

Disjoint-Set Forests

Thanks for Showing Up!

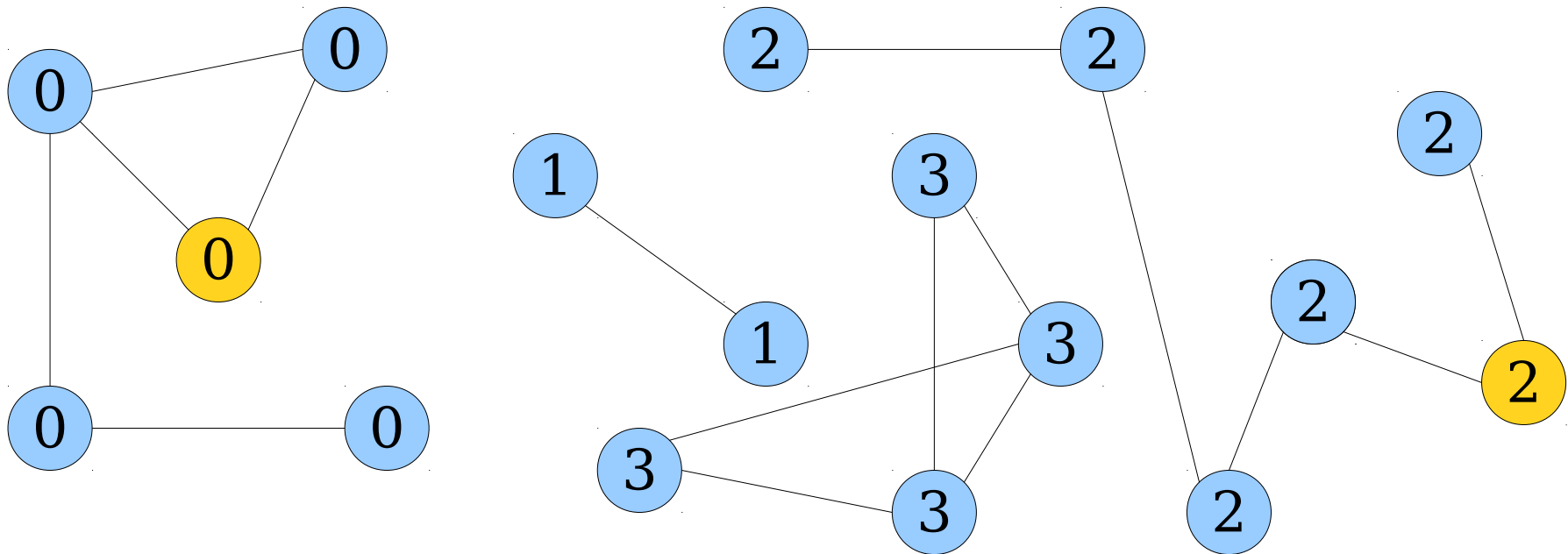
Outline for Today

- ***Incremental Connectivity***
 - Maintaining connectivity as edges are added to a graph.
- ***Disjoint-Set Forests***
 - A simple data structure for incremental connectivity.
- ***Union-by-Rank and Path Compression***
 - Two improvements over the basic data structure.
- ***Forest Slicing***
 - A technique for analyzing these structures.
- ***The Ackermann Inverse Function***
 - An unbelievably slowly-growing function.

The Dynamic Connectivity Problem

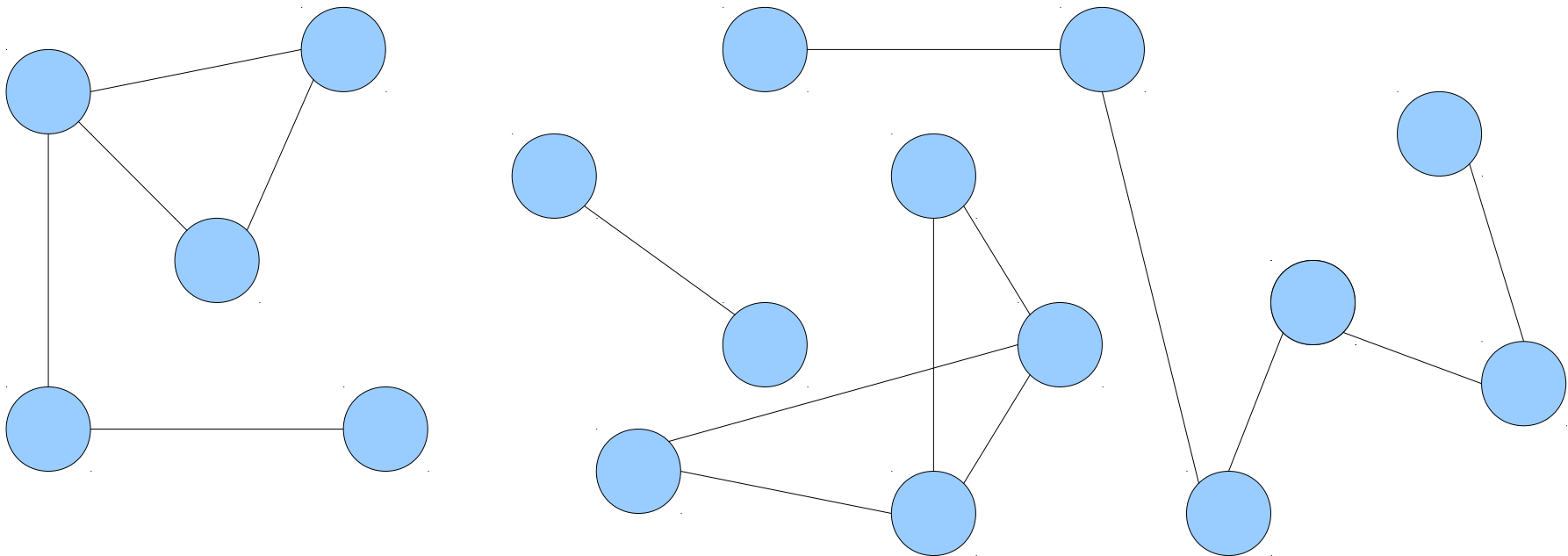
The Connectivity Problem

- The **graph connectivity problem** is the following:
Given an undirected graph G , preprocess the graph so that queries of the form “are nodes u and v connected?”
- Using $\Theta(m + n)$ preprocessing, can preprocess the graph to answer queries in time $O(1)$.



Dynamic Connectivity

- The ***dynamic connectivity problem*** is the following:
Maintain an undirected graph G so that edges may be inserted and deleted and connectivity queries may be answered efficiently.
- This is a *much* harder problem!



Dynamic Connectivity

- Today, we'll focus on the ***incremental dynamic connectivity problem***: maintaining connectivity when edges can only be added, not deleted.
- Has applications to Kruskal's MST algorithm and to many other online connectivity settings.
 - Look up ***percolation theory*** for an example.
- These data structures are also used as building blocks in other algorithms:
 - Speeding up Edmond's blossom algorithm for finding maximum matchings.
 - As a subroutine in Tarjan's offline lowest common ancestors algorithm.
 - Building meldable priority queues out of non-meldable queues.

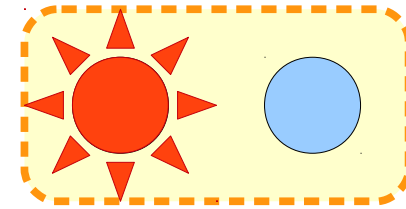
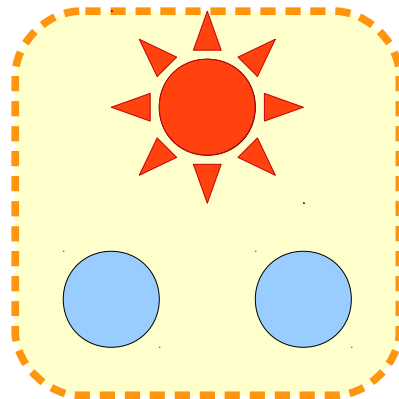
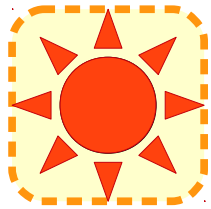
Incremental Connectivity and Partitions

Set Partitions

- The incremental connectivity problem is equivalent to maintaining a partition of a set.
- Initially, each node belongs to its own set.
- As edges are added, the sets at the endpoints become connected and are merged together.
- Querying for connectivity is equivalent to querying for whether two elements belong to the same set.

Representatives

- Given a partition of a set S , we can choose one **representative** from each of the sets in the partition.
- Representatives give a simple proxy for which set an element belongs to: two elements are in the same set in the partition iff their set has the same representative.



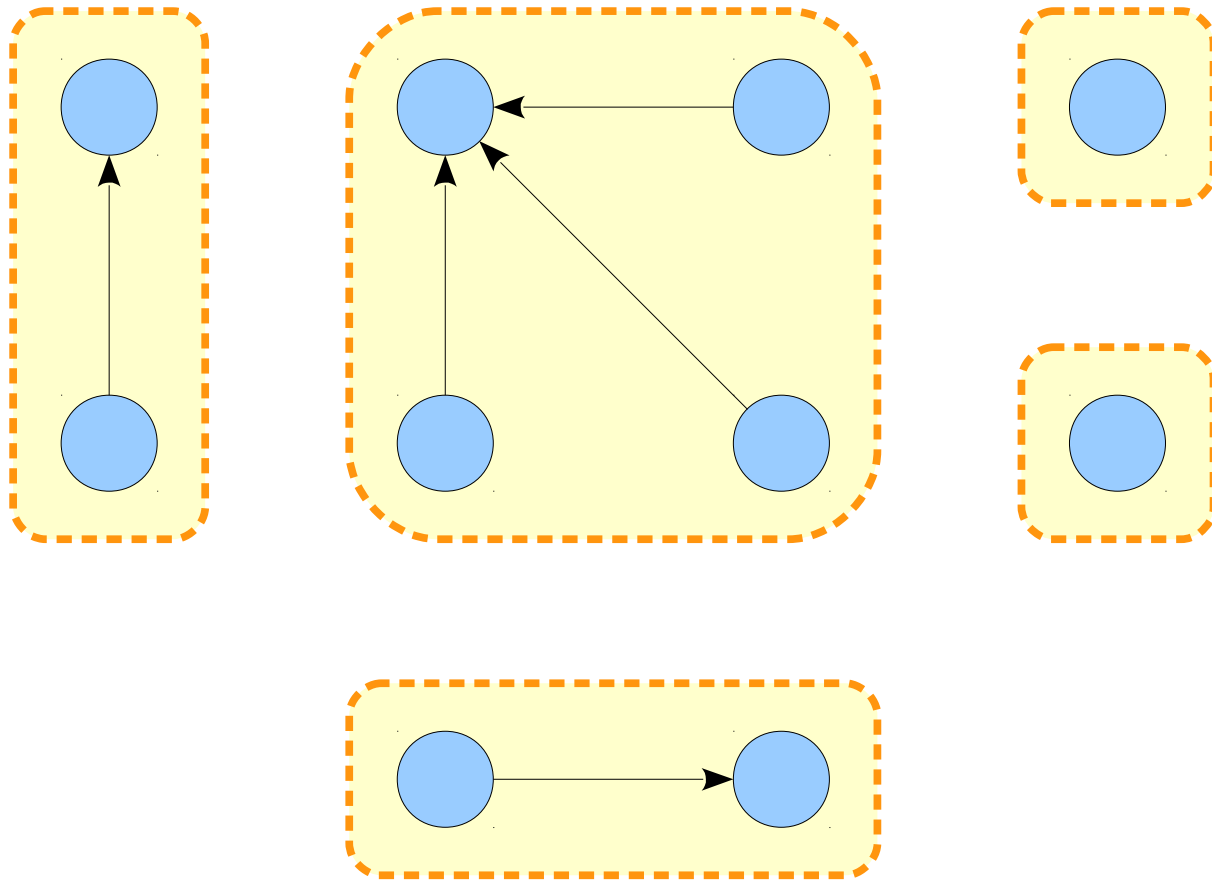
Union-Find Structures

- A ***union-find structure*** is a data structure supporting the following operations:
 - ***find***(x), which returns the representative of the set containing node x , and
 - ***union***(x, y), which merges the sets containing x and y into a single set.
- We'll focus on these sorts of structures as a solution to incremental connectivity.

Data Structure Idea

- **Idea:** Have each element store a pointer directly to its representative.
- To determine if two nodes are in the same set, check if they have the same representative.
- To link two sets together, change all elements of the two sets so they reference a single representative.

