



CSE373: Data Structures & Algorithms

Lecture 11: Implementing Union-Find

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Announcements

- Homework 3 due in ONE week...Wednesday April 29th!
- TA sessions
- Catie will be back on Monday.

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

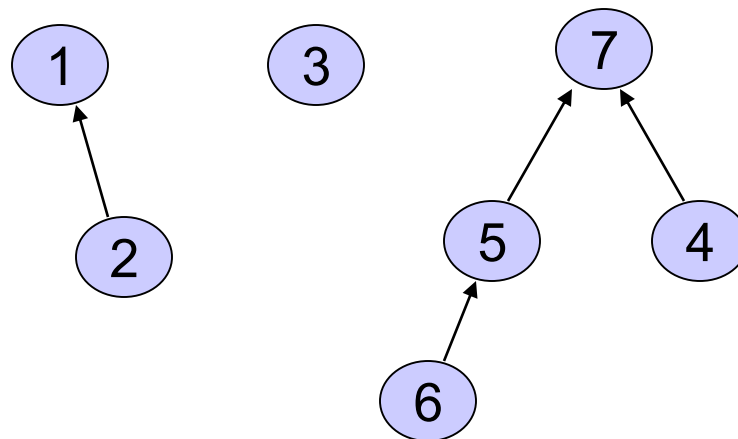
- Start with an initial partition of n subsets
 - Often 1-element sets, e.g., $\{1\}$, $\{2\}$, $\{3\}$, ..., $\{n\}$
- May have m **find** operations
- May have up to $n-1$ **union** operations in any order
 - After $n-1$ **union** operations, every **find** returns same 1 set

