```

```
def union(u, v):
    p[find-set(v)] ← find-set(u)
    # v's root points to u's root
```



# Analysis of Basic Version

---

## □ Space complexity

- It takes  $\Theta(n)$  space because of the array p.

## □ Time complexity

- make-set(u) takes  $\Theta(1)$  time.
- find-set(u) takes  $\Theta(h_u)$  time.
  - $h_u$  is the height of the tree having u
- union(u, v) takes  $h_v + h_u + c$  time.

## □ For a worst case, find-set(u) takes $\Theta(n)$ time.

- When the tree of  $n$  nodes becomes degenerate
- Can we improve this even for such a worst case?

# Outline

---

- ❑ Definition of disjoint set
- ❑ Disjoint set using non-binary tree
- ❑ How to improve efficiency?
- ❑ Analysis of disjoint set

# How To Improve Efficiency?

---

□ The efficiency of disjoint set can be improved

- By reducing the height of each tree
- Because main operations totally depends on the tree height

□ Two techniques can be used for the purpose

⇒ Union by rank

- Idea: smaller tree is merged into taller tree in union

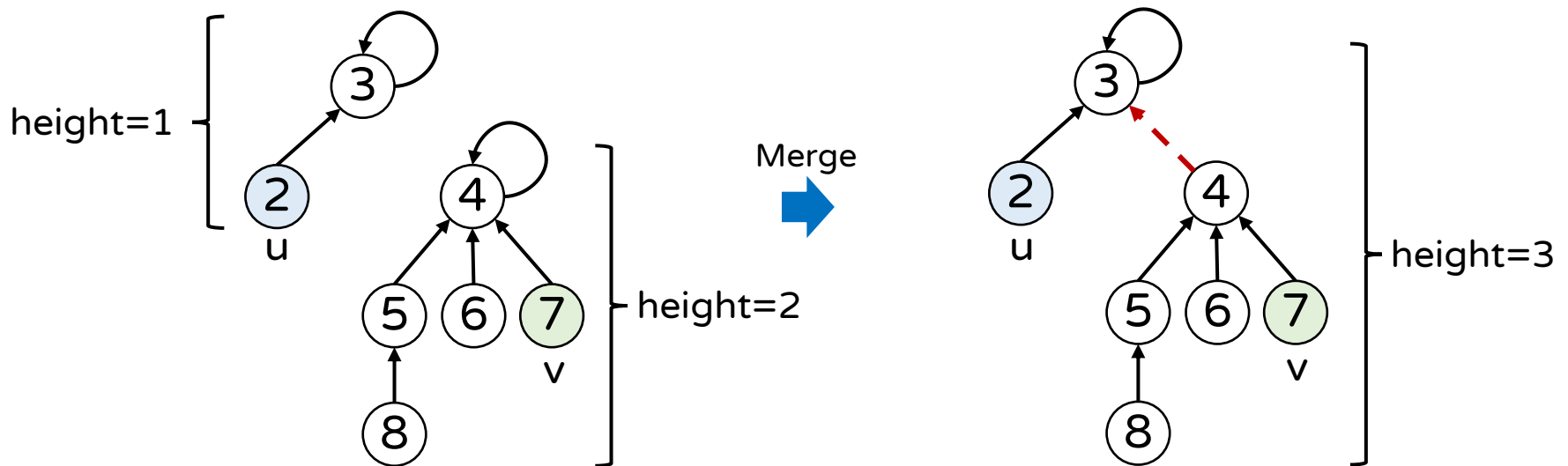
▪ Path compression

- Idea: flatten the tree while walking up to the root in find-set

# Union By Rank (1)

## □ When does the tree's height increase?

- It increases while we merge two disjoint sets –  $\text{union}(u, v)$
- Suppose the right tree is merged into the left one.