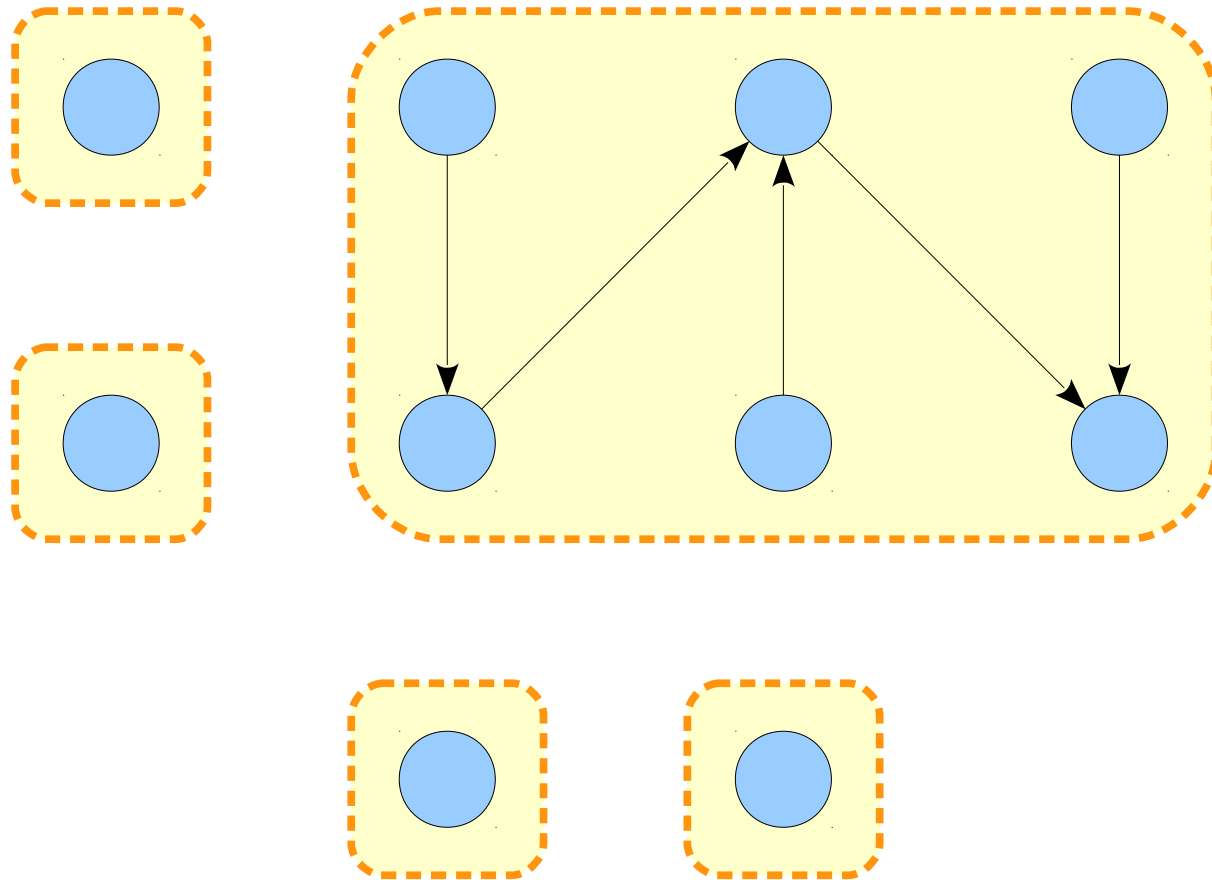
Using Representatives



Using Representatives

- If we update all the representative pointers in a set when doing a **union**, we may spend time $O(n)$ per **union** operation.
 - If you're clever with how you change the pointers, you can make it amortized $O(\log n)$ per operation. Do you see how?
- Can we avoid paying this cost?

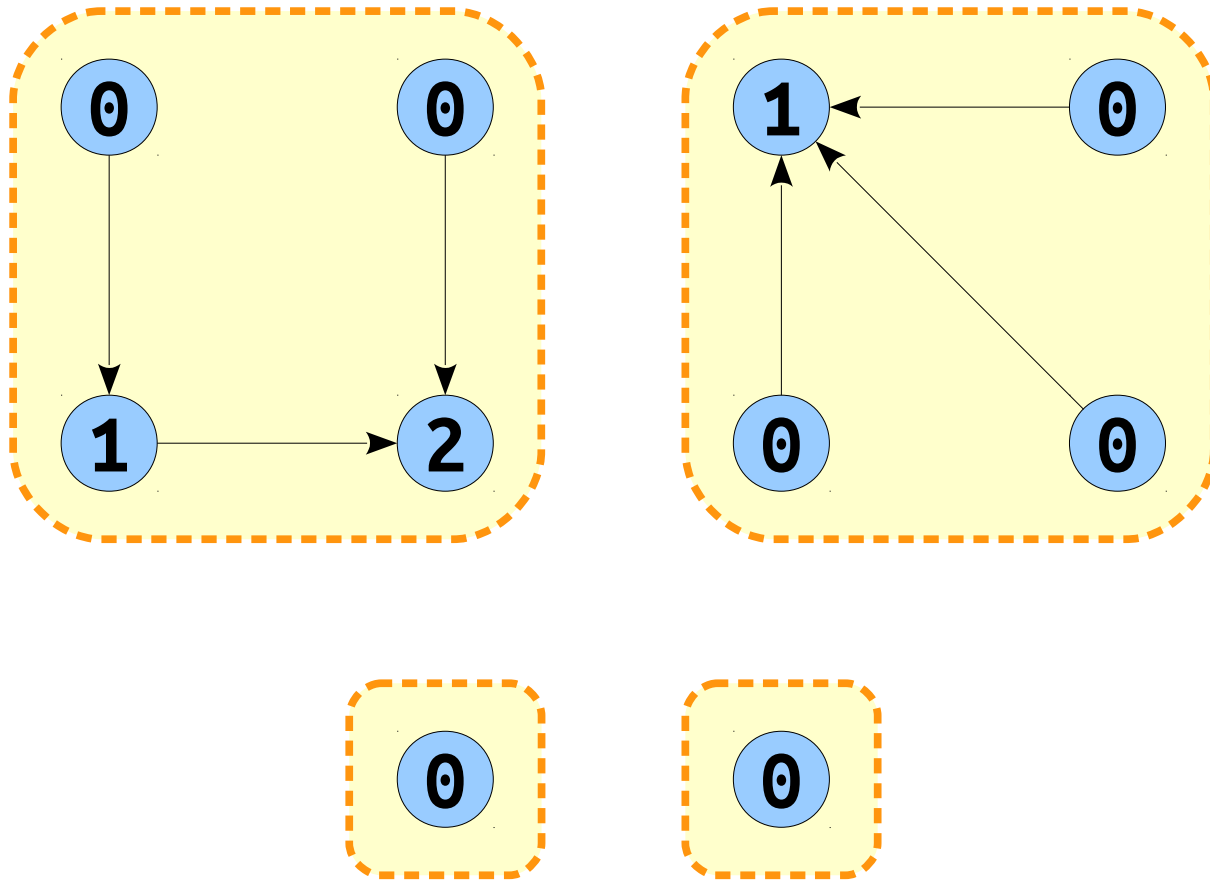
Hierarchical Representatives



Hierarchical Representatives

- In a degenerate case, a hierarchical representative approach will require time $\Theta(n)$ for some *find* operations.
- Therefore, some *union* operations will take time $\Theta(n)$ as well.
- Can we avoid these degenerate cases?

Union by Rank



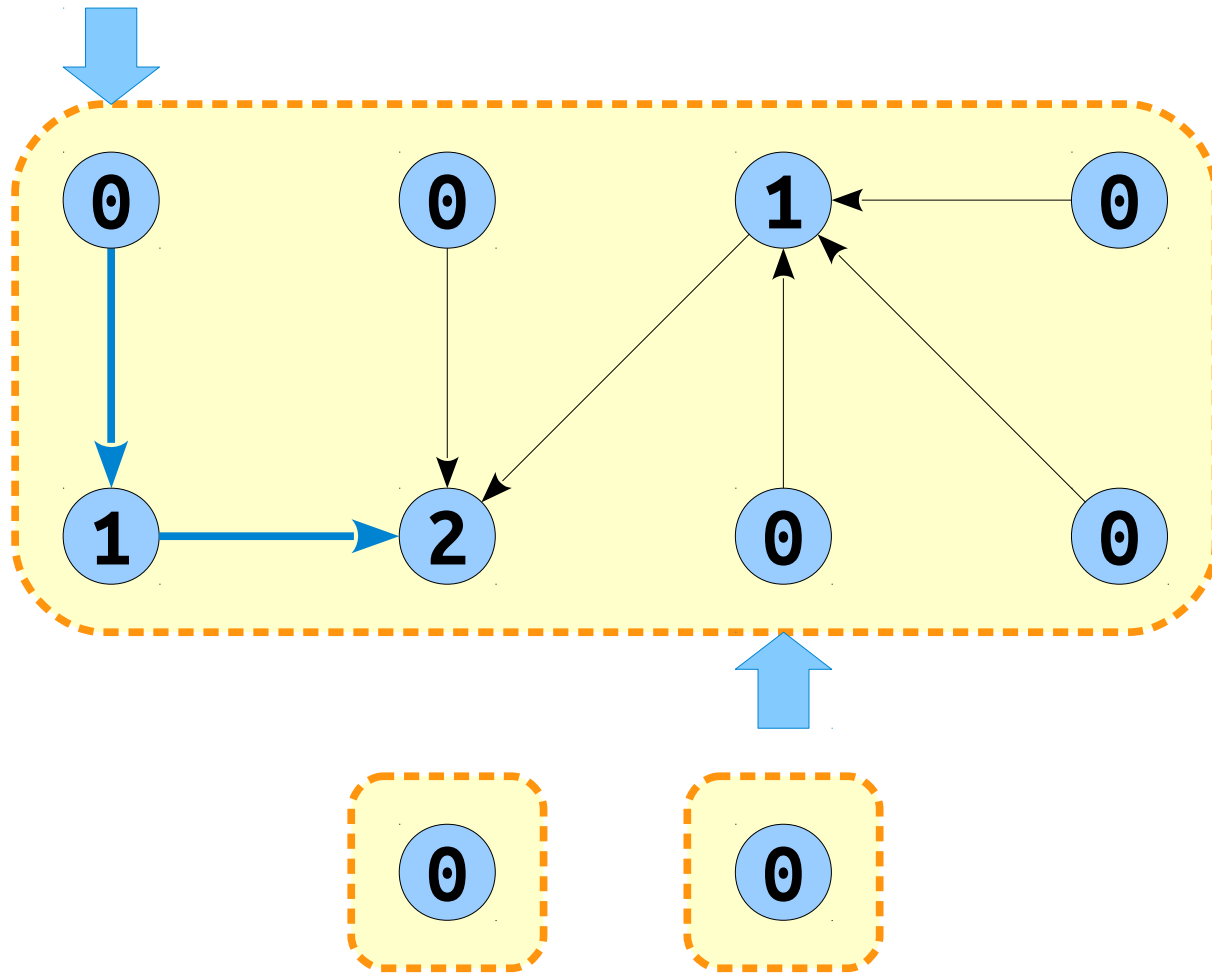
Union by Rank

- Assign to each node a *rank* that is initially zero.
- To link two trees, link the tree of the smaller rank to the tree of the larger rank.
- If both trees have the same rank, link one to the other and increase the rank of the other tree by one.

