



# **Chapter 4**

# **Data Structures**

**Algorithm Theory**  
**WS 2013/14**

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# Fibonacci Heaps: Marks

## Cycle of a node:

1. Node  $v$  is removed from root list and linked to a node  
 **$v.mark = false$**
2. Child node  $u$  of  $v$  is cut and added to root list  
 **$v.mark = true$**
3. Second child of  $v$  is cut  
**node  $v$  is cut as well and moved to root list**

The boolean value  $v.mark$  indicates whether node  $v$  has lost a child since the last time  $v$  was made the child of another node.

# Potential Function

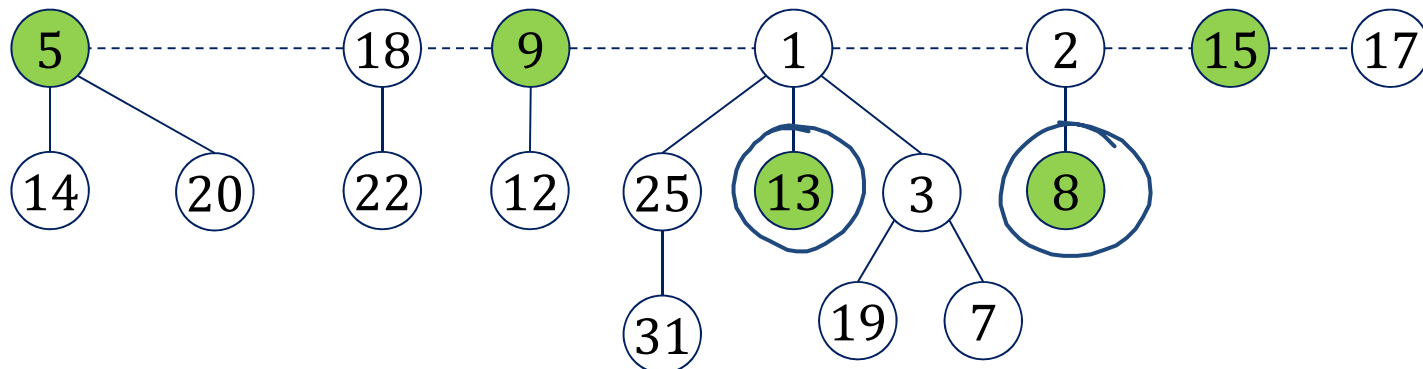
System state characterized by two parameters:

- **$R$** : number of trees (length of  $H.rootlist$ )
- **$M$** : number of marked nodes that are not in the root list

Potential function:

$$\Phi := R + 2M$$

Example:



- $R = 7, \underline{M = 2} \rightarrow \Phi = 11$

# Actual Time of Operations

- Operations: ***initialize-heap, is-empty, insert, get-min, merge***

actual time:  $O(1)$

- Normalize unit time such that

$$t_{init}, t_{is-empty}, \underline{t_{insert}}, \underline{t_{get-min}}, \underline{t_{merge}} \leq 1$$

- Operation ***delete-min***:

- Actual time:  $O(\text{length of } H.\text{rootlist} + D(n))$

- Normalize unit time such that

$$t_{del-min} \leq \underline{D(n)} + \underline{\text{length of } H.\text{rootlist}}$$

- Operation ***decrease-key***:

- Actual time:  $O(\text{length of path to next unmarked ancestor})$

- Normalize unit time such that

$$t_{decr-key} \leq \underline{\text{length of path to next unmarked ancestor}}$$

# Amortized Time of Delete-Min

Assume that operation  $i$  is a *delete-min* operation:

**Actual time:**  $t_i \leq D(n) + |H.rootlist|$

**Potential function  $\Phi = R + 2M$ :**

$$a_i = t_i + \Phi_i - \Phi_{i-1}$$

- $R$ : changes from  $H.rootlist$  to at most  $D(n)$
- $M$ : (# of marked nodes that are not in the root list)
  - no new marks
  - if node  $v$  is moved away from root list,  $v.mark$  is set to false  
→ value of  $M$  does not increase!

$$\left[ \begin{array}{l} M_i \leq M_{i-1}, \quad R_i \leq R_{i-1} + D(n) - |H.rootlist| \\ \Phi_i \leq \Phi_{i-1} + D(n) - |H.rootlist| \end{array} \right]$$

**Amortized Time:**  $a_i = t_i + \Phi_i - \Phi_{i-1} \leq 2D(n)$

# Amortized Time of Decrease-Key

Assume that operation  $i$  is a *decrease-key* operation at node  $u$ :

**Actual time:**  $t_i \leq$  length of path to next unmarked ancestor  $v$

**Potential function  $\Phi = R + 2M$ :**  $\ell_i \leq k+1$

- Assume, node  $u$  and nodes  $u_1, \dots, u_k$  are moved to root list
  - $u_1, \dots, u_k$  are marked and moved to root list,  $v$ . mark is set to true
- $\geq k$  marked nodes go to root list,  $\leq 1$  node gets newly marked
- $R$  grows by  $\leq k + 1$ ,  $M$  grows by 1 and is decreased by  $\geq k$



$$R_i \leq R_{i-1} + k + 1, \quad M_i \leq M_{i-1} + 1 - k$$

$$\Phi_i \leq \Phi_{i-1} + (k + 1) - 2(k - 1) = \Phi_{i-1} + 3 - k$$

**Amortized time:**

$$a_i = \underbrace{t_i}_{\leq k+1} + \underbrace{\Phi_i - \Phi_{i-1}}_{\leq 3-k} \leq k + 1 + 3 - k = \underline{\underline{4}}$$

