



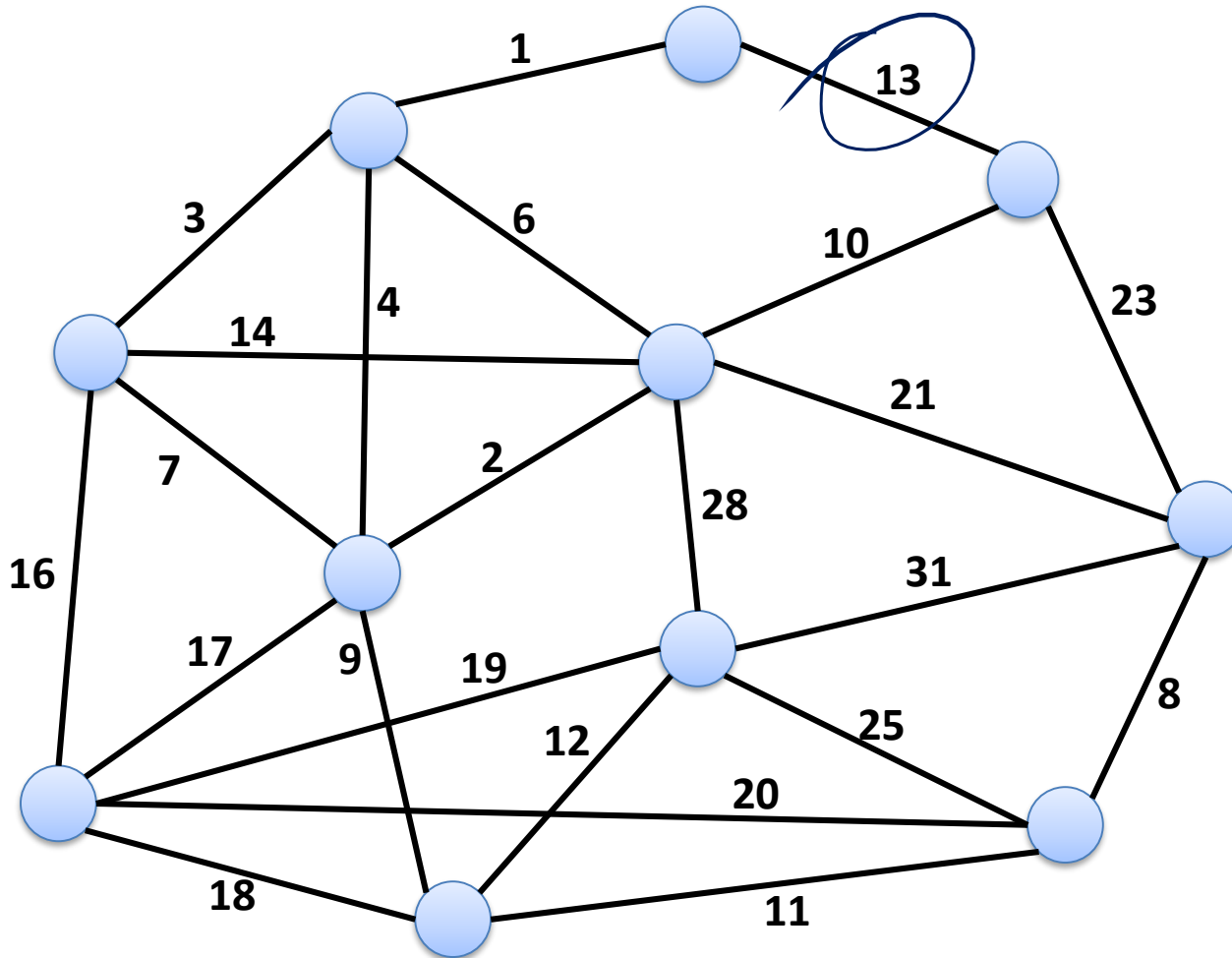
Chapter 5

Data Structures

Algorithm Theory
WS 2018/19

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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

1. Go through edges in order of increasing weights

sort edges by weight

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

(if weights are nice,
this might be faster)