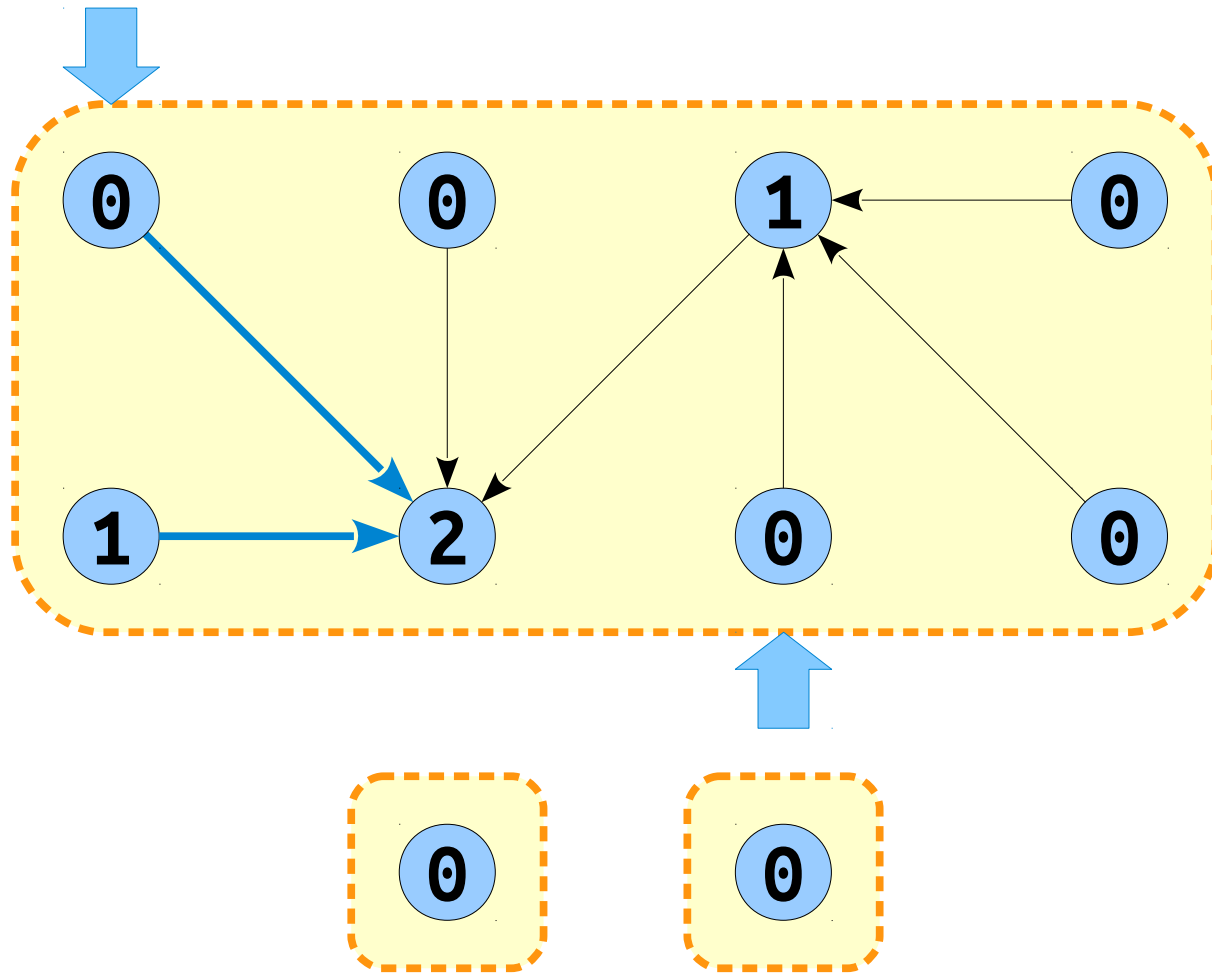
Union by Rank

- **Claim:** The number of nodes in a tree of rank r is at least 2^r .
 - Proof is by induction; intuitively, need to double the size to get to a tree of the next order.
 - Fun fact: the smallest tree with a root of rank r is a binomial tree of order r . Crazy!
- **Claim:** Maximum rank of a node in a graph with n nodes is $O(\log n)$.
- Runtime for **union** and **find** is now $O(\log n)$.
- **Useful fact for later on:** The number of nodes of rank r or higher in a disjoint set forest with n nodes is at most $n / 2^r$.

Path Compression



Path Compression



Path Compression

- ***Path compression*** is an optimization to the standard disjoint-set forest.
- When performing a ***find***, change the parent pointers of each node found along the way to point to the representative.
- Purely using path compression, each operation has amortized cost $O(\log n)$.
- What happens if we combine this with union-by-rank?

The Claim

- **Claim:** The runtime of performing m **union** and **find** operations on an n -node disjoint-set forest using path compression and union-by-rank is $O(n + m\alpha(n))$, where α is an *extremely* slowly-growing function.
- The original proof of this result (which is included in CLRS) is due to Tarjan and uses a complex amortized charging scheme.
- Today, we'll use an aggregate analysis due to Seidel and Sharir based on a technique called **forest-slicing**.

Where We're Going

- First, we're going to define our cost model so we know how to analyze the structure.
- Next, we'll introduce the forest-slicing approach and use it to prove a key lemma.
- Finally, we'll use that lemma to build recurrence relations that analyze the runtime.

Our Cost Model

- The cost of performing a *union* or *find* depends on the length of the paths followed.
- The cost of any one operation is

$$\Theta(1 + \#ptr\text{-changes-made})$$

because each time we visit a node that doesn't immediately point to its representative, we change where it points.

- Therefore, the cost of m operations is

$$\Theta(m + \#ptr\text{-changes-made})$$

- We will analyze the number of pointers changed across the life of the data structure to bound the overall cost.

Some Accounting Tricks

- To perform a **union** operation, we need to first perform two **finds**.
- After that, only $O(1)$ time is required to perform the **union** operation.
- Therefore, we can replace each **union**(x, y) with three operations:
 - A call to **find**(x).
 - A call to **find**(y).
 - A linking step between the nodes found this way.
- Going forward, we will assume that each **union** operation will take worst-case time $O(1)$.

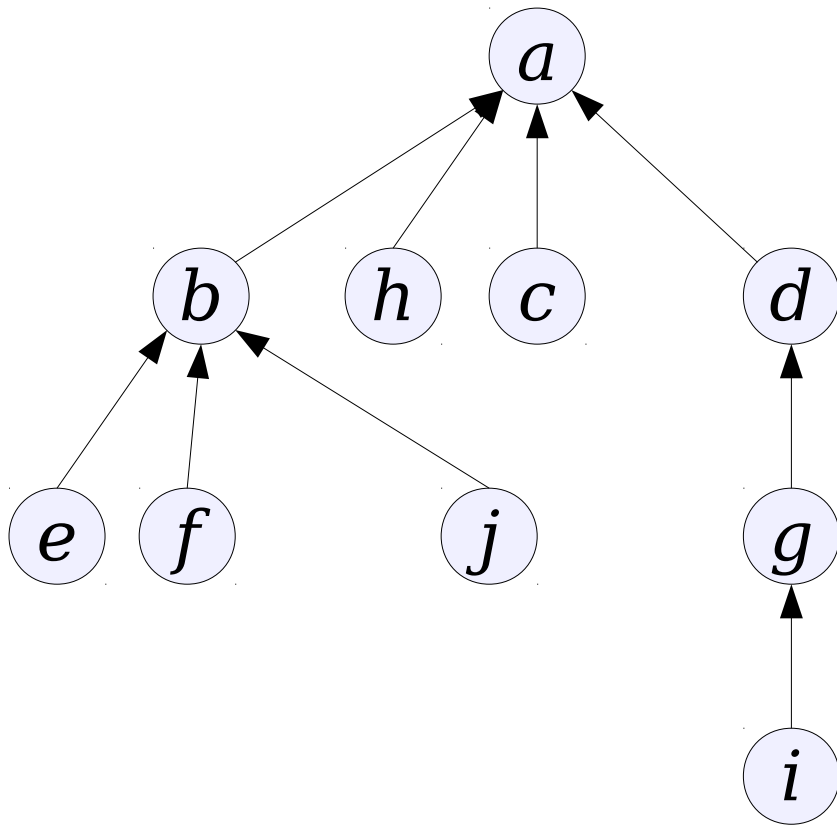
A Slight Simplification

- Currently, ***find***(x) compresses from x up to its ancestor.
- For mathematical simplicity, we'll introduce an operation ***compress***(x, y) that compresses from x upward to y , assuming that y is an ancestor of x .
- Our analysis will then try to bound the total cost of the ***compress*** operations.

Removing the Interleaving

- We will run into some trouble in our analysis because *unions* and *compresses* can be interleaved.
- To address this, we will will remove the interleaving by pretending that all *unions* come before all *compresses*.
- This does not change the overall work being done.

Removing the Interleaving



compress(j, b)
union(b, a)
compress(h, a)

union(b, a)
compress(j, b)
compress(h, a)

$f \rightarrow b$
 $h \rightarrow b$
 $j \rightarrow b$
 $b \rightarrow a$
 $h \rightarrow a$

$b \rightarrow a$
 $f \rightarrow b$
 $h \rightarrow b$
 $j \rightarrow b$
 $h \rightarrow a$

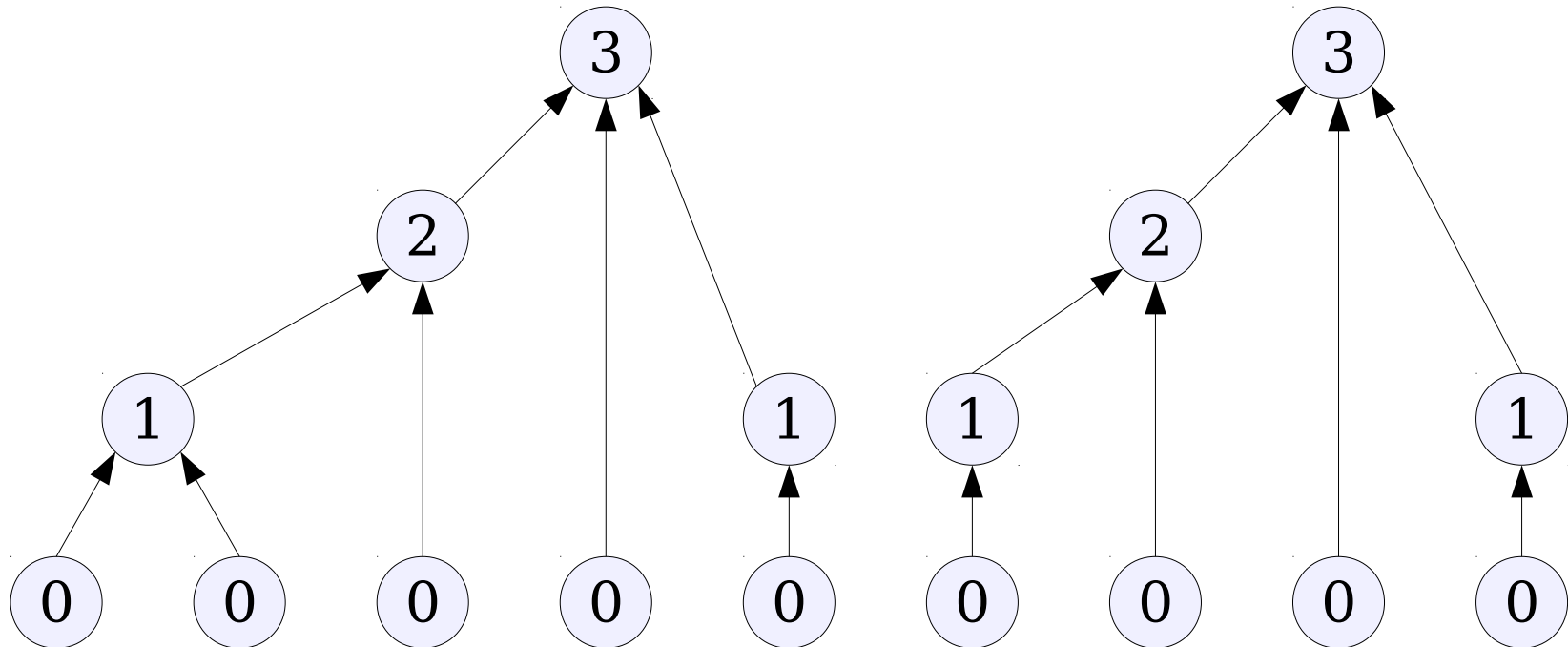
Recap: The Setup

- Transform any sequence of *unions* and *finds* as follows:
 - Replace all *union* operations with two *finds* and a *union* on the ancestors.
 - Replace each *find* operation with a *compress* operation indicating its start and end nodes.
 - Move all *union* operations to the front.
- Since all *unions* are at the front, we build the entire forest before we begin compressing.
- Can analyze *compress* assuming the forest has already been created for us.