# Complexities Fibonacci Heap

- Initialize-Heap:  $O(1)$
- Is-Empty:  $O(1)$
- Insert:  $O(1)$
- Get-Min:  $O(1)$
- → Delete-Min:  $O(D(n))$
- → Decrease-Key:  $O(1)$
- Merge (heaps of size  $m$  and  $n, m \leq n$ ):  $O(1)$
- How large can  $D(n)$  get?

$$O(n + n_d \cdot D(n))$$

amortized

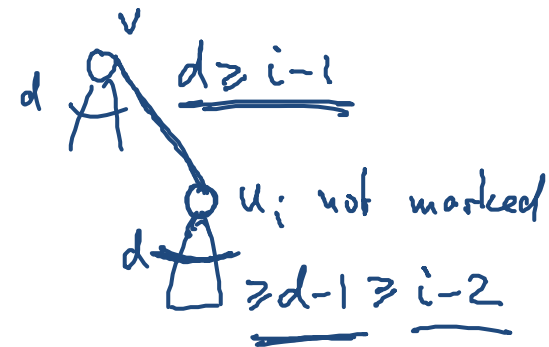
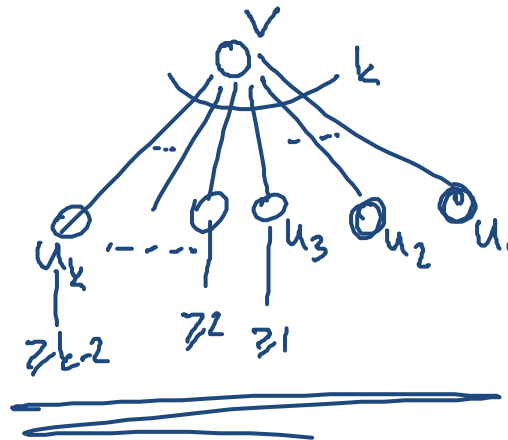
# Rank of Children

## Lemma:

Consider a node  $v$  of rank  $k$  and let  $u_1, \dots, u_k$  be the children of  $v$  in the order in which they were linked to  $v$ . Then,

$$\underline{\underline{\text{rank}(u_i) \geq i - 2.}}$$

## Proof:





# Size of Trees

## Fibonacci Numbers:

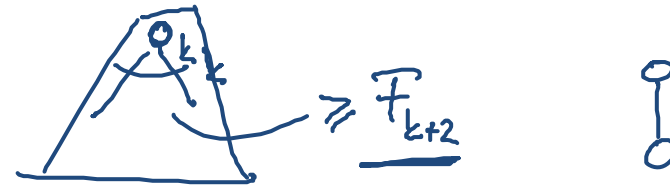
$$F_0 = 0, \quad F_1 = 1, \quad \forall k \geq 2: F_k = F_{k-1} + F_{k-2}$$

$F_0, F_1, F_2$

0, 1, 1, 2, 3, 5, 8, 13, 21, ...

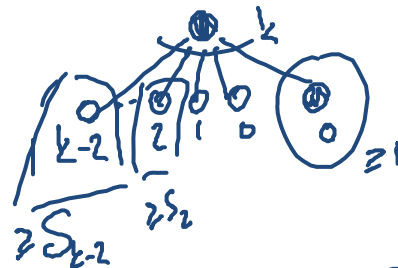
## Lemma:

In a Fibonacci heap, the size of the sub-tree of a node  $v$  with rank  $k$  is at least  $F_{k+2}$ .



## Proof:

- $S_k$ : minimum size of the sub-tree of a node of rank  $k$   
 $S_0 = 1, S_1 = 2$



$$S_k \geq 2 + \sum_{i=0}^{k-2} S_i$$

# Size of Trees

$$\underline{S_0 = 1}, \quad \underline{S_1 = 2}, \quad \forall k \geq 2: S_k \geq 2 + \sum_{i=0}^{k-2} S_i$$

- Claim about Fibonacci numbers:

$$\forall k \geq 0: \underline{F_{k+2}} = 1 + \sum_{i=0}^k F_i$$

$$F_0 = 0, F_1 = 1, F_k = F_{k-1} + F_{k-2}$$

$$k=0: F_2 = 1 + \sum_{i=0}^0 F_i = 1 + F_0 = 1 \quad \checkmark$$

~~step~~  
step:  $F_{k+2} = F_{k+1} + F_k$   
 $= \overset{\text{(I.H.)}}{F_k} + 1 + \sum_{i=0}^{k-1} F_i = 1 + \sum_{i=0}^k F_i$

# Size of Trees

$$\underline{S_0 = 1, S_1 = 2, \forall k \geq 2: S_k \geq 2 + \sum_{i=0}^{k-2} S_i}, \quad \underline{F_{k+2} = 1 + \sum_{i=0}^k F_i}$$

- Claim of lemma:  $S_k \geq F_{k+2}$

Ind. on k:

base:  $S_0 \geq F_2 = 1 \checkmark \quad S_1 \geq F_3 = 2 \checkmark$

step:  $S_k \geq 2 + \sum_{i=0}^{k-2} S_i \stackrel{\text{i.H.}}{\geq} 2 + \sum_{i=0}^{k-2} F_{i+2}$

$\begin{matrix} \downarrow & \downarrow \\ F_0 + F_1 & F_2 + F_3 + \dots + F_k \end{matrix}$

$= 2 + \sum_{j=2}^k F_j$

$= 1 + \sum_{j=0}^k F_j = \underline{F_{k+2}}$



# Size of Trees

## Lemma:

In a Fibonacci heap, the size of the sub-tree of a node  $v$  with rank  $k$  is at least  $F_{k+2}$ .

