



# CSE 332: Data Structures & Parallelism

## Lecture 23: Disjoint Sets

Ruth Anderson  
Autumn 2016

## *Aside: Union-Find aka Disjoint Set ADT*

- **Union(x,y)** – take the union of two sets named x and y
  - Given sets: {3,5,7} , {4,2,8}, {9}, {1,6}
  - **Union(5,1)**  
Result: {3,5,7,1,6}, {4,2,8}, {9},
  - To perform the union operation, we replace sets x and y by  $(x \cup y)$
- **Find(x)** – return the name of the set containing x.
  - Given sets: {3,5,7,1,6}, {4,2,8}, {9},
  - **Find(1)** returns 5
  - **Find(4)** returns 8
- We can do Union in constant time.
- We can get Find to be ***amortized*** constant time (worst case  $O(\log n)$  for an individual Find operation).

# Implementing the DS ADT

- $n$  elements,  
Total Cost of:  $m$  finds,  $\leq n-1$  unions

*can there be  
more unions?*

- Target complexity:  $O(m+n)$   
i.e.  $O(1)$  amortized
- $O(1)$  worst-case for find as well as union would be great, but...  
*Known result:* both find and union *cannot* be done in worst-case  $O(1)$  time

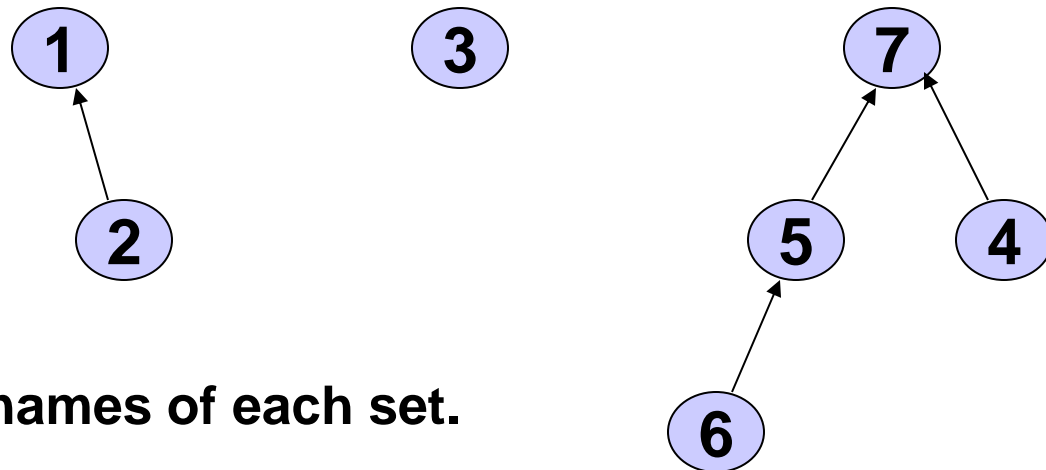
# *Data Structure for the DS ADT*

- **Observation:** trees let us find many elements given one root...
- **Idea:** if we reverse the pointers (make them point up from child to parent), we can find a single root from many elements...
- **Idea:** Use one tree for each equivalence class. The name of the class is the tree root.

# *Up-Tree for Disjoint Union/Find*

Initial state: ① ② ③ ④ ⑤ ⑥ ⑦

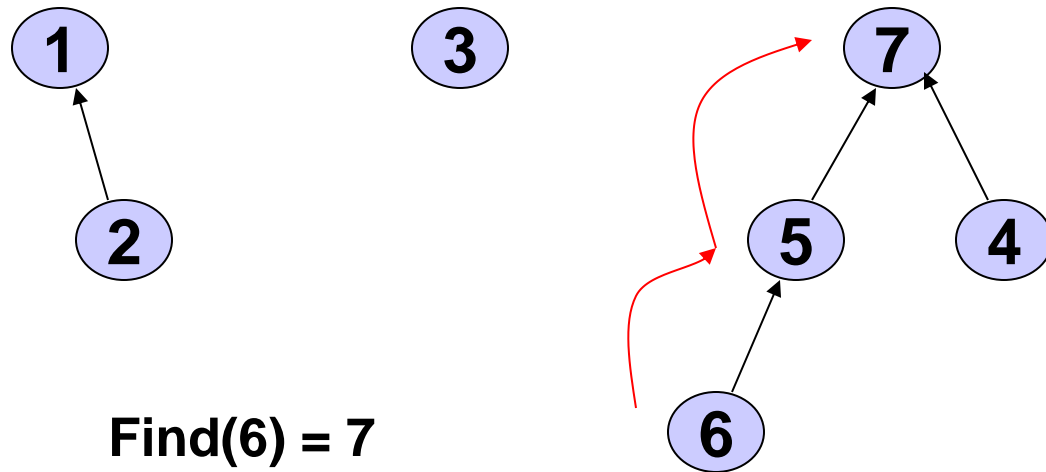
After several  
Unions:



Roots are the names of each set.