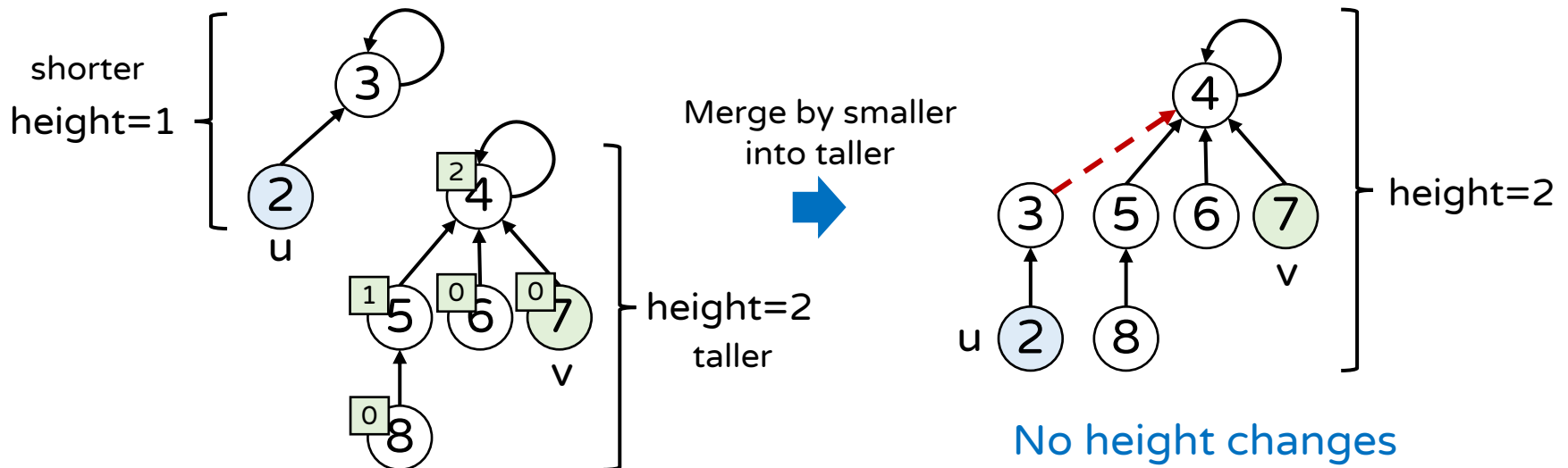


- What if the left tree is merged into the right one?
  - Then, the height doesn't change.

# Union By Rank (2)

## □ Smaller into taller strategy

- Let's merge the shorter tree into the taller tree



- To check the tree's height quickly, let's store a variable for each node, called **rank**.

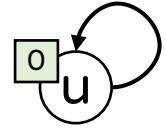
# Union By Rank (3)

---

## □ make-set(u)

- Make one disjoint set of u

```
def make-set(u):  
    p[u] ← u  
    rank[u] ← 0
```



## □ union(u, v)

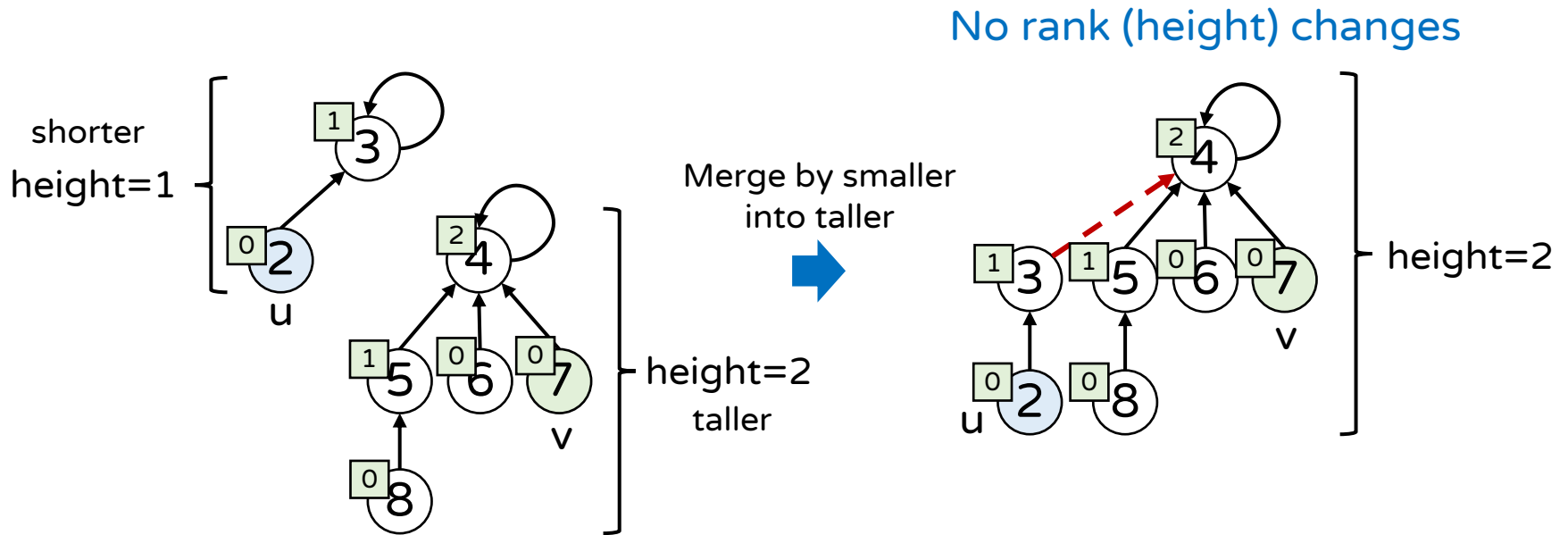
- Merge the set having u and the set having v by smaller into larger strategy

