Disjoint-Set Forests

Outline for Today

- Iterated Functions
 - Making an implicit idea explicit.
- Incremental Connectivity
 - Finding connected nodes as a graph changes.
- Disjoint-Set Forests
 - A surprisingly simple and subtle data structure.
- Analyzing Disjoint-Set Forests
 - A clever, nuanced analysis with a surprising result.

- Recursive functions work by converting a problem of size *n* into one or more subproblems of a smaller size.
- How much smaller those subproblems are indicates how many levels of recursion we'll have.
 - $n \rightarrow n 1$: $\Theta(n)$ levels.
 - $n \rightarrow n/2$: $\Theta(\log n)$ levels

Let f be a function. The *iterated function of* f, denoted f* is a function defined as follows:

$$f^{*}(n) = \begin{cases} 0 & \text{if } f(n) \le 1\\ 1 + f^{*}(f(n)) & \text{otherwise} \end{cases}$$

- Intuitively, f*(n) is (roughly) the number of times you need to apply f to n to reduce it to a sufficiently small constant.
 - If $f(n) \le 1$, no steps are needed.
 - Otherwise, you need one step to turn n into f(n), then f*(f(n)) more steps from there.

	<i>f</i> *(<i>n</i>)	As seen in
f(n)=n-1	Θ(n)	Linear search
f(n) = n/2	$\Theta(\log n)$	Binary search
$f(n) = n^{1/2}$	$\Theta(\log \log n)$	Rabin's closest pair of points algorithm
$f(n) = \log n$	$\Theta(\log^* n)$	Succinct binary rank

Iterated Logarithms

- **Intuition:** The log function is incredibly effective at shrinking down large quantities.
 - Number of protons in the known universe: $\approx 2^{240}$.
 - log⁽⁰⁾ 2²⁴⁰ = 1,766,847,[... 57 digits ...],292,619,776
 - $\log^{(1)} 2^{240} = 240$
 - $\log^{(2)} 2^{240} \approx 7.91$
 - $\log^{(3)} 2^{240} \approx 2.98$
 - $\log^{(4)} 2^{240} \approx 1.58$
 - $\log^{(5)} 2^{240} \approx 0.66$
- So $\log^* 2^{240} = 4$.
- The *iterated logarithm of n*, denoted **log*** *n*, grows *much* more slowly than log *n*.

Intuiting log* n

• What is log* *n* for the value of *n* shown below?



- **Answer**: $\log^* n = 16$.
- The value of n is inconceivably large, and yet log* n is small enough to hold in your hand. The log* function grows very, very slowly!

Iterated Iterates

• What is the value of this expression?



- After taking one log*, we're left with $16 = 2^{2^2}$.
- After taking another log*, we're left with 2.
- After taking another \log^* , we're left with 0.
- So the above expression evaluates to **2**.
- How big of an input do we need to get log** n to be 3?

Just how slowly can a function grow?

Incremental Dynamic Connectivity

Kruskal's Algorithm

- **Kruskal's Algorithm** finds an MST of a graph. It works as follows:
 - Remove all edges from the graph and sort them from lowest to highest.
 - Repeatedly insert edges back into the graph, as long as their endpoints aren't already reachable from each other.



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Incremental Connectivity

- Kruskal's algorithm needs a data structure that solves *incremental connectivity*.
 - We begin with an empty graph.
 - We need to be able to add new edges to the graph and check whether arbitrary pairs of nodes are connected.
- *Question:* How efficiently can we do this?



- Idea: Assign a
 representative to each
 CC in the graph.
- To see if two nodes are in the same CC, check if they have the same representative.
- To link together two different CCs, change the representative of all the nodes in one CC to be the representative of the other CC.



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- Here's how we'll implement this idea.
 - Each node has a *parent pointer*.
 - Representatives' parent pointers are null.
 - Other nodes' parent pointers form chains leading to the representative.
- Although the original graph is undirected, parent pointers are directed.
- This data structure is called a *disjoint-set forest*.



- We'll support two operations.
- *find*(x) returns x's representative. It works by following parent pointers until we hit the representative.
- union(x, y) merges the clusters containing x and y. It works by finding x and y's representatives. If they aren't equal, it assigns one of those representatives the other as a parent.