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

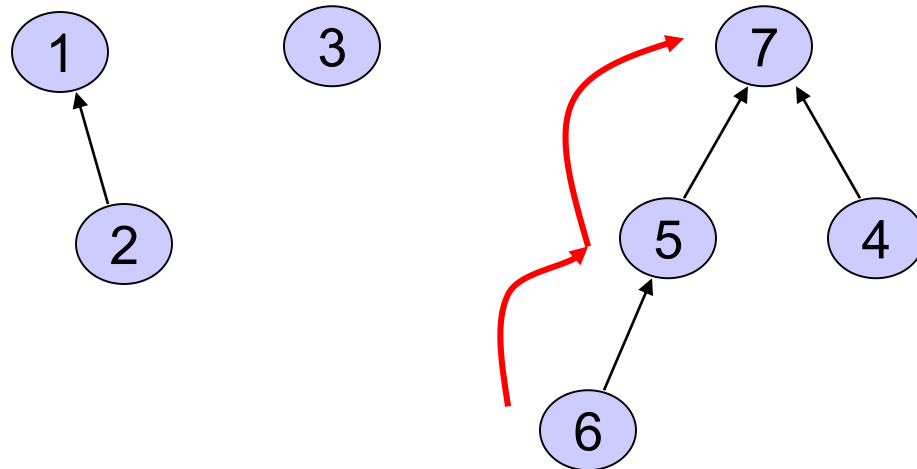


Find

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

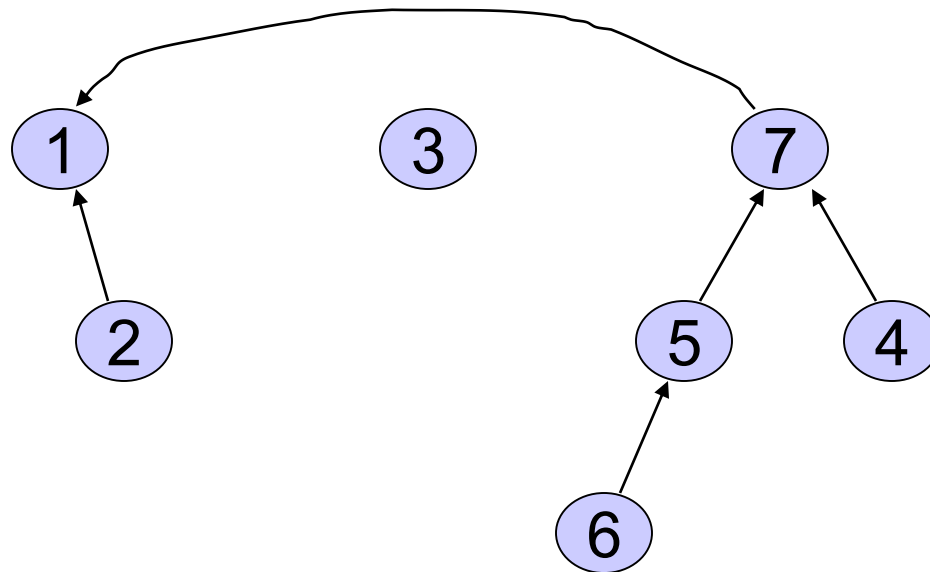


Union

`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

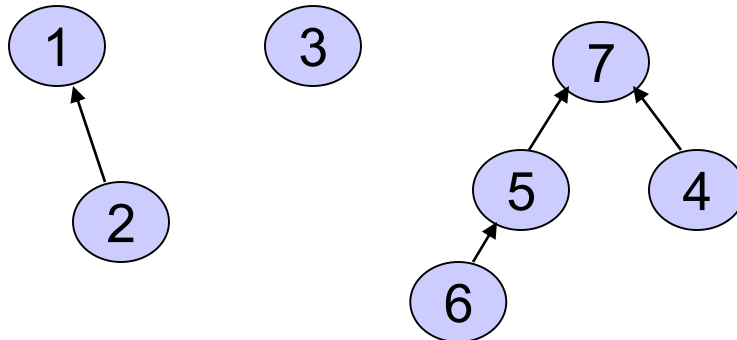
- If set elements are contiguous numbers (e.g., $1, 2, \dots, n$), use an array of length n called **up**
 - Starting at index 1 on slides
 - Put in array index of parent, with 0 (or -1, etc.) for a root

• Example:



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

• Example:



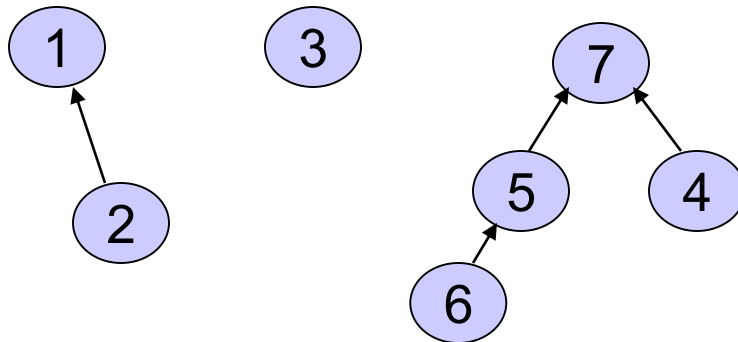
	1	2	3	4	5	6	7
up	0	1	0	7	7	5	0

- 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;
}
```



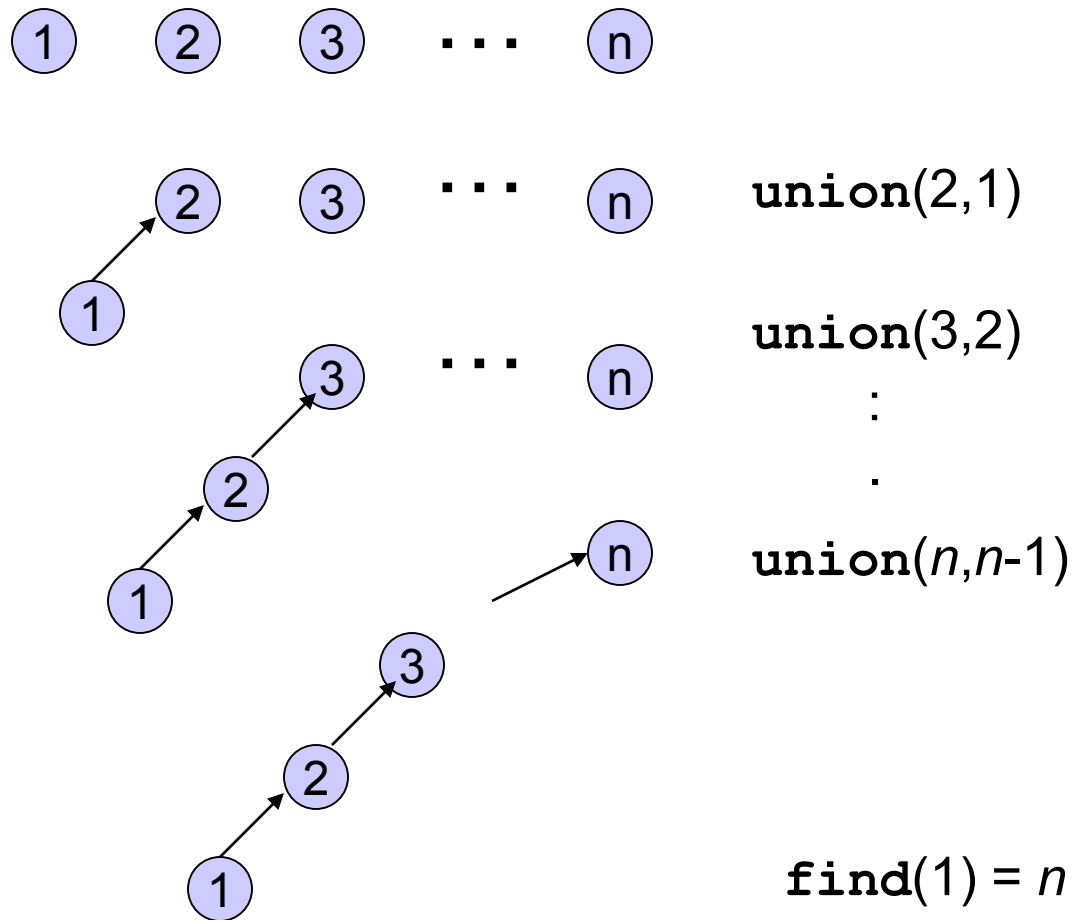
	1	2	3	4	5	6	7
up	0	1	0	7	7	5	0

- 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

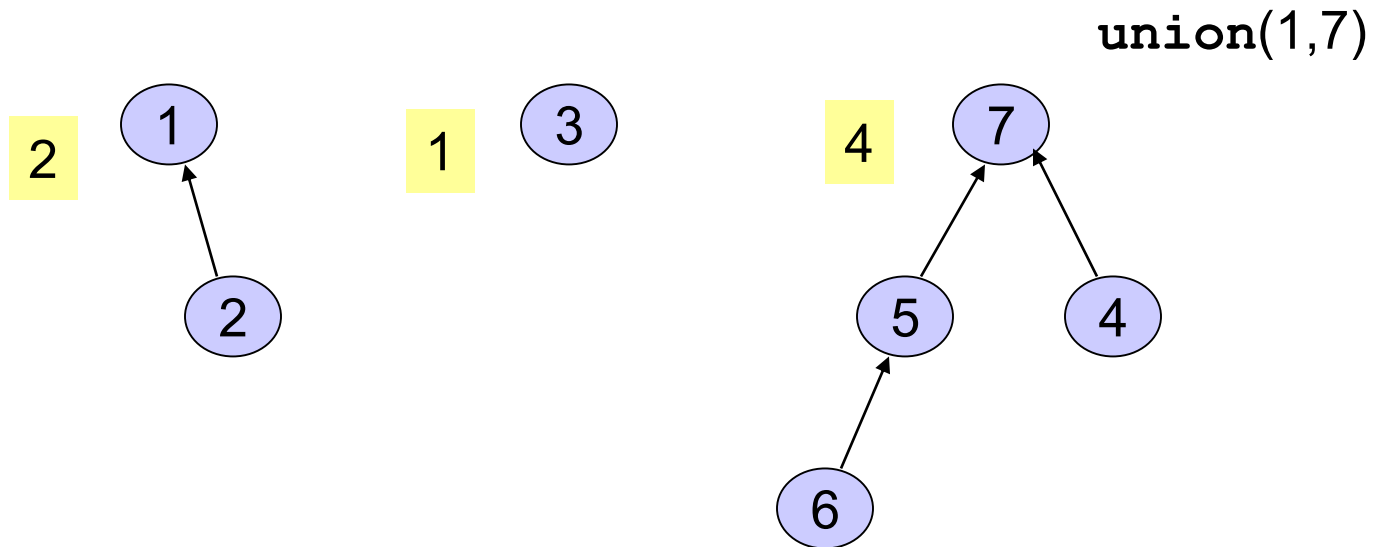
The bad case to avoid



Union-by-size

Union-by-size:

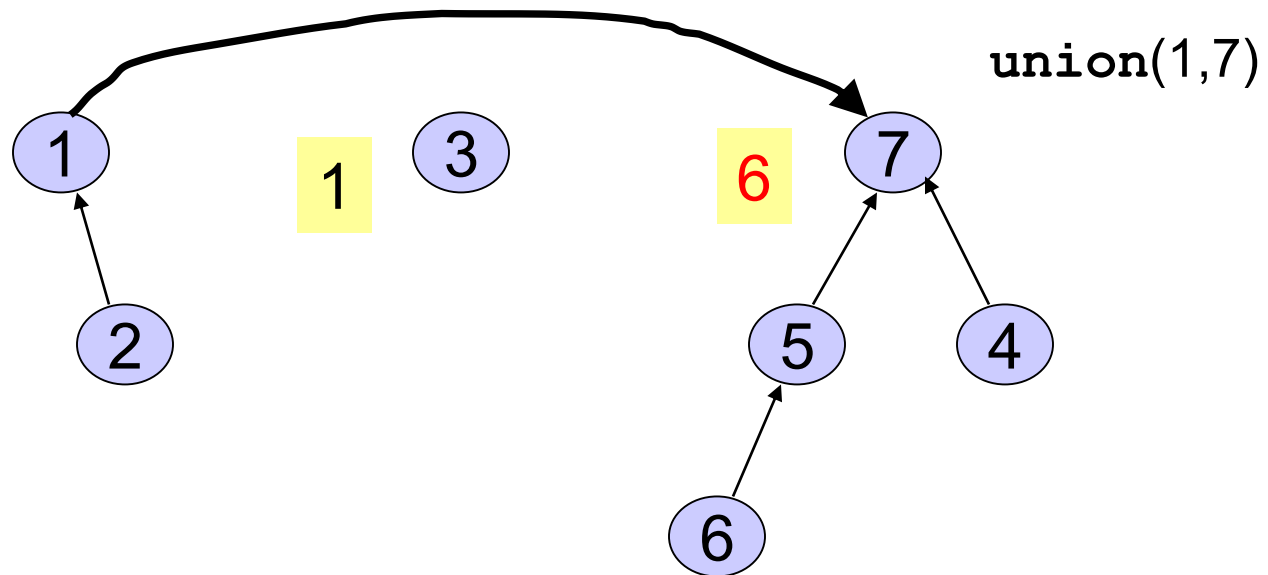
- Always point the *smaller* (total # of nodes) tree to the root of the larger tree



Union-by-size

Union-by-size:

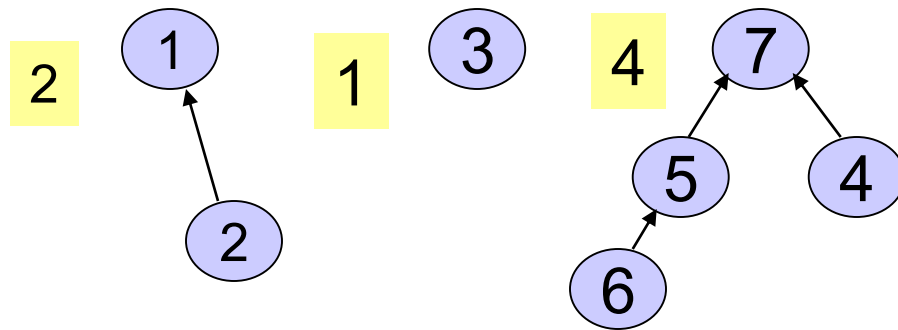
- Always point the *smaller* (total # of nodes) tree to the root of the larger tree



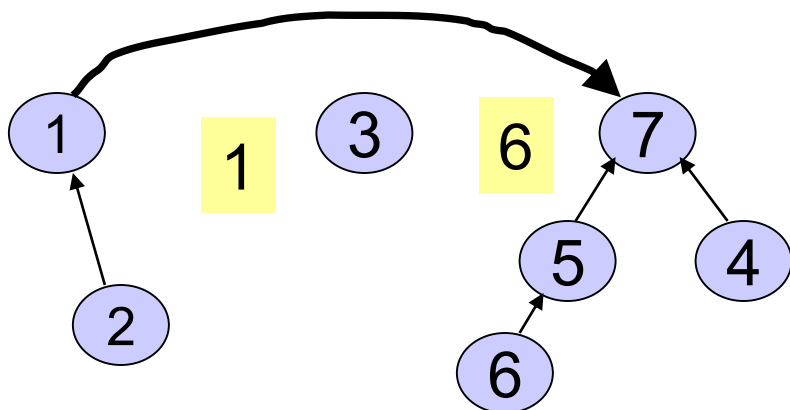