```
def union(u, v):  
    ur ← find-set(u) # ur is the root node of the set having u  
    vr ← find-set(v)  
    if rank[ur] > rank[vr]: # the tree of ur is taller than that of vr  
        p[vr] ← ur # the tree of vr is merged into that of ur  
    else:  
        p[ur] ← vr  
        if rank[ur] == rank[vr]: # the tree of ur has the same height as that of vr  
            rank[vr] ← rank[vr] + 1 # the resulting height increases by 1
```

# Examples (1)

## □ When the height does not change after merge

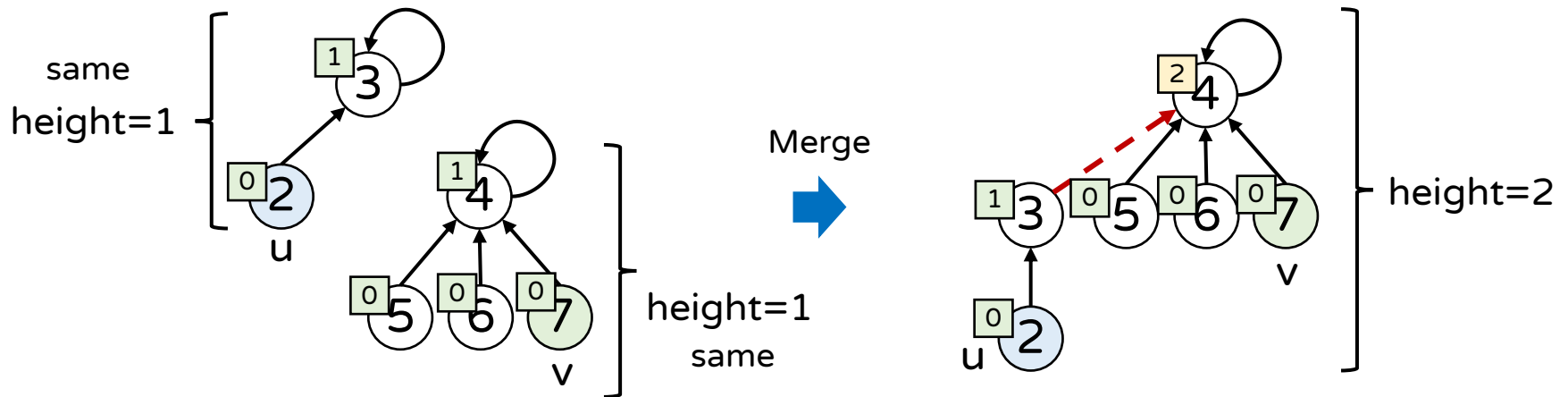
- If their heights are different, the height of the merged tree don't change.



