



# CSE373: Data Structures & Algorithms

## Lecture 11: Implementing Union-Find

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Winter 2015

# *Announcements*

- Homework 3 due in Friday
- Help sessions: Do we need them?

# *The plan*

Last lecture:

- Disjoint sets
- The union-find ADT for disjoint sets

Today's lecture:

- Basic implementation of the union-find ADT with “up trees”
- Optimizations that make the implementation much faster

# Union-Find ADT

- Given an unchanging set  $S$ , **create** an initial partition of a set
  - Typically each item in its own subset:  $\{a\}$ ,  $\{b\}$ ,  $\{c\}$ , ...
  - Give each subset a “name” by choosing a *representative element*
- Operation **find** takes an element of  $S$  and returns the representative element of the subset it is in
- Operation **union** takes two subsets and (permanently) makes one larger subset
  - A different partition with one fewer set
  - Affects result of subsequent **find** operations
  - Choice of representative element up to implementation

## Implementation – our goal

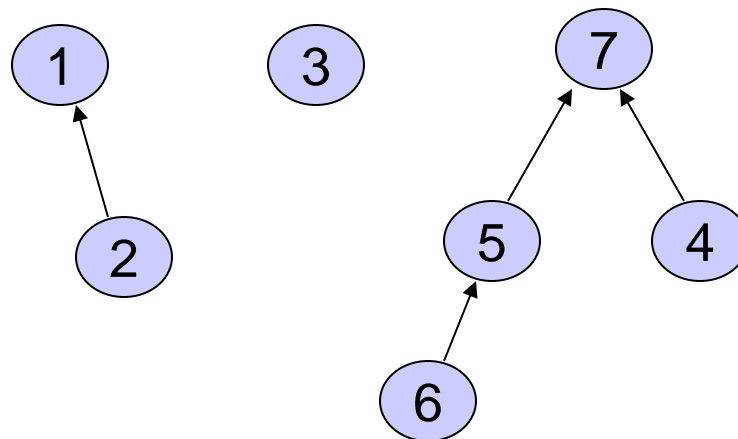
- Start with an initial partition of  $n$  subsets
  - Often 1-element sets, e.g.,  $\{1\}$ ,  $\{2\}$ ,  $\{3\}$ , ...,  $\{n\}$
- May have *any number of* **find** operations
- May have *up to  $n-1$*  **union** operations in any order
  - After  $n-1$  **union** operations, every **find** returns same *single* set

# Up-tree data structure

- Tree with:
  - No limit on branching factor
  - References from **children** to **parent**
- Start with *forest* of 1-node trees



- Possible forest after several unions:
  - Will use roots for set names

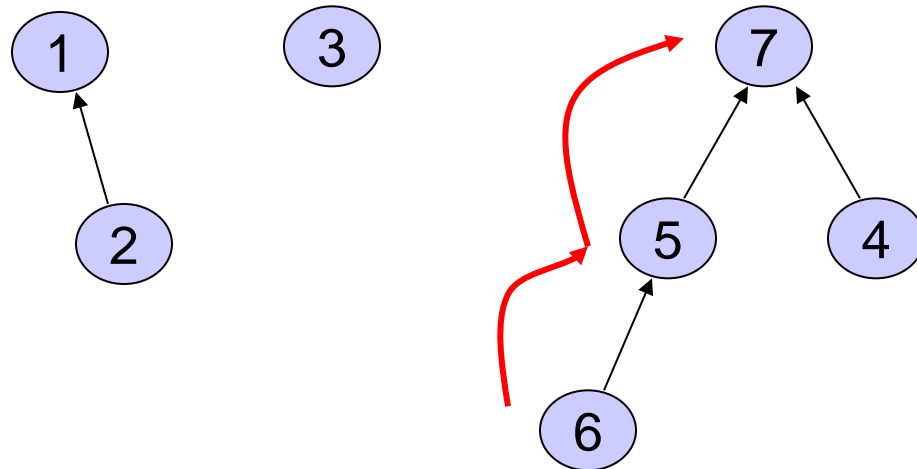


# Find

**find(x):**

- Assume we have  $O(1)$  access to each node
  - Will use an array where index  $i$  holds node  $i$
- Start at  $x$  and follow parent pointers to root
- Return the root