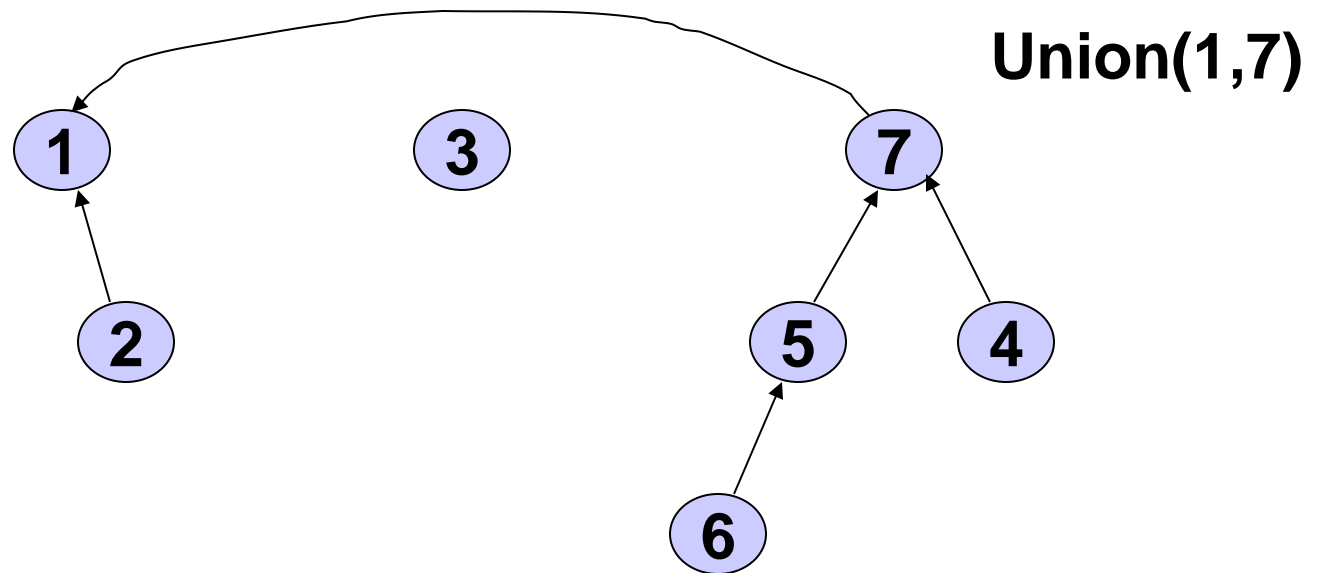
# *Find Operation*

**Find(x)** - follow x to the root and return the root



# *Union Operation*

$\text{Union}(x,y)$  - assuming  $x$  and  $y$  are roots, point  $y$  to  $x$ .

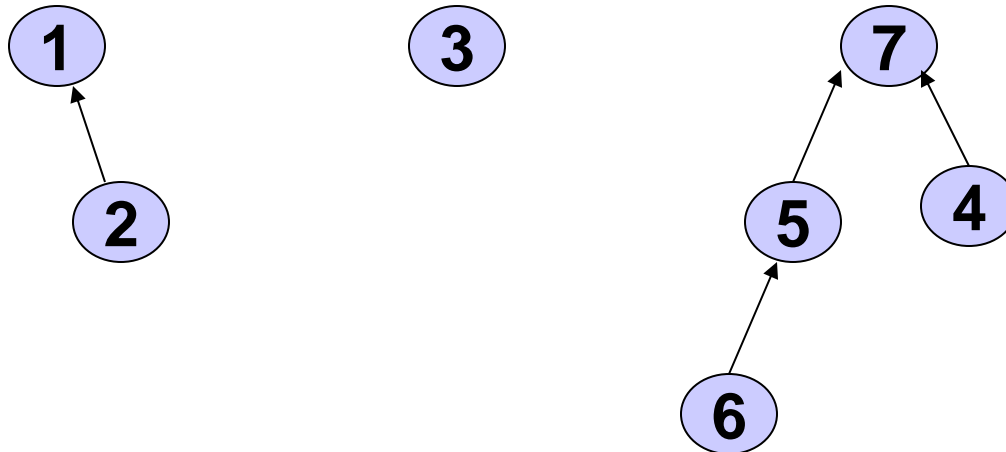


# Simple Implementation

- Array of indices

	1	2	3	4	5	6	7
up	0	1	0	7	7	5	0

**Up[x] = 0 means  
x is a root.**





# *Implementation*

```
int Find(int x) {  
  
    while (up[x] != 0) {  
        x = up[x];  
    }  
  
    return x;  
}
```

```
void Union(int x, int y) {  
    up[y] = x;  
}
```