# Examples (2)

## □ When the height changes

- If their heights are the same, the height of the merged tree increases by 1.





# How To Improve Efficiency?

---

## ❑ The efficiency of disjoint set can be improved

- By reducing the height of each tree
- Because main operations totally depends on the tree height

## ❑ Two techniques can be used for the purpose

- Union by rank
  - Idea: smaller tree is merged into taller tree in union

## ⇒ Path compression

- Idea: flatten the tree while walking up to the root in find-set

# Path Compression

---

❑ Even though we use union-by-rank, the tree's height can increase during the union operation

- When the height of the sets to be merged is the same
- Where else can we reduce the tree's height?

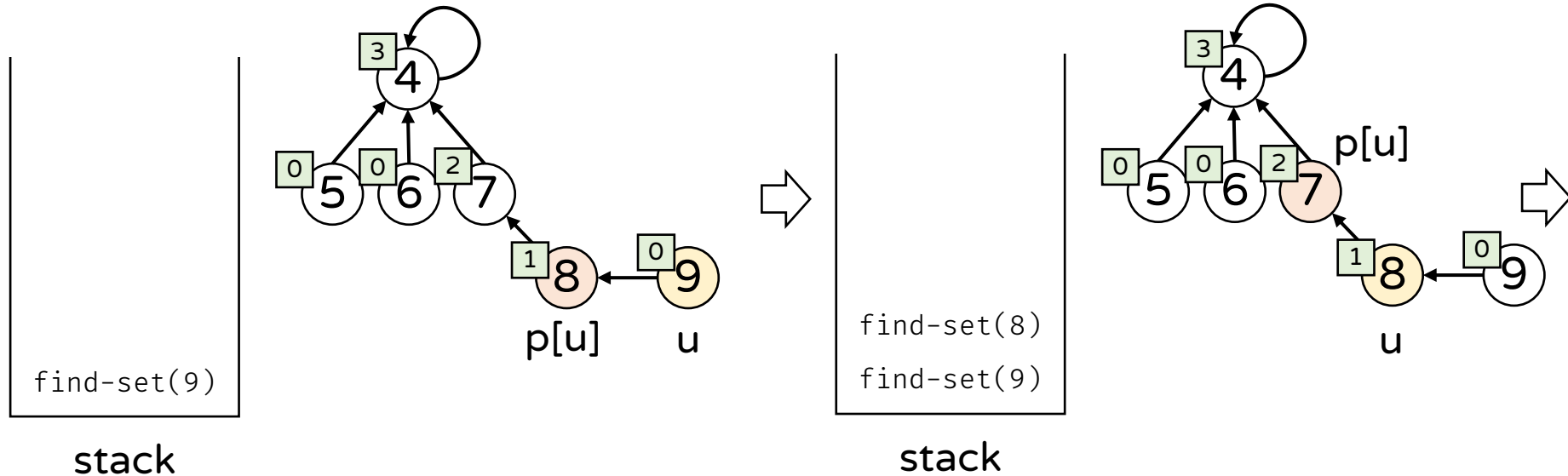
❑ Path compression's idea: Let's flatten the tree

- Every time we walk up the tree during find-set, let's re-assign parent pointers to make each node we pass a direct child of the root

# Examples (1)

□ Flatten the tree during the find-set operation!