A Quick Initial Analysis

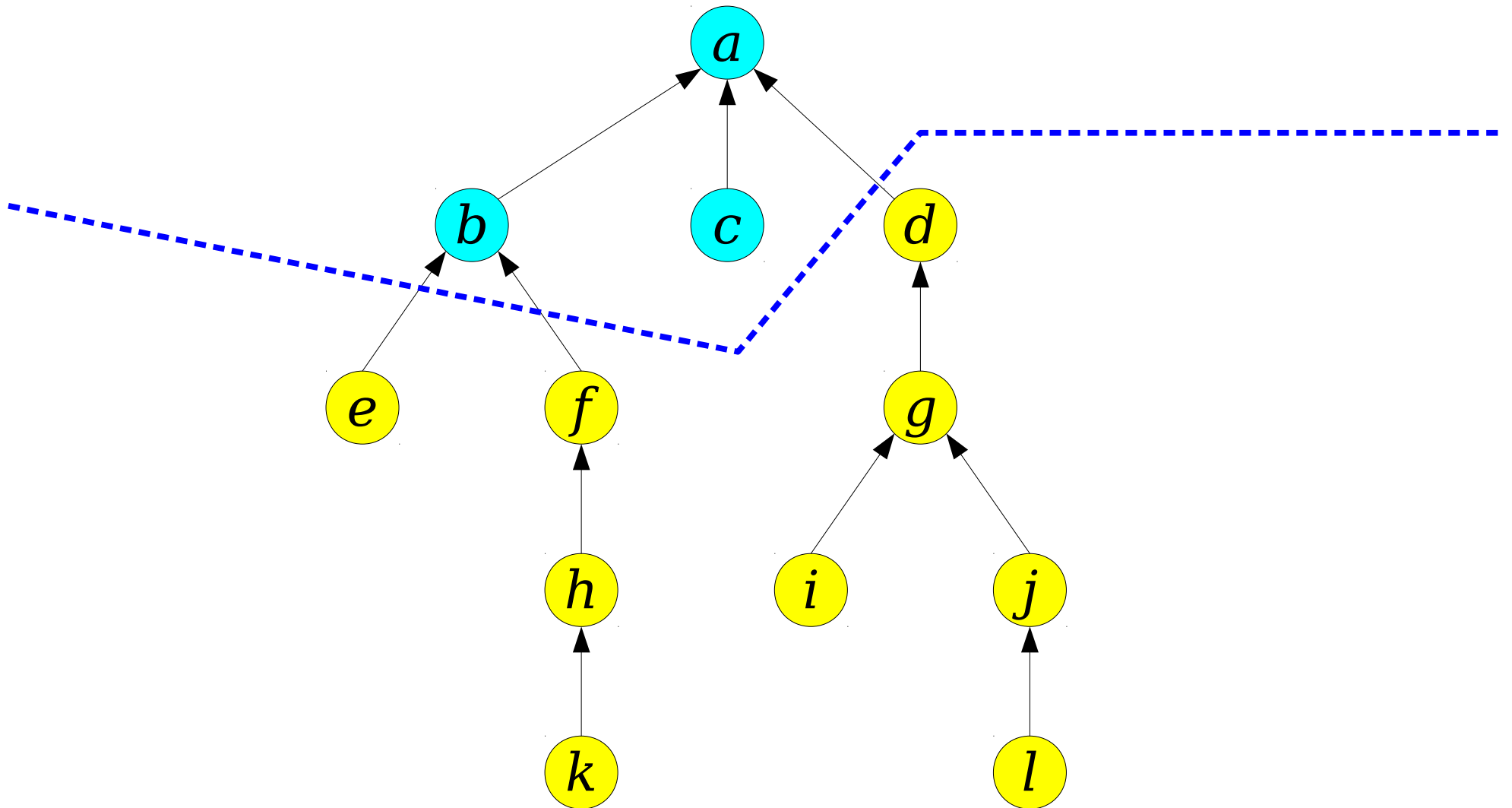
An Initial Analysis

- **Lemma:** Any series of m **compress** operation on a forest \mathcal{F} with n nodes and maximum rank r makes at most nr pointer changes.
- **Proof:** Every time a node's representative change, the rank of that representative increases. The maximum number of times this can happen per node is r , giving an upper bound of nr . ■



The Forest-Slicing Approach

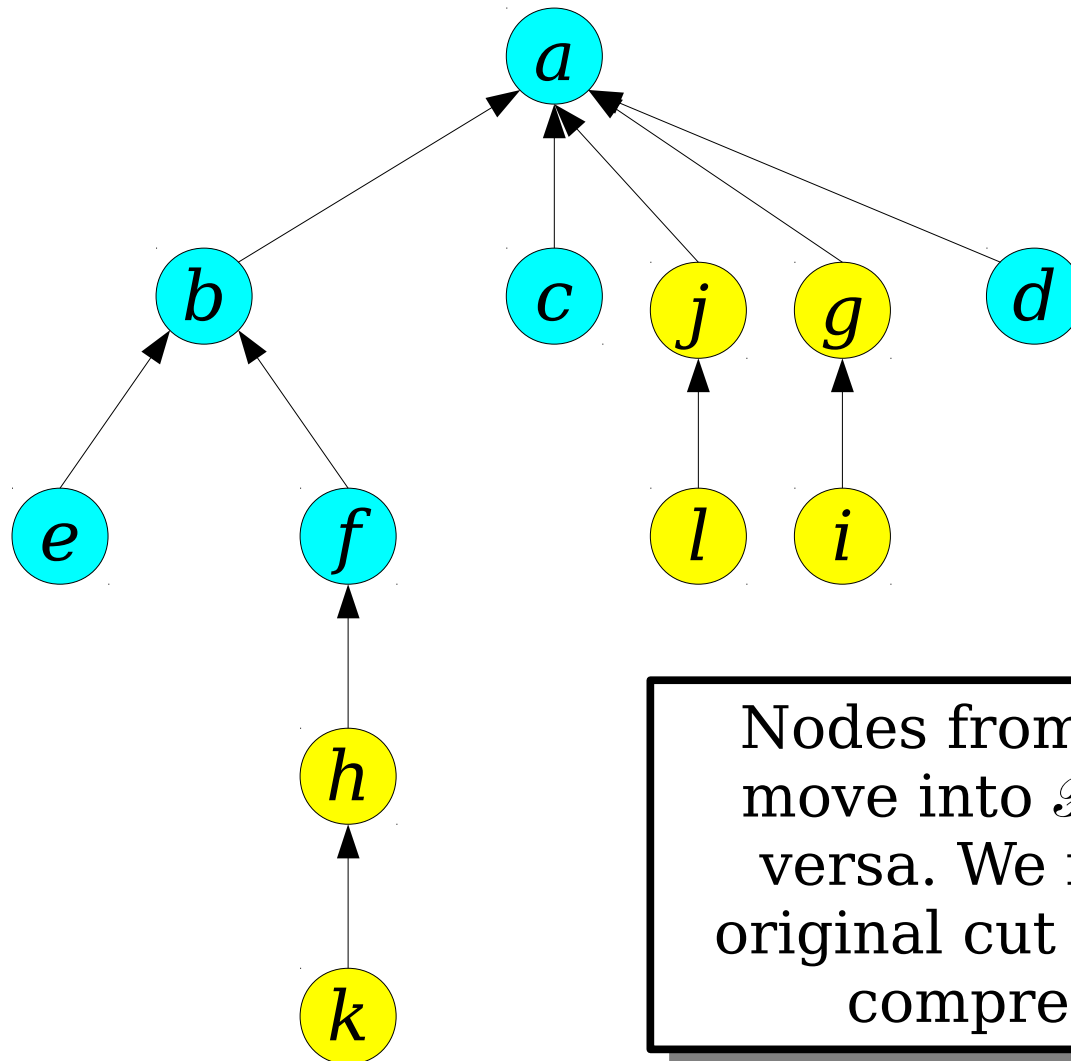
Forest-Slicing



Forest-Slicing

- Let \mathcal{F} be a disjoint-set forest.
- Consider splitting \mathcal{F} into two forests \mathcal{F}_+ and \mathcal{F}_- with the following properties:
 - \mathcal{F}_+ is **upward-closed**: if $x \in \mathcal{F}_+$, then any ancestor of x is also in \mathcal{F}_+ .
 - \mathcal{F}_- is **downward-closed**: if $x \in \mathcal{F}_-$, then any descendant of x is also in \mathcal{F}_- .
- We'll call \mathcal{F}_+ the **top forest** and \mathcal{F}_- the **bottom forest**.

Forest-Slicing

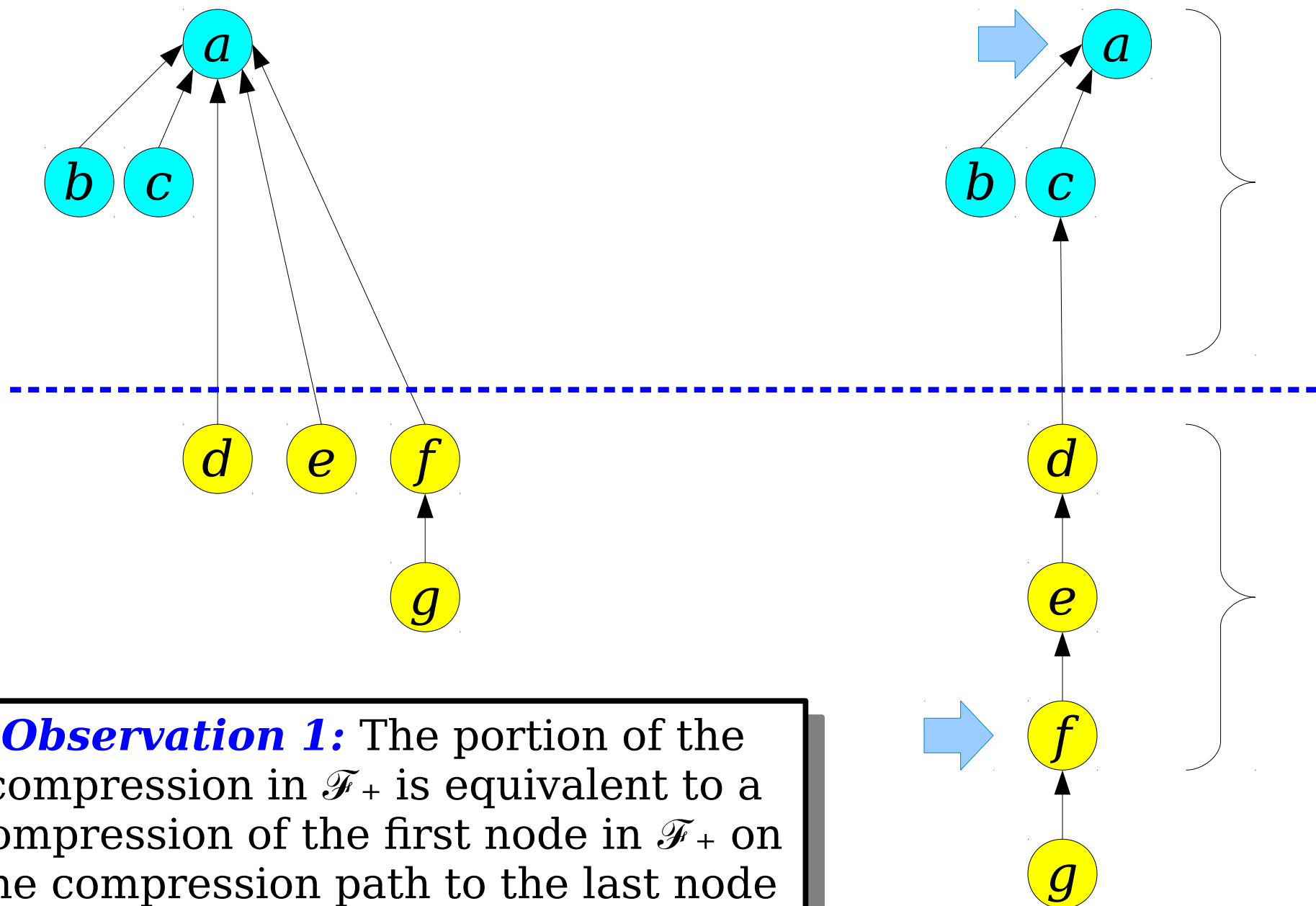


Nodes from \mathcal{F}_- never move into \mathcal{F}_+ or vice-versa. We retain the original cut after doing compressions.

