



Algorithms: Design  
and Analysis, Part II

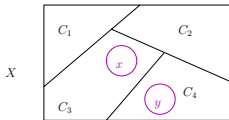
# Advanced Union-Find

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Lazy Unions

# The Union-Find Data Structure

**Raison d'être:** Maintain a partition of a set  $X$ .

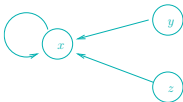


**FIND:** Given  $x \in X$ , return name of  $x$ 's group.

**UNION:** Given  $x$  &  $y$ , merge groups containing them.

**Previous solution (for Kruskal's MST algorithm)**

- Each  $x \in X$  points directly to the "leader" of its group.

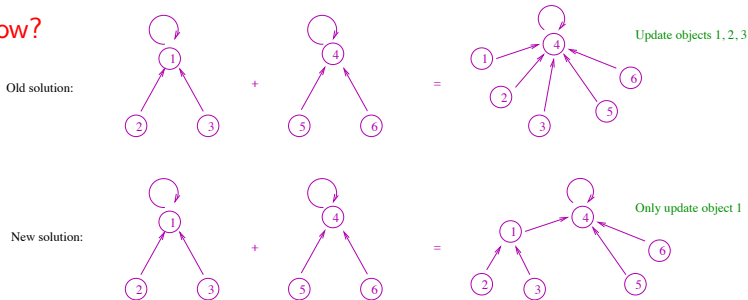


- $O(1)$  FIND [just return  $x$ 's leader]
- $O(n \log n)$  total work for  $n$  UNIONS [when 2 groups merge, smaller group inherits leader of larger one]

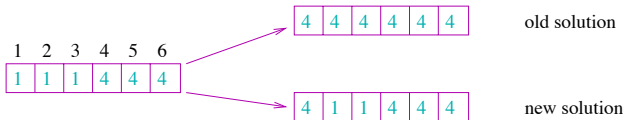
# Lazy Unions

New idea: Update only one pointer each merge!

How?

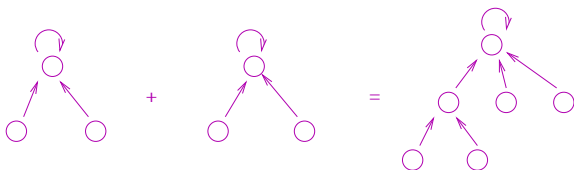


In array representation: (Where  $A[i] \leftrightarrow$  name of  $i$ 's parent)



# How to Merge?

**In general:** When two groups merge in a UNION, make one group's leader (i.e., root of the tree) a child of the other one.



**Pro:** UNION reduces to 2 FINDS [ $r_1 = \text{FIND}(x)$ ,  $r_2 = \text{FIND}(y)$ ] and  $O(1)$  extra work [link  $r_1$ ,  $r_2$  together]

**Con:** To recover leader of an object, need to follow a path of parent pointers [not just one!]

⇒ Not clear if FIND still takes  $O(1)$  time.



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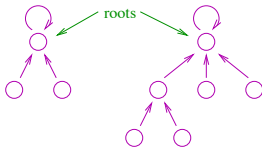
# Advanced Union-Find

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Union by Rank

# The Lazy Union Implementation

**New implementation:** Each object  $x \in X$  has a parent field.



**Invariant:** Parent pointers induce a collection of directed trees on  $X$ . ( $x$  is a root  $\iff$   $\text{parent}[x]=x$ )

**Initially:** For all  $x$ ,  $\text{parent}[x]=x$



**FIND( $x$ ):** Traverse parent pointers from  $x$  until you hit the root.

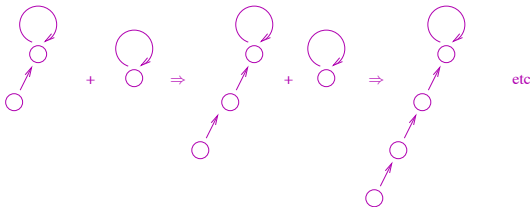
**UNION( $x, y$ ):**  $s_1 = \text{FIND}(x)$ ;  $s_2 = \text{FIND}(y)$ ; Reset parent of one of  $s_1, s_2$  to be the other.