## Theorem:

The maximum rank of a node in a Fibonacci heap of size  $n$  is at most

$$D(n) = O(\log n).$$

## Proof:

- The Fibonacci numbers grow exponentially:

$$\rightarrow F_k = \frac{1}{\sqrt{5}} \cdot \left( \left( \frac{1 + \sqrt{5}}{2} \right)^k - \left( \frac{1 - \sqrt{5}}{2} \right)^k \right)$$

- For  $D(n) \geq k$ , we need  $n \geq F_{k+2}$  nodes.

# Summary: Binomial and Fibonacci Heaps



	Binomial Heap	Fibonacci Heap
<i>initialize</i>	$O(1)$	$O(1)$ $O(n)$
<i>insert</i>	$O(\log n)$	$O(1)$ $O(n)$
<i>get-min</i>	$O(1)$	$O(1)$ $O(n)$
<i>delete-min</i>	$O(\log n)$	$O(\log n)$ * $O(n)$ ←
<i>decrease-key</i>	$O(\log n)$	$O(1)$ * $O(m)$
<i>merge</i>	$O(\log n)$	$O(1)$
<i>is-empty</i>	$O(1)$	$O(1)$ $O(n)$

\* amortized time

Dijkstra:  $O(m + n \log n)$

$O(m \log n)$

# Minimum Spanning Trees

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## Prim Algorithm:

1. Start with any node  $v$  ( $v$  is the initial component)
2. In each step:  
Grow the current component by adding the minimum weight edge  $e$  connecting the current component with any other node

## Kruskal Algorithm:

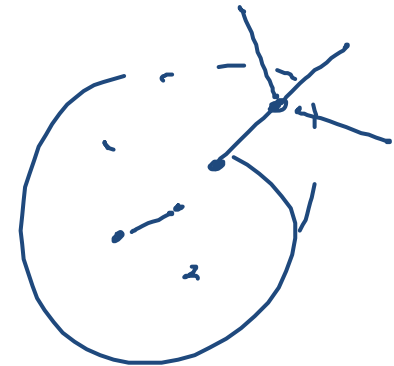
1. Start with an empty edge set
2. In each step:  
Add minimum weight edge  $e$  such that  $e$  does not close a cycle

# Implementation of Prim Algorithm

Start at node  $s$ , very similar to Dijkstra's algorithm:  $s$ .

1. Initialize  $d(s) = 0$  and  $d(v) = \infty$  for all  $v \neq s$
2. All nodes  ~~$s$  and  $v$~~  are unmarked

3. Get unmarked node  $u$  which minimizes  $d(u)$ :



4. For all  $e = \{u, v\} \in E$ ,  $d(v) = \min\{d(v), w(e)\}$

using Fib. heaps:

5. mark node  $u$

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

$\uparrow$                        $\uparrow$   
 #edges                      #nodes

6. Until all nodes are marked

# Implementation of Prim Algorithm

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## Implementation with Fibonacci heap:

- Analysis identical to the analysis of Dijkstra's algorithm:

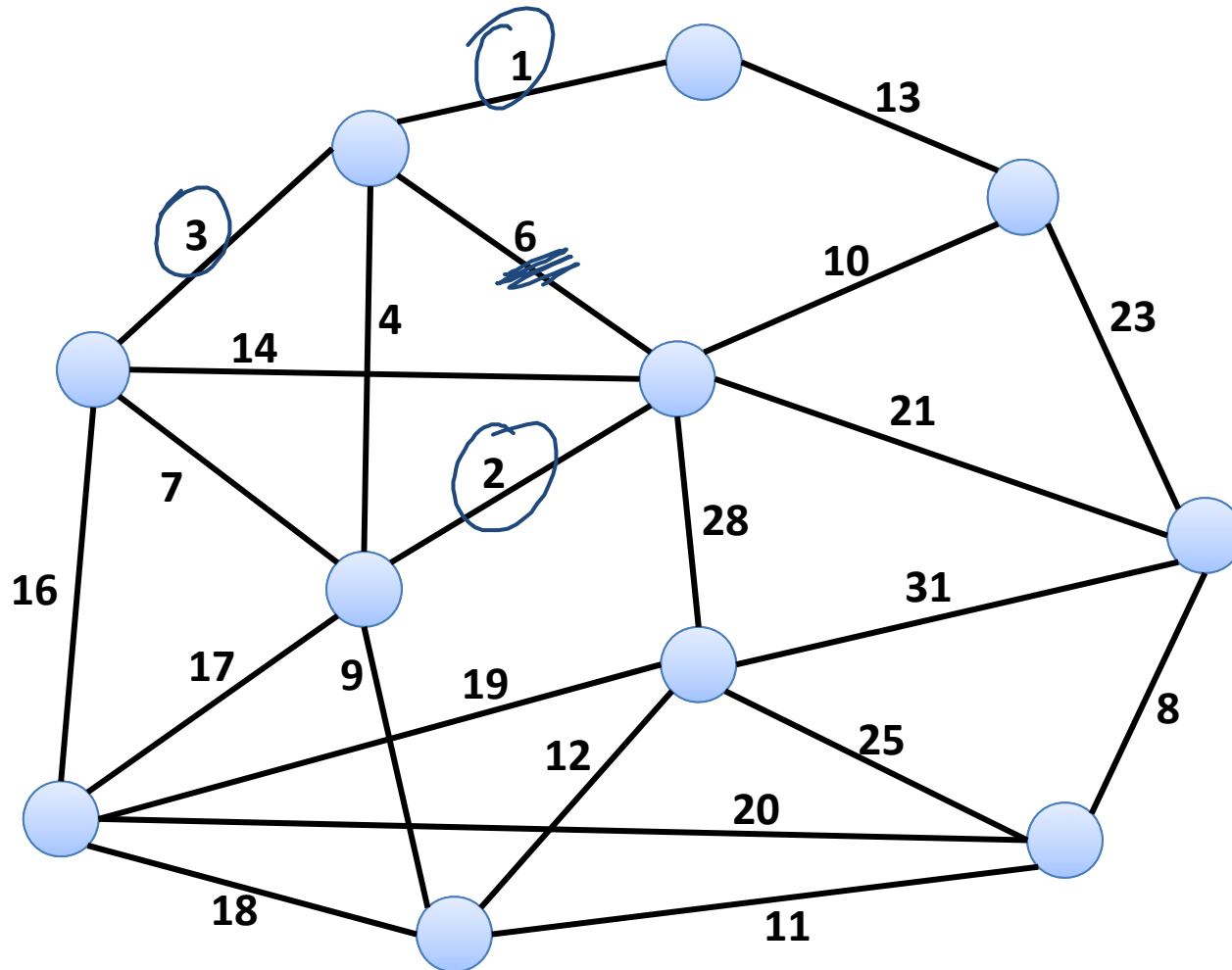
$O(n)$  insert and delete-min operations