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- Unfortunately, this system can be very slow.
- If we aren't careful with how we link trees, the cost of a *find* or *union* can grow to $\Theta(n)$, where *n* is the number of nodes in the graph.
- Can we do better?



- Assign each node a *rank*, initially 0.
- When linking two representatives *x* and *y*:
 - If one representative has a lower rank than the other, set its parent to the other.
 - Otherwise, arbitrarily set x's parent to y, then increment y's rank.
- This keeps the lengths of parent chains low.





- Lemma: A node of rank r has children of ranks 0, 1, 2, ..., and r – 1.
- **Proof:** Induction!
 - A node of rank 0 has no children.
 - A node v of rank r + 1, at the time its rank was increased, was a tree of rank r that got another tree of rank r as a child.
 - By the IH v already had children of ranks
 0, 1, 2, ..., r 1. Now it also has a child of rank r. ■



- Our lemma tells us, indirectly, that the "simplest" tree whose root has rank r is a binomial tree of order r.
- A nice consequence of this is that all trees in a forest of n nodes have height O(log n), so each union and find takes time O(log n).
- Can we do better?



An Observation

- Suppose we call find(x) multiple times.
- Each time we do that, we may have to traverse a chain of O(log n) nodes to find its representative.
- Do we really need to scan things so many times?



Path Compression

- Path compression is an optimization on the find operation.
- After figuring out x's representative, change the parent pointers of all of x's ancestors to point directly to x's representative.
- This makes it a lot faster to find representatives across multiple operations.



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Path Compression

• The resulting code for our data structure is surprisingly simple:

```
Node* find(Node* source) {
    if (source->parent == nullptr) return source;
    /* Path compression: update parent before returning. */
    source->parent = find(source→parent);
    return source->parent;
}
void doUnion(Node* one, Node* two) {
    /* Find the representatives. */
    one = find(one);
    two = find(two);
    if (one->rank > two->rank) swap(one, two);
    /* Link and update ranks if needed. */
    one->parent = two:
    if (one->rank == two->rank) two->rank++;
```

• Now, all we have to do is analyze the runtime.

Analyzing Disjoint-Set Forests

History

- The analysis of union-by-rank plus path compression has a long history.
- For a while, its actual efficiency was an open problem!
- In 1979 Tarjan proved a tight upper bound on the runtime using a clever and nuanced analysis, and provided a matching lower bound.
- In 2003 Seidel and Sharir arrived at the same upper bound using a totally different technique.
- Both analyses require a careful analysis of the costs of the operations and result in a very surprising result.
- The analysis I'll share comes from Seidel and Sharir and is based on <u>this set of lecture slides</u> from an algorithms course at Harvard.

Our Analysis

- We're going to analyze a slightly simplified version of this problem.
- We'll be given a forest F formed purely from union-byrank, then do a series of path compressions on it.
 - These don't have to go all the way from a node to its representative.
- Our goal will be to bound the total amount of work done.
- *Great Exercise:* Show that this analysis carries over to the case of interleaved *union*s and *find*s.



Our Analysis

- Some notation we'll use throughout this analysis:
 - Let **n** be the number of nodes in the disjoint-set forest.
 - Let **m** be the number of operations performed.
 - Let **r** be the maximum rank of any node in the forest. (We know $r = O(\log n)$, but could be lower.)
- In practice, we'll have $m = \Omega(n)$, and we'll assume this in our analysis.



Our Analysis

- We will specifically focus on the number of times a node's parent changes.
- Why?
 - Each operation does O(1) work, plus work proportional to the number of parents changed.
- The total work done is then $\Theta(m + \#$ changes).


A Starting Analysis

- **Lemma:** The number of pointer changes is at most $m + n \cdot r / 2$.
- **Proof Sketch:** Consider nodes of zero and nonzero rank.

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- **Nodes of rank 0:** A node of rank 0 only has its parent change if it is the start node of a compress. There are *m* compresses, so these pointers change at most *m* times.
- **Nodes of nonzero rank:** When a parent changes, the new parent's rank is bigger than the old parent's rank, so a node's rank can increase at most r times. There are at most n / 2 nodes of nonzero rank. This gives a bound of $n \cdot r / 2$.