# Quiz on Lazy Unions

**Question:** Suppose, in the UNION operation, we choose the new root arbitrarily from the two old ones. What is the worst-case running time of the FIND and UNION operations, respectively?

- A)  $\Theta(1), \Theta(1)$
- B)  $\Theta(\log n), \Theta(1)$
- C)  $\Theta(\log n), \Theta(\log n)$
- D)  $\Theta(n), \Theta(n)$

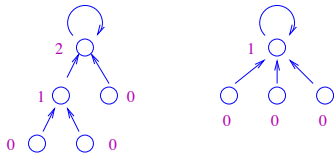
**Issue:** Scraggly trees:



# Union by Rank

**Ranks:** For each  $x \in X$ , maintain field  $\text{rank}[x]$ .

[In general  $\text{rank}[x] = 1 + (\max \text{rank of } x\text{'s children})$ ]



**Invariant (for now):** For all  $x \in X$ ,  $\text{rank}[x] = \text{maximum number of hops from some leaf to } x$ .

[Initially,  $\text{rank}[x] = 0$  for all  $x \in X$ ]

**To avoid scraggly trees ("Union by Rank"):** Given  $x$  &  $y$ :

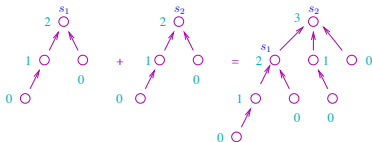
- $s_1 = \text{FIND}(x)$ ,  $s_2 = \text{FIND}(y)$
- If  $\text{rank}[s_1] > \text{rank}[s_2]$  then set  $\text{parent}[s_2]$  to  $s_1$  else set  $\text{parent}[s_1]$  to  $s_2$ .

**To-do:** Update ranks to restore Invariant.



# Quiz on Rank Updates

**Question:** Recall  $s_1 = \text{FIND}(x)$ ,  $s_2 = \text{FIND}(y)$ . How do the ranks of  $s_1$  &  $s_2$  change after  $\text{UNION}(x, y)$ ?



- A) Unchanged
- B) The one with larger rank goes up by 1
- C) The one with smaller rank goes up by 1
- D) No change unless ranks of  $s_1, s_2$  were equal, in which case  $s_2$ 's rank goes up by 1



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Union by Rank -  
Analysis

# Properties of Ranks

**Recall:** Lazy Unions.

**Invariant (for now):**  $\text{rank}[x] = \max \# \text{ of hops from a leaf to } x$ .

[Note  $\max_x \text{rank}[x] \approx \text{worst-case running time of FIND.}$ ]

**Union by Rank:** Make old root with smaller rank child of the root with the larger rank.

[Choose new root arbitrarily in case of a tie, and add 1 to its rank.]



**Immediate from Invariant/Rank Maintenance:**

- (1) For all objects  $x$ ,  $\text{rank}[x]$  only goes up over time
- (2) Only ranks of roots can go up  
[once  $x$  a non-root,  $\text{rank}[x]$  frozen forevermore]
- (3) Ranks strictly increase along a path to the root

# Rank Lemma

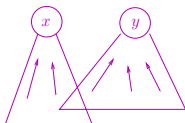
**Rank Lemma:** Consider an arbitrary sequence of UNION (+FIND) operations. For every  $r \in \{0, 1, 2, \dots\}$ , there are at most  $n/2^r$  objects with rank  $r$ .

**Corollary:** Max rank always  $\leq \log_2 n$

**Corollary:** Worst-case running time of FIND, UNION is  $O(\log n)$ .  
[With Union by Rank.]

# Proof of Rank Lemma

**Claim 1:** If  $x, y$  have the same rank  $r$ , then their subtrees (objects from which can reach  $x, y$ ) are disjoint.