$O(m)$  decrease-key operations

- Running time:  $O(m + n \log n)$



# Kruskal Algorithm



1. Start with an empty edge set
2. In each step: Add minimum weight edge  $e$  such that  $e$  does not close a cycle

# Implementation of Kruskal Algorithm

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1. Go through edges in order of increasing weights

sort edges by weight  $O(m \log n)$

2. For each edge  $e$ :

**if  $e$  does not close a cycle then**

need an efficient way to check whether  
 $e$  closes a cycle

**add  $e$  to the current solution**

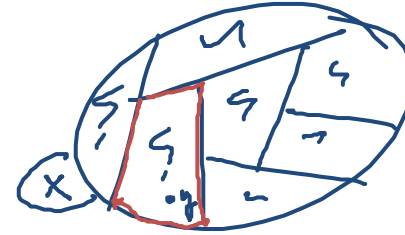
update data struct.

# Union-Find Data Structure

Also known as Disjoint-Set Data Structure...

Manages partition of a set of elements

- set of disjoint sets



**Operations:**

- **make\_set( $x$ ):** create a new set that only contains element  $x$
- **find( $x$ ):** return the set containing  $x$
- **union( $x, y$ ):** merge the two sets containing  $x$  and  $y$



# Managing Connected Components

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- Union-find data structure can be used more generally to manage the connected components of a graph
  - ... if edges are added incrementally
- **make\_set( $v$ )** for every node  $v$
- **find( $v$ )** returns component containing  $v$
- **union( $u, v$ )** merges the components of  $u$  and  $v$   
(when an edge is added between the components)
- Can also be used to manage biconnected components

# Basic Implementation Properties

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## Representation of sets:

- Every set  $S$  of the partition is identified with a representative, by one of its members  $x \in S$

## Operations:

- make\_set( $x$ ):  $x$  is the representative of the new set  $\{x\}$
- find( $x$ ): return representative of set  $S_x$  containing  $x$
- union( $x, y$ ): unites the sets  $S_x$  and  $S_y$  containing  $x$  and  $y$  and returns the new representative of  $S_x \cup S_y$

# Observations

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## Throughout the discussion of union-find:

- $n$ : total number of make\_set operations
- $m$ : total number of operations (make\_set, find, and union)

## Clearly:

- $m \geq n$
- There are at most  $n - 1$  union operations

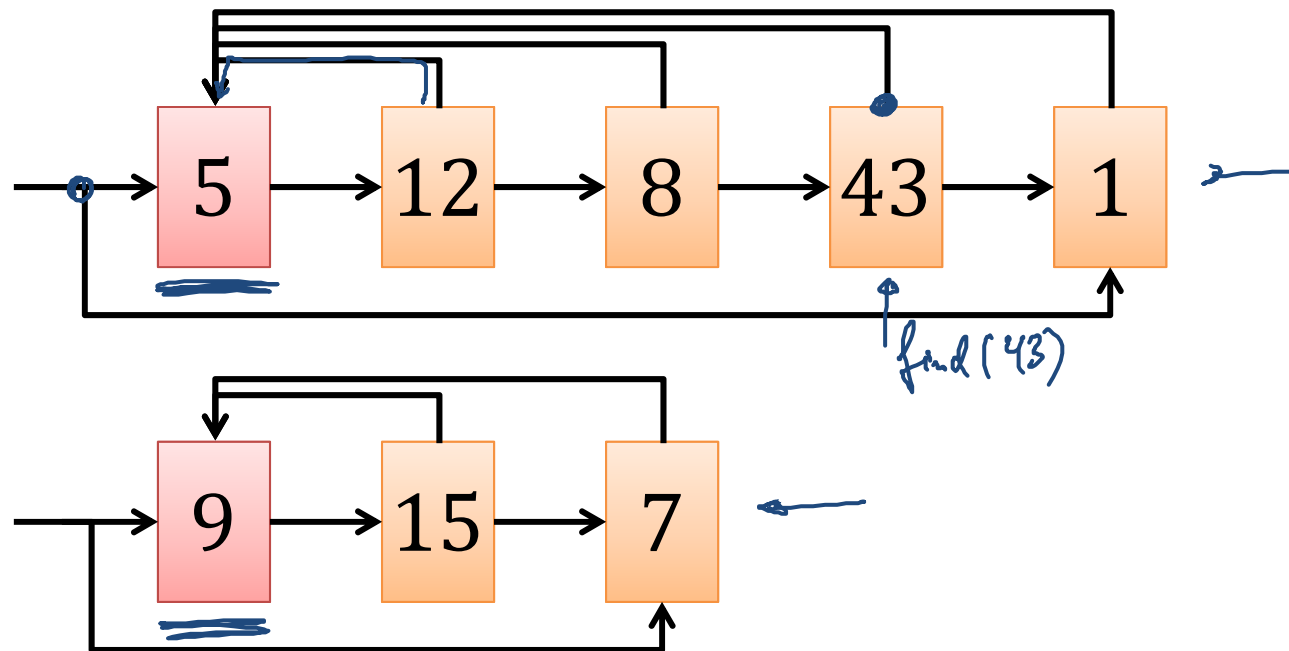
## Remark:

- We assume that the  $n$  make\_set operations are the first  $n$  operations
  - Does not really matter...