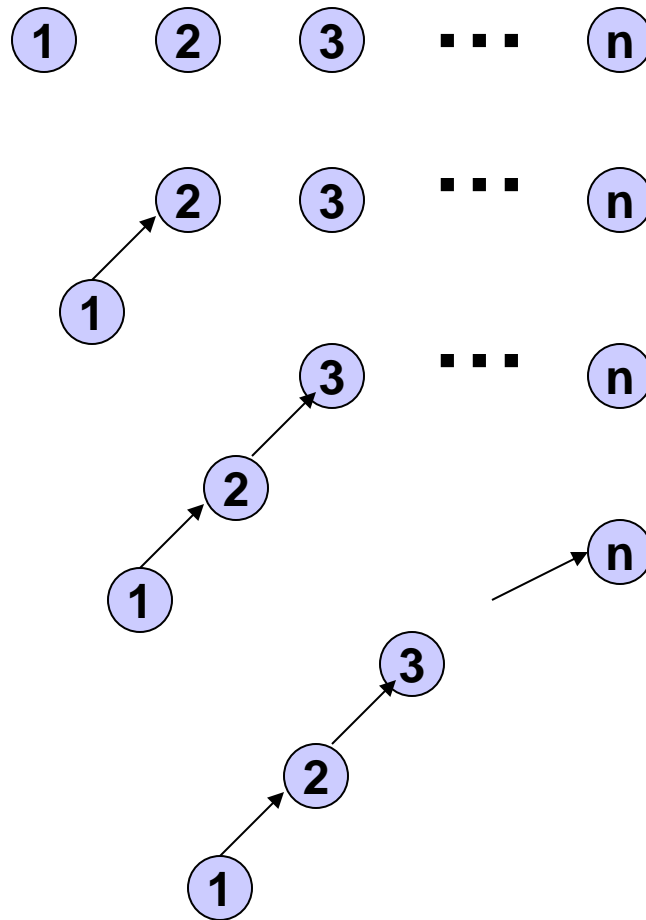
*runtime for Union():*

*runtime for Find():*

*runtime for m Finds and n-1 Unions:*

# A Bad Case

**Union(x,y)** – “point y to x”



**Union(2,1)**

**Union(3,2)**

**:**  
**:**

**Union(n,n-1)**

**Find(1) n steps!!**

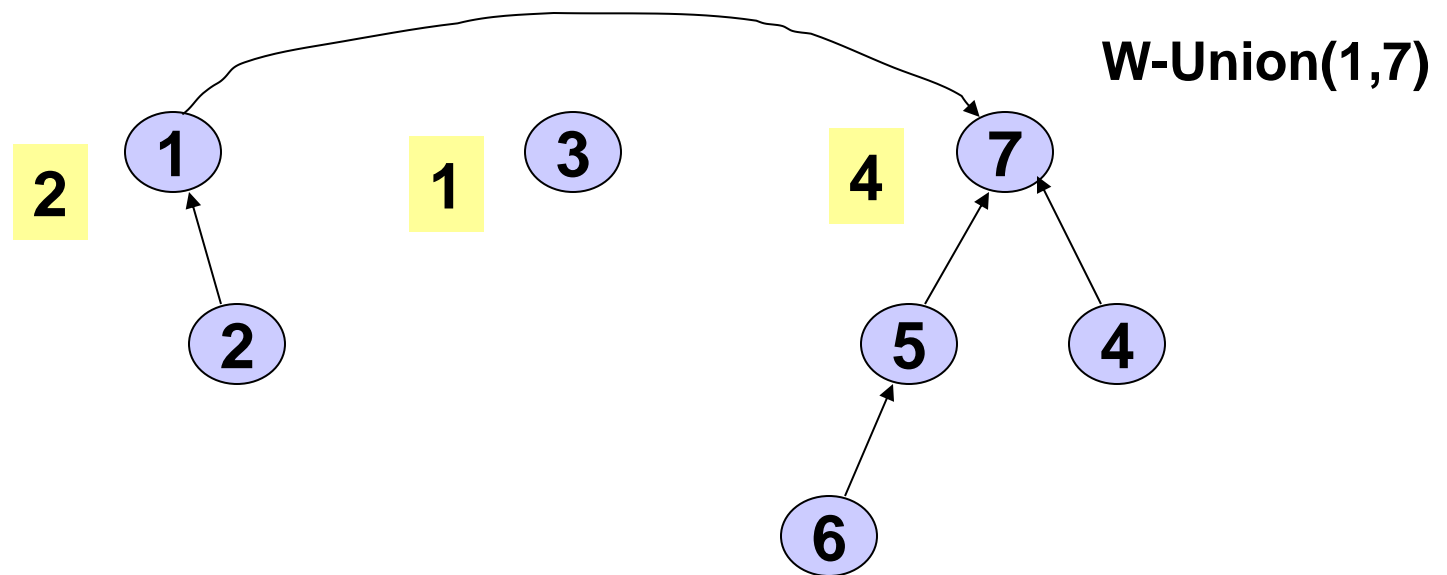
*Now this doesn't look good ☹️*

Can we do better?      Yes!

1. Improve **union** so that **find** only takes  $\Theta(\log n)$ 
  - Union-by-size
  - Reduces complexity to  $\Theta(m \log n + n)$
2. Improve **find** so that it becomes even better!
  - Path compression
  - Reduces complexity to almost  $\Theta(m + n)$

# Weighted Union/Union by Size

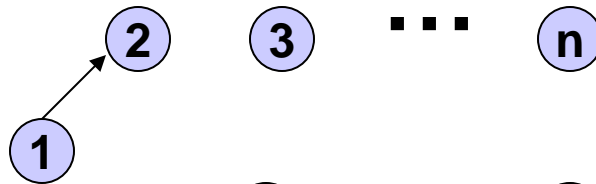
- Weighted Union
  - Always point the *smaller* (total # of nodes) tree to the root of the larger tree