2. For each edge e : $e = \{u, v\}$

if e does not close a cycle then

need to check whether e closes a cycle

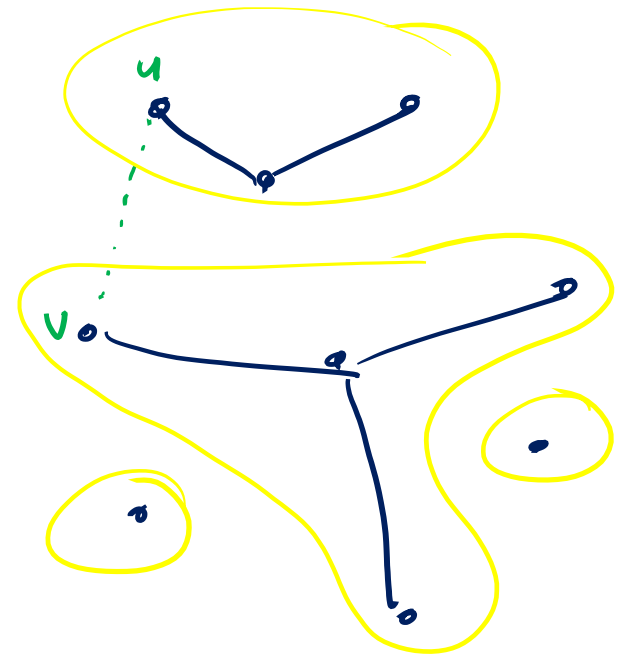


check whether u & v are in the
same component

add e to the current solution

add $\{u, v\}$

need to merge components of u & v

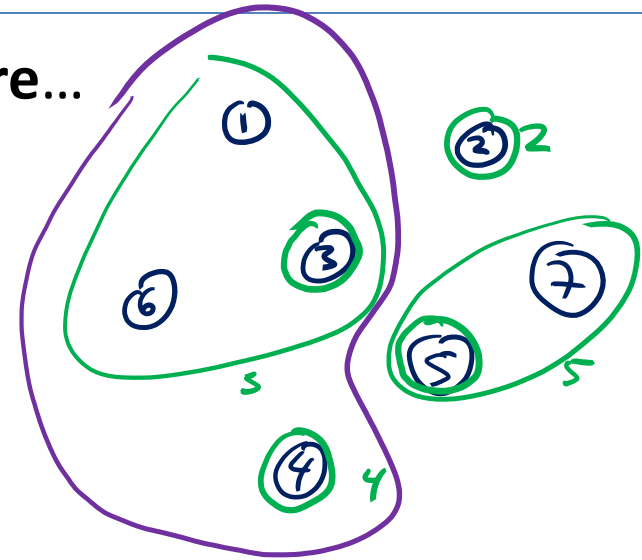


Union-Find Data Structure

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 (sets have identifiers)
- **union(x, y)**: merge the two sets containing x and y

Implementation of Kruskal Algorithm

1. Initialization:

For each node v : make_set(v)

2. Go through edges in order of increasing weights:
Sort edges by edge weight

3. For each edge $e = \{u, v\}$:

if find(u) \neq find(v) then

add e to the current solution

union(u, v)

Managing Connected Components

- 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

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 $S_x \cup S_y$

Throughout the discussion of union-find:

- n : total number of make_set operations u : # elements
- m : total number of operations (make_set, find, and union)

Clearly:

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

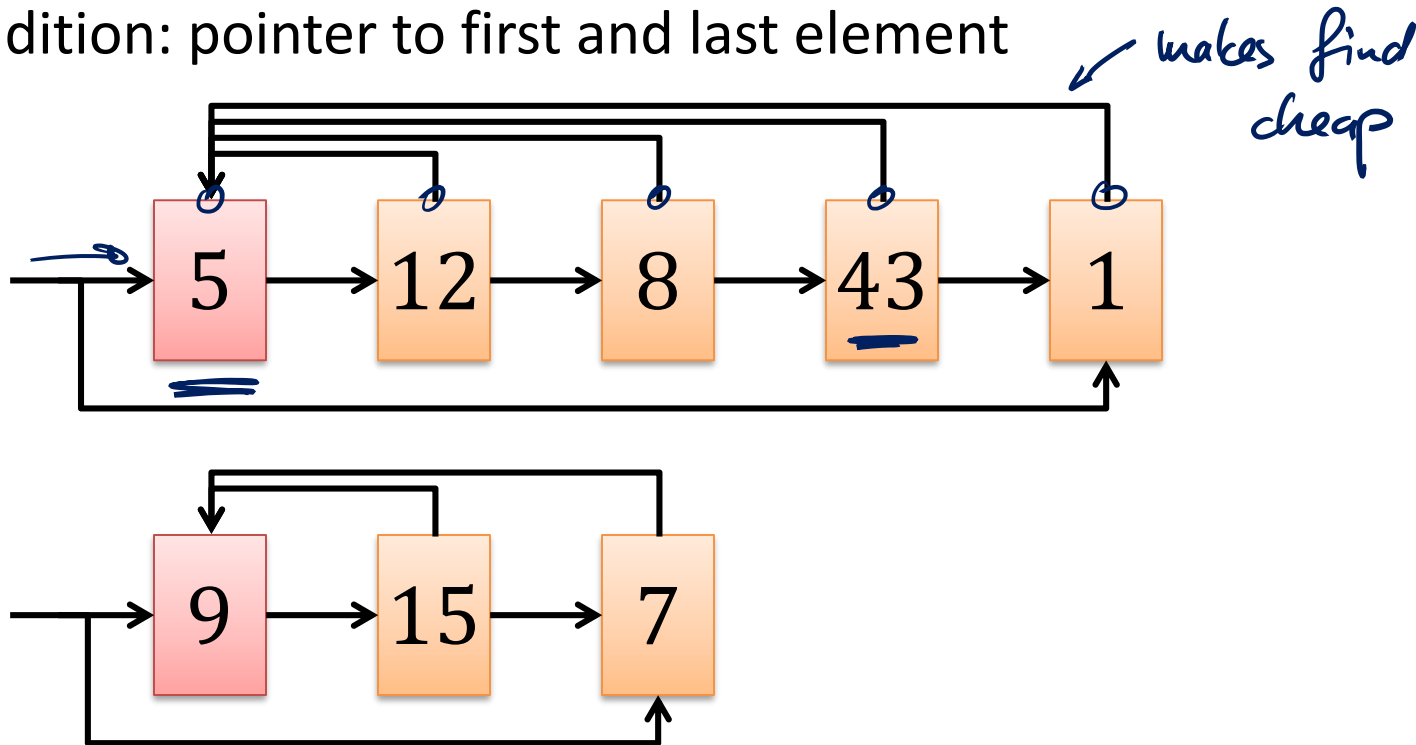
Remark:

- 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



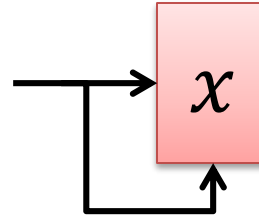
- sets: $\{1, 5, 8, 12, 43\}$, $\{7, 9, 15\}$; representatives: 5, 9

Linked List Implementation

make_set(x):

- Create list with one element:

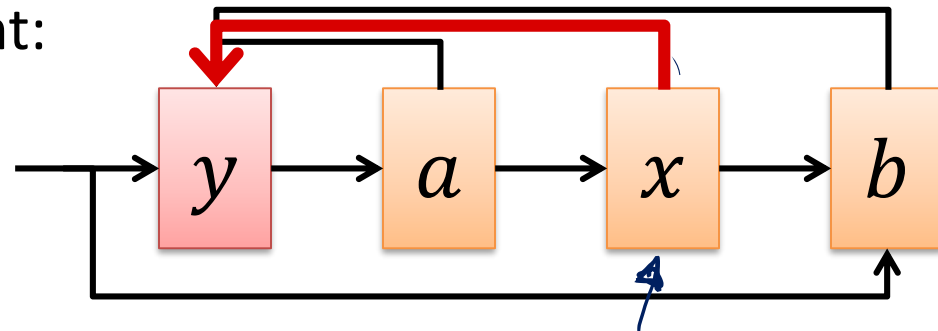
time: $O(1)$



find(x):

- Return first list element:

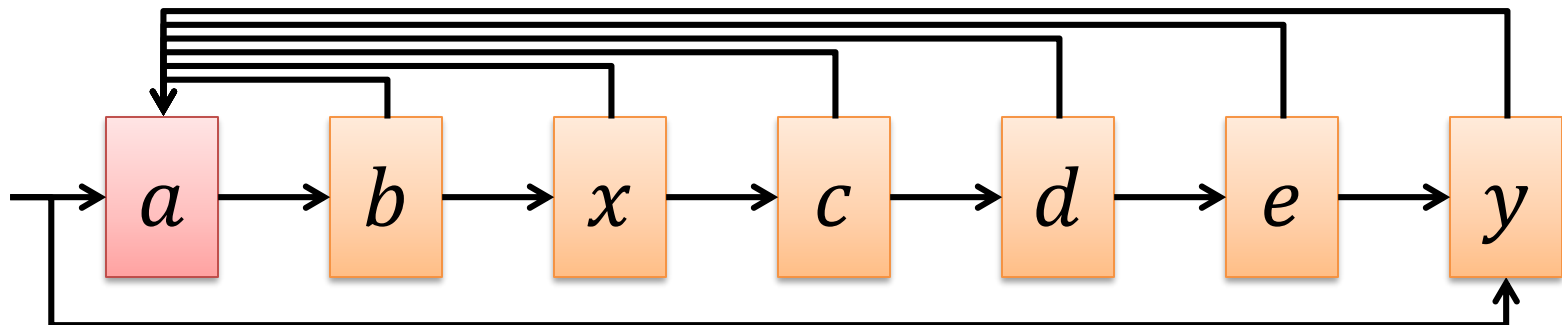
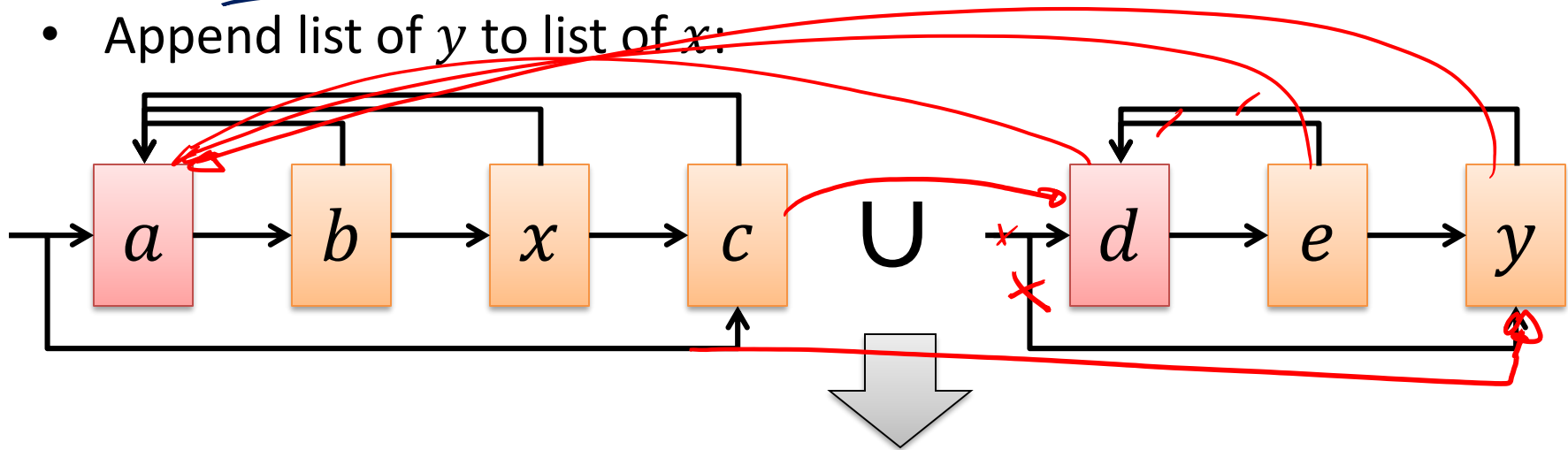
time: $O(1)$



Linked List Implementation

union(x, y):

- Append list of y to list of x :



cost: $O(1) + \#$ pointer redirections to representative

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

Cost of Union (Linked List Implementation)

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

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



pointer redirection: $1 + 2 + 3 + \dots + n - 2 = \Theta(n^2)$

\Rightarrow avg. cost per union: $\Theta(n)$

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

Weighted-Union Heuristic

Theorem: The overall cost of m operations of which at most n are make_set operations is $\underline{O(m + n \log n)}$.