# Linked List Implementation

Each set is implemented as a linked list:

- representative: first list element (all nodes point to first elem.)  
in addition: pointer to first and last element



- sets: {1,5,8,12,43}, {7,9,15}; representatives: 5, 9

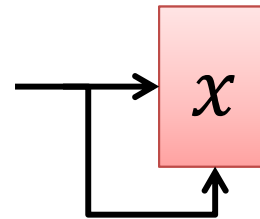


# Linked List Implementation

## make\_set( $x$ ):

- Create list with one element:

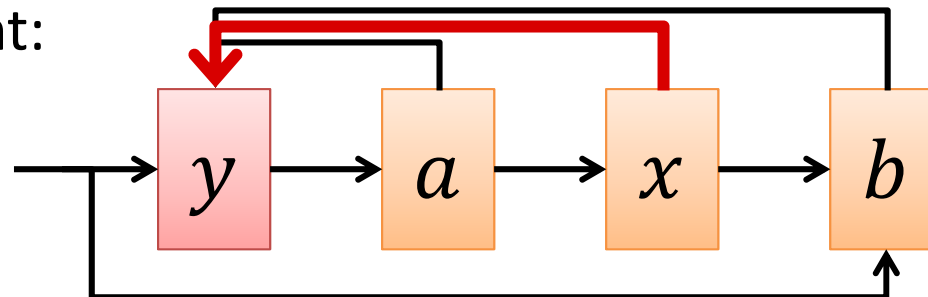
**time:  $O(1)$**



## find( $x$ ):

- Return first list element:

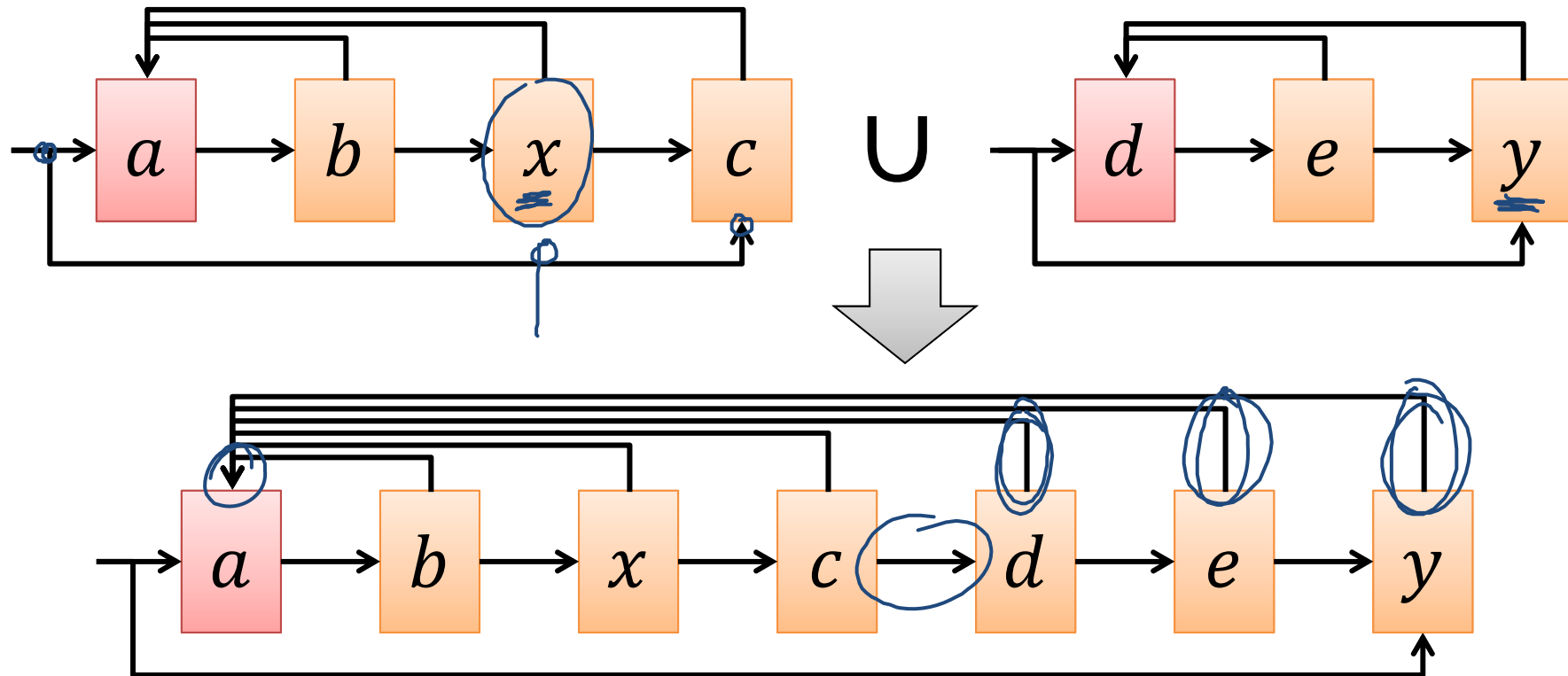
**time:  $O(1)$**



# Linked List Implementation

**union( $x, y$ ):**

- Append list of  $y$  to list of  $x$ :