**Claim 2:** The subtree of a rank- $r$  object has size  $\geq 2^r$ .  
[Note Claim 1 + Claim 2 imply the Rank Lemma.]

**Proof of Claim 1:** Will show contrapositive. Suppose subtrees of  $x, y$  have object  $z$  in common  $\Rightarrow \exists$  paths  $z \rightarrow x, z \rightarrow y$   
 $\Rightarrow$  One of  $x, y$  is an ancestor of the other  
 $\Rightarrow$  The ancestor has strictly larger rank. [By property (3)]

QED (Claim 1)

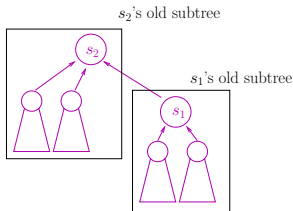
# Proof of Claim 2

Rank  $r \Rightarrow$  Subtree size  $\geq 2^r$

**Base case:** Initially all ranks = 0, all subtree sizes = 1

**Inductive step:** Nothing to prove unless the rank of some object changes (subtree sizes only go up).

**Interesting case:** UNION( $x, y$ ), with  $s_1 = \text{FIND}(x)$ ,  $s_2 = \text{FIND}(y)$ , and  $\text{rank}[s_1] = \text{rank}[s_2] = r \Rightarrow s_2$ 's new rank =  $r + 1$   
 $\Rightarrow s_2$ 's new subtree size =  $s_2$ 's old subtree size +  $s_1$ 's old subtree size (each at least  $2^r$  by the inductive hypothesis)  $\geq 2^{r+1}$ . QED!





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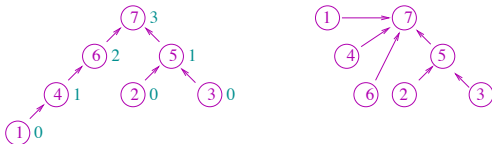
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Path Compression

# Path Compression

**Idea:** Why bother traversing a leaf-root path multiple times?

**Path compression:** After  $\text{FIND}(x)$ , install shortcuts (i.e., revise parent pointers) to  $x$ 's root all along the  $x \rightarrow \text{root}$  path.



**In array representation:**



**Con:** Constant-factor overhead to  $\text{FIND}$  (from “multitasking”).

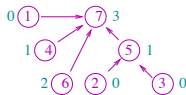
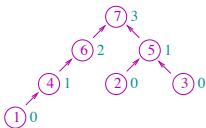
**Pro:** Speeds up subsequent  $\text{FIND}$ s. [But by how much?]



# On Ranks

**Important:** Maintain all rank fields EXACTLY as without path compression.

- Ranks initially all 0
- In UNION, new root = old root with bigger rank
- When merging two nodes of common rank  $r$ , reset new root's rank to  $r + 1$



**Bad news:** Now  $\text{rank}[x]$  is only an upper bound on the maximum number of hops on a path from a leaf to  $x$

(which could be much less)

**Good news:** Rank Lemma still holds ( $\leq n/2^r$  objects with rank  $r$ )

**Also:** Still always have  $\text{rank}[\text{parent}[x]] > \text{rank}[x]$  for all non-roots  $x$

# Hopcroft-Ullman Theorem

**Theorem:** [Hopcroft-Ullman 73] With Union by Rank and path compression,  $m$  Union+Find operations take  $O(m \log^* n)$  time, where  $\log^* n =$  the number of times you need to apply log to  $n$  before the result is  $\leq 1$ .

# Quiz on $\log^*$

Question: What is  $\log^*(2^{65536})$ ?

A) 2

B) 5

C) 16

D) 65536

In general:  $\log^*(2^{2^{\dots t \text{ times } \dots^2}}) = t$

# Measuring Progress



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Path Compression: The  
Hopcroft-Ullman Analysis

# Hopcroft-Ullman Theorem

**Theorem:** [Hopcroft-Ullman 73] With Union by Rank and path compression,  $m$  UNION+FIND operations take  $O(m \log^* n)$  time, where  $\log^* n$  = the number of times you need to apply log to  $n$  before the result is  $\leq 1$ .

[Will focus on interesting case where  $m = \Omega(n)$ ]

# Measuring Progress

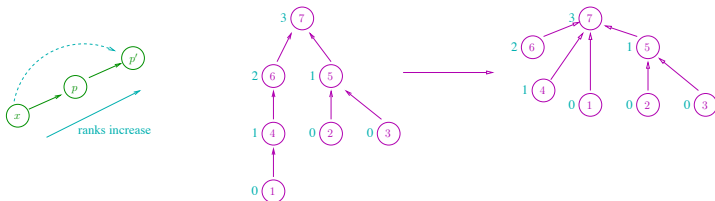
**Intuition:** Installing shortcuts should significantly speed up subsequent FINDs+UNIONS.