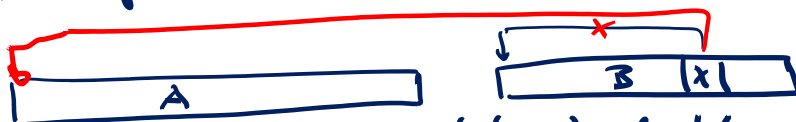
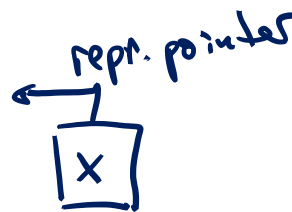
Proof:

total cost of make-set & find operations : $O(m)$
need to bound the total cost of the union operations

= # pointer redirections

consider a fixed element x

How often do we need to redirect the representative pointer of x



Size of set cont. x at least doubles

$\implies \leq \log_2 n$ redirections of repr. pointer of x

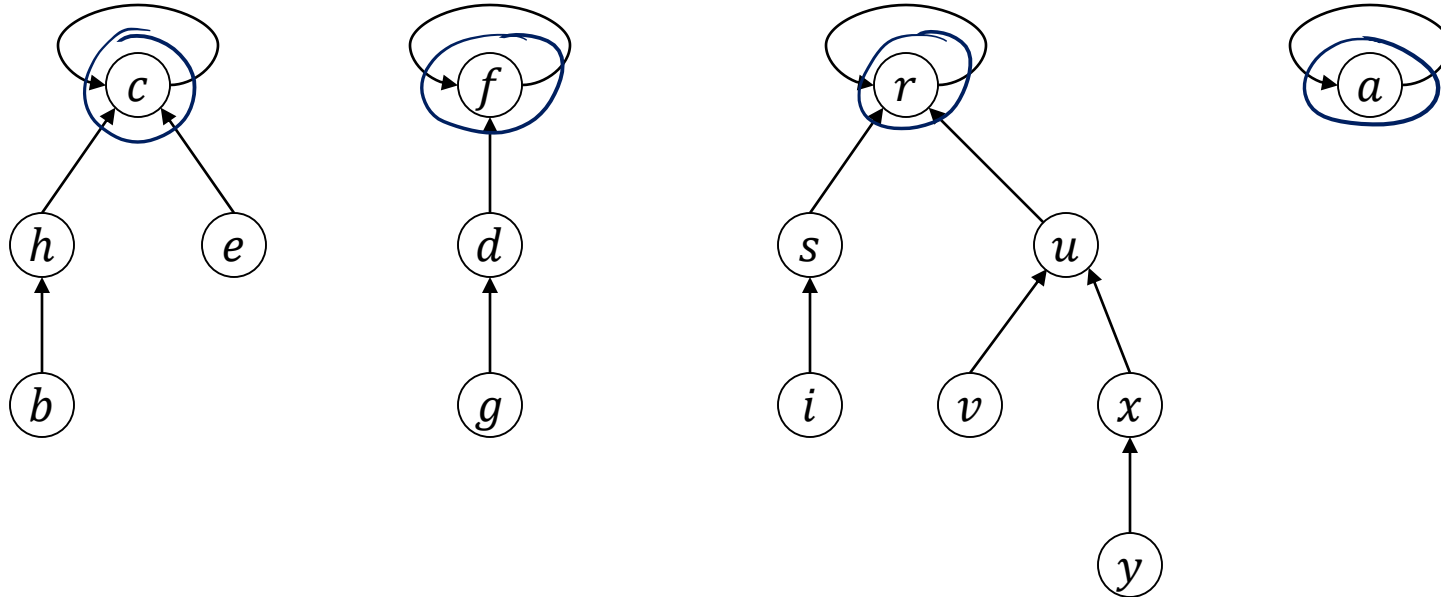
Kruskal's alg:

sorting : $O(m \log n)$

union-find part :

$O(m + n \log n)$

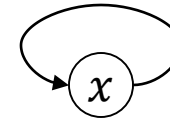
Disjoint-Set Forests



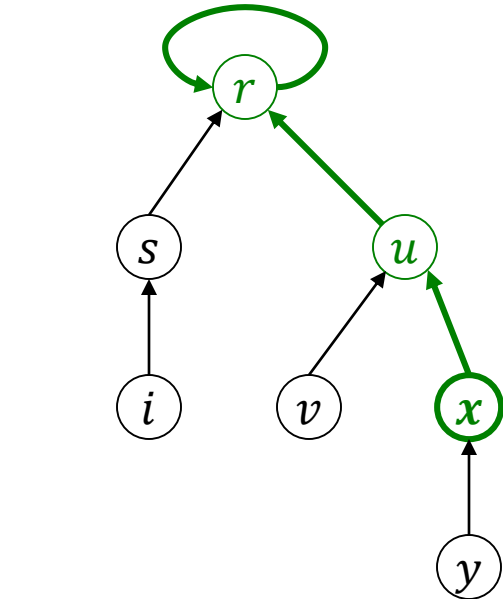
- Represent each set by a tree
- Representative of a set is the root of the tree