**Time:  $O(\text{length of list of } y)$**

# Cost of Union (Linked List Implementation)

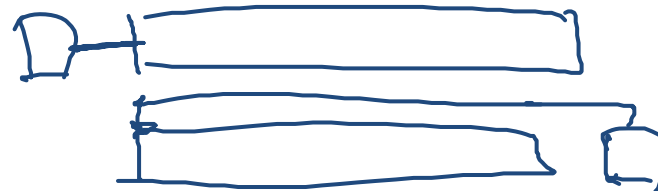


Total cost for  $n - 1$  union operations can be  $\Theta(n^2)$ :

- $\text{make\_set}(x_1), \text{make\_set}(x_2), \dots, \text{make\_set}(x_n),$   
 $\text{union}(x_{n-1}, x_n), \text{union}(x_{n-2}, x_{n-1}), \dots, \text{union}(x_1, x_2)$



redirect  $1 + 2 + \dots + n - 1 = \Theta(n^2)$



# Weighted-Union Heuristic

- In a bad execution, **average cost per union** can be  $\Theta(n)$
- Problem: The longer list is always appended to the shorter one

## Idea:

- In each union operation, append shorter list to longer one!

Cost for union of sets  $S_x$  and  $S_y$ :  $O(\min\{|S_x|, |S_y|\})$

**Theorem:** The overall cost of  $m$  operations of which at most  $n$  are make\_set operations is  **$O(m + n \log n)$** .

$$\underline{O(m + n \log n)}$$

# Weighted-Union Heuristic

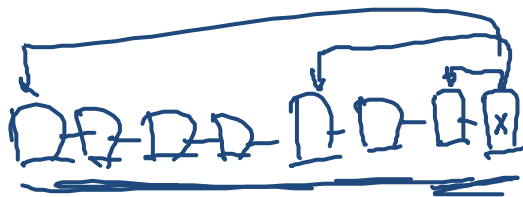
**Theorem:** The overall cost of  $m$  operations of which at most  $n$  are make\_set operations is  $O(m + n \log n)$ .

**Proof:**

make-set, find:  $O(1)$

total union cost  $\approx O(\text{total \# of pointer redirections})$

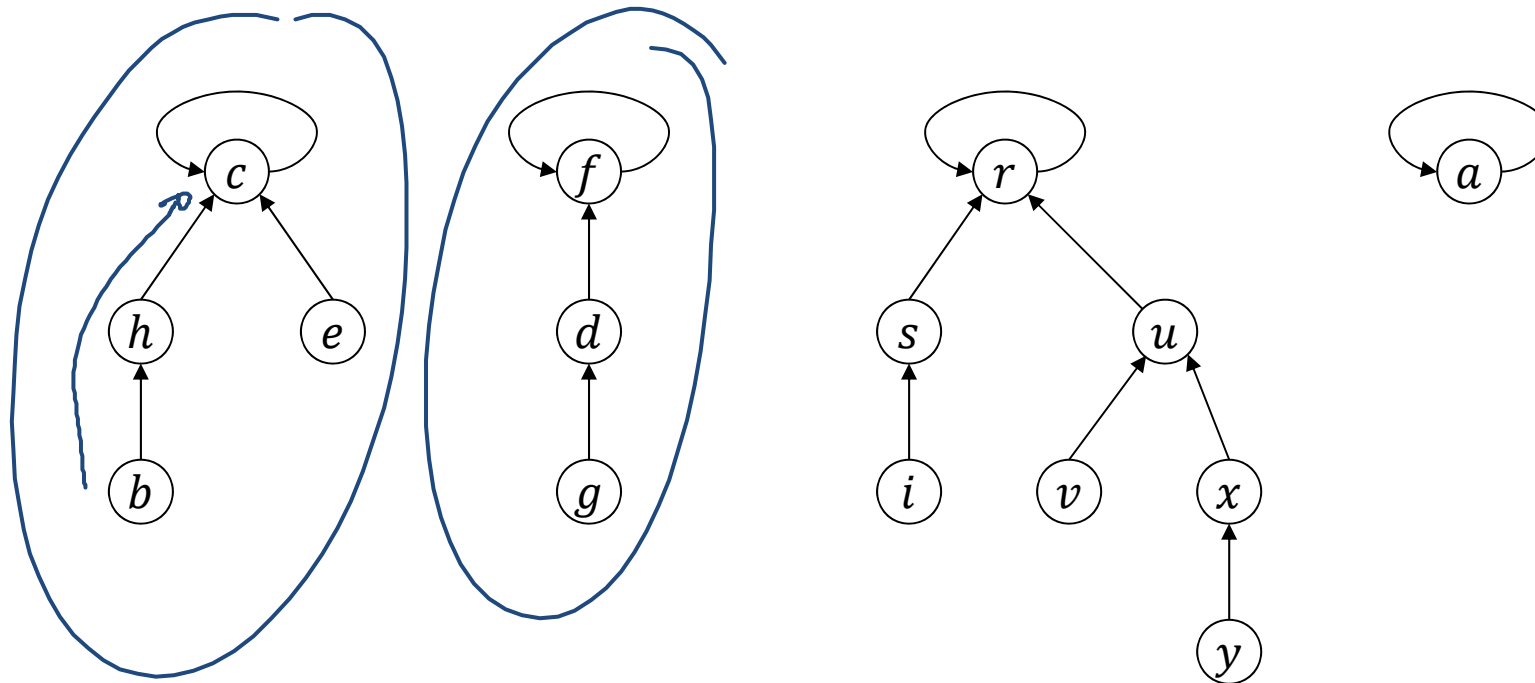
$O(n \cdot \# \text{ redir. per element})$   
 $\leq \log n$



after  $k$  redir. of  $x$ 's pointer  
 set of  $x$  has  $\geq 2^k$  elements

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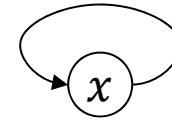
# Disjoint-Set Forests