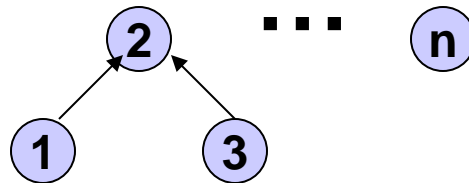
# *Example Again*



**W-Union(2,1)**

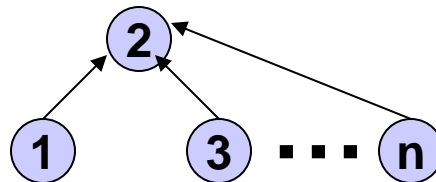


**W-Union(3,2)**



**:**  
**:**

**W-Union(n,2)**



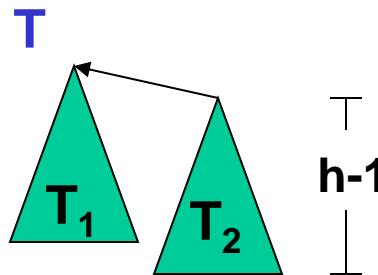
**Find(1) constant time**

# Analysis of Weighted Union

With weighted union an up-tree of height  $h$  has weight *at least*  $2^h$ .

- Proof by induction
  - **Basis**:  $h = 0$ . The up-tree has one node,  $2^0 = 1$
  - **Inductive step**: Assume true for all  $h' < h$ .

Minimum weight  
up-tree of height  $h$   
formed by  
weighted unions



$$W(T_1) \geq W(T_2) \geq 2^{h-1}$$

Weighted union      Induction hypothesis

$$W(T) \geq 2^{h-1} + 2^{h-1} = 2^h$$

## *Analysis of Weighted Union (cont)*

Let  $T$  be an up-tree of weight  $n$  formed by weighted union. Let  $h$  be its height.

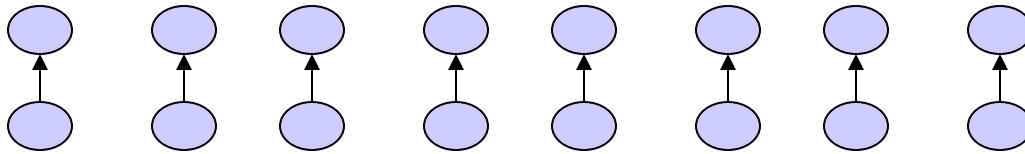
$$n \geq 2^h$$

$$\log_2 n \geq h$$

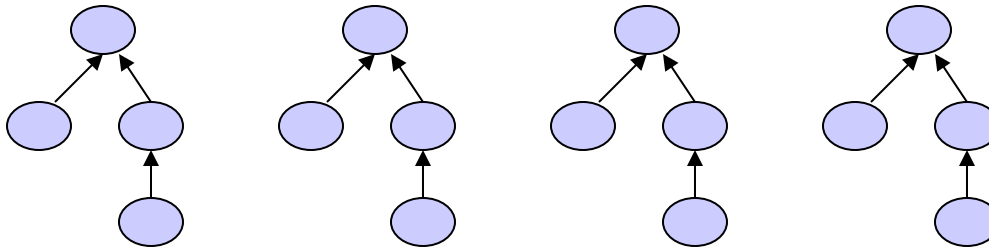
- Find( $x$ ) in tree  $T$  takes  $O(\log n)$  time.
  - Can we do better?