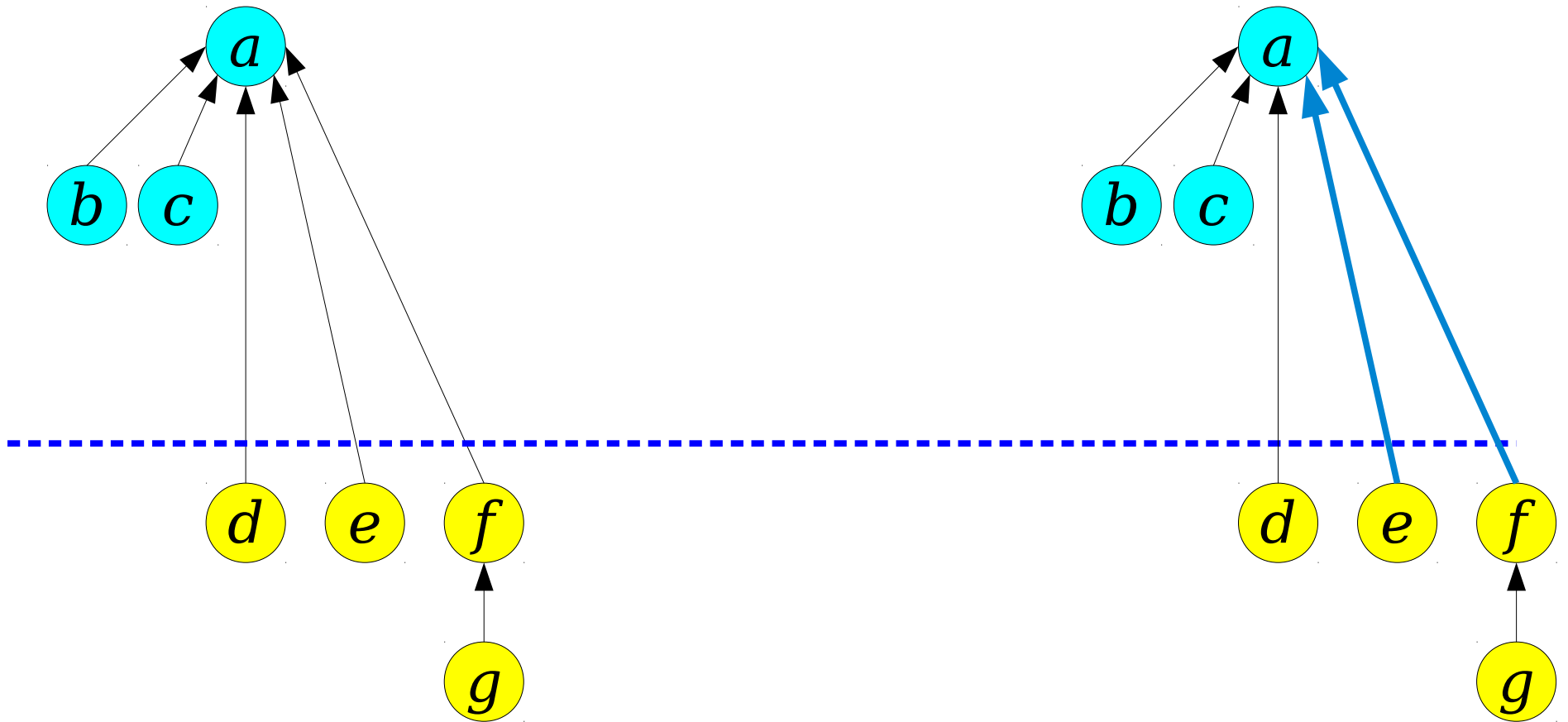
Why Slice Forests?

Forest-Slicing

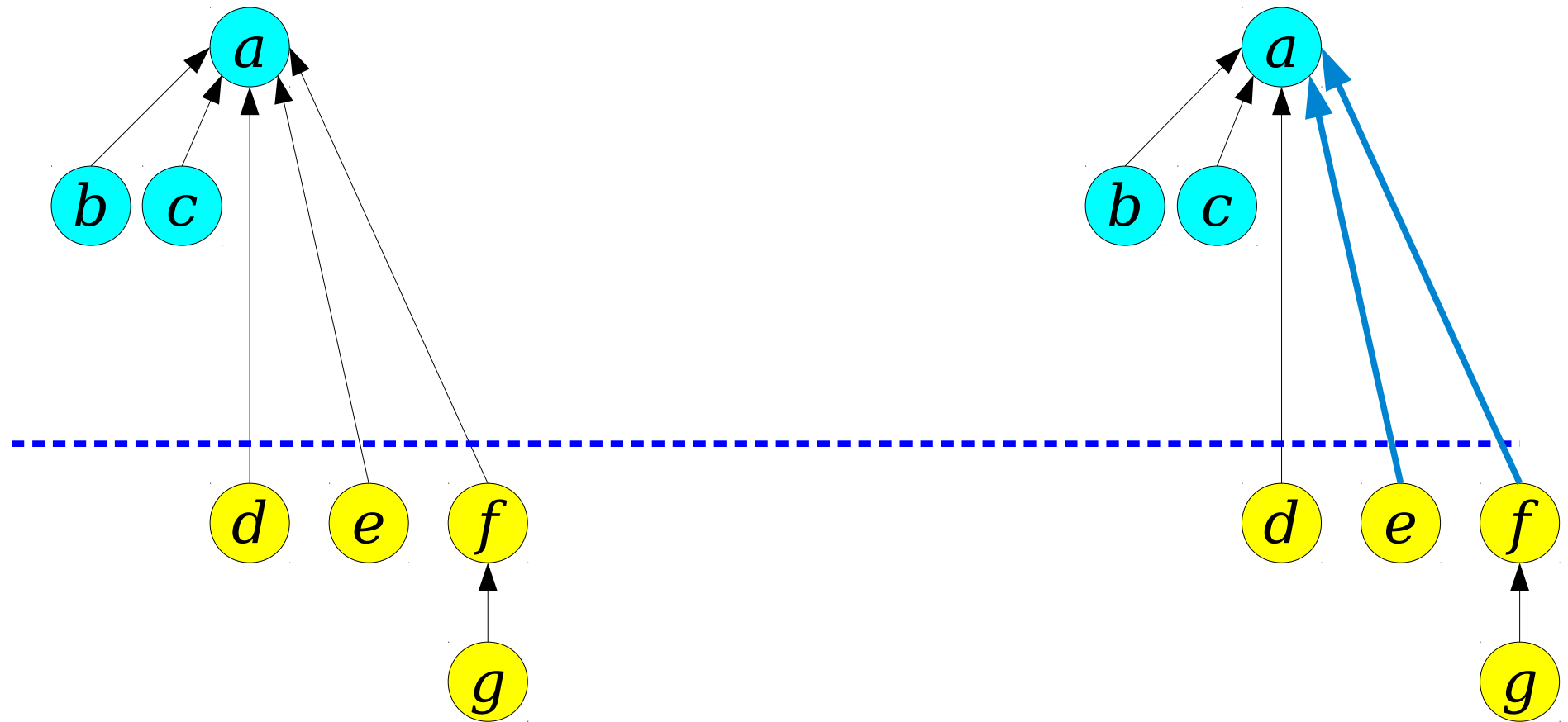
- **Key insight:** Each **compress** operation is either
 - purely in \mathcal{F}_+ ,
 - purely in \mathcal{F}_- , or
 - crosses from \mathcal{F}_- into \mathcal{F}_+ .
- If we can bound the cost of **compress** operations that cross from \mathcal{F}_- to \mathcal{F}_+ , we can try to set up a recurrence relation to analyze the cost of those **compresses**.



Observation 1: The portion of the compression in \mathcal{F}_+ is equivalent to a compression of the first node in \mathcal{F}_+ on the compression path to the last node in \mathcal{F}_+ on the compression path.



Observation 2: The effect of the compression on \mathcal{F}_- is *not* the same as the effect of compressing from the first node in \mathcal{F}_- to the last node in \mathcal{F}_- .



Observation 3: The cost of the compress in \mathcal{F}_- is the number of nodes in \mathcal{F}_- that got a parent in \mathcal{F}_+ , plus (possibly) one more for the topmost node in \mathcal{F}_- on the compression path.

The Cost of Crossing Compressions

- Suppose we do m compressions, of which m_+ of them cross from \mathcal{F}_- into \mathcal{F}_+ .
- We can upper bound the cost of these compressions as the sum of the following:
 - the cost of all the tops of those compressions, which occur purely in \mathcal{F}_+ ;
 - the number of nodes in \mathcal{F}_- , since each node in \mathcal{F}_- gets a parent in \mathcal{F}_+ for the first time at most once; and
 - m_+ , since each compression may change the pointer of the topmost node on the path in \mathcal{F}_- .

Theorem: Let \mathcal{F} be a disjoint-set forest and let \mathcal{F}_+ and \mathcal{F}_- be a partition of \mathcal{F} into top and bottom forests.

Then for any series of m compressions C , there exist two sequences of compressions

- C_+ , a series of m_+ compressions purely in \mathcal{F}_+ ; and
- C_- , a series of m_- compressions purely in \mathcal{F}_- ,

such that

- $m_+ + m_- = m$
- $\text{cost}(C) \leq \text{cost}(C_+) + \text{cost}(C_-) + n + m_+$

Compressions that appear purely in \mathcal{F}_+ or purely in \mathcal{F}_- , plus the tops of crossing compressions.

Nodes in \mathcal{F}_- getting their first parent in \mathcal{F}_+

Nodes in \mathcal{F}_- having their parent in \mathcal{F}_+ change.

Time-Out for Announcements!

The midterm is tonight
from 7PM - 10PM
in room 320-105.

Good luck!

Back to CS166!

The Main Analysis

Where We Are

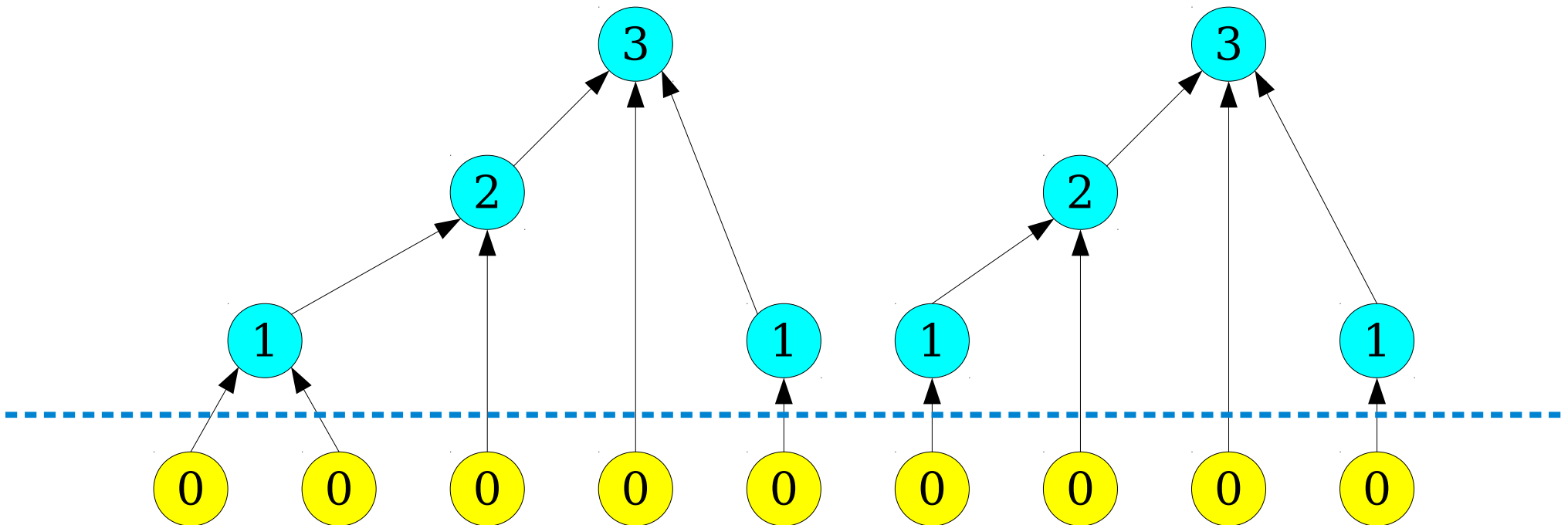
- We now have a sort of recurrence relation for evaluating the runtime of a series C of m **compresses** on an n -node forest \mathcal{F} sliced into \mathcal{F}_+ and \mathcal{F}_- :

$$\text{cost}(C) \leq \text{cost}(C_+) + \text{cost}(C_-) + n + m_+$$

- This recurrence relation assumes that we already know how we've sliced \mathcal{F} into \mathcal{F}_+ and \mathcal{F}_- .
- To complete the analysis, we're going to need to precisely quantify what happens if we slice the forest in a number of different ways.