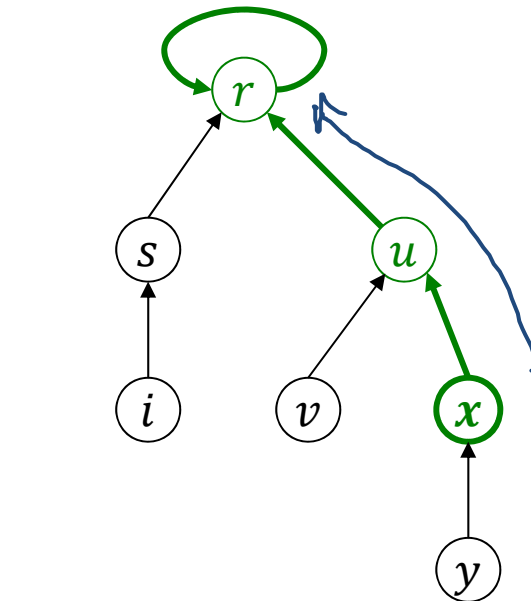
- Represent each set by a tree
- Representative of a set is the root of the tree

# Disjoint-Set Forests

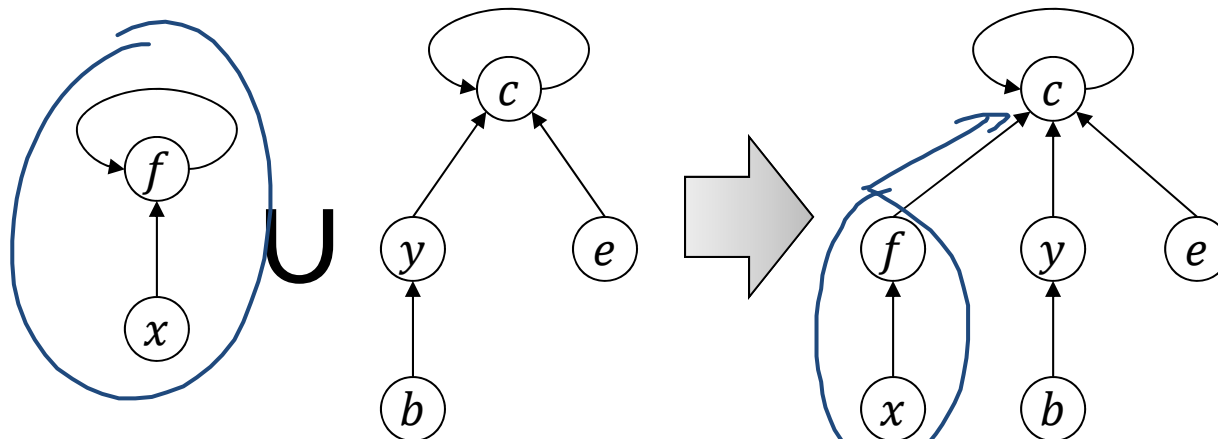
**make\_set(x)**: create new one-node tree



**find(x)**: follow parent point to root  
(parent pointer to itself)



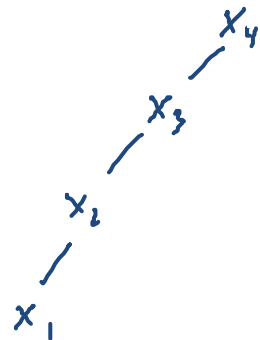
**union(x, y)**: attach tree of x to tree of y



# Bad Sequence

Bad sequence leads to tree(s) of depth  $\Theta(n)$

- $\text{make\_set}(x_1), \text{make\_set}(x_2), \dots, \text{make\_set}(x_n),$   
 $\text{union}(x_1, x_2), \text{union}(x_1, x_3), \dots, \text{union}(x_1, x_n)$





# Union-By-Size Heuristic

## Union of sets $S_1$ and $S_2$ :

- Root of trees representing  $S_1$  and  $S_2$ :  $r_1$  and  $r_2$
- W.l.o.g., assume that  $|S_1| \geq |S_2|$
- **Root of  $S_1 \cup S_2$ :  $r_1$**  ( $r_2$  is attached to  $r_1$  as a new child)