# *Worst Case for Weighted Union*

## **$n/2$ Weighted Unions**



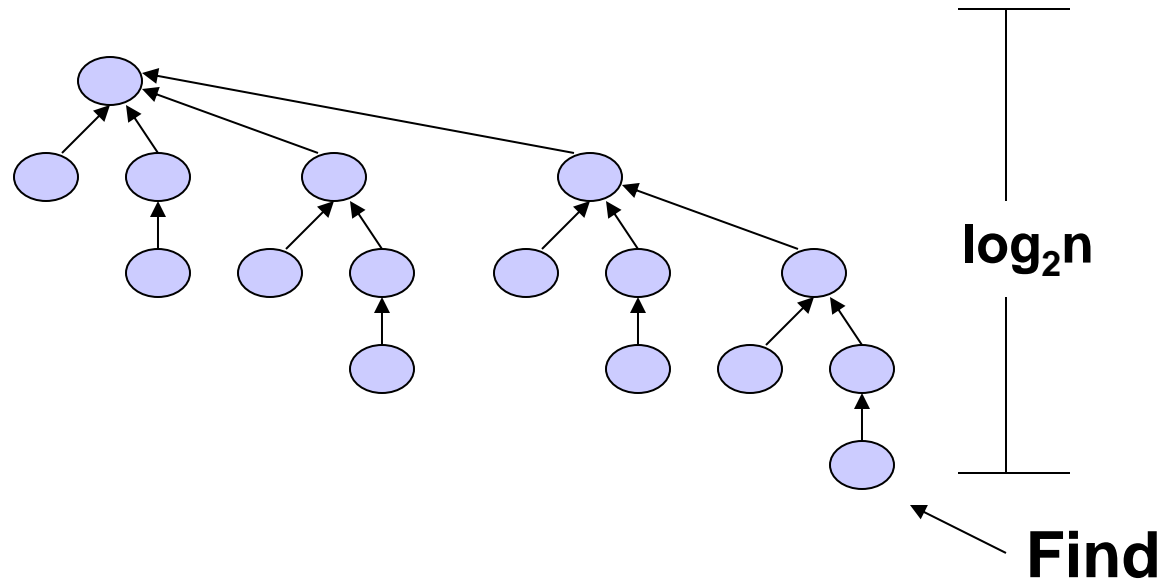
## **$n/4$ Weighted Unions**





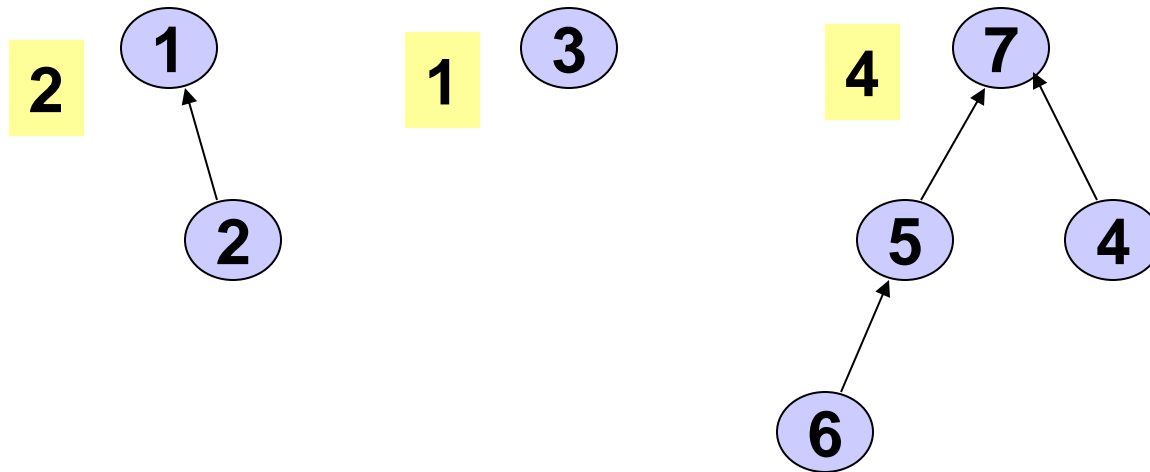
## *Example of Worst Cast (cont')*

**After  $n/2 + n/4 + \dots + 1$  Weighted Unions:**



**If there are  $n = 2^k$  nodes then the longest path from leaf to root has length  $k$ .**

# *Array Implementation*



	1	2	3	4	5	6	7
up	-1	1	-1	7	7	5	-1
weight	2		1				4

# Weighted Union

```
W-Union(i, j : index) {  
    //i and j are roots  
    wi := weight[i];  
    wj := weight[j];  
    if wi < wj then  
        up[i] := j;  
        weight[j] := wi + wj;  
    else  
        up[j] := i;  
        weight[i] := wi + wj;  
}
```

*new runtime for Union():*

*new runtime for Find():*

*runtime for m finds and n-1 unions =*

# *Nifty Storage Trick*

- Use the same array representation as before
- Instead of storing **-1** for the root,  
simply store **-size**

[Read section 8.4]

## *How about Union-by-height?*

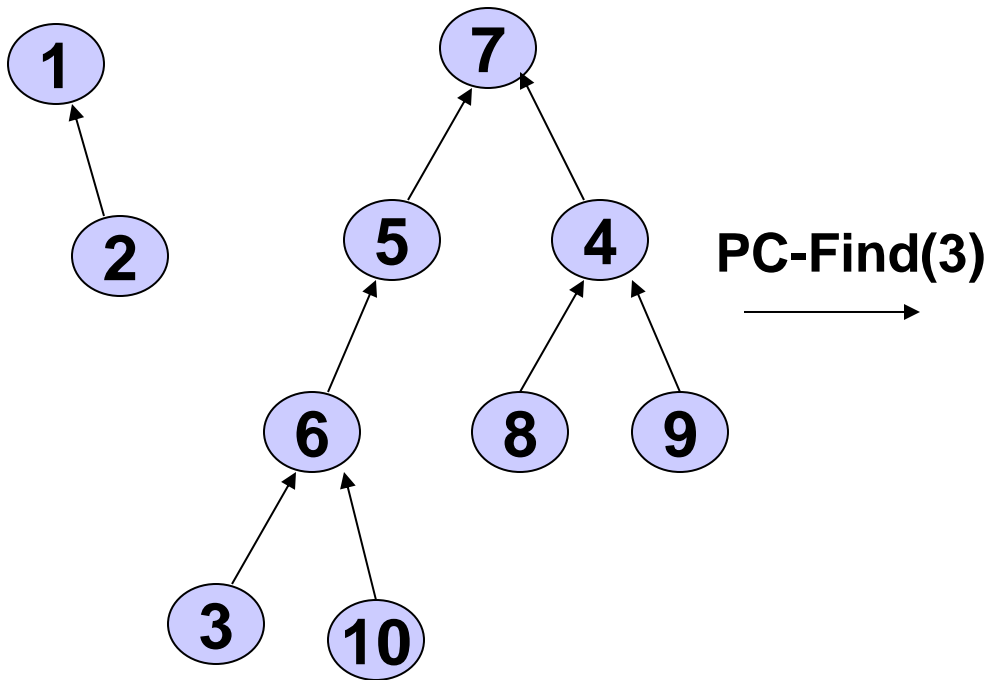
- Can still guarantee  $O(\log n)$  worst case depth