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A Starting Analysis

- Our starting analysis is weak.
 - Compressing a path impacts other nodes not on that path.
 - Nodes with high starting rank have can't have their parents change too many times.
- These effects work differently in different parts of the tree.
 - The first effect is more pronounced at the bottom of the forest.
 - The second effect is more pronounced at the top.
- **Idea:** Split the forest into a "top forest" and "bottom forest," and analyze the costs in each forest separately.



- As before, let r be the maximum rank in \mathcal{F} .
- Suppose that, somehow, we pick a rank s(r) as a separating rank.
- Then, split our forest *F* into two forests:
 - \mathcal{F} consists of all nodes of rank s(r) or below.
 - \mathcal{F}_+ consists of all nodes of rank above s(r).
- **Goal:** Split the cost of compressions across \mathscr{F}_{-} and \mathscr{F}_{+} .





Some Terminology

- Let C(m, n, r) be the maximum number of pointer changes that can be made if there are m compresses, n nodes, and the maximum rank is r.
- Using this notation, our earlier result is that

 $C(m, n, r) \leq m + n \cdot r / 2.$

- **Question:** What is C(m, n, 0)? What's C(m, n, 1)?
- **Goal:** Split \mathscr{F} into \mathscr{F}_+ and \mathscr{F}_- , find a way to write a recurrence relation for C(m, n, r), then solve the recurrence to get a tight bound on the cost of any series of **unions** and **finds**.

- Focus on any one compression from x to y. Let's see how it interacts with *F* – and *F*+.
- **Case 1:** x and y are both in \mathcal{F}_+ .



- Focus on any one compression from x to y. Let's see how it interacts with *F* – and *F*+.
- **Case 1:** x and y are both in \mathcal{F}_+ .
- We can recursively handle this compression when bounding the work done purely in \mathcal{F}_+ .



• **Case 2:** x and y are both in \mathcal{F}_{-} .



- **Case 2:** x and y are both in \mathcal{F}_{-} .
- We can recursively handle this compression when bounding the work done purely in *F*_.



• **Case 3:** x is in \mathcal{F}_- and y is in \mathcal{F}_+ .



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- **Case 3:** x is in \mathcal{F}_- and y is in \mathcal{F}_+ .
- We compress from *b* to *y*, purely in \mathcal{F}_+ .
- a, whose parent was already in F+, gets a new parent in F+.
- Every node from x

 (inclusive) and a (exclusive),
 whose parents were in F-,
 now has a parent in F+.

Case 3: x is in \mathcal{F}_- and y is in \mathcal{F}_+ .

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 (inclusive) and a (exclusive),
 whose parents were in F-,
 now has a parent in F+.

Recursively handle this when processing \mathcal{F}_+ .

Happens once per compression from \mathcal{F} - to \mathcal{F} +.

Happens once per non-root node in F-, counting across all compressions

- Claim: The cost of compressions crossing from *F* – to *F* + can be bounded by
 - the cost of some compressions done purely in *F* + (the top parts of the compressions),
 - the total number of compressions from F - to F + (changing the parents of nodes in F - whose parents are already in F +), and
 - the number of nodes in *F* whose parents are in *F* (each of which may get a parent in *F*+ for the first time at most once).

Recursively handle this when processing \mathcal{F}_+ .

Happens once per compression from \mathcal{F} – to \mathcal{F} +.

Happens once per non-root node in F-, counting across all compressions

Putting It All Together

- **Claim:** The cost of *all* the compressions performed in *F* is bounded by the following:
 - The cost of some compressions purely in \mathcal{F}_{-} .
 - The cost of some compressions purely in \mathcal{F}_+ .
 - This includes compressions originally in \mathcal{F}_+ , plus the "tops" of compressions from \mathcal{F}_- to \mathcal{F}_+ .
 - The number of compresses from \mathscr{F} to \mathscr{F} +.
 - This accounts for changing the parents of nodes in \mathscr{F}_- whose parents are already in \mathscr{F}_+ .
 - The number of nodes in \mathcal{F}_- with parents in \mathcal{F}_- .
 - Each of these nodes may get a parent in \mathscr{F}_+ for the first time once.