Disjoint-Set Forests

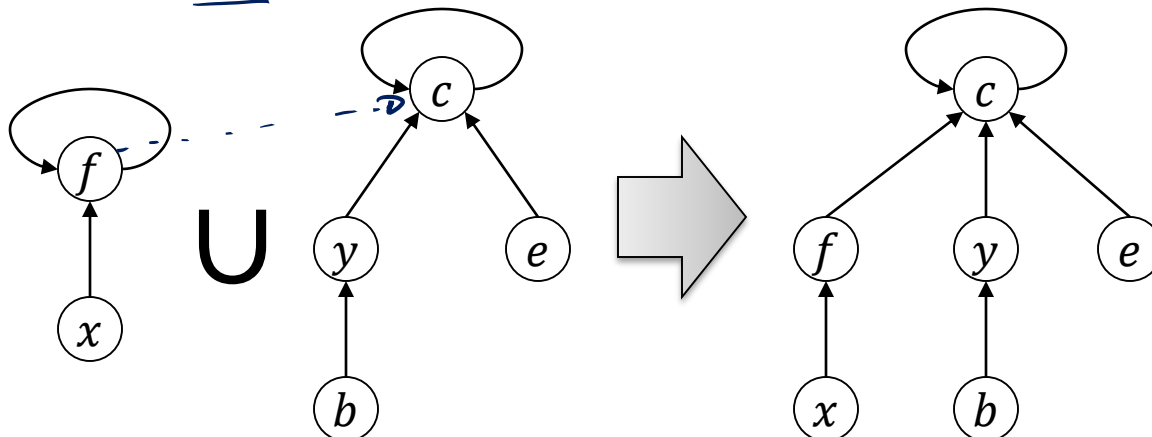
make_set(x): create new one-node tree



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



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



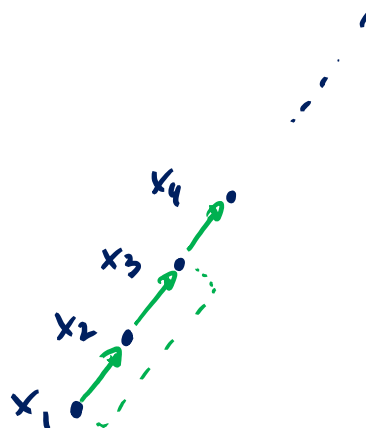
Bad Sequence

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

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



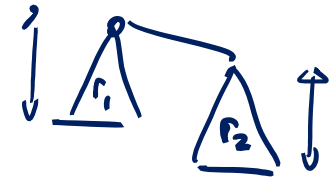
$x_n \cdot$



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: depth of a tree with k nodes is $\leq \log_2 k$
 depth of element x is $d_x \Rightarrow$ size of tree cont. $x \geq 2^{d_x}$
 $d_x = 0 \checkmark$ how can d_x grow?



when d_x grows by 1,
 size of tree cont. x at
 least doubles

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:

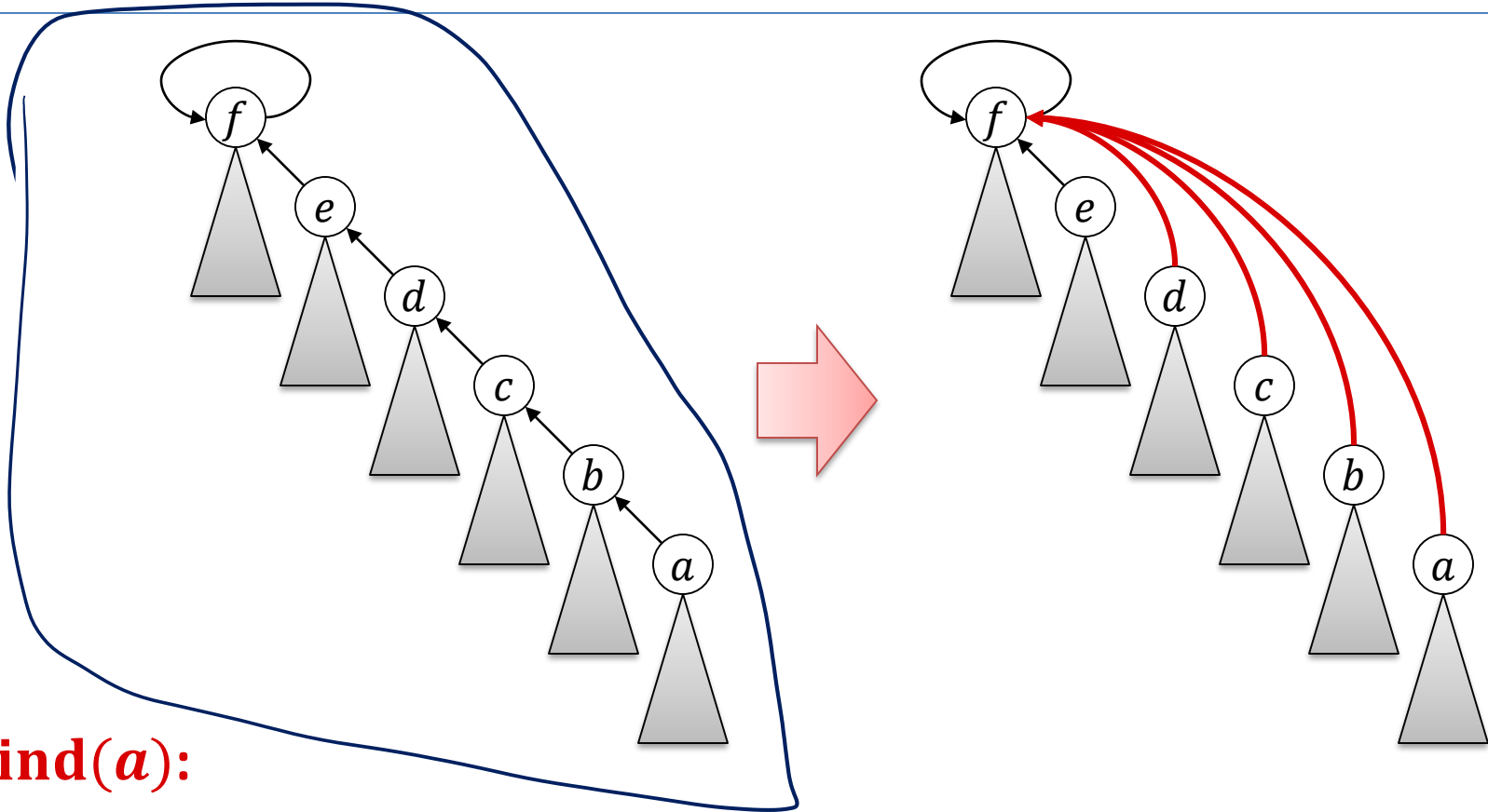
- make_set: **worst-case** cost $O(1)$
- find : **worst-case** cost $O(1)$
- union : **amortized** worst-case cost $O(\log n)$

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

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

$m \gg n$

f large $\Rightarrow f \approx m$

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

if $m \gg n$ (e.g., $m = n^{1.1}$)

$O(m \cdot \underbrace{\log_{n^{0.1}} n}_{= 10})$

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

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

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

↑
grows extremely slowly

↑
grows extremely fast

in practice: $\alpha(m, n) \leq 4$

Kruskal:

sorting: $O(m \log n)$

union-find: $O(m \cdot \alpha(m, n))$

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

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

$$m \geq n \Rightarrow A(k, \lfloor m/n \rfloor) \geq A(k, 1) \Rightarrow \alpha(m, n) \leq \min\{k \geq 1 \mid A(k, 1) > \log n\}$$

- $A(1, \ell) = 2^\ell, \quad A(k, 1) = A(k - 1, 2),$
 $A(k, \ell) = A(k - 1, A(k, \ell - 1))$

$$A(2, 1) = A(1, 2) = 4$$

$$A(3, 1) = A(2, 2) = A(1, A(2, 1)) = A(1, 4) = 2^4 = 16$$

$$\begin{aligned} \underline{A(4, 1)} &= A(3, 2) = A(2, A(3, 1)) = A(2, 16) = A(1, A(2, 15)) = 2^{A(2, 15)} \\ & \qquad \qquad \qquad (A(2, 15) = A(1, A(2, 14))) \qquad \left| \begin{aligned} &= 2^{2^{A(2, 14)}} \\ &\vdots \\ &= 2^{2^{\dots^2}} \end{aligned} \right. \left. \vphantom{A(2, 15)}} \right\} 16 \end{aligned}$$

$$10^{80} \approx 2^{250}$$