**Question:** How to track this progress and quantify the benefit?

**Idea:** Consider a non-root object  $x$ .  $\longrightarrow$  Recall:  $\text{rank}[x]$  frozen


**Progress measure:**  $\text{rank}[\text{parent}[x]] - \text{rank}[x]$

**Path compression increases this progress measure:** If  $x$  has old parent  $p$ , new parent  $p' \neq p$ , then  $\text{rank}[p'] > \text{rank}[p]$ .



# Proof Setup

**Rank blocks:**  $\{0\}, \{1\}, \{2, 3, 4\}, \{5, \dots, 2^4\}, \{17, 18, \dots, 2^{16}\},$   
 $\{65537, \dots, 2^{65536}\}, \dots, \{\dots, n\}$



**Note:** There are  $O(\log^* n)$  different rank blocks.

**Semantics:** Traversal  $x \rightarrow \text{parent}(x)$  is “fast progress”  $\iff$   
 $\text{rank}[\text{parent}[x]]$  in larger block than  $\text{rank}[x]$

**Definition:** At a given point in time, call object  $x$  good if

- (1)  $x$  or  $x$ 's parent is a root OR
- (2)  $\text{rank}[\text{parent}[x]]$  in larger block than  $\text{rank}[x]$

$x$  is bad otherwise.



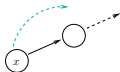
# Proof of Hopcroft-Ullman

**Point:** Every FIND visits only  $O(\log^* n)$  good nodes  $[2 + \# \text{ of rank blocks} = O(\log^* n)]$

**Upshot:** Total work done during  $m$  operations  $= O(m \log^* n)$   
(visits to good objects) + total  $\#$  of visits to bad nodes (need to bound globally by separate argument)

**Consider:** A rank block  $\{k + 1, k + 2, \dots, 2^k\}$ .

**Note:** When a bad node is visited



its parent is changed to one with strictly larger rank  $\Rightarrow$  Can only happen  $2^k$  times before  $x$  becomes good (forevermore).

# Proof of Hopcroft-Ullman II

**Total work:**  $O(m \log^* n) + O(\text{\# visits to bad nodes})$ .

$\leq n$  for each of  $O(\log^* n)$  rank blocks

**Consider:** A rank block  $\{k+1, k+2, \dots, 2^k\}$ .

**Last slide:** For each object  $x$  with final rank in this block,  $\text{\# visits to } x \text{ while } x \text{ is bad}$  is  $\leq 2^k$ .

**Rank Lemma:** Total number of objects  $x$  with final rank in this rank block is  $\sum_{i=k+1}^{2^k} n/2^i \leq n/2^k$ .

$\leq n$  visits to bad objects in this rank block.

**Recall:** Only  $O(\log^* n)$  rank blocks.

**Total work:**  $O((m+n) \log^* n)$ .



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The Ackermann Function

# Tarjan's Bound

**Theorem:** [Tarjan 75] With Union by Rank and path compression,  $m$  UNION+FIND operations take  $O(m\alpha(n))$  time, where  $\alpha(n)$  is the inverse Ackerman function (will define in this video)

Proof in next video.

# The Ackermann Function

**Aside:** Many different definitions, all more or less equivalent.

Will define  $A_k(r)$  for all integers  $k$  and  $r \geq 1$ . (recursively)

**Base case:**  $A_0(r) = r + 1$  for all  $r \geq 1$ .

**In general:** For  $k, r \geq 1$ :

$$\begin{aligned} A_k(r) &= \text{Apply } A_{k-1} \text{ } r \text{ times to } r \\ &= (A_{k-1} \circ A_{k-1} \circ \dots \circ A_{k-1})(r) \end{aligned}$$

$r$ -fold composition



## Quiz: $A_1$

Quiz:  $A_1(r)$  corresponds to what function of  $r$ ?