**find(6) = 7**

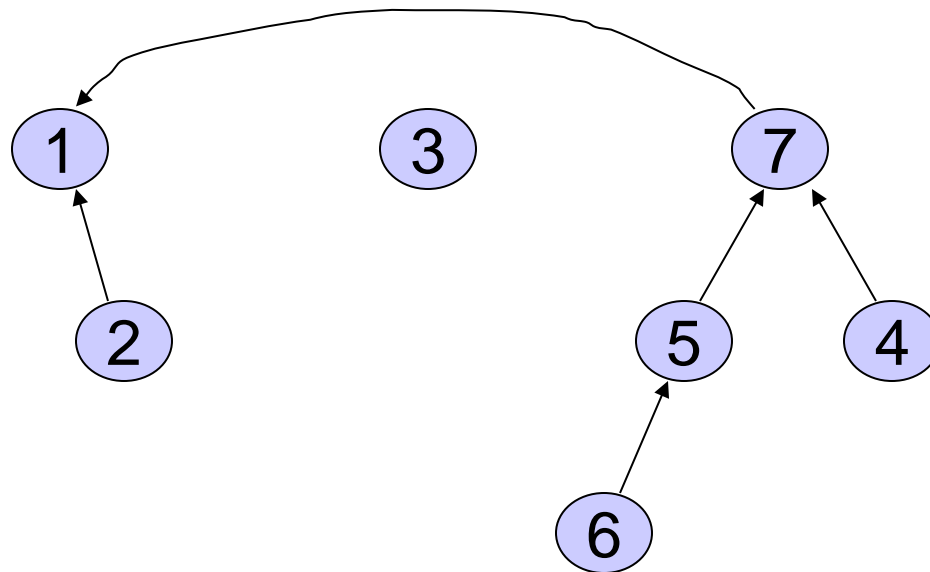


# Union

`union(x, y):`

- Assume **x** and **y** are roots
  - Else **find** the roots of their trees
- Assume distinct trees (else do nothing)
- Change root of one to have parent be the root of the other
  - Notice no limit on branching factor

`union(1,7)`





# Simple implementation

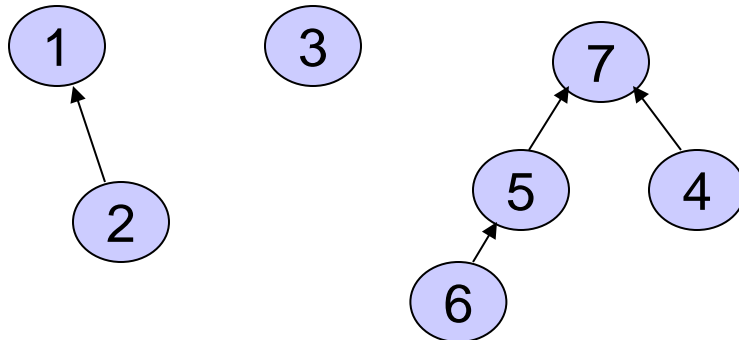
- If set elements are contiguous numbers (e.g.,  $1, 2, \dots, n$ ), use an array of length  $n$  called `up`
  - Starting at index 1 on slides
  - Put in array index of parent, with 0 (or -1, etc.) for a root

• Example:



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

• Example:



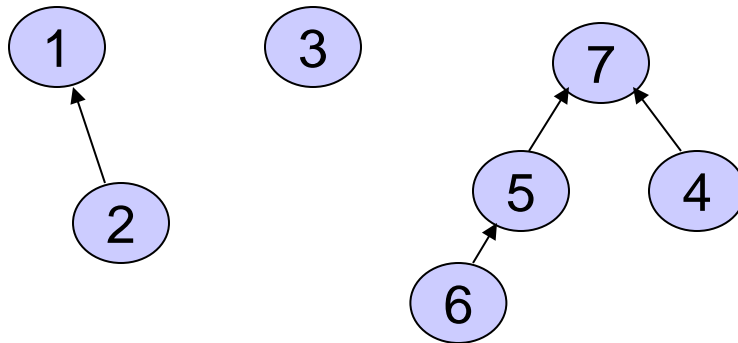
	1	2	3	4	5	6	7
up	0	1	0	7	7	5	0

- If set elements are not contiguous numbers, could have a separate dictionary to map elements (keys) to numbers (values)

# Implement operations

```
// assumes x in range 1,n
int find(int x) {
    while(up[x] != 0) {
        x = up[x];
    }
    return x;
}
```

```
// assumes x,y are roots
void union(int x, int y){
    up[y] = x;
}
```