*Left as an exercise!*

- Problem: Union-by-height doesn't combine very well with the new find optimization technique we'll see next

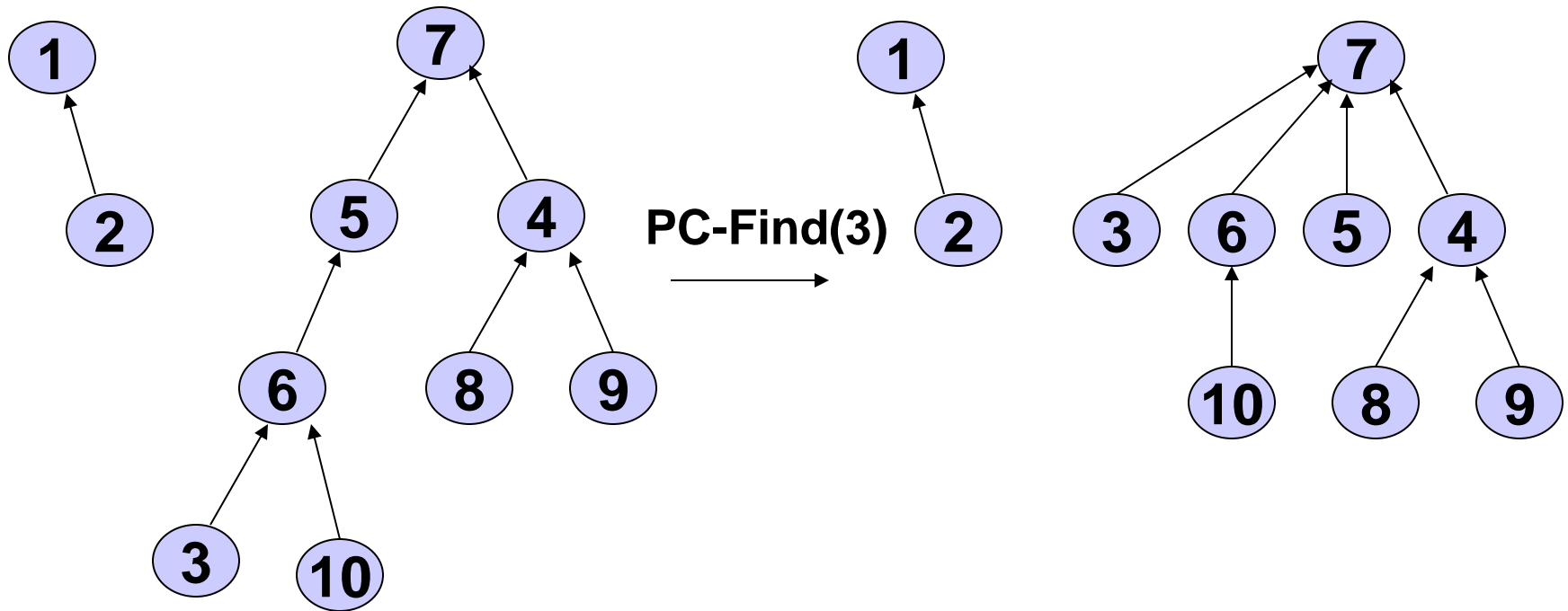
# *Path Compression*

- On a Find operation point all the nodes on the search path directly to the root.