- A) Successor ( $r \mapsto r + 1$ )
- B) Doubling ( $r \mapsto 2r$ )
- C) Exponentiation ( $r \mapsto 2^r$ )
- D) Tower function ( $r \mapsto 2^{2^{\dots r \text{ times } \dots^2}}$ )

$$A_1(r) = (A_0 \circ A_0 \circ \dots \circ A_0)(r) = 2r$$

( $r$ -fold composition, add 1 each time)

## Quiz: $A_2$

**Quiz:** What function does  $A_2(r)$  correspond to?

A)  $r \mapsto 4r$

B)  $r \mapsto 2^r$

B)  $r \mapsto r2^r$

D)  $r \mapsto 2^{2 \dots r \text{ times} \dots 2}$

$A_2(r) = (A_1 \circ A_1 \circ \dots \circ A_1)(r) = r2^r$   
( $r$ -fold composition, doubles each time)

## Quiz: $A_3$

Quiz: What is  $A_3(2)$ ? Recall  $A_2(r) = r2^r$

- A) 8
- B) 1024
- B) 2048
- D) Bigger than 2048

$$A_3(2) = A_2(A_2(2)) = A_2(8) = 82^8 = 2^{11} = 2048$$

In general:  $A_3(r) = (A_2 \circ A_2 \circ \dots (r \text{ times}) \dots \circ A_2)(r) \geq$  a tower  
of  $r$  2's  $= 2^{2^{\dots r \text{ times} \dots 2}}$



$$A_4$$

$$A_4(2) = A_3(A_3(2)) = A_3(2048) \geq 2^{2^{\dots \text{height } 2048} \dots^2}$$

In general:  $A_4(r) = (A_3 \circ \dots \text{ } r \text{ times } \dots \circ A_3)(r) \approx$  iterated tower function (aka “wowzer” function)

# The Inverse Ackermann Function

**Definition:** For every  $n \geq 4$ ,  $\alpha(n)$  = minimum value of  $k$  such that  $A_k(2) \geq n$ .

$$\alpha(n) = 1, n = 4 \text{ (} A_1(2) = 4 \text{)}$$

$$\alpha(n) = 2, n = 5, \dots, 8 \text{ (} A_2(2) = 8 \text{)}$$

$$\alpha(n) = 3, n = 9, 10, \dots, 2048$$

$$\alpha(n) = 4, n \text{ up to roughly a tower}$$

of 2's of height 2048

$$\alpha(n) = 5 \text{ for } n \text{ up to ???}$$

$$\log^* n = 1, n = \underline{2}$$

$$\log^* n = 2, n = 3, \underline{4}$$

$$\log^* n = 3, n = 5, \dots, \underline{16}$$

$$\log^* n = 4, n = 17, \underline{65536}$$

$$\log^* n = 5, n = 65537, \underline{2^{65536}}$$

$$\log^* n = 2048 \text{ for such } n$$



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Tarjan's Analysis

# Tarjan's Bound

**Theorem:** [Tarjan 75] With Union by Rank and path compression,  $m$  UNION+FIND operations take  $O(m\alpha(n))$  time, where  $\alpha(n)$  is the inverse Ackerman function

**Acknowledgement:** Kozen, "Design and Analysis of Algorithms"

# Building Blocks of Hopcroft-Ullman Analysis

**Block #1:** Rank Lemma (at most  $n/2^r$  objects of rank  $r$ )

**Block #2:** Path compression  $\Rightarrow$  If  $x$ 's parent pointer updated from  $p$  to  $p'$ , then  $\text{rank}(p') \geq \text{rank}(p) + 1$

**New idea:** Stronger version of building block #2. In most cases, rank of new parent much bigger than rank of old parent (not just by 1).