Array implementation

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

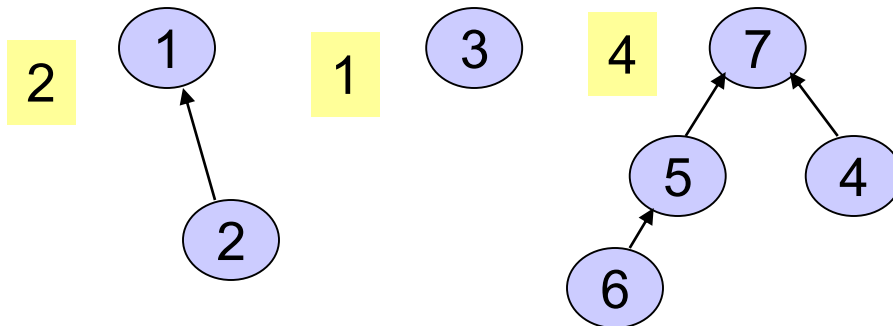


	1	2	3	4	5	6	7
up	7	1	0	7	7	5	0
weight			1				6

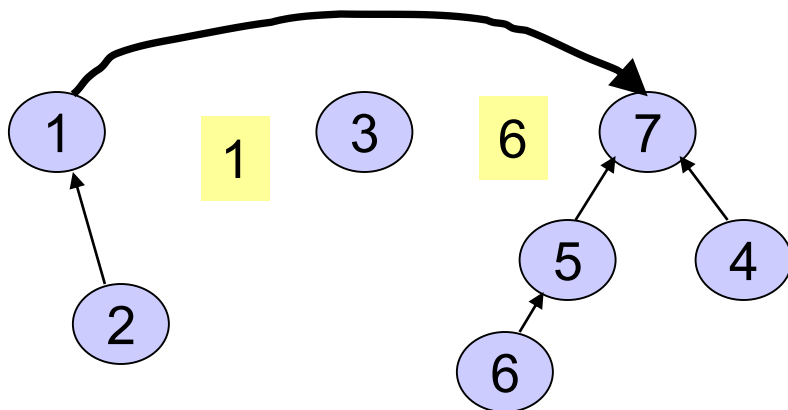
Nifty trick

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

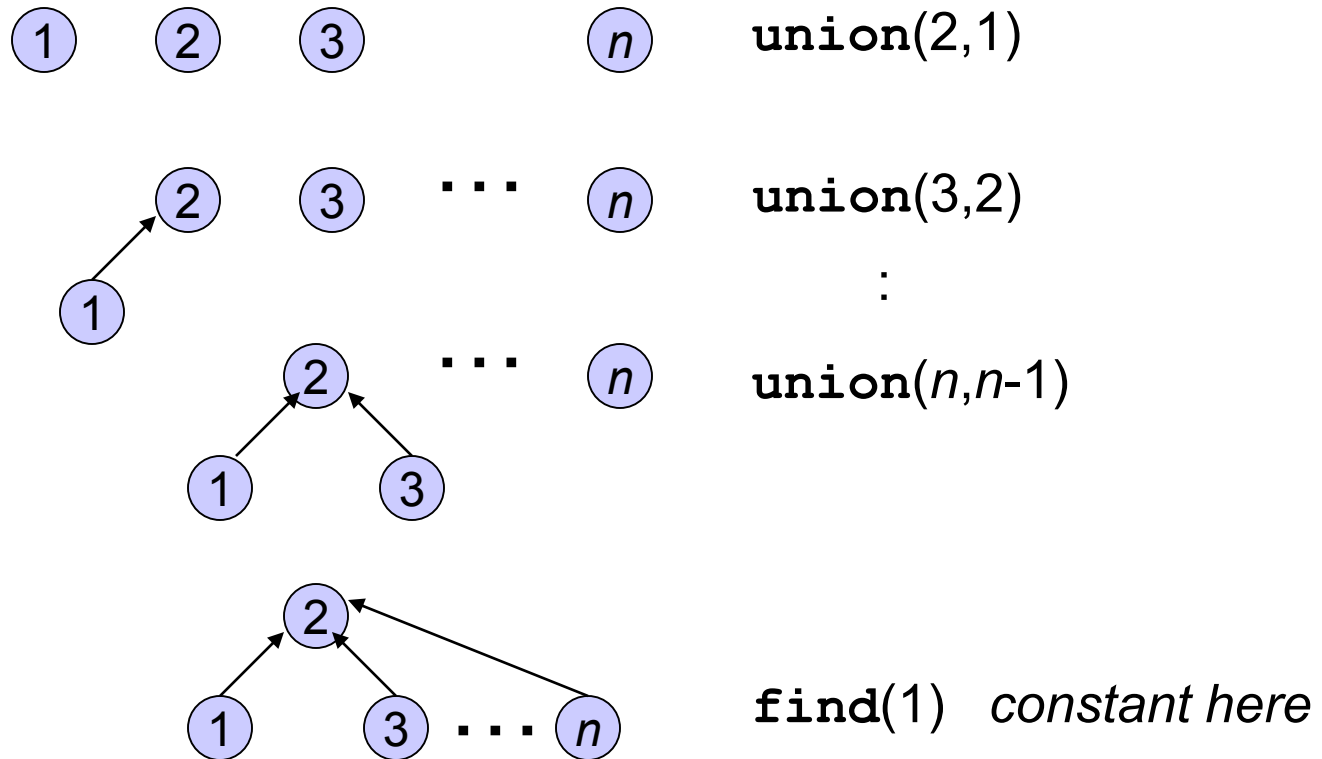


	1	2	3	4	5	6	7
up	-2	1	-1	7	7	5	-4



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

The Bad case? Now a Great case...



General analysis

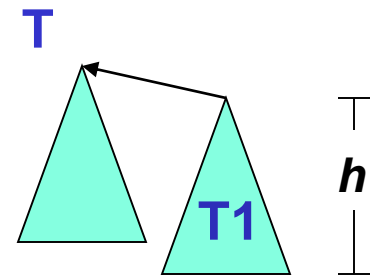
- Showing one worst-case example is now good is *not* a proof that the worst-case has improved
- So let's prove:
 - **union** is still $O(1)$ – this is “obvious”
 - **find** is now $O(\log n)$