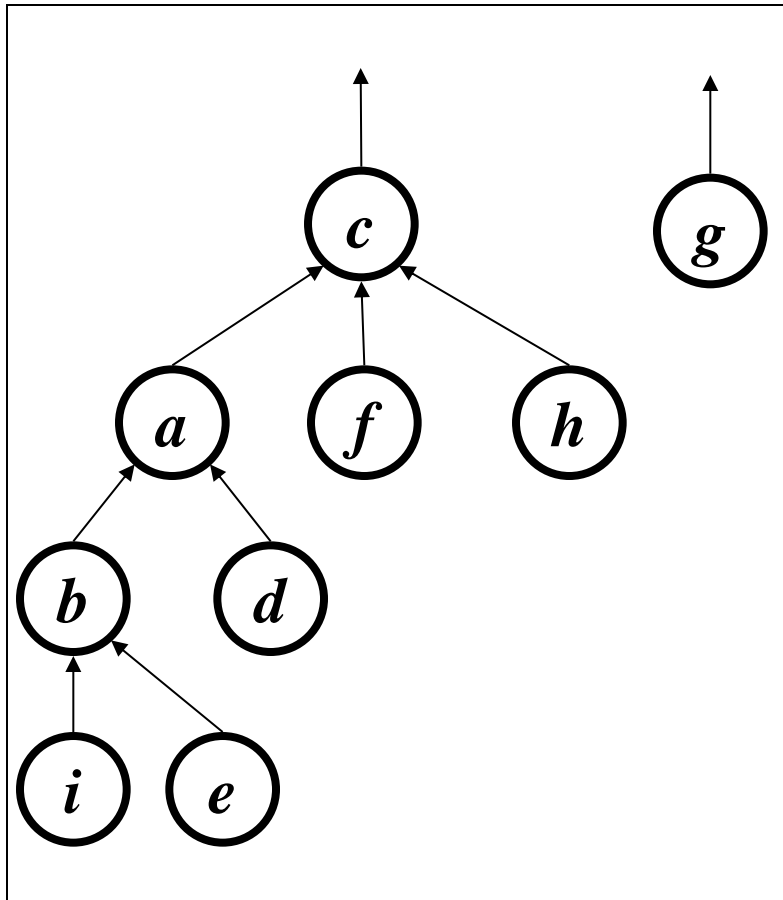
# *Path Compression*

- On a Find operation point all the nodes on the search path directly to the root.



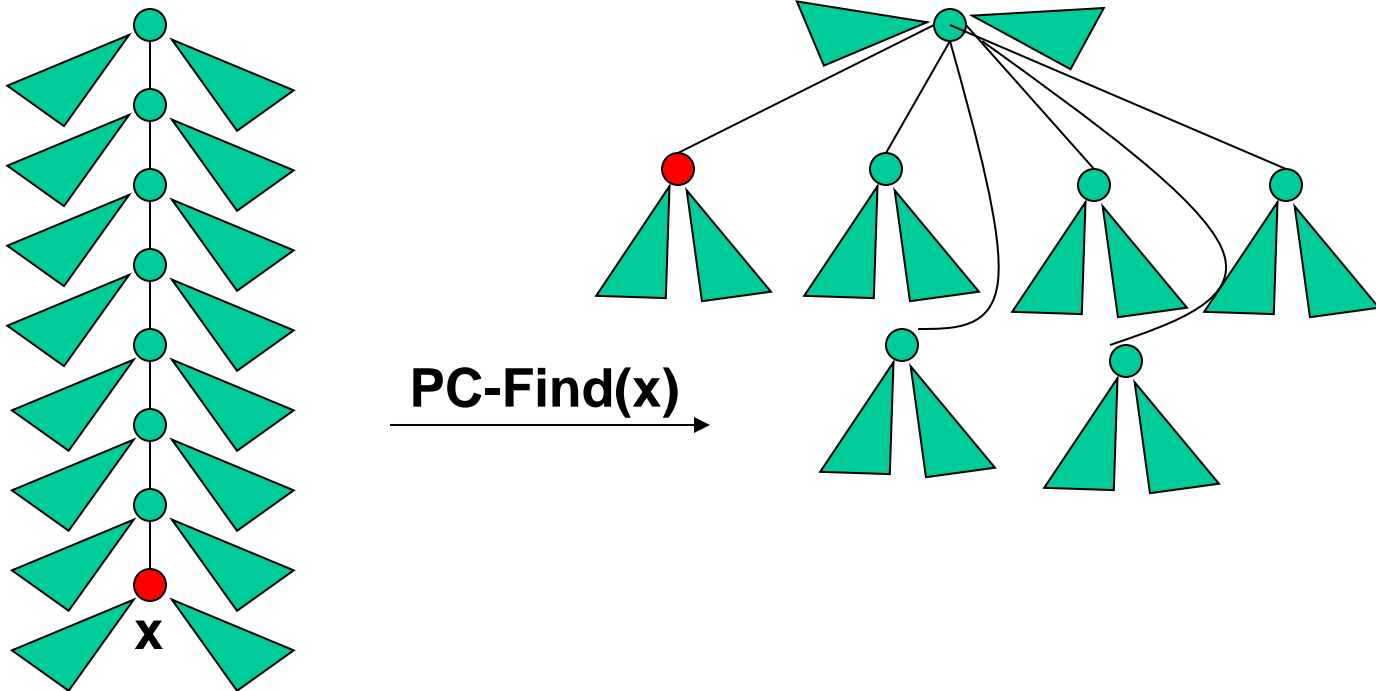
## Student Activity

*Draw the result of Find(e):*





# *Self-Adjustment Works*



# *Path Compression Find*

```
PC-Find(i : index) {  
    r := i;  
    while up[r] ≠ -1 do //find root//  
        r := up[r];  
    if i ≠ r then //compress path//  
        k := up[i];  
        while k ≠ r do  
            up[i] := r;  
            i := k;  
            k := up[k]  
    return(r)  
}
```

# Path Compression: Code

```
int Find(Object x) {  
    // x had better be in  
    // the set!  
    int xID = hTable[x];  
    int i = xID;  
  
    // Get the root for  
    // this set  
    while(up[xID] != -1)  
    {  
        xID = up[xID];  
    }
```

```
        // Change the parent for  
        // all nodes along the path  
        while(up[i] != -1) {  
            temp = up[i];  
            up[i] = xID;  
            i = temp;  
        }  
    return xID;  
}
```

*(New?) runtime for Find:*

## *Interlude: A Really Slow Function*

**Ackermann's function** is a really big function  $A(x, y)$  with inverse  $\alpha(x, y)$  which is really small

How fast does  $\alpha(x, y)$  grow?