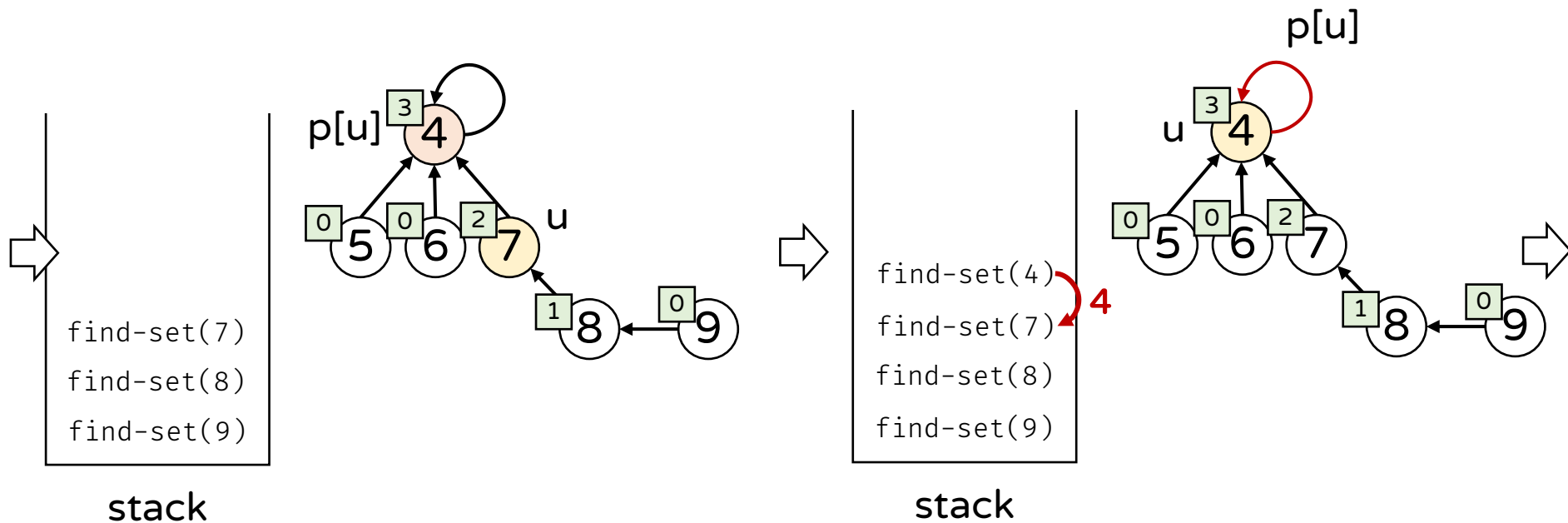
```
def find-set(u):  
    if p[u] != u:  
        p[u] ← find-set(p[u])  
    return p[u]
```



# Examples (2)

## □ Flatten the tree during the find-set operation!

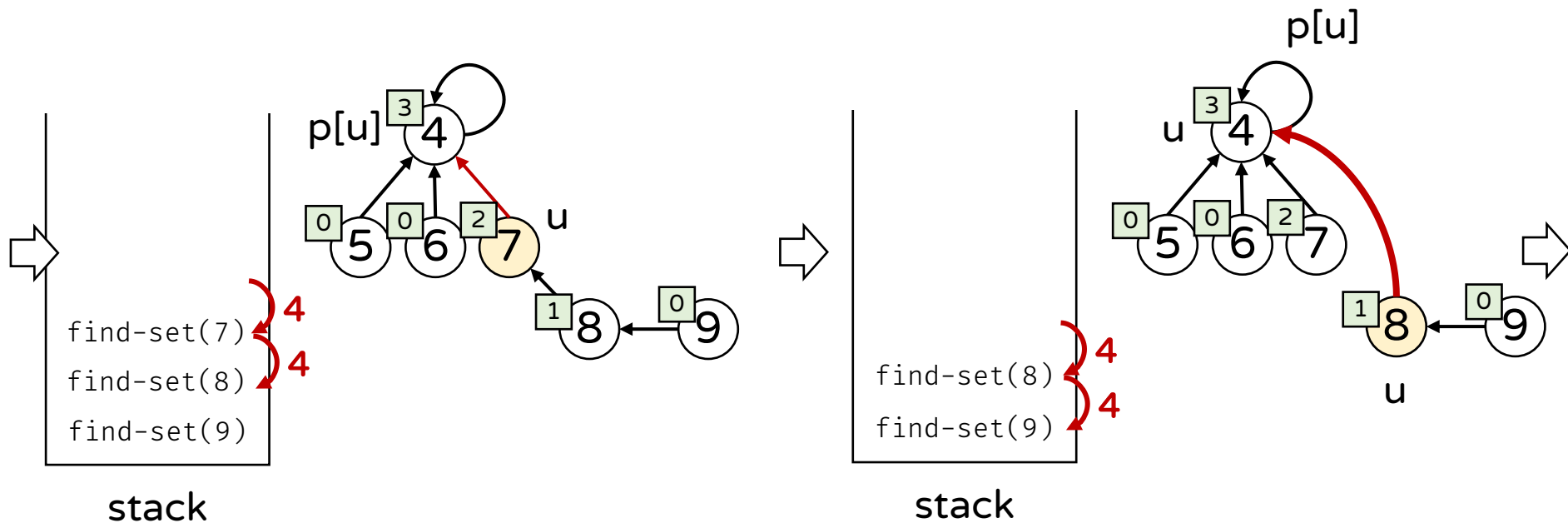
```
def find-set(u):  
    if p[u] != u:  
        p[u] ← find-set(p[u])  
    return p[u]
```



# Examples (3)

## □ Flatten the tree during the find-set operation!

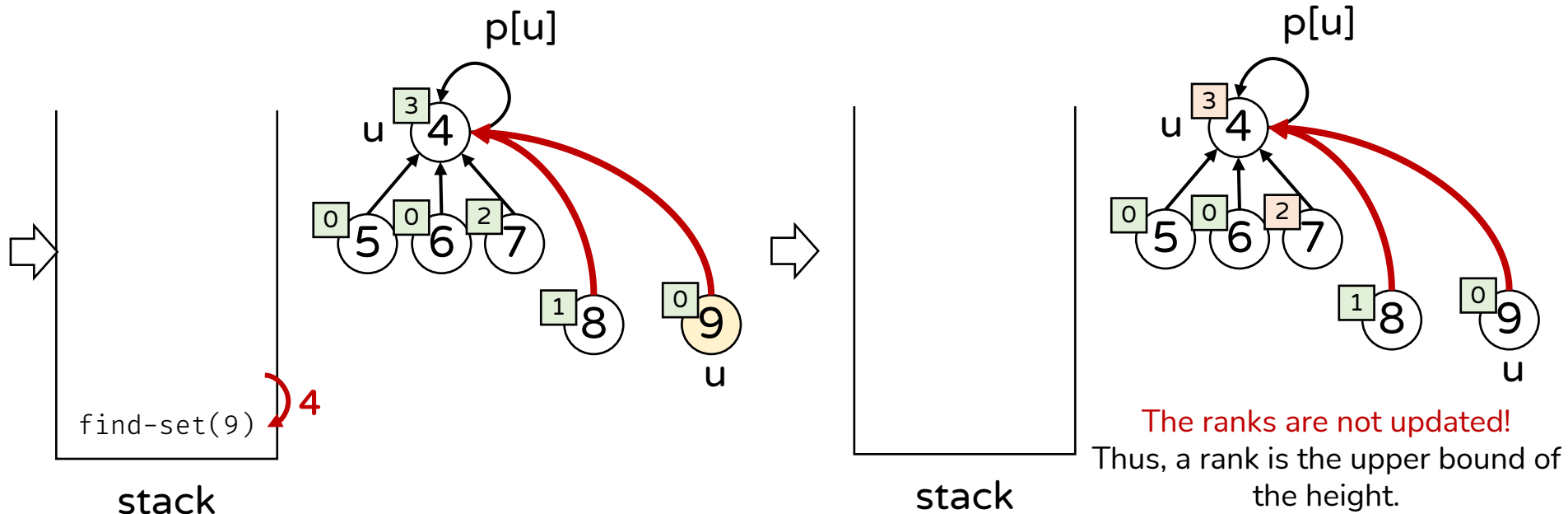
```
def find-set(u):  
    if p[u] != u:  
        p[u] ← find-set(p[u])  
    return p[u]
```



# Examples (4)

## □ Flatten the tree during the find-set operation!

```
def find-set(u):  
    if p[u] != u:  
        p[u] ← find-set(p[u])  
    return p[u]
```



# Outline

---

- ❑ Definition of disjoint set
- ❑ Disjoint set using non-binary tree
- ❑ How to improve efficiency?