**Theorem:** If the union-by-size heuristic is used, the **worst-case cost of a find-operation is  $O(\log n)$**

**Proof:**

tree with  $k$  elements has depth  $\leq O(\log k)$

**Similar Strategy: union-by-rank**

- rank: essentially the depth of a tree

# Union-Find Algorithms

Recall:  $m$  operations,  $n$  of the operations are make\_set-operations

## Linked List with Weighted Union Heuristic:

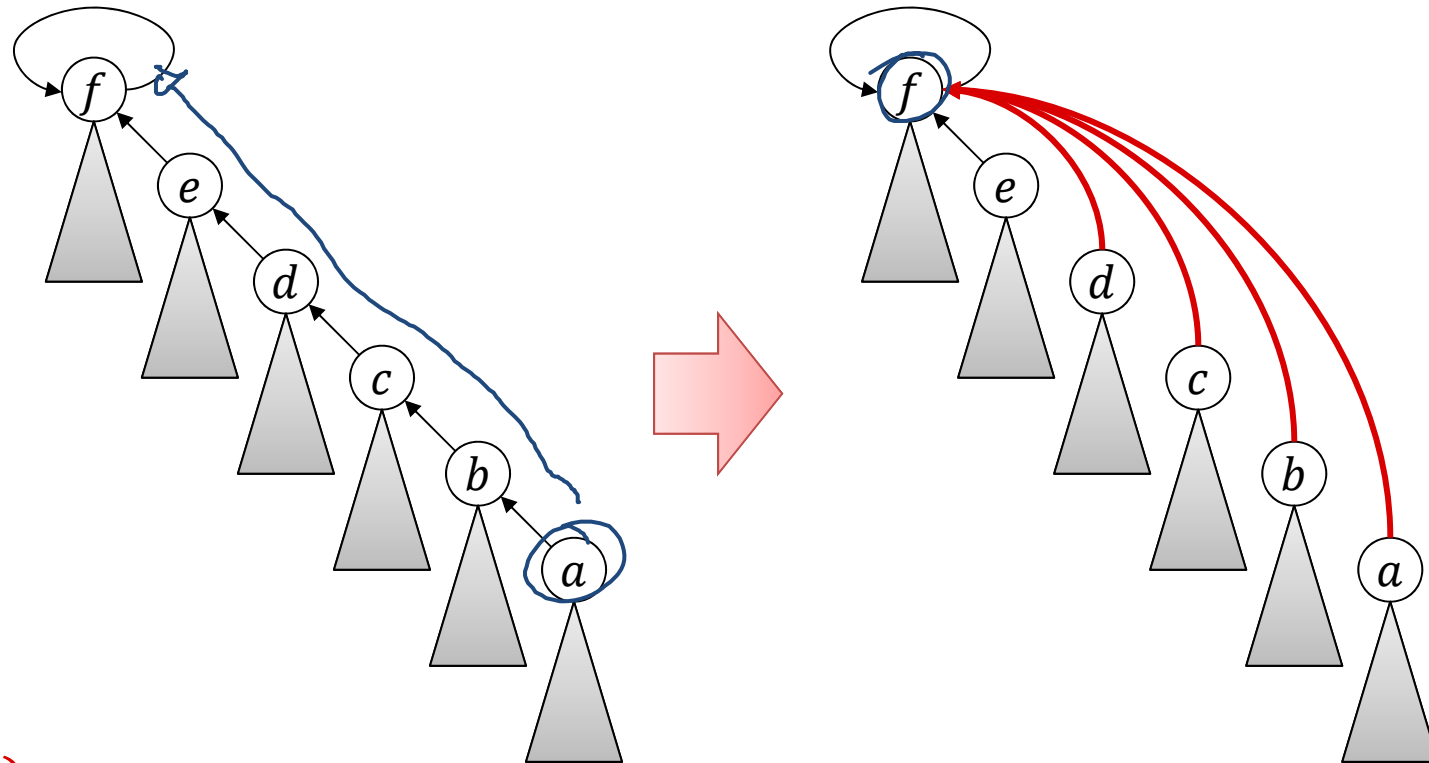
- make\_set: **worst-case** cost  $O(1)$
- find : **worst-case** cost  $O(1)$
- union : **amortized** worst-case cost  $O(\log n)$

## Disjoint-Set Forest with Union-By-Size Heuristic:

- make\_set: **worst-case** cost  $O(1)$
- find : **worst-case** cost  $O(\log n)$
- union : **worst-case** cost  $O(\log n)$

Can we make this faster?

# Path Compression During Find Operation



**find(*a*):**

1. **if**  $a \neq a.\text{parent}$  **then**
2.      $a.\text{parent} := \text{find}(a.\text{parent})$
3. **return**  $a.\text{parent}$

# Complexity With Path Compression

When using only path compression (without union-by-rank):

$m$ : total number of operations

- $f$  of which are find-operations
- $n$  of which are make\_set-operations  
→ at most  $n - 1$  are union-operations

$$\text{Total cost: } O\left(m + f \cdot \underbrace{\left\lceil \log_{2+f/n} n \right\rceil}_{\text{handwritten underline}}\right) = O\left(m + f \cdot \log_{2+m/n} n\right)$$

# Union-By-Size and Path Compression

## Theorem:

Using the combined union-by-rank and path compression heuristic, the running time of  $m$  disjoint-set (union-find) operations on  $n$  elements (at most  $n$  make\_set-operations) is

$$\Theta(\underline{m \cdot \alpha(m, n)}),$$

Where  $\alpha(m, n)$  is the inverse of the Ackermann function.

*grows extremely slowly*  
*in practice  $\leq 4$*

# Ackermann Function and its Inverse

## Ackermann Function:

For  $k, \ell \geq 1$ ,

$$A(k, \ell) := \begin{cases} 2^\ell, & \text{if } k = 1, \ell \geq 1 \\ A(k - 1, 2), & \text{if } k > 1, \ell = 1 \\ A(k - 1, A(k, \ell - 1)), & \text{if } k > 1, \ell > 1 \end{cases}$$

## Inverse of Ackermann Function:

$$\alpha(m, n) := \min\{k \geq 1 \mid A(k, \lfloor m/n \rfloor) > \log_2 n\}$$

# Inverse of Ackermann Function

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- $\alpha(m, n) := \min\{k \geq 1 \mid A(k, \lfloor m/n \rfloor) > \log_2 n\}$   
 $m \geq n \Rightarrow A(k, \lfloor m/n \rfloor) \geq A(k, 1) \Rightarrow \alpha(m, n) \leq \min\{k \geq 1 \mid A(k, 1) > \log n\}$
- $A(1, \ell) = 2^\ell, \quad A(k, 1) = A(k - 1, 2),$   
 $A(k, \ell) = A(k - 1, A(k, \ell - 1))$