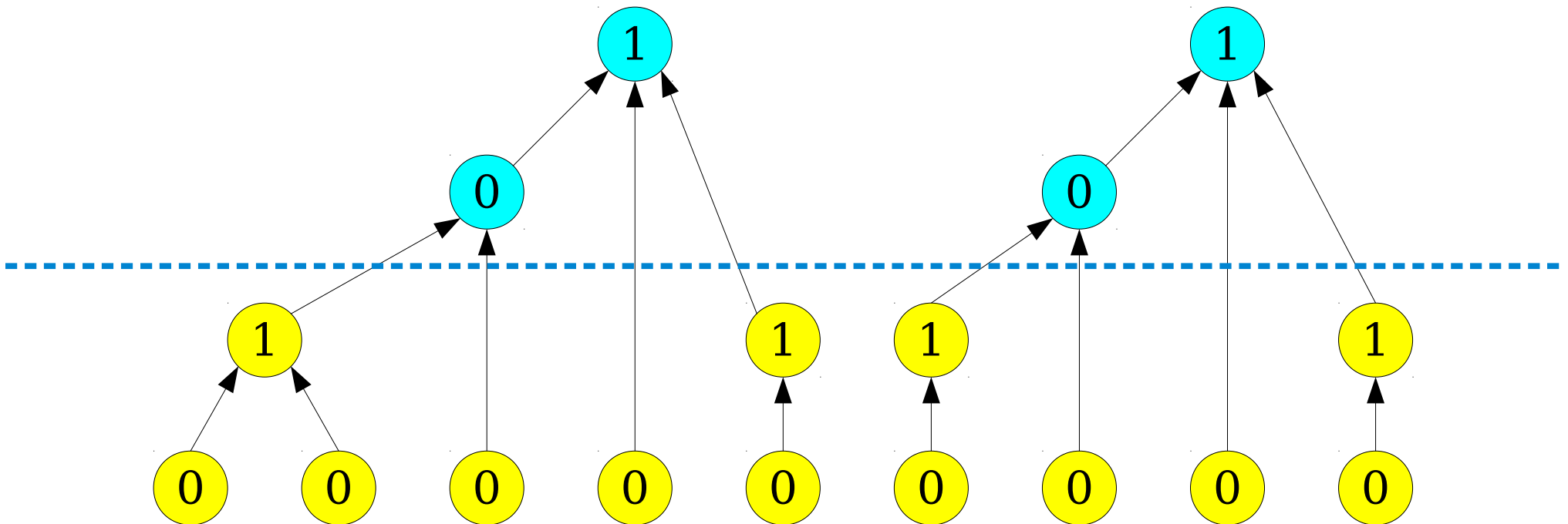
Natural Slices

- One “natural” way to slice a forest \mathcal{F} into \mathcal{F}_+ and \mathcal{F}_- is to pick some threshold rank. We then choose \mathcal{F}_+ to be all the nodes whose rank is above the threshold and \mathcal{F}_- to be all the other nodes.



Natural Slices

- If our initial forest has maximum rank r and we slice the forest at rank r' , the bottom forest has maximum rank r' and the top forest is (essentially) a forest of rank $r - r'$.



Slicing our Forest

- Imagine that we have our forest \mathcal{F} of maximum rank r .
- Suppose we cut slice the forest into \mathcal{F}_+ and \mathcal{F}_- at some rank r' .
- We know that

$$\text{cost}(C) \leq \text{cost}(C_+) + \text{cost}(C_-) + n + m_+.$$

- Let's investigate $\text{cost}(C_+)$ and $\text{cost}(C_-)$ independently.

The Top Forest

- Let's begin by thinking about $\text{cost}(C_+)$, the cost of compresses in the top forest \mathcal{F}_+ .
- **Recall:** \mathcal{F}_+ consists of all nodes of rank r or higher.
- Intuitively, we'd expect there to not be “too many” nodes in the top forest, since it's exponentially harder to get nodes of progressively harder orders.
- Using our lemma from before, we know that there can be at most $n / 2^r$ nodes in \mathcal{F}_+ .
- Therefore, using our (weak) bound from before, we see that

$$\text{cost}(C_+) \leq nr / 2^r.$$

Slicing our Forest

- Imagine that we have our forest \mathcal{F} of maximum rank r .
- Suppose we cut slice the forest into \mathcal{F}_+ and \mathcal{F}_- at some rank r' .
- We know that

$$\text{cost}(C) \leq \text{cost}(C_+) + \text{cost}(C_-) + n + m_+.$$

- Therefore

$$\text{cost}(C) \leq nr / 2^{r'} + \text{cost}(C_-) + n + m_+.$$

- Let's now go investigate $\text{cost}(C_-)$.

Improving our Recurrence

$$\text{cost}(C) \leq nr / 2^{r'} + \text{cost}(C_-) + n + m_+.$$

- Notice that $\text{cost}(C)$ is the cost of
 - doing m **compresses**,
 - in an n -node forest, with
 - maximum rank r .
- We now have $\text{cost}(C_-)$, which is the cost of
 - doing m_- **compresses**,
 - in a forest with at most n nodes, with
 - maximum rank r' .
- Let's make these dependencies more explicit.

Improving our Recurrence

$$\text{cost}(C) \leq nr / 2^{r'} + \text{cost}(C_-) + n + m_+.$$

- Define $T(m, n, r)$ to be the cost of
 - performing m **compress** operations,
 - in a forest of at most n nodes, where
 - the maximum rank is r .

- The above recurrence can be rewritten as

$$T(m, n, r) \leq T(m_-, n, r') + nr / 2^{r'} + n + m_+$$

- Now, we “just” need to solve this recurrence.
Don't worry... it's not too bad!

Finalizing our Recurrence

$$T(m, n, r) \leq T(m_-, n, r') + nr / 2^{r'} + n + m_+$$

- The above recurrence is dependent on having a choice of r' based on our choice of r .
- If we make r' too large, then the recurrence relation takes too long to bottom out and we'll expect a higher runtime.
- If we make r' too small, the $nr / 2^{r'}$ term will be too large and our analysis won't be tight.
- How do we balance these terms out?

Finalizing our Recurrence

$$T(m, n, r) \leq T(m_-, n, r') + nr / 2^{r'} + n + m_+$$

- **Idea:** Choose $r' = \lg r$. Then

$$T(m, n, r) \leq T(m_-, n, \lg r) + 2n + m_+.$$

- Imagine that this recurrence expands out L times before it bottoms out. Think about what happens:
 - The $2n$ term gets summed in L times.
 - The m_+ term - the number of compresses in the top forest - sums up to at most m across all compressions.
- Overall, we get $T(m, n, r) \leq 2nL + m$.

Iterated Logarithms

- We now have

$$T(m, n, r) \leq 2nL + m.$$

- The quantity L represents the number of layers in the recurrence, and at each step we have r dropping to $\lg r$.
- The ***iterated logarithm***, denoted $\mathbf{lg}^* n$, is the number of times we can apply \lg to n before it drops to some constant (say, 2). Therefore:

$$T(m, n, r) \leq 2n \lg^* r + m.$$

- And since the maximum rank is at most $\lg n$, we see that the cost of performing m operations on an n -node forest is $\mathbf{O}(n \lg^* n + m)$.

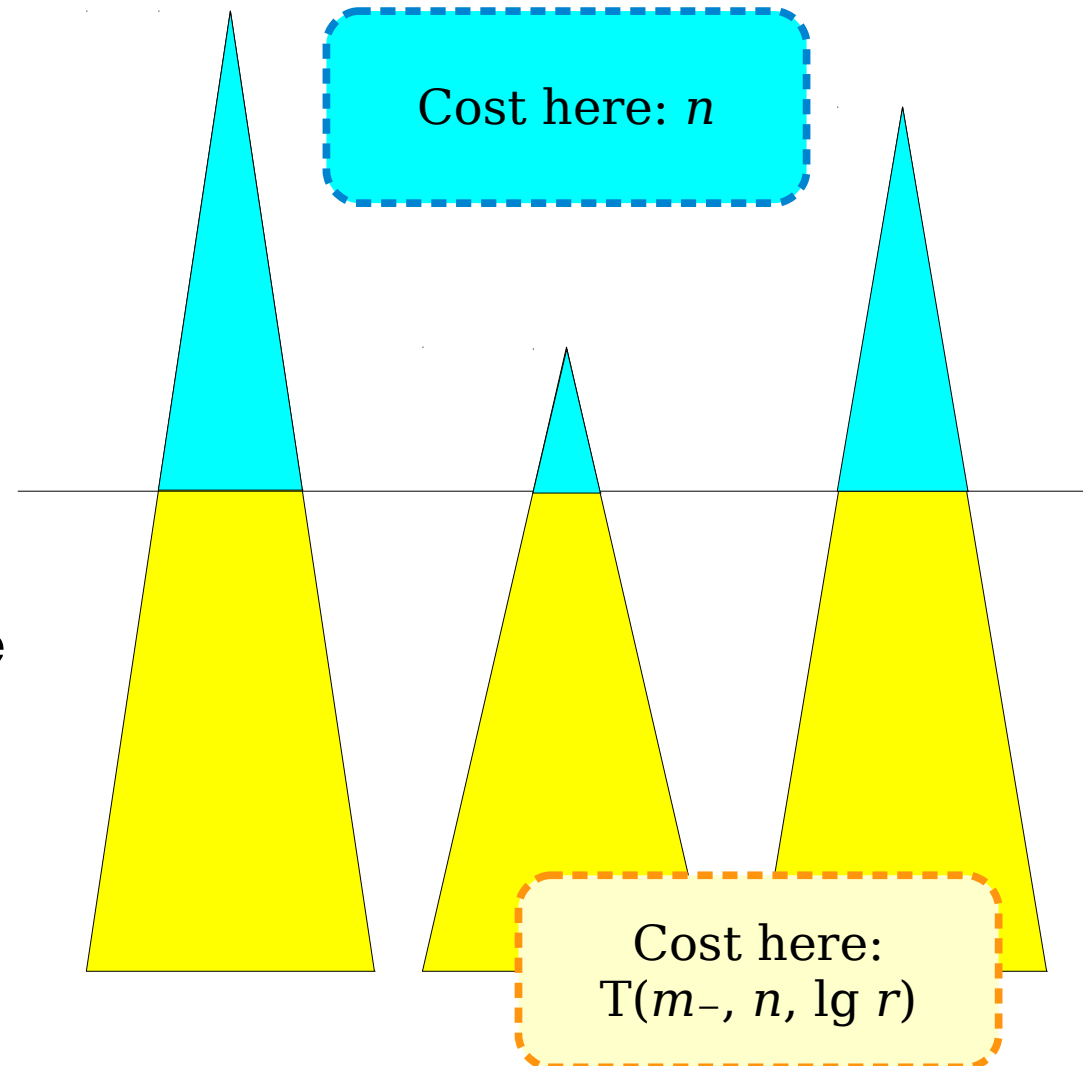
Iterated Logarithms

- The function $\lg n$ is the inverse of the function 2^n ; that is, $2 \times 2 \times \dots \times 2$, n times.
- The **tetration** operation, denoted ${}^n 2$, is given by ${}^n 2 = 2^{2^{\dots^2}}$, with n copies of 2 in the tower of exponents. It grows *extremely* quickly!
- The function $\lg^* n$ is the inverse of tetration. It grows *extremely* slowly!
- **Useful fact:** $\lg^* n \leq 5$ for any n less than or equal to the number of atoms in the universe.

Our Strategy

- Let's recap, how we got here.
- We begin with a forest \mathcal{F} of maximum rank r .
- We sliced \mathcal{F} at rank $\lg r$.
- We (directly) obtained a weak bound on the cost of the compressions in the (small) forest \mathcal{F}_+ .
- We recursively obtained a (good) bound on the cost of the compressions in the (larger) forest \mathcal{F}_- .
- We solved the recurrence to get the bound