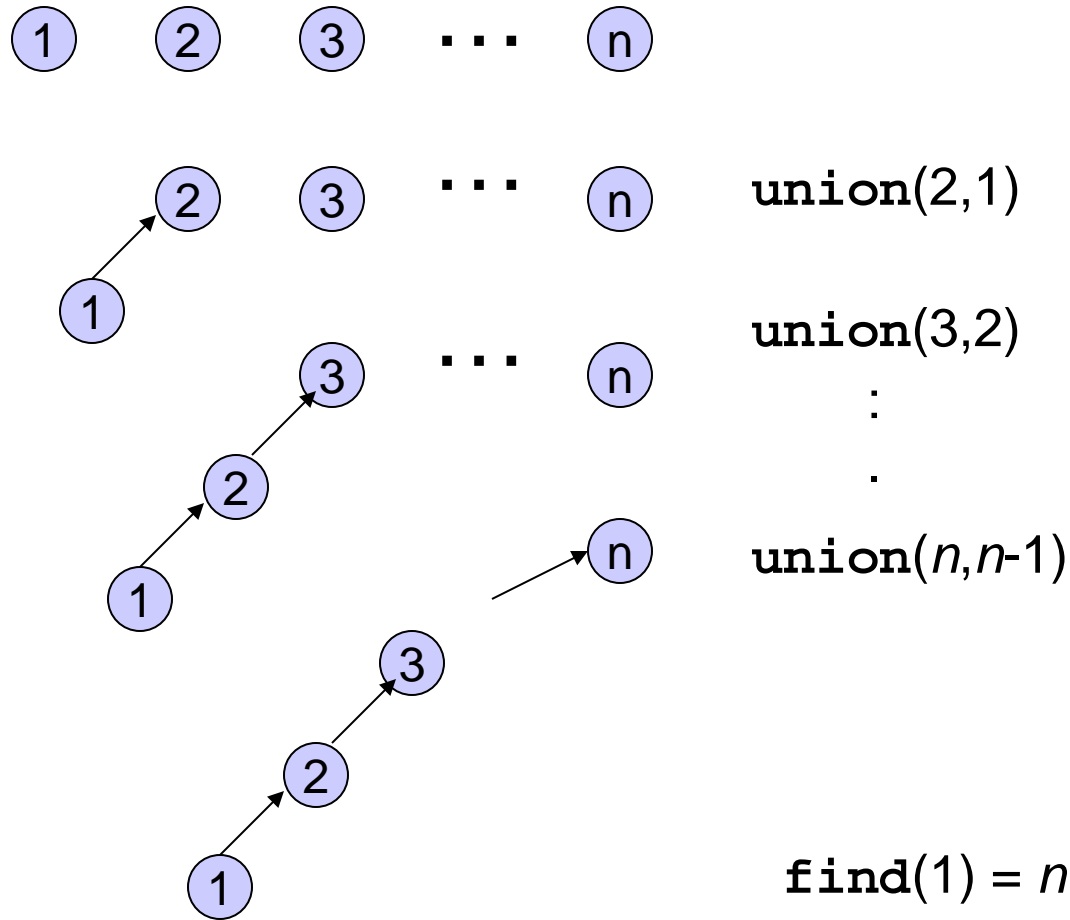
	1	2	3	4	5	6	7
up	0	1	0	7	7	5	0

- Worst-case run-time for `union`?  $O(1)$
- Worst-case run-time for `find`?  $O(n)$
- Worst-case run-time for  $m$  `finds` and  $n-1$  `unions`?  $O(m*n)$

# *Two key optimizations*

1. Improve **union** so it stays  $O(1)$  but makes **find**  $O(\log n)$ 
  - So  $m$  **finds** and  $n-1$  **unions** is  $O(m \log n + n)$
  - *Union-by-size: connect smaller tree to larger tree*
2. Improve **find** so it becomes even faster
  - Make  $m$  **finds** and  $n-1$  **unions** **almost**  $O(m + n)$
  - *Path-compression: connect directly to root during finds*

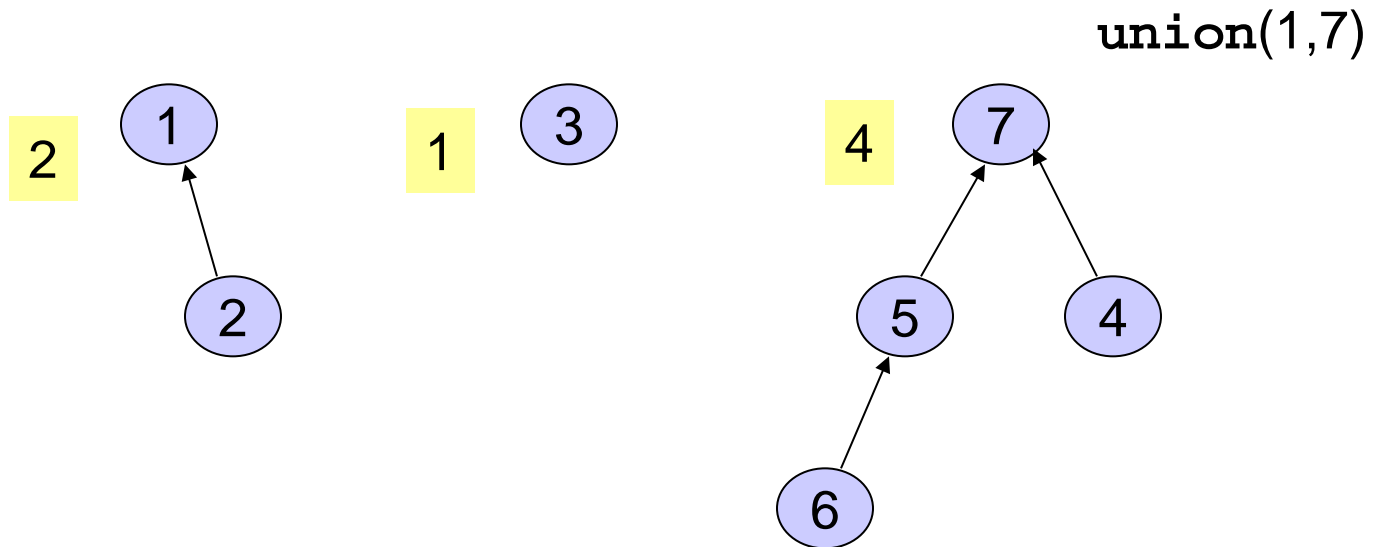
# The bad case to avoid



# Union-by-size

Union-by-size:

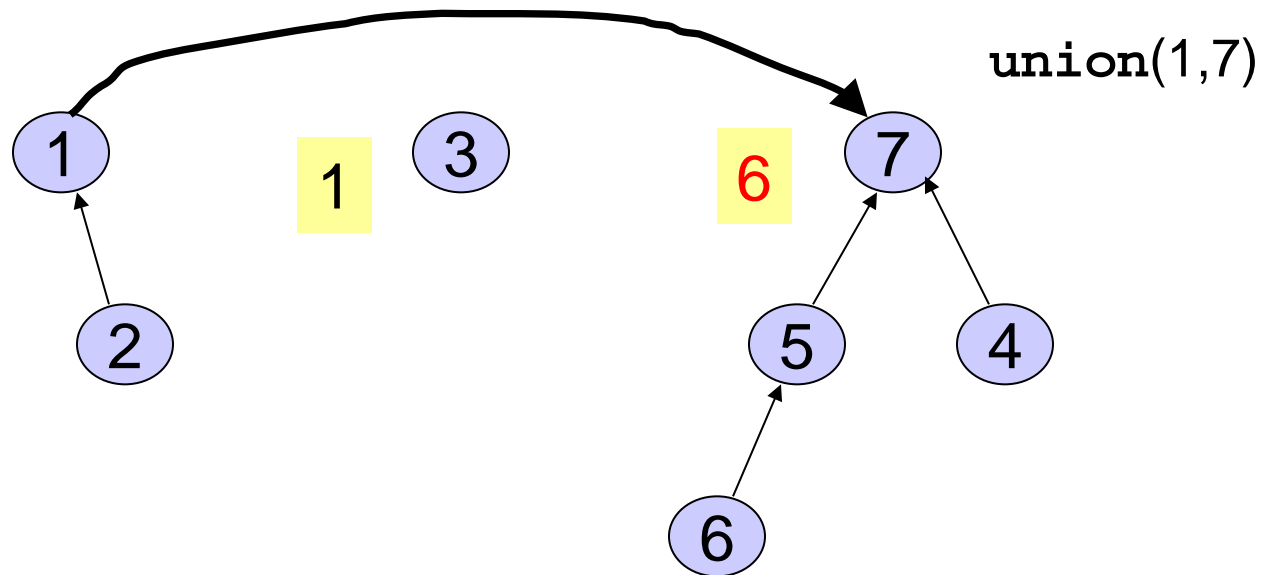
- Always point the *smaller* (total # of nodes) tree to the root of the larger tree



# Union-by-size

Union-by-size:

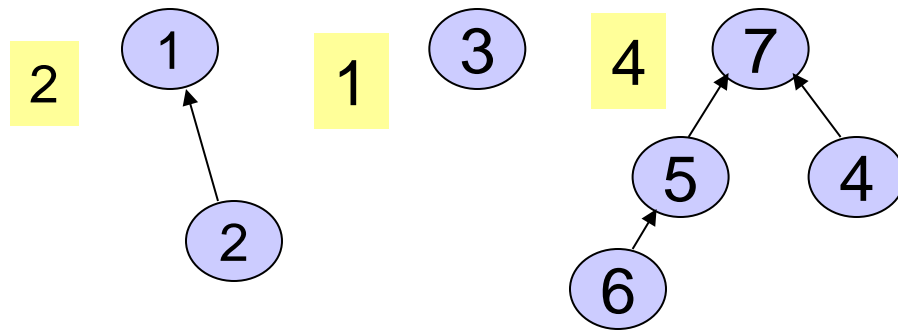
- Always point the *smaller* (total # of nodes) tree to the root of the larger tree