- Claim: If we use union-by-size, an up-tree of height h has at least 2^h nodes
 - Proof by induction on h ...

Exponential number of nodes

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

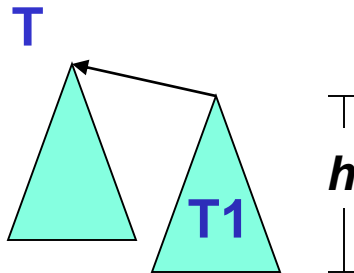
Proof by induction on h ...

- 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 (assumption)
 - 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

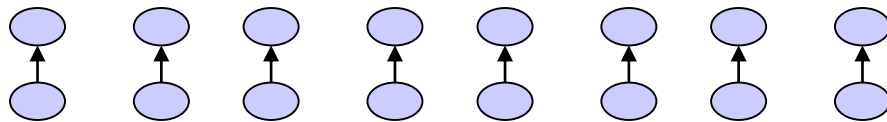


So, as usual, if number of nodes is exponential in height, then height is logarithmic in number of nodes

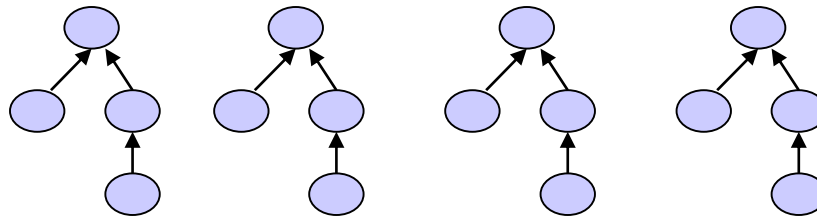
So **find** is $O(\log n)$

The new worst case

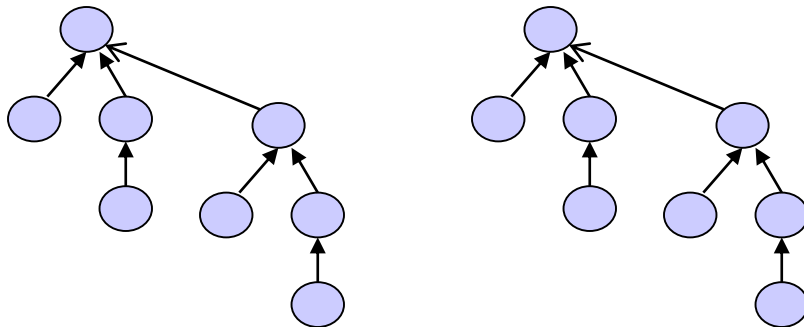
$n/2$ Unions-by-size