Cost of compressions purely in F+.	C(m +, n +, r)
Cost of compressions purely in F	The "tops" of all compressions running from \mathscr{F} to \mathscr{F}_+ are
The number of compresses from F – to F+	handled in this bunch. Let m_+ be the number of compressions charged to \mathscr{F}_+ , including both compressions
Number of nodes in \mathcal{F} with parents in \mathcal{F} .	purely within \mathscr{F}_+ and the "tops" of compressions crossing from \mathscr{F} to \mathscr{F}_+ .

Cost of compressions purely in F+.	C(m+, n+, r)	
Cost of compressions purely in F–.	C(m - m +, n , s(r))	
The number of compresses from F – to F+	rank r	
Number of nodes in \mathcal{F} with parents in \mathcal{F} .	rank s(r)	

Cost of compressions purely in F+.	C(m+,	n+, r)
Cost of compressions purely in F–.	C(m – m+	., n , s(r))
The number of compresses from F– to F+	<i>m</i> +	
Number of nodes in \mathcal{F} with parents in \mathcal{F} .		(Since this includes all compresses from F – to F+).



Cost of compressions purely in F+.	C(m+, n+, r)	
Cost of compressions purely in <i>F</i>	$C(m - m_+, n, s(r))$	
The number of compresses from F – to F +	<i>m</i> +	
Number of nodes in \mathcal{F} with parents in \mathcal{F} .	$n - n_+ \cdot s(r)$	

Putting it all together:

 $C(m, n, r) \leq C(m - m_+, n, s(r)) +$ $C(m_+, n_+, r) +$ $m_+ + n - n_+ \cdot s(r).$

- Now, "all" we need to do is solve this.
- Don't panic! This is indeed tricky, but it's not as bad as it looks.

 $C(m, n, r) \leq C(m - m_+, n, s(r)) + C(m_+, n_+, r) + m_+ + n - n_+ \cdot s(r).$

• **Recall:** $C(m, n, r) \le m + n \cdot r / 2$.

 $C(m, n, r) \leq C(m - m_+, n, s(r)) +$ $m_+ + n_+ \cdot r / 2 +$ $m_+ + n - n_+ \cdot s(r).$

• **Recall:** $C(m, n, r) \le m + n \cdot r / 2$.

 $C(m, n, r) \leq C(m - m_+, n, s(r)) + 2m_+ + n + n_+ + (r / 2 - s(r)).$

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 $C(m, n, r) \leq C(m - m_+, n, s(r)) + 2m_+ + n + n_+ + (r / 2 - s(r)).$

• **Clever Decision:** Set s(r) = r / 2.

$C(m, n, r) \le C(m - m_+, n, r/2) + 2m_+ + n.$

• **Clever Decision:** Set s(r) = r / 2.







Where We Are

• We've just proven that

$C(m, n, r) \leq 2m + n \lg r.$

- The maximum rank in an *n*-node forest is $r = O(\lg n)$.
- This gives a bound of O(m + n log log n) for any series of operations.
- That's a lot better than the O(m log n) we started with – and it's just due to better accounting, rather than a fundamental reenvisioning of the data structure.
- Is this a tight bound, or can we do better?

 $C(m, n, r) \leq C(m - m_+, n, s(r)) + C(m_+, n_+, r) + m_+ + n - n_+ \cdot s(r).$

- Recall: $C(m, n, r) \leq m + n \cdot r / 2$.
- **Recall:** $C(m, n, r) \leq 2m + n \lg r$.

$$C(m, n, r) \leq C(m - m_+, n, s(r)) + 2m_+ + n_+ \cdot \lg r + m_+ + n - n_+ \cdot s(r).$$

- Recall: $C(m, n, r) \leq m + n \cdot r / 2$.
- **Recall:** $C(m, n, r) \leq 2m + n \lg r$.

 $C(m, n, r) \leq C(m - m_+, n, s(r)) +$ $3m_+ + n +$ $n_+ \cdot (\lg r - s(r)).$

- Recall: $C(m, n, r) \leq m + n \cdot r / 2$.
- **Recall:** $C(m, n, r) \leq 2m + n \lg r$.

 $C(m, n, r) \le C(m - m_+, n, s(r)) +$ $3m_+ + n +$ $n_+ \cdot (\lg r - s(r)).$

• **Clever Decision:** Set $s(r) = \lg r$.

$C(m, n, r) \le C(m - m_+, n, \lg r) +$ $3m_+ + n$

• **Clever Decision:** Set $s(r) = \lg r$.