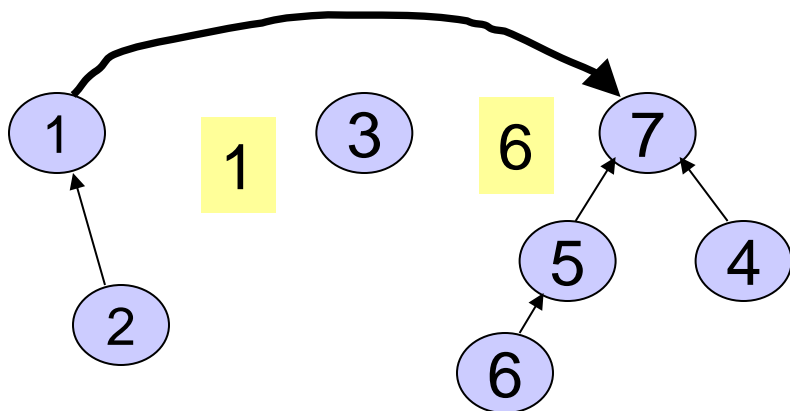
# Array implementation

Keep the size (number of nodes in a second array)

- Or have one array of objects with two fields



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

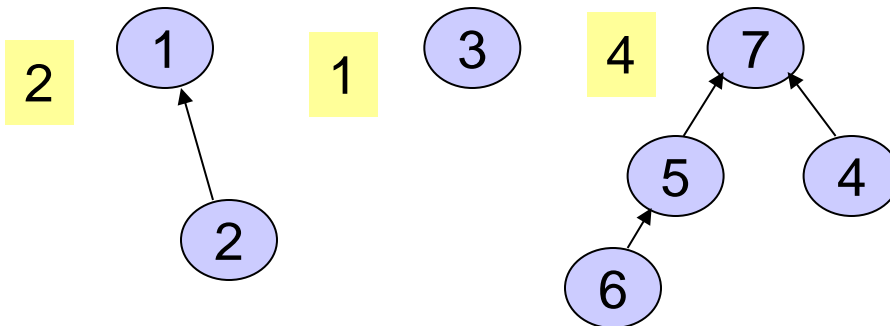


	1	2	3	4	5	6	7
up	7	1	0	7	7	5	0
weight			1				6

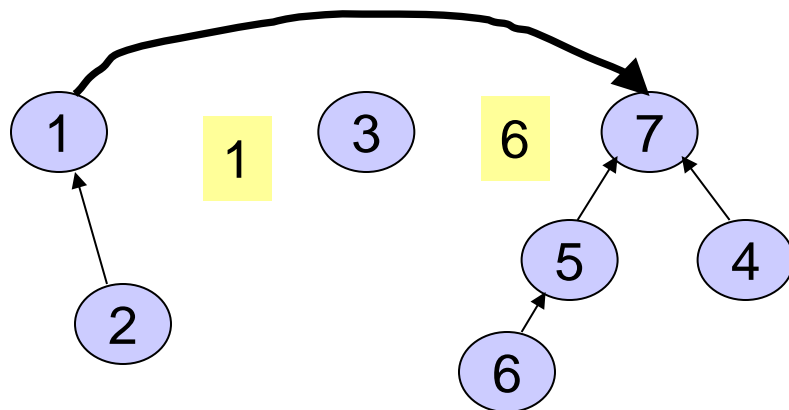
# Nifty trick

Actually we do not need a second array...

- Instead of storing 0 for a root, store negation of size
- So up value  $< 0$  means a root