$n/4$ Unions-by-size

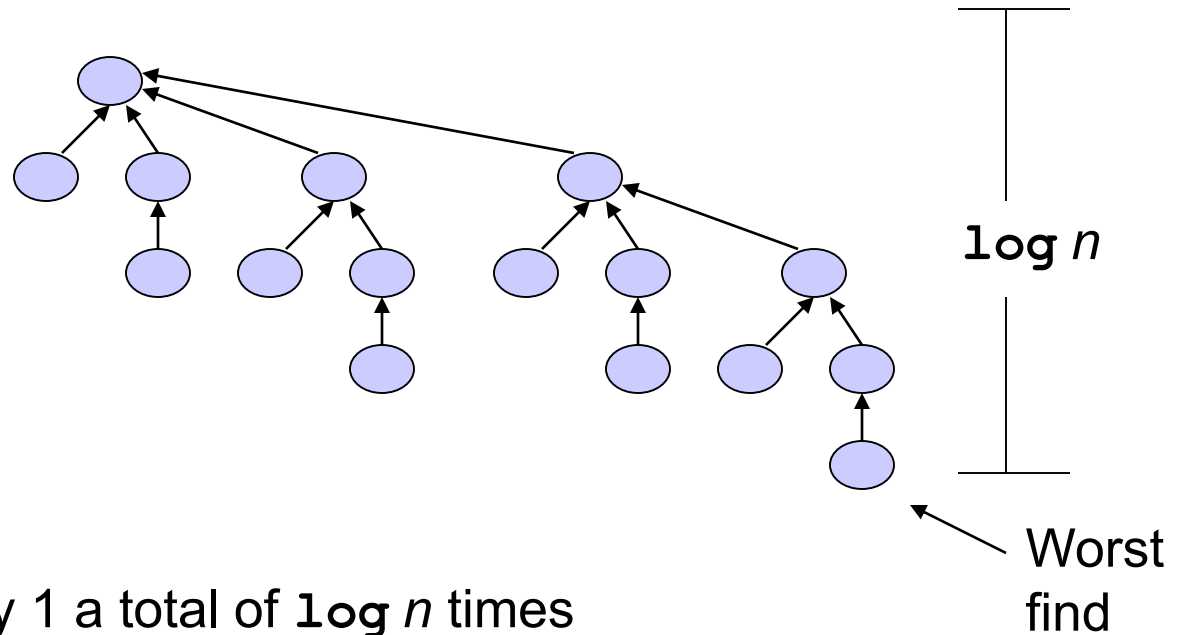


$n/8$ Unions-by-size



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

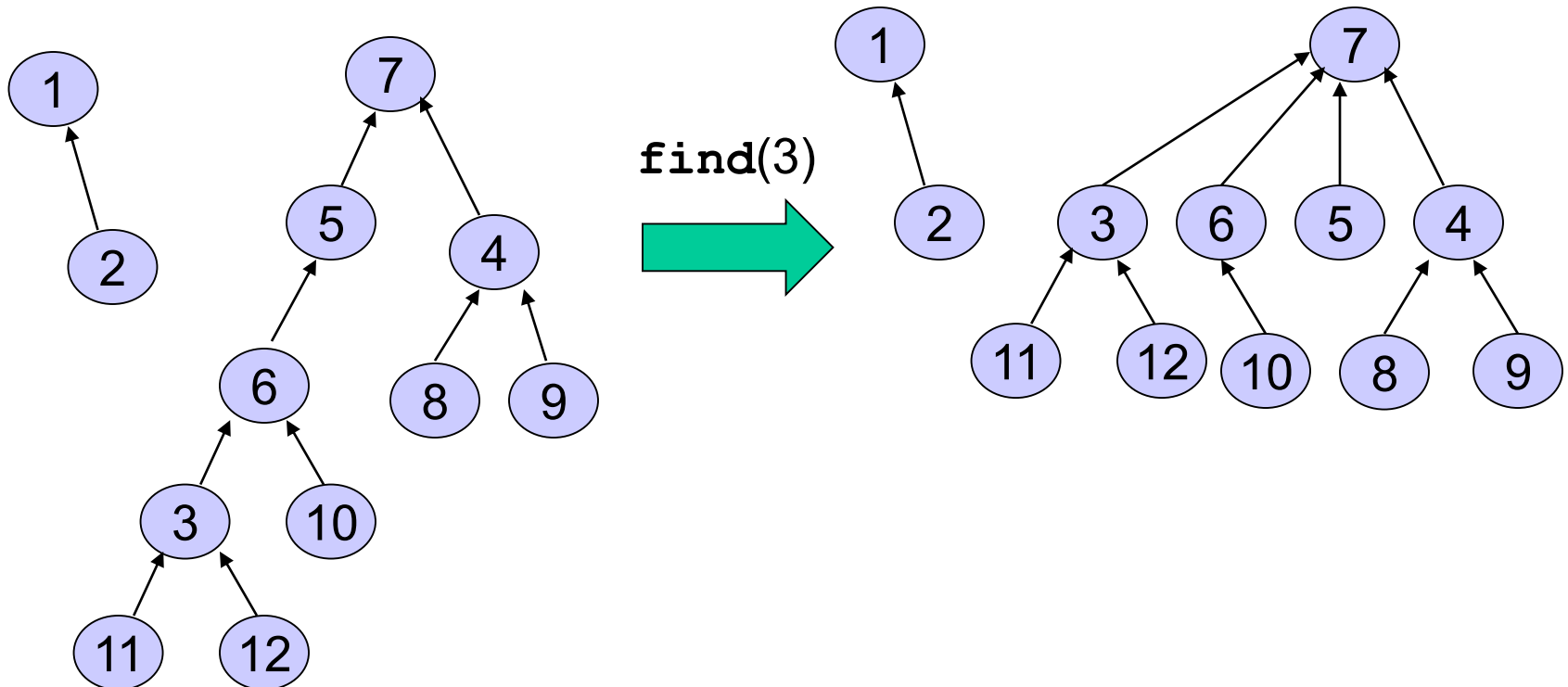
- Still guarantees logarithmic worst-case find
 - Proof left as an exercise if interested
- But does not work well with our next optimization
 - Maintaining height becomes inefficient, but maintaining size still easy

Two key optimizations

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

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

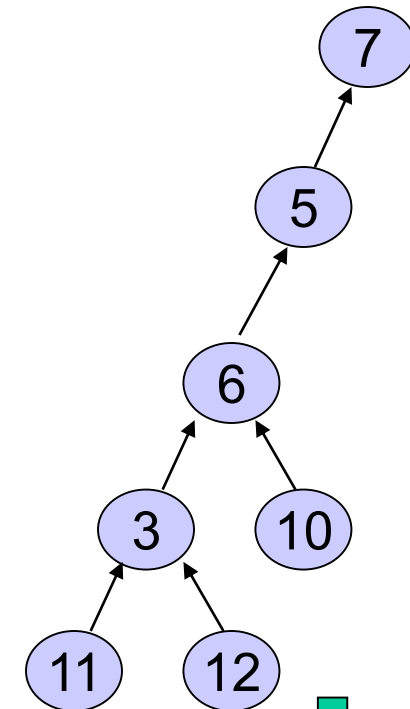
$i=3$

$r=3$

$r=6$

$r=5$

$r=7$



$old_parent=6$

↓ find(3)

$up[3]=7$

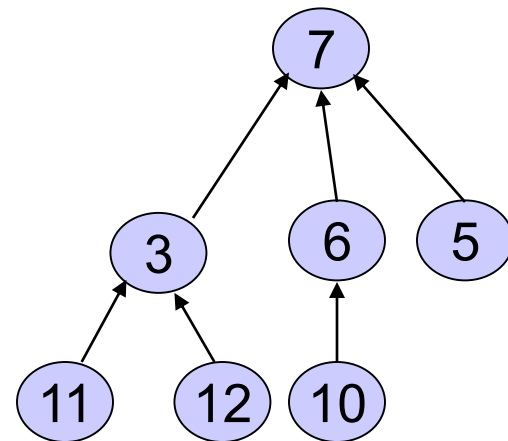
$i=6$

