

CSE373: Data Structures and Algorithms

Implementing the UNION-FIND ADT

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Winter 2016

This lecture material represents the work of multiple instructors at the University of Washington. Thank you to all who have contributed!

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

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Union-Find ADT

- Given an unchanging set S , **create** an initial partition of a set
 - Typically each item in its own subset: $\{a\}, \{b\}, \{c\}, \dots$
 - 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


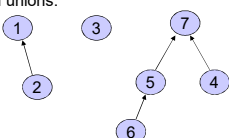
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Implementation – our goal

- Start with an initial partition of n subsets
 - Often 1-element sets, e.g., $\{1\}, \{2\}, \{3\}, \dots, \{n\}$
- 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

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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

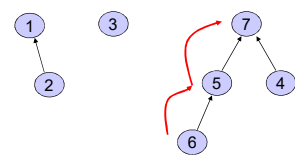
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Find

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

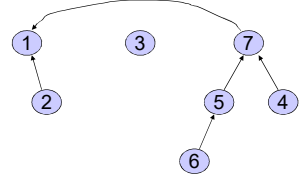


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Union

`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



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Simple implementation

- If set elements are contiguous numbers (e.g., 1,2,...,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
0	0	0	0	0	0	0
- Example:

1	2	3	4	5	6	7
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)

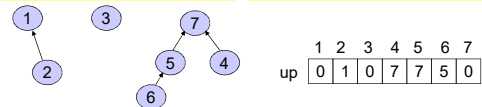
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Implement operations

```

// 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;
}
    
```



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

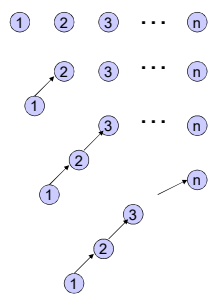
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Two key optimizations

1. Improve `union` so it stays $O(1)$ but makes `find` $O(\log n)$
 - So `m` `find`s 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` `find`s and `n-1` `unions` **almost** $O(m + n)$
 - *Path-compression*: connect directly to root during finds

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The bad case to avoid



`union(2,1)`
`union(3,2)`
 \vdots
`union(n,n-1)`

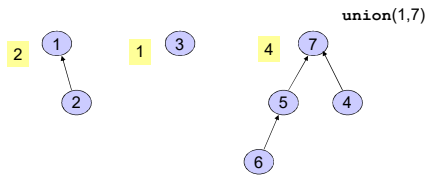
`find(1) = n` steps!!

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Union-by-size

Union-by-size:

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

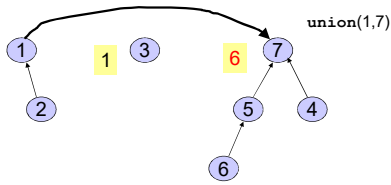


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Union-by-size

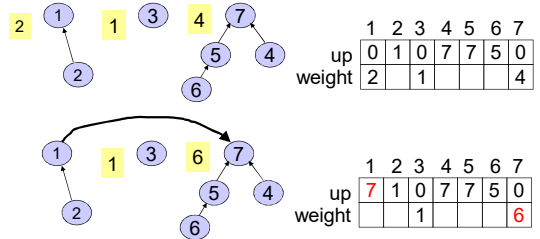
Union-by-size:

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



Array implementation

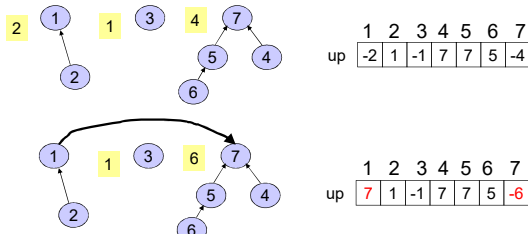
Keep the size (number of nodes in a second array)
- Or have one array of objects with two fields



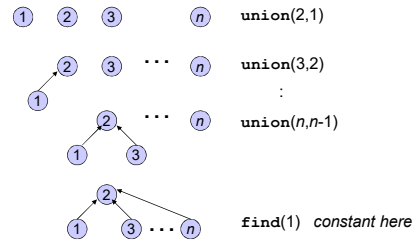
Nice trick

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



The Bad case? Now a Great case...



General analysis

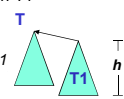
- Showing one worst-case example is now good is *not* a proof that the worst-case has improved
- So let's prove:
 - union is still $O(1)$ - this is "obvious"
 - find is now $O(\log n)$
- Claim: If we use union-by-size, an up-tree of height h has at least 2^h nodes
 - Proof by induction on h ...

Exponential number of nodes

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

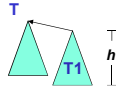
Proof by induction on h ...

- 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 T_1
 - T_1 has at least 2^h nodes by induction
 - And T has *at least* as many nodes not in T_1 than in T_1
 - Else union-by-size would have had T point to T_1 , not T_1 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

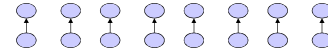


So, as usual, if number of nodes is exponential in height, then height is logarithmic in number of nodes

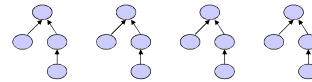
So `find` is $O(\log n)$

The new worst case

$n/2$ Unions-by-size

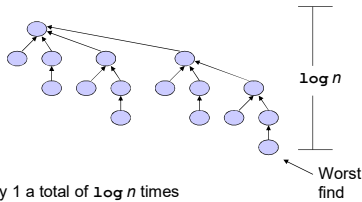


$n/4$ Unions-by-size



The new worst case (continued)

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



Height grows by 1 a total of $\log n$ times

What about union-by-height

We could store the height of each root rather than size

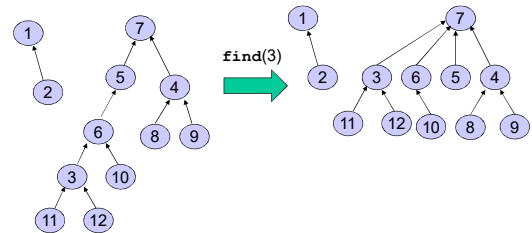
- Still guarantees logarithmic worst-case find
 - Proof left as an exercise if interested
- But does not work well with our next optimization
 - Maintaining height becomes inefficient, but maintaining size still easy

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

$i=3$
 $r=3$

$r=6$
 $r=5$
 $r=7$

$old_parent=6$

$up[3]=7$
 $i=6$
 $old_parent=5$

$up[6]=7$
 $i=5$
 $old_parent=7$

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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* $O(m+n)$

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

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Almost linear

- Turns out total time for m finds and $n-1$ unions is $O((m+n)(\log^* (m+n)))$
 - Remember, if $m+n < 2^{65536}$ then $\log^* (m+n) < 5$ so effectively we have $O(m+n)$
- Because \log^* grows sooooo 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”

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