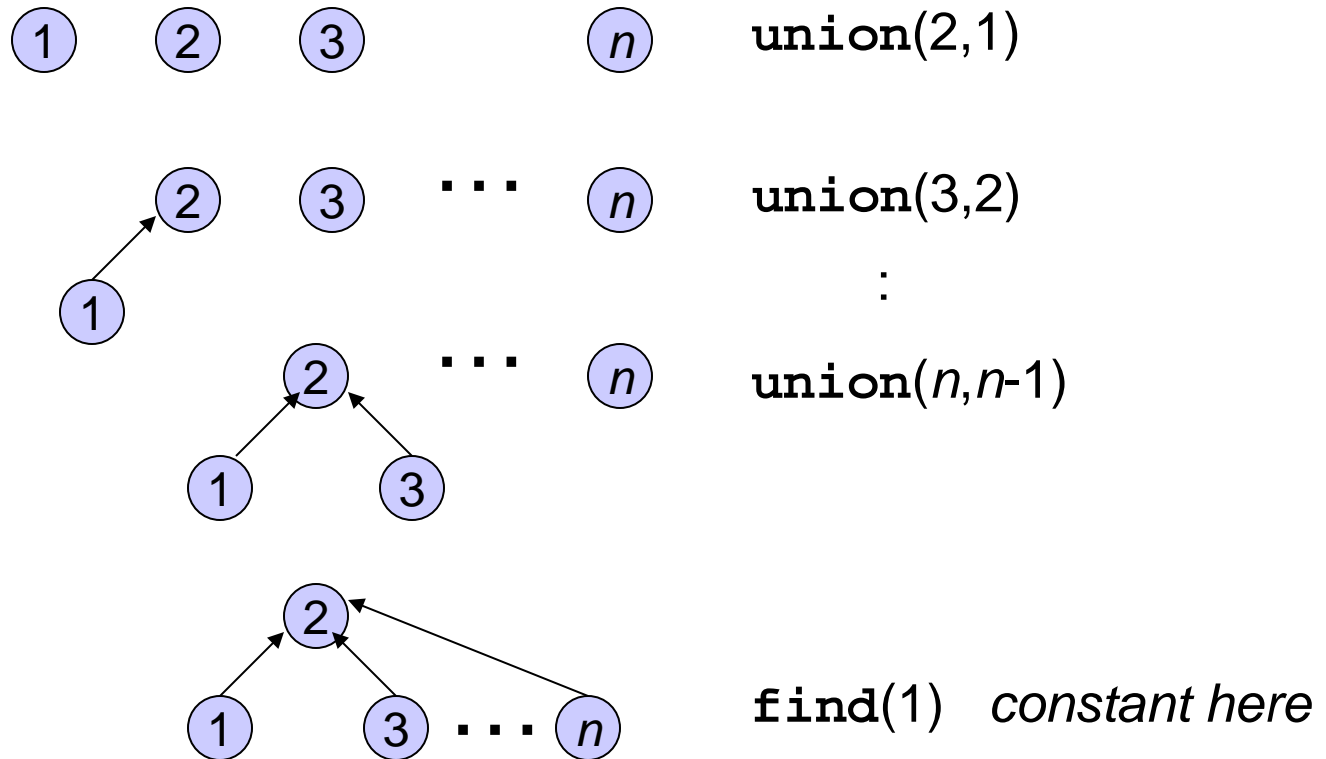
	1	2	3	4	5	6	7
up	-2	1	-1	7	7	5	-4



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



# The Bad case? Now a Great case...



# General analysis

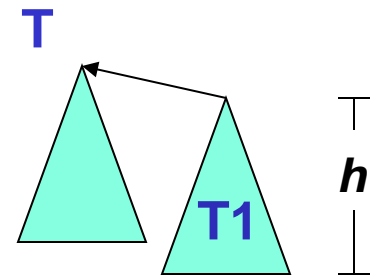
- Showing one worst-case example is now good is *not* a proof that the worst-case has improved
- So let's prove:
  - **union** is still  $O(1)$  – this is “obvious”
  - **find** is now  $O(\log n)$
- Claim: If we use union-by-size, an up-tree of height  $h$  has at least  $2^h$  nodes
  - Proof by induction on  $h$ ...

# Exponential number of nodes

$P(h)$  = With union-by-size, up-tree of height  $h$  has at least  $2^h$  nodes

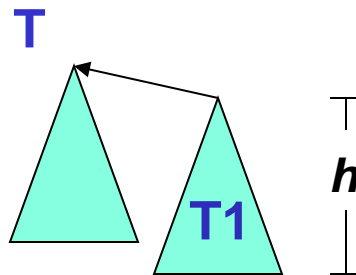
Proof by induction on  $h$ ...

- Base case:  $h = 0$ : The up-tree has 1 node and  $2^0 = 1$
- Inductive case: Assume  $P(h)$  and show  $P(h+1)$ 
  - A height  $h+1$  tree  $T$  has at least one height  $h$  child  $T1$
  - $T1$  has at least  $2^h$  nodes by induction
  - And  $T$  has *at least* as many nodes not in  $T1$  than in  $T1$ 
    - Else union-by-size would have had  $T$  point to  $T1$ , not  $T1$  point to  $T$  (!!)
  - So total number of nodes is *at least*  $2^h + 2^h = 2^{h+1}$



# The key idea

Intuition behind the proof: No one child can have more than half the nodes

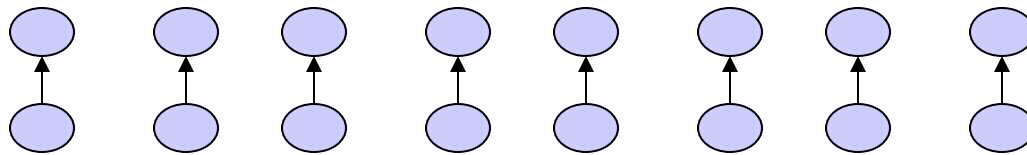


So, as usual, if number of nodes is exponential in height, then height is logarithmic in number of nodes