# Quantifying Rank Gaps

**Definition:** Consider a non-root object  $x$  (so  $\text{rank}[x]$  fixed forevermore)

Define  $\delta(x) = \max$  value of  $k$  such that  
 $\text{rank}[\text{parent}[x]] \geq A_k(\text{rank}[x])$

(Note  $\delta(x)$  only goes up over time)

**Examples:** Always have  $\delta(x) \geq 0$

$$\delta(x) \geq 1 \iff \text{rank}[\text{parent}[x]] \geq 2 \text{ rank}[x]$$

$$\delta(x) \geq 2 \iff \text{rank}[\text{parent}[x]] \geq \text{rank}[x] 2^{\text{rank}[x]}$$

**Note:** For all objects  $x$  with  $\text{rank}[x] \geq 2$ , then  $\delta(x) \leq \alpha(n)$   
[Since  $A_{\alpha(n)}(2) \geq n$ ]

# Good and Bad Objects

**Definition:** An object  $x$  is bad if all of the following hold:

- (1)  $x$  is not a root
- (2)  $\text{parent}(x)$  is not a root
- (3)  $\text{rank}(x) \geq 2$
- (4)  $x$  has an ancestor  $y$  with  $\delta(y) = \delta(x)$

$x$  is good otherwise.

# Quiz

**Question:** What is the maximum number of good objects on an object-root path?

- A)  $\Theta(1)$
- B)  $\Theta(\alpha(n))$
- C)  $\Theta(\log^* n)$
- D)  $\Theta(\log n)$

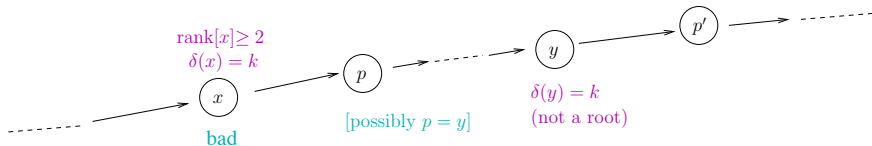
$\leq$  1 root + 1 child of root  
+ 1 object with rank 0  
+ 1 object with rank 1  
+ 1 object with  $\delta(x) = k$   
for each  $k = 0, 1, 2, \dots, \alpha(n)$



# Proof of Tarjan's Bound

**Upshot:** Total work of  $m$  operations =  $O(m\alpha(n))$  (visits to good objects) + total # of visits to bad objects (will show is  $O(n\alpha(n))$ )

**Main argument:** Suppose a FIND operation visits a bad object  $x$ :



**Path compression:**  $x$ 's new parent will be  $p'$  or even higher.

$$\Rightarrow \text{rank}[x\text{'s new parent}] \geq \text{rank}[p'] \geq A_k(\text{rank}[y]) \geq A_k(\text{rank}[p])$$

ranks only go up

since  $\delta(y) = k$

ranks only go up

# Proof of Tarjan's Bound II

**Point:** Path compression (at least) applies the  $A_k$  function to  $\text{rank}[x\text{'s parent}]$

**Consequence:** If  $r = \text{rank}[x]$  ( $\geq 2$ ), then after  $r$  such pointer updates we have

$$\text{rank}[x\text{'s parent}] \geq (A_k \circ \dots \text{ } r \text{ times } \dots \circ A_k)(r) = A_{k+1}(r)$$

Definition of Ackermann function

**Thus:** While  $x$  is bad, every  $r$  visits increases  $\delta(x)$   
 $\Rightarrow \leq r\alpha(n)$  visits to  $x$  while it's bad

# Proof of Tarjan's Bound III

**Recall:** Total work of  $m$  operations is  $O(m\alpha(n))$  (visits to good objects) + total # of visits to bad objects.

$$\leq \sum_{\text{objects } x} \text{rank}[x] \alpha(n)$$

$$= \alpha(n) \sum_{r \geq 0} r \cdot (\# \text{ of objects with rank } r)$$

$\leq n/2^r$  for each  $r$  by the Rank Lemma

$$= n\alpha(n) \sum_{r \geq 0} r/2^r \longrightarrow = O(1)$$

$$= O(n\alpha(n)). \quad \text{QED!}$$

# Epilogue

“This is probably the first and maybe the only existing example of a simple algorithm with a very complicated running time. . . . I conjecture that there is no linear-time method, and that the algorithm considered here is optimal to within a constant factor.”

-Tarjan, “Efficiency of a Good But Non Linear Set Union Algorithm”, Journal of the ACM, 1975.

Conjecture proved by [Fredman/Saks 89]!