- ❑ Analysis of disjoint set

# Analysis of Union By Rank

---

- **Claim:** using union by rank, # of elements in a set represented by a root having rank  $k$  is at least  $2^k$ .
  - **Base case:** If rank = 0,  $2^0 = 1$  element in the set.
  - **Inductive step**
    - Assume the claim holds for rank  $r$ ; then, is it true for rank  $r + 1$ .
    - The rank becomes  $r + 1$  when both ranks of two sets are  $r$ .
    - By the assumption, each set has at least  $2^r$  elements.
    - Thus, the merged set of rank  $r + 1$  has at least  $2^r + 2^r = 2^{r+1}$  elements.
- **Claim:** using union by rank, if the set has  $n$  nodes, then the root of the set for has  $O(\log n)$  rank.
  - Let  $k$  be the root's rank;  $n \geq 2^k \Leftrightarrow k \leq \log_2 n = O(\log n)$ 
    - The height of the tree  $\leq$  rank  $k \leq \log_2 n$



# Analysis of Union By Rank

---

## □ Time complexity of basic version + union-by-rank

- $\text{make-set}(u)$  takes  $O(1)$  time.
- $\text{find-set}(u)$  takes  $O(\log n)$  time.
- $\text{union}(u, v)$  takes  $O(\log n)$  time.

## □ (Amortized) Analysis in a sequence of operations

- Among  $m$  operations consisting of make-set, find-set, and union, let  $n$  be the number of make-set operations.
- Then, the total complexity is  $O(m \log n)$ .
  - Because after  $n$  make-set operations, there are  $n$  nodes; thus, the height of a tree cannot exceed  $O(\log n)$ .
  - Thus,  $m$  times of the above operations takes  $O(m \log n)$

# Analysis of Path Compression

---

## □ (Amortized) Analysis in on a sequence of operations

- Among  $m$  operations consisting of make-set, find-set, and union, let  $n$  be the number of make-set operations.
- The total complexity is  $O(m \log^* n)$  (proof is out-of-scope)
  - $\log^* n = \min\{k \mid \underbrace{\log \log \cdots \log n}_k \leq 1\}$  (repeatedly apply  $\log()$  to  $n$ ,  $k$  times)
  - $\log^* n$  is very small for extremely large  $n$  (e.g.,  $\log^* 2^{65536} = 5$ ).
- $\Rightarrow$  After  $m$  operations, it takes  $O(m)$  time for a worst case.
  - On average, each operation takes  $O(1)$  time!
- **Disjoint-set with union-by-rank and path-compression supports very fast operations.**

# What You Need To Know

---

## ❑ Disjoint set (a.k.a. union-find)

- Data structure managing such non-overlapping sets
- **Main operations**: make-set, find-set, and union
- Represented by a non-binary **parent pointer tree**
  - For positive integer elements, 1D-array is enough for the purpose
- Disjoint set is improved by
  - **Union by rank**: smaller into taller strategy
  - **Path compression**: flatten the tree while walking up to the root
- Disjoint set with both techniques is very fast
  - By amortized analysis, each operation takes  $O(1)$  time!

# In Next Lecture

---

## □ Minimum spanning tree on a graph

- Prim's algorithm
- Kruskal's algorithm

## □ Should review graph representation & basic graph searches

- See the graph section in data structure
  - Adjacency matrix
  - Adjacency list
  - Depth first search
  - Breadth first search

Thank You