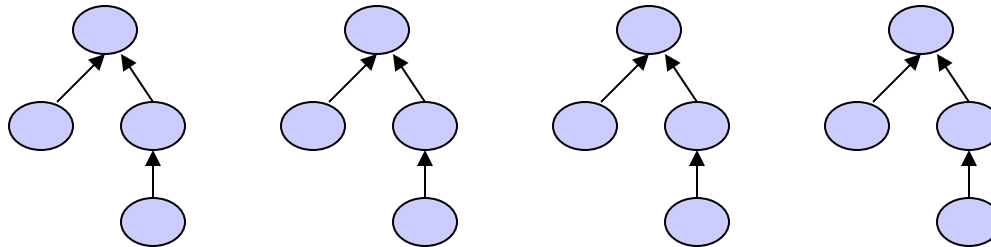
So **find** is  $O(\log n)$

# *The new worst case*

n/2 Unions-by-size

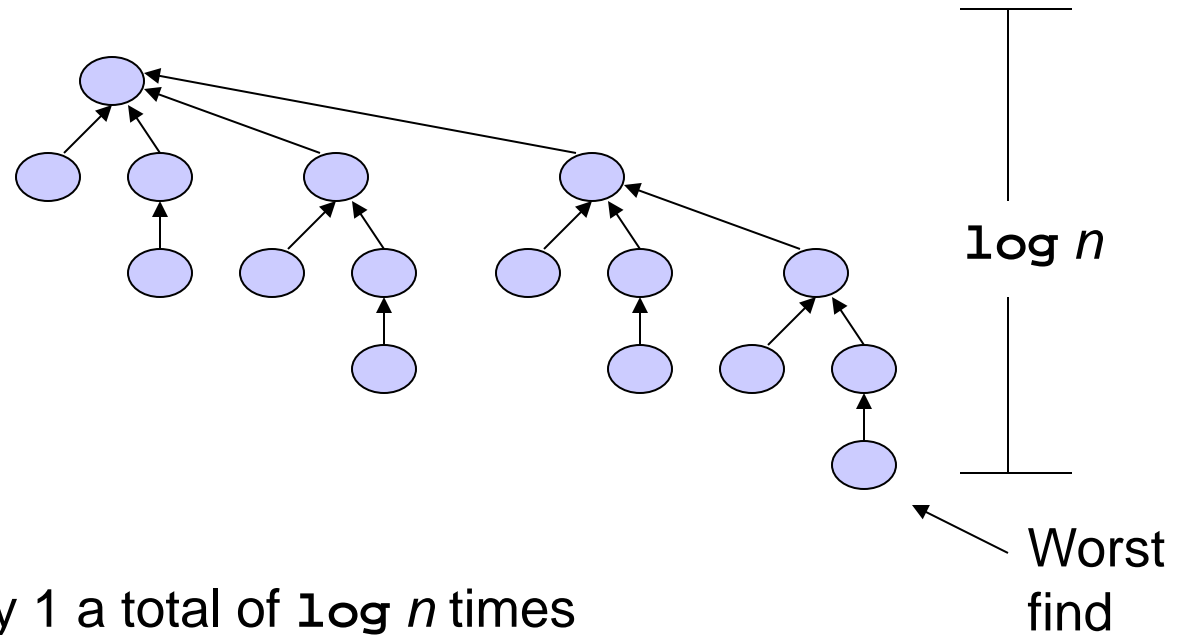


n/4 Unions-by-size



# The new worst case (continued)

After  $n/2 + n/4 + \dots + 1$  Unions-by-size:



## *What about union-by-height*

We could store the height of each root rather than size

- Still guarantees logarithmic worst-case find
  - Proof left as an exercise if interested
- But **does not work well with our next optimization**
  - Maintaining height becomes inefficient, but maintaining size still easy

