

## Implementing Union-Find

A set is an unordered collection of things.

One way to implement a set is with a doubly-linked list.

- **makeset(x)**
    - create a linked list with a single node containing x
    - $O(1)$
  - **union(x,y)**
    - append y's list to x's list
    - $O(1)$  if you have tail pointers – set x's tail's next to point to y's head
  - **find(x)**
    - how to identify a set? – could use the head node as the representative of the set
    - given a node, it is  $O(\text{size of list})$  to find the head of its list – follow prev pointers backwards from node to the head
- total:  $O(n \times \text{makeset} + m \log n + m \times \text{find} + n \times \text{union})$   
 $= O(n + m \log n + nm + n) = O(nm)$
- (this is much worse than graph traversal, can we do better?)

## Implementing Union-Find

Can we do better? Find is the slow part...

- what's better than  $O(n)$ ? →  $O(1)$ 
  - if every node also had a pointer directly to the head, find(x) could be done in constant time!

New implementation: singly-linked list with tail pointer and each node also pointing directly to the head.

- **makeset(x)** (can actually store head pointers instead of prev pointers since the only reason to back up was to find the head)
    - $O(1)$
  - **union(x,y)**
    - $O(1)$  to append...but  $O(\text{size of } y)$  to update all head pointers in y
  - **find(x)**
    - $O(1)$
- total:  $O(n \times \text{makeset} + m \log n + m \times \text{find} + n \times \text{union})$   
 $= O(n + m \log n + m + n^2) = O(m \log n + n^2)$
- (somewhat better...)

## Implementing Union-Find

union by rank list implementation

Can we do better? Union is the slow part – what if we updated as few head pointers as possible?

- **union(x,y)**
    - $O(1)$  to append the smaller list to the larger list...but still  $O(\text{size of smaller list})$  to update head pointers in the smaller list
      - can store list sizes so it is possible to find the smaller list in  $O(1)$
- total:  $O(n \times \text{makeset} + m \log n + m \times \text{find} + n \times \text{union})$
- observation: in the worst case  $O(\text{size of smaller list})$  is  $O(n)$ , but we know something about the series of unions
- each time we union and the head pointer for a node is updated, the node is going into a set at least twice as big as it came from
  - this can happen at most  $\log n$  times if there's a total of  $n$  elements
- thus  $n$  unions with appending the smaller list is  $O(n \log n)$  instead of  $O(n^2)$
- $= O(n + m \log n + m + n \log n) = O((n+m) \log n)$
- an improvement for sparse graphs!

## Implementing Union-Find

$O((n+m) \log n)$  for Kruskal's algorithm is pretty good – can we do better?

Observation.

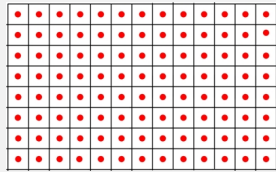
- sorting the edges by weight requires  $O(m \log n)$ , which will dominate  $O(n \log n)$  as long as the graph is connected
  - improving the data structure will result in elapsed time gains, but not change the big-Oh

However...

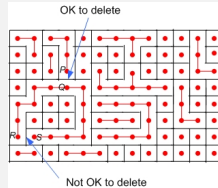
- union-find has applications beyond Kruskal's algorithm
  - greater efficiency in union-find operations may make a difference there

## Maze Creation

In a good maze, every room is reachable from every other and there's only one possible path from start to goal.  
How to generate a random maze?



start with all of the walls and every room in a separate set



while there is more than one set left  
- choose a random wall  
- if the rooms on either side belong to different sets, knock down the wall

## Implementing Union-Find

We improved the implementation to speed up Kruskal's algorithm by making union the slow part instead of find.

We've seen  $O(n) \leftrightarrow O(1)$  tradeoffs before...and sometimes could compromise on  $O(\log n)$  for both.

Will that work here?

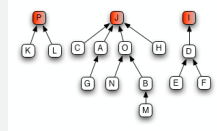
Observation.

- trees are often associated with  $O(\log n)$  run times
  - the height can be as good as  $O(\log n)$

## Implementing Union-Find

Idea:

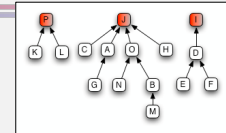
Represent a set as a directed tree.



- **makeset(x)**
  - create a tree with a single node containing x
  - $O(1)$
- **union(x,y)**
  - find the roots of x's and y's trees
  - make y's root point to x's root
  - $O(\text{find})$
- **find(x)**
  - use the root as the representative element
  - given a node, it is  $O(\text{height of tree})$  to find the root of its tree

## Implementing Union-Find

union by rank tree implementation



How tall are the trees?

- could be  $n$  – can we do better?
  - **union(x,y)**
    - find the roots of x's and y's trees
    - make the root of the shorter tree point to the root of the taller tree
      - store  $\text{rank}(v)$  = height of subtree rooted at  $v$  for each node so the shorter tree can be found in  $O(1)$  time
      - only the rank of the taller tree's root may change as a result of the union –  $O(1)$  to update
  - result is  $O(\log n)$  height for each tree – so find is  $O(\log n)$ 
    - idea: height of merged tree only increases if the two trees are equally tall – that merge doubles the size of the tree so it can happen at most  $\log n$  times
- total:  $O(n \times \text{makeset} + m \log n + m \times \text{find} + n \times \text{union})$   
 $= O(n + m \log n + m \log n + n \log n) = O((n+m) \log n)$   
 - (no improvement over union-by-rank lists)

## Implementing Union-Find

$O(\log n)$  find(x) isn't bad, but  $O(1)$  is still better...

Observation:

- could get  $O(1)$  find(x) if each node had a direct pointer to the root

But:

- updating these pointers during union is too expensive

Observation:

- find(x) locates the root for every node between x and the root

It seems a waste to throw that information away!

## Implementing Union-Find

The best of both worlds: (*path compression*)

- union(x,y)
  - find the roots of x's and y's trees
  - make the root of the shorter tree point to the root of the taller tree
- find(x)
  - locate the root
  - update the pointers for every node on the path  $x \rightarrow$  root to point directly to the root

## Implementing Union-Find

Running time?

- find(x) and thus union(x,y) are still  $O(\text{height of tree})$

What's the height of the trees?

- path compression keeps the height of the trees short
- find(x) and union(x,y) are effectively  $O(1)$

## Implementing Union-Find

“Short”?! “Effectively”?!

- based on *amortized time*, not worst case
  - an individual operation may take longer, but if a sequence of  $k$  operations takes a total of  $O(k f(n))$  time, we can say each operation is  $O(f(n))$  amortized
- total time for  $m$  find(x) operations is  $O((m+n) \log^* n)$ 
  - on average,  $O(n/m \log^* n)$  per find
  - with  $m > n$  (typical), this is  $O(\log^* n)$ 
    - $\log^* n$  = the number of successive log operations to bring  $n$  down to 1
    - extremely slow growing! (value  $< 5$  for any value of  $n$  you might encounter, and thus is effectively constant time)

## Union-Find Summary

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- union-by-rank list implementation yields  $O((n+m) \log n)$  for Kruskal's algorithm
  - $O(1)$  makeset(x)
  - $O(1)$  find(x)
  - $O(n \log n)$  for a series of  $n$  union(x,y)
- union-by-rank tree implementation with path compression yields  $O(m \log n)$  for Kruskal's algorithm
  - $O(1)$  makeset(x)
  - effectively  $O(1)$  find(x) and union(x,y)
    - the tree height is a very slow-growing  $\log^*$
    - *amortized* over a series of operations

Both are an improvement over our initial  $O(nm)$  algorithm.

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