$\alpha(x, y) = 4$  for  $x$  **far** larger than the number of atoms in the universe ( $2^{300}$ )

$\alpha$  shows up in:

- Computation Geometry (surface complexity)
- Combinatorics of sequences

# *A More Comprehensible Slow Function*

**$\log^* x$  = number of times you need to compute  
log to bring value down to at most 1**

E.g.  $\log^* 2 = 1$

$$\log^* 4 = \log^* 2^2 = 2$$

$$\log^* 16 = \log^* 2^{2^2} = 3 \quad (\log \log \log 16 = 1)$$

$$\log^* 65536 = \log^* 2^{2^{2^2}} = 4 \quad (\log \log \log \log 65536 =$$

1)

$$\log^* 2^{65536} = \dots\dots\dots = 5$$

Take this:  $\alpha(m,n)$  grows even slower than  $\log^* n$  !!

# *Complex Complexity of Union-by-Size + Path Compression*

Tarjan proved that, with these optimizations,  $p$  union and find operations on a set of  $n$  elements have worst case complexity of  $O(p \cdot \alpha(p, n))$

For *all practical purposes* this is amortized constant time:

$O(p \cdot 4)$  for  $p$  operations!

- Very complex analysis

## *Disjoint Union / Find with Weighted Union and PC*

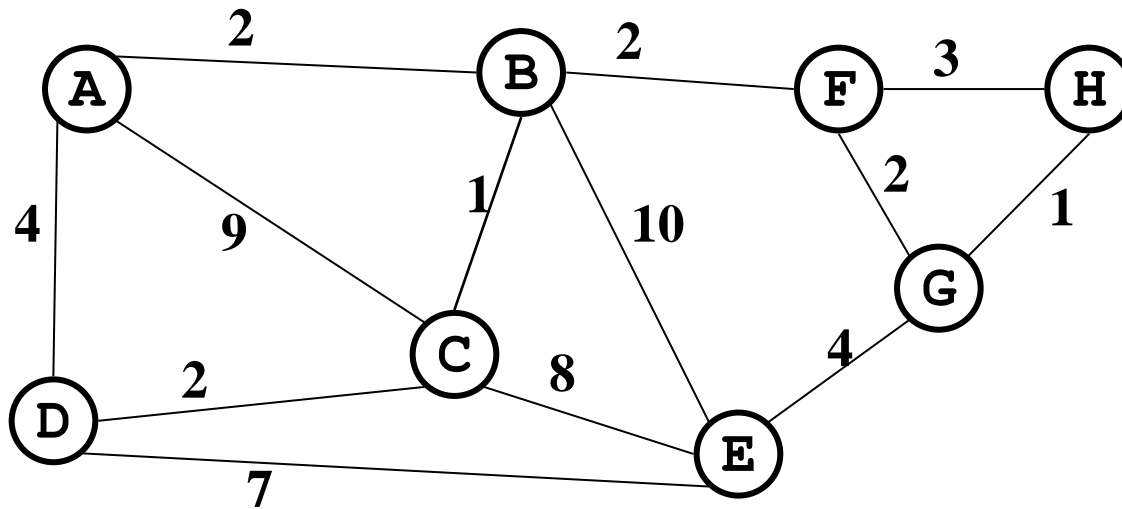
- Worst case time complexity for a W-Union is  $O(1)$  and for a PC-Find is  $O(\log n)$ .
- Time complexity for  $m \geq n$  operations on  $n$  elements is  $O(m \log^* n)$  where  $\log^* n$  is a very slow growing function.
  - $\log^* n < 7$  for all reasonable  $n$ . Essentially constant time per operation!
- Using “ranked union” gives an even better bound theoretically.

# *Amortized Complexity*

- For disjoint union / find with weighted union and path compression.
  - average time per operation is essentially a constant.
  - worst case time for a PC-Find is  $O(\log n)$ .
- An individual operation can be costly, but over time the average cost per operation is not.



*Find MST using Kruskal's*

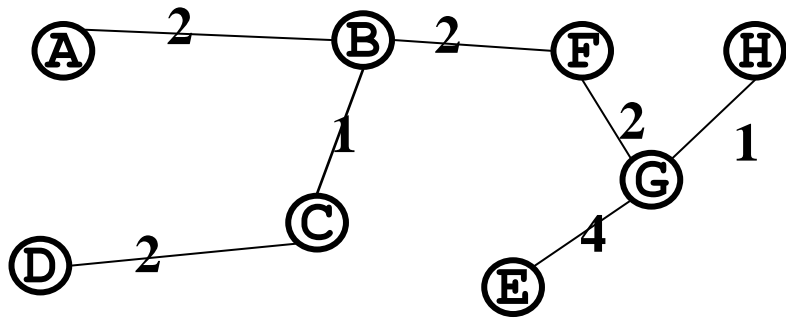


**Total Cost:**

- **Now find the MST using Prim's method.**
- **Under what conditions will these methods give the same result?**

*Draw the UpTree*

<b>Nodes</b>	<b>A</b>	<b>B</b>	<b>C</b>	<b>D</b>	<b>E</b>	<b>F</b>	<b>G</b>	<b>H</b>
Parent								
Size								



*Draw the UpTree*

Nodes	A	B	C	D	E	F	G	H
Parent								
Size								