$old_parent=5$

$up[6]=7$

$i=5$

$old_parent=7$



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 $O(\log n)$

- We won't *prove* it – see text if curious
- But we will *understand* it:
 - How it is *almost* $O(1)$
 - Because total for m `finds` and $n-1$ `unions` is *almost* $O(m+n)$

A really slow-growing function

$\log^* x$ is the minimum number of times you need to apply “ \log of \log of \log of” to go from x to a number ≤ 1

For just about every number we care about, $\log^* x$ is less than or equal to 5 (!)

If $x \leq 2^{65536}$ then $\log^* x \leq 5$

- $\log^* 2 = 1$
- $\log^* 4 = \log^* 2^2 = 2$
- $\log^* 16 = \log^* 2^{(2^2)} = 3$ ($\log \log \log 16 = 1$)
- $\log^* 65536 = \log^* 2^{((2^2)^2)} = 4$ ($\log \log \log \log 65536 = 1$)
- $\log^* 2^{65536} = \dots\dots\dots = 5$

Almost linear

- Turns out total time for m finds and $n-1$ unions is $O((m+n) \cdot \log^*(m+n))$
 - Remember, if $m+n < 2^{65536}$ then $\log^*(m+n) < 5$ so effectively we have $O(m+n)$
- Because \log^* grows soooo slowly
 - For all practical purposes, amortized bound is constant, i.e., cost of find is constant, total cost for m finds is linear
 - We say “near linear” or “effectively linear”
- Need union-by-size and path-compression for this bound
 - Path-compression changes height but not weight, so they interact well
- As always, asymptotic analysis is separate from “coding it up”