$$\mathbf{T(m, n, r) \leq 2n \lg^* r + m.}$$

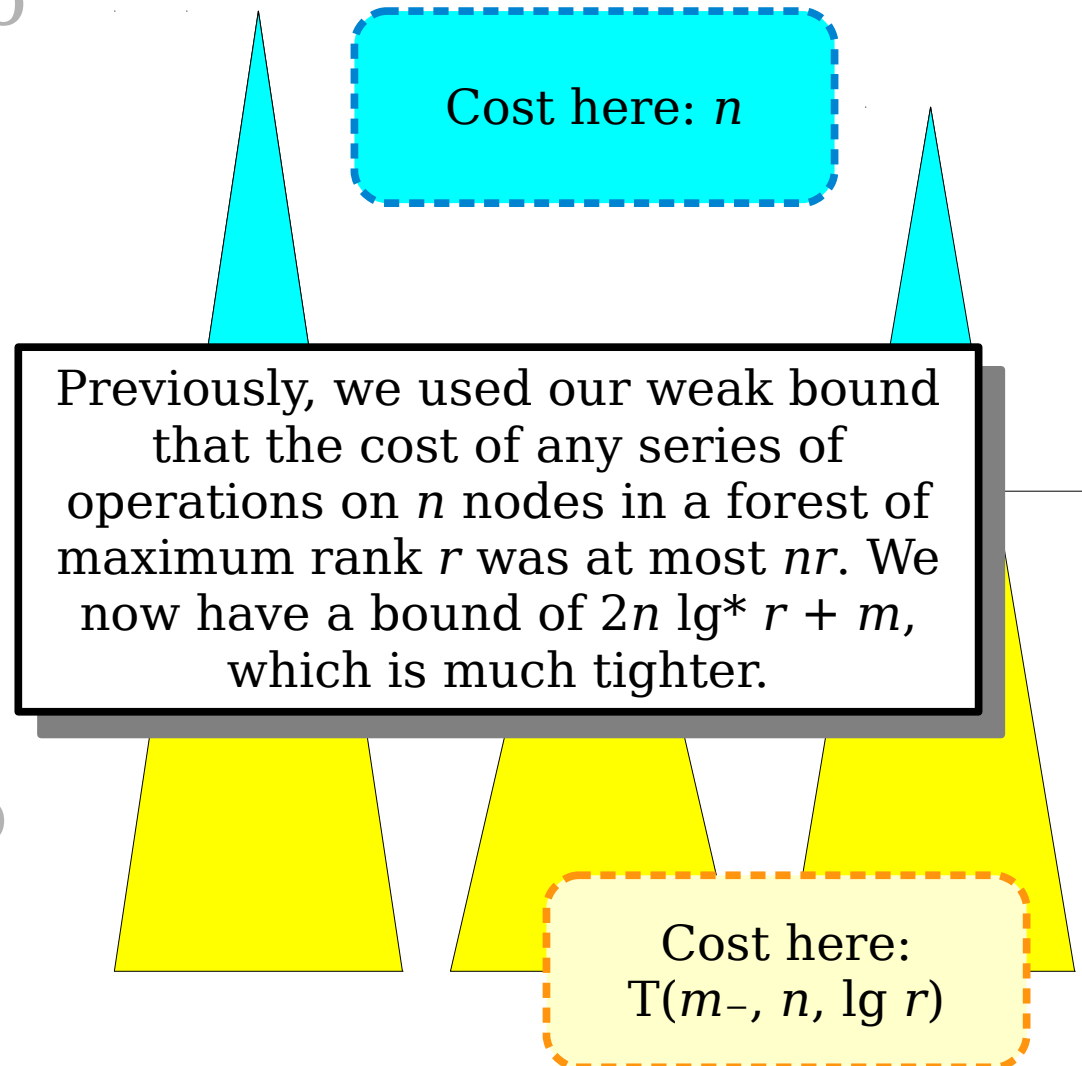


Our Strategy

What could we do to tighten the runtime bound?

- **Option 1:** Tighten the bound on the cost of the top forest.

Option 2: Slice the forest even lower to make the recursion tree shorter.

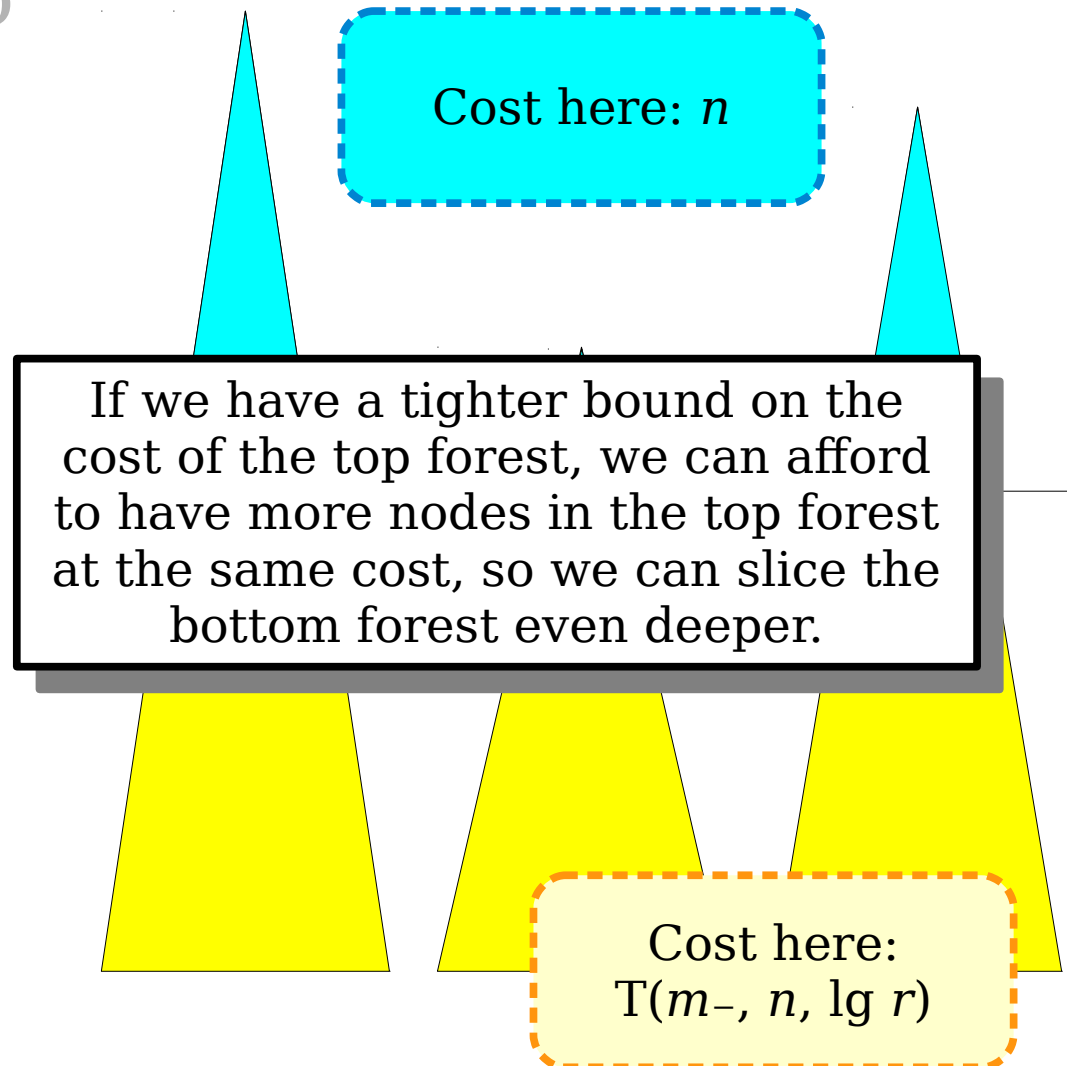


Our Strategy

What could we do to tighten the runtime bound?

Option 1: Tighten the bound on the cost of the top forest.

- **Option 2:** Slice the forest even lower to make the recursion tree shorter.



Slicing our Forest, Again

- Imagine that we have a forest \mathcal{F} of maximum rank r .
- Suppose we cut slice the forest into \mathcal{F}_+ and \mathcal{F}_- at some rank r' .

- We know that

$$\text{cost}(C) \leq \text{cost}(C_+) + \text{cost}(C_-) + n + m_+.$$

- Therefore

$$T(m, n, r) \leq \text{cost}(C_+) + T(m, n, r') + n + m_+.$$

- Let's investigate $\text{cost}(C_+)$ using our previous analysis.

The Top Forest

- **Lemma:** In an n -node forest \mathcal{F} of maximum rank r , if we split \mathcal{F} into \mathcal{F}_+ and \mathcal{F}_- by cutting the forest at rank r' , then $\text{cost}(C_+) \leq 2n \lg^* r / 2^{r'} + m_+$.
- **Proof:** There are $n / 2^{r'}$ nodes in this forest and the maximum rank is at most r . The cost of performing m_+ compress operations here is therefore

$$2(n / 2^{r'}) \lg^* r + m_+.$$

- **Observation:** Our previous bound was

$$rn / 2^{r'}.$$

We previously set $r' = \lg r$ because that was as low as we could go without $\text{cost}(C_+)$ being too high. With our new bound, we can afford to make r' much lower.

Our Recurrence

- We had

$$T(m, n, r) \leq \text{cost}(C_+) + T(m_-, n, r') + n + m_+.$$

- So we now have

$$T(m, n, r) \leq T(m_-, n, r') + 2n \lg^* r / 2^{r'} + n + 2m_+.$$

- Previously, we picked $r' = \lg r$ and ended up with a bound in terms of $\lg^* r$.
- Now, we pick $r' = \lg^* r$. Then we have

$$T(m, n, r) \leq T(m_-, n, \lg^* r) + 2n + 2m_+.$$

- Using a similar analysis as before, if L is the number of layers in the recurrence, this solves to

$$T(m, n, r) \leq \mathbf{2nL + 2m}.$$

Iterated Iteration

- We have

$$T(m, n, r) \leq 2nL + 2m,$$

where L is the number of layers in the iteration.

- At each step, we shrink r to $\lg^* r$. The maximum number of times we can do this is denoted $\lg^{**} r$, so we have

$$T(m, n, r) \leq 2n \lg^{**} r + 2m.$$

- So the cost of any m operations is $O(n \lg^{**} n + m)$.

Iterated Iterated Logarithms

- The ***pentation*** operation is next in the family of fast-growing functions.
- Just as tetration is iterated exponentiation, pentation is iterated tetration, so 2 pentated to the n th power, denoted $_n 2$, is

$$\left(2^{2^{2^{\dots^2}}}\right) \dots \left(2^{2^{2^{\dots^2}}}\right)$$

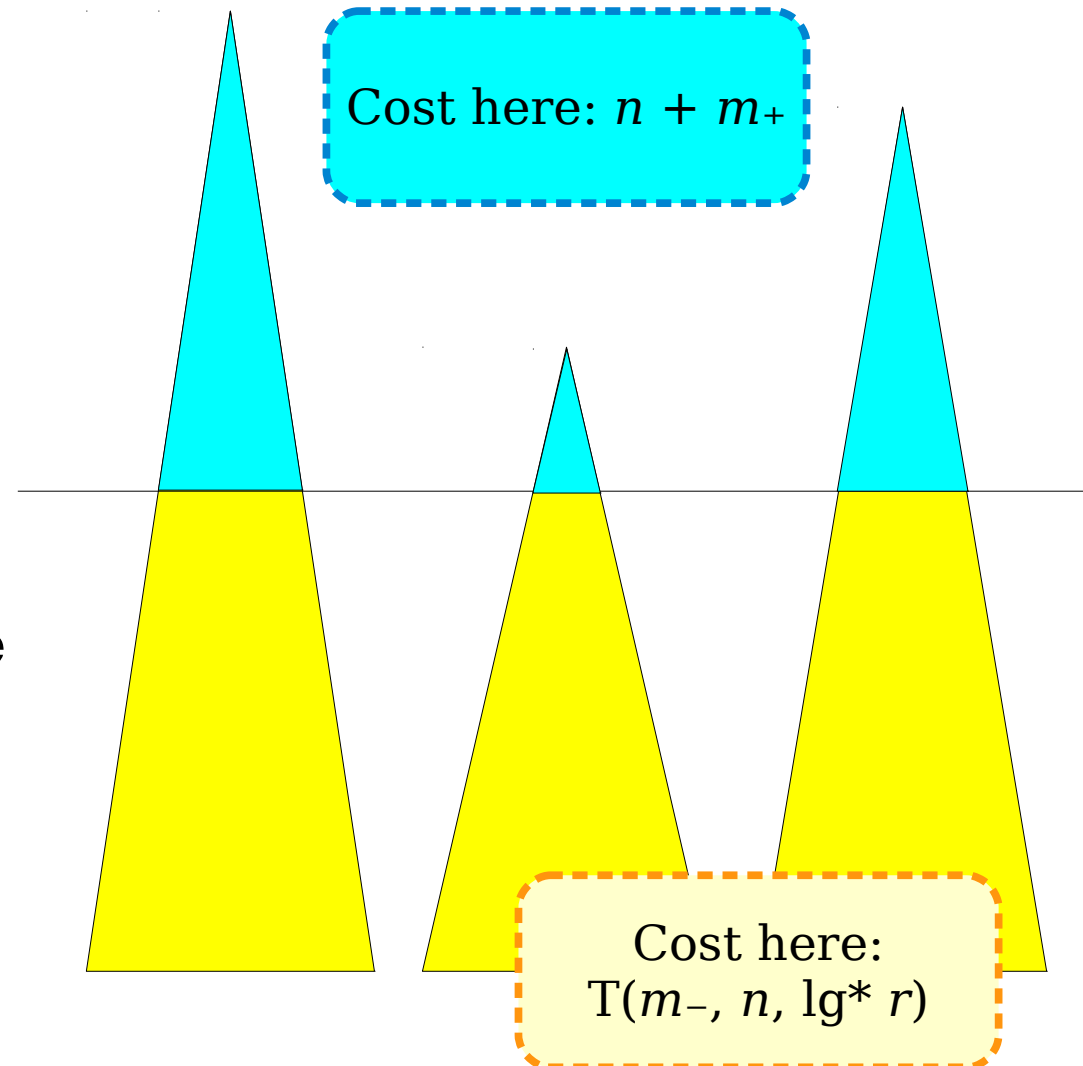
where there are $_n 2$ copies of the exponential towers.