Where We Are

• We've just proven that

$C(m, n, r) \leq 3m + n \lg^* r.$

- Since r = O(log n) this gives a bound of
 O(m + n log* n) for any series of operations.
- That's a *substantial* improvement over our previous bound – and all we did was feed the analysis back into itself!
- Can we do better?
Notice Something?

If we start with this bound on <i>C</i> (<i>m</i> , <i>n</i> , r)	then we get this stronger bound on <i>C</i> (<i>m</i> , <i>n</i> , <i>r</i>):
$m + n \cdot r / 2$	$2m + n \lg r$
$2m + n \lg r$	3 <i>m</i> + <i>n</i> lg* <i>r</i>
km + n f(r)	$(k+1)m + n f^*(r)$

 $C(m, n, r) \leq C(m - m_+, n, s(r)) + C(m_+, n_+, r) + m_+ + n - n_+ \cdot s(r).$

• **Assume:** $C(m, n, r) \leq km + n \cdot f(r)$

 $C(m, n, r) \leq C(m - m_+, n, s(r)) + (k+1)m_+ + n + n + n_+ \cdot (f(r) - s(r)).$

• **Assume:** $C(m, n, r) \leq km + n \cdot f(r)$

 $C(m, n, r) \le C(m - m_+, n, s(r)) + (k+1)m_+ + n + n + n_+ \cdot (f(r) - s(r)).$

• **Clever Idea:** Set s(r) = f(r).

$C(m, n, r) \le C(m - m_+, n, f(r)) + (k+1)m_+ + n.$

• **Clever Idea:** Set s(r) = f(r).



 $\leq (k+1)m + nf^*(r)$

Interpreting This Result

• We now have a family of bounds on the cost of operations on a disjoint-set forest:

 $m + n \cdot (r/2)$ $2m + n \lg r$ $3m + n \lg^* r$ $4m + n \lg^{**} r$ $5m + n \lg^{***} r$

• Which of these is the "best" bound?

Interpreting This Result

• For now, focus on these bounds:

 $2m + n \lg r$ $3m + n \lg^* r$ $4m + n \lg^{**} r$ $5m + n \lg^{***} r$ $6m + n \lg^{****} r$

• More generally, we have bounds of the form

```
(k+2)m + n \lg^{*(k)} r.
```

- There's some point at which making k larger makes that first term larger without decreasing the second term.
- What is it?

The Ackermann Inverse

• The Ackermann inverse function, denoted $\alpha(z)$, is defined as follows:

 $\alpha(z) = \min\{ k \in \mathbb{N} \mid \log^{*(k)} z \leq 1 \}$

- Intuitively, this counts how many times you have to put stars on $\log^{***...**} z$ before it drops to 1.
- This function grows more slowly than anything in the iterated logarithm family – and that should give you a sense of just how slowly this function grows!
- Worthwhile Activity: find the smallest natural numbers where α produces 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, and 10.

The Ackermann Inverse

• We have a bound of

 $(k + 2)m + n \log^{*(k)} r.$

• Picking $k = \alpha(r) = \alpha(\log n)$, and the bound on the cost of any series of *m* operations is.

 $O(m\alpha(\log n) + n).$

• This is *essentially* "O(m + n)," because that α term is a constant for any input that could ever be fed in with the resources we know about in the universe. But technically speaking it's superlinear.

A Tighter Analysis

- By being a bit more clever with the analysis, we can tighten the bound as follows.
- Define $\alpha(m, n)$ as $\alpha(m, n) = \min\{k \in \mathbb{N} \mid \log^{*(k)}(m / n) \leq \log n\}.$
- Then the cost of *m* operations on an *n*-element forest can be shown to be $O(m\alpha(m, n))$, a slight improvement over what we just did here.

Major Ideas for Today

- Iterated functions generalize the idea of "how many times can you divide by two before you run out of things?"
- Iterated logarithms are a family of very slowlygrowing functions, each of which grows more slowly than the previous one.
- The Ackermann inverse function grows slower than any number of iterated logarithms and essentially count what level of iteration is needed to clear a number.

Next Time

- Euler Tour Trees
 - Fully dynamic connectivity in forests.
- Augmented Dynamic Trees
 - Figuring out information about connected components in sublinear time.