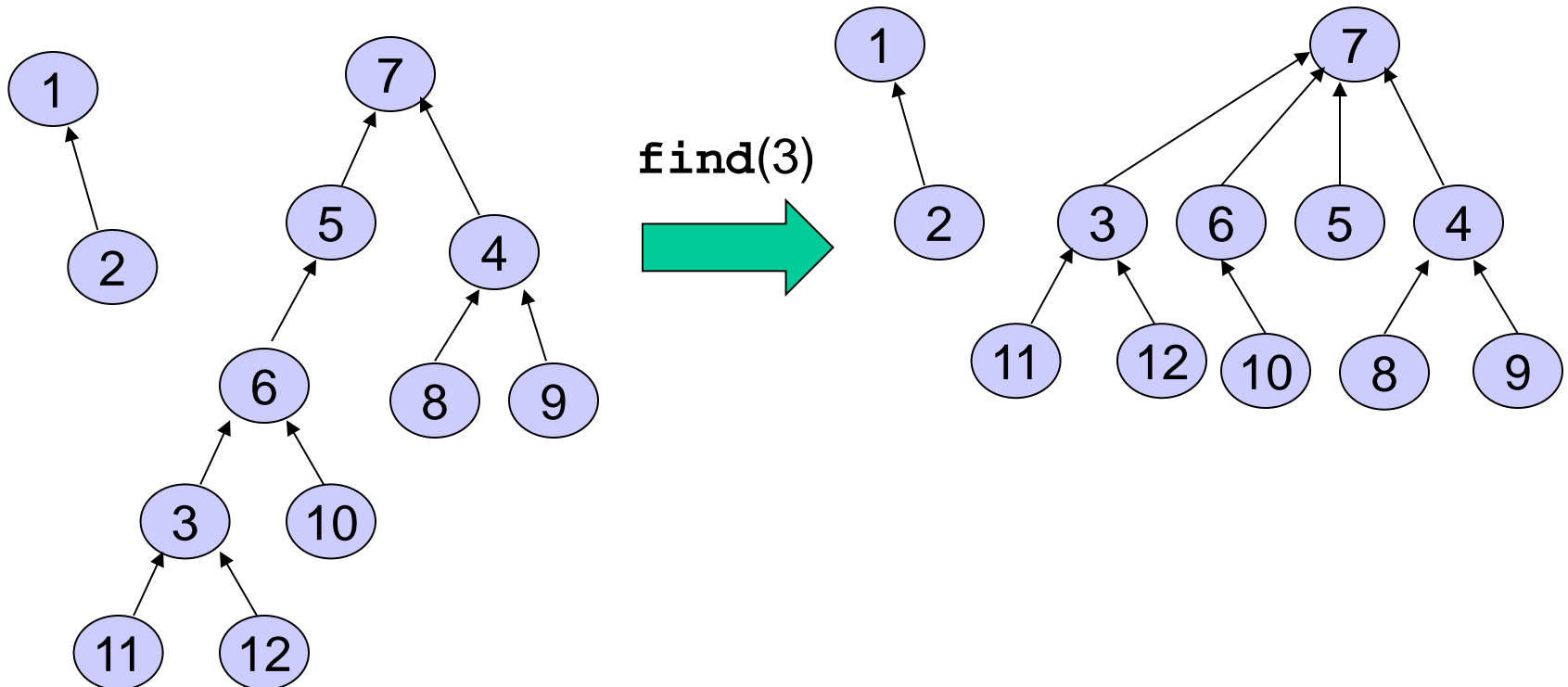
# *Two key optimizations*

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  - So  $m$  **finds** and  $n-1$  **unions** is  $O(m \log n + n)$
  - *Union-by-size*: connect smaller tree to larger tree
2. Improve **find** so it becomes even faster
  - Make  $m$  **finds** and  $n-1$  **unions** **almost**  $O(m + n)$
  - *Path-compression*: connect directly to root during finds



# Path compression

- Simple idea: As part of a `find`, change each encountered node's parent to point directly to root
  - Faster future `finds` for everything on the path (and their descendants)



# Pseudocode

```
// performs path compression
int find(i) {
    // find root
    int r = i
    while(up[r] > 0)
        r = up[r]

    // compress path
    if i==r
        return r;
    int old_parent = up[i]
    while(old_parent != r) {
        up[i] = r
        i = old_parent;
        old_parent = up[i]
    }
    return r;
}
```

## Example

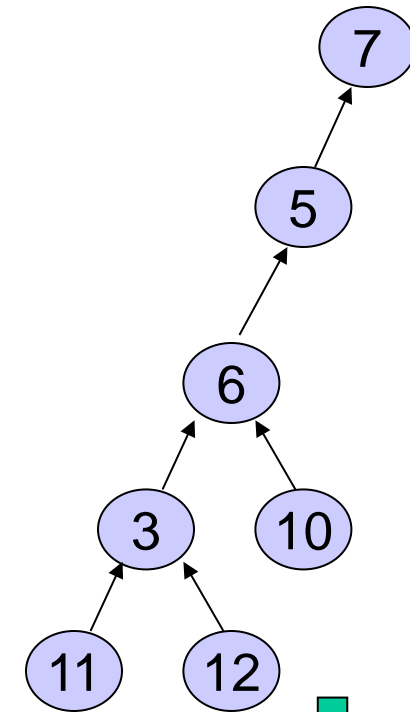
$i=3$

$r=3$

$r=6$

$r=5$

$r=7$



$old\_parent=6$

↓ find(3)

$up[3]=7$

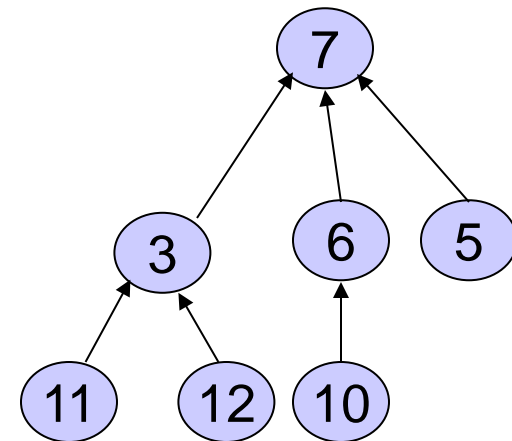
$i=6$

$old\_parent=5$

$up[6]=7$

$i=5$

$old\_parent=7$



# So, how fast is it?

A single worst-case **find** could be  $O(\log n)$

- But only if we did a lot of worst-case unions beforehand
- And path compression will make future finds faster

Turns out the **amortized worst-case bound** is much better than  $O(\log n)$

- total for  $m$  **finds** and  $n-1$  **unions** is *almost*  $O(m+n)$

- We won't *prove* it – **see text if curious**