- The function $\text{lg}^{**} n$ is the inverse of pentation. It grows *unbelievably* slowly!

Our Strategy

- Let's recap, how we got here.
- We begin with a forest \mathcal{F} of maximum rank r .
- We sliced \mathcal{F} at rank $\lg^* r$.
- We (directly) obtained a weak bound on the cost of the compressions in the (small) forest \mathcal{F}_+ .
- We recursively obtained a (good) bound on the cost of the compressions in the (larger) forest \mathcal{F}_- .
- We solved the recurrence to get the bound

$$\mathbf{T(m, n, r) \leq 2n \lg^{**} r + 2m.}$$

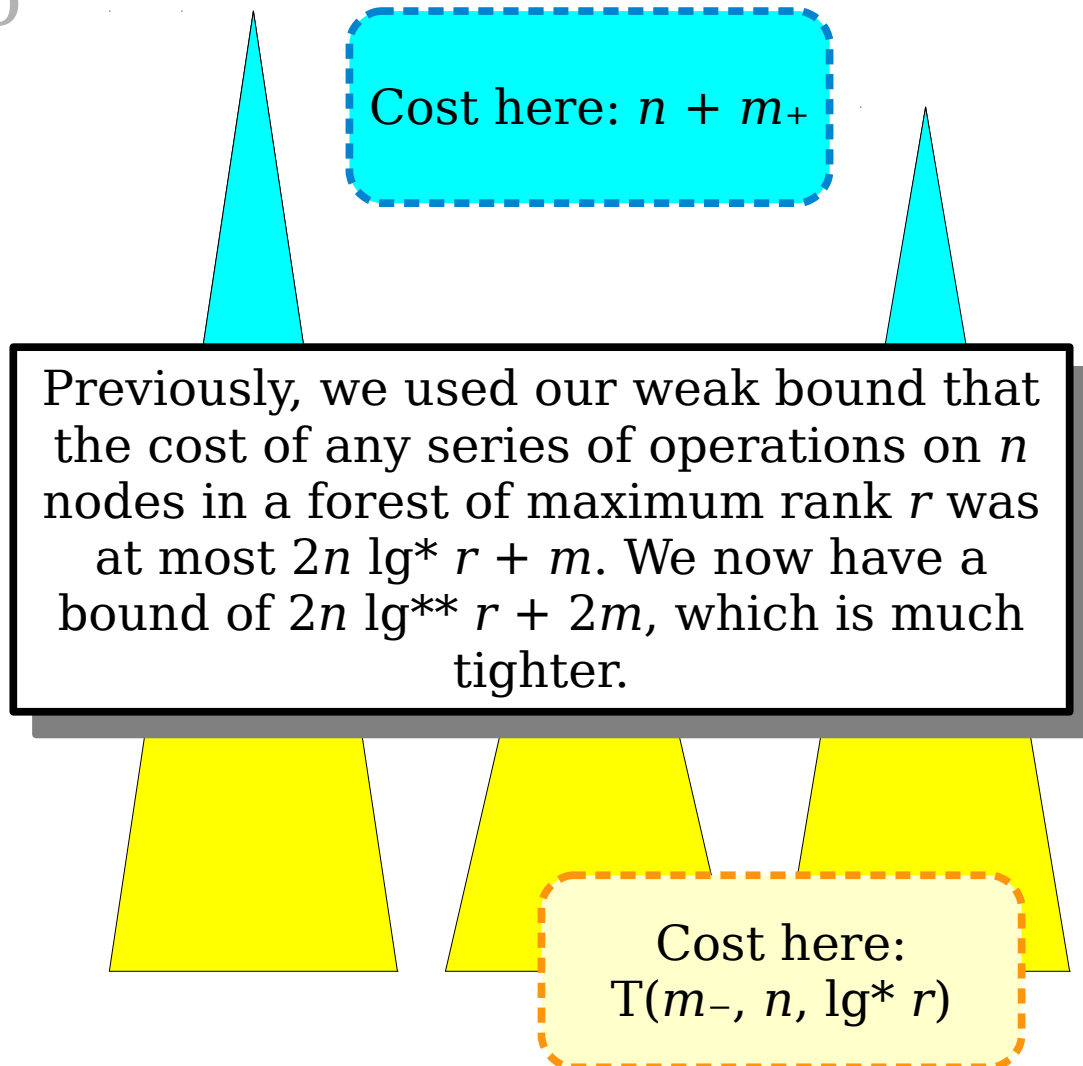


Our Strategy

What could we do to tighten the runtime bound?

- **Option 1:** Tighten the bound on the cost of the top forest.

Option 2: Slice the forest even lower to make the recursion tree shorter.

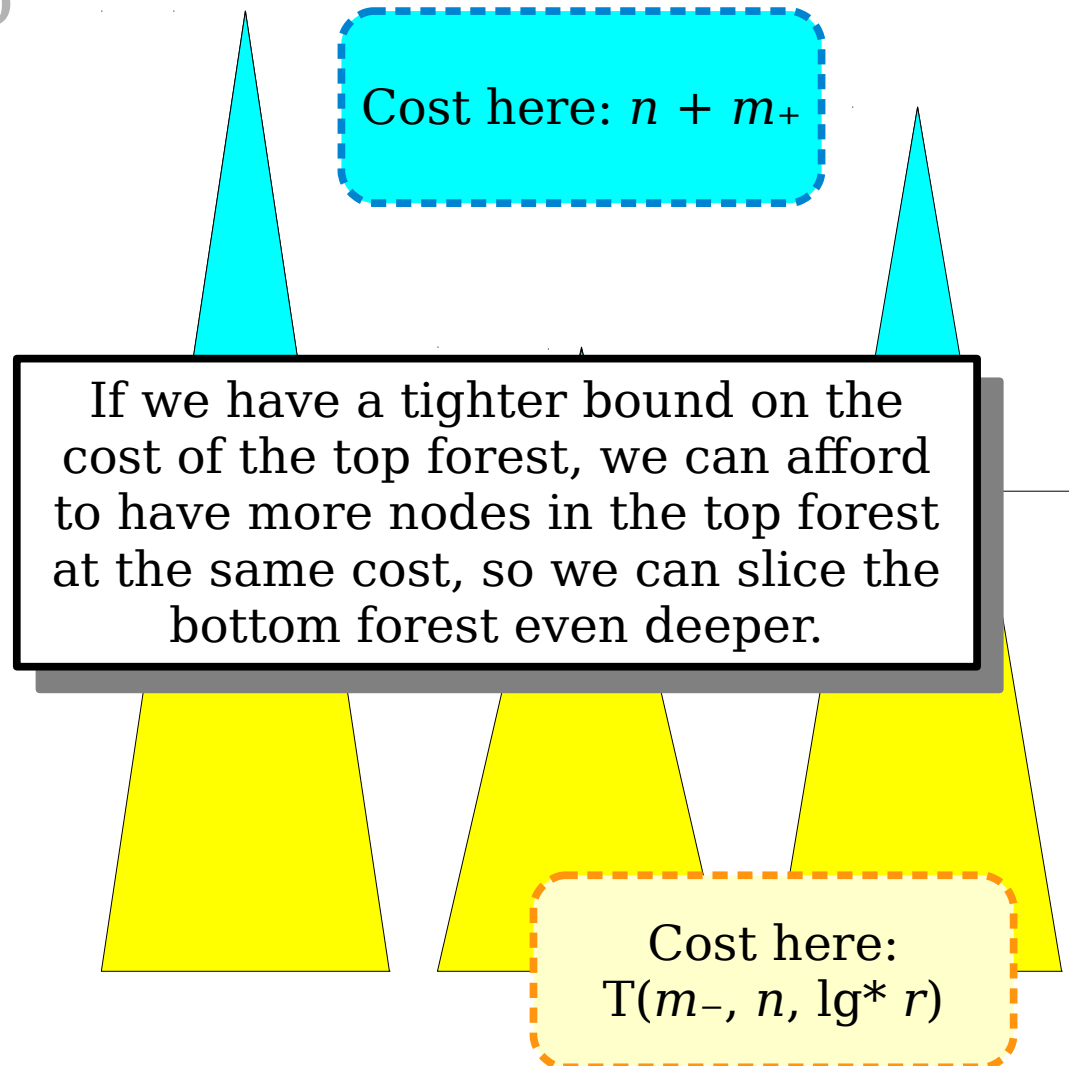


Our Strategy

What could we do to tighten the runtime bound?

Option 1: Tighten the bound on the cost of the top forest.

- **Option 2:** Slice the forest even lower to make the recursion tree shorter.



The Feedback Lemma

- **Lemma:** Suppose we know that

$$T(m, n, r) \leq 2n \lg^{*(k)} n + km.$$

Then

$$T(m, n, r) \leq 2n \lg^{*(k+1)} n + (k+1)m.$$

- **Proof:** Induction! Use the previous proof as a template: split the forest at rank $\lg^{*(k)} r$, use the known bound to bound the cost of the top forest, and use recursion to bound the cost of the bottom forest. ■

The Final Steps

- For any $k \in \mathbb{N}$, we have

$$T(m, n, r) \leq 2n \lg^{*(k)} r + km.$$

- We can upper-bound r at $\log n$, so we have

$$T(m, n) \leq 2n \lg^{*(k)} n + km.$$

- As n gets larger and larger, we can increase the value of k to make the $\lg^{*(k)} n$ term at most some constant value.
- **Question:** What is that k , as a function of n ?
- The **Ackermann inverse function**, denoted $\alpha(n)$, is
$$\alpha(n) = \min\{ k \in \mathbb{N} \mid \lg^{*(k)} n \leq 3 \}$$
- **Theorem:** The cost of performing any m operations on any n -node disjoint set forest using union-by-rank and path compression is $O(n + m\alpha(n))$.

Intuiting $\alpha(n)$

- Imagine we want to define some function A such that
 - $A(n, 0) = 2$
 - $A(n, 1) = 2 + 2 + \dots + 2 = 2n$
 - $A(n, 2) = 2 \times 2 \times \dots \times 2 = 2^n$.
 - $A(n, 3) = 2^{2^{\dots^2}} = {}_n2$. (tetration)
 - $A(n, 4) = {}^2 \dots {}^2 2 = {}_n2$. (pentation)
 - $A(n, 5)$ doesn't have a name, but scares children.
- The function A is called an **Ackermann-type function**. There are a number of different functions in this family, but they all (fundamentally) apply higher and higher orders of functions to the arguments.

Intuiting $\alpha(n)$

- **Theorem:** Asymptotically, the function $\alpha(n)$ is the inverse of $A(n, n)$, hence the name “Ackermann inverse”.
- **Intuition:**
 - $\lg n$ is the inverse of 2^n , which is $A(n, 2)$.
 - $\lg^* n$ is the inverse of ${}^n 2$ (tetration), which is $A(n, 3)$.
 - $\lg^{**} n$ is the inverse of ${}_{n^2} 2$ (pentation), which is $A(n, 4)$.
 - $\alpha(n)$ tells you how many stars you need to make $\lg^{*(k)} n$ drop to a constant, which essentially asks for which essentially asks for what order of operation you need to invert.
- This function grows more slowly than *any* of the iterated logarithm families. It's so slowly-growing that an input to it that would make it more than, say, 10 can't even be expressed without inventing special notation for fast-growing numbers.

Intuiting $\alpha(n)$

- If you keep dividing by two, you should expect a \log term.
- If you keep taking logs, you should expect a \log^* term.
- If you keep taking \log^* s, you should expect a \log^{**} term.
- If you keep adding stars to your logs, you should expect an α term.

Some Notes on $\alpha(n)$

- The term $\alpha(n)$ arises in many different algorithms:
 - Range semigroup queries: there's a lower bound of $\alpha(n)$ on the cost of a query under certain algebraic assumptions.
 - Minimum spanning trees: the fastest known deterministic MST algorithm runs in time $O(m\alpha(n))$ due to a connection to the above topic.
 - Splay trees: imagine you treat a splay tree as a deque. hilariously, the best bound we have on the runtime of performing n deque operations is $O(n\alpha^*(n))$. It's suspected to be $O(n)$, but this hasn't been proven.
- $\alpha(n)$ and its variants are the slowest-growing functions that are routinely encountered in algorithms and data structures. ***And now you know where it comes from!***

Next Time

- ***Euler Tour Trees***
 - Fully dynamic connectivity in forests.
- ***Dynamic Graphs***
 - Fully dynamic connectivity in general graphs (ITA).