



ALBERT-LUDWIGS-
UNIVERSITÄT FREIBURG

Algorithm Theory

17 Union Find

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Union-Find Structures

- ▶ **Problem:**
 - Maintain a collection of disjoint sets while supporting the following operations:
 - **e.make-set():**
 - Creates a new set whose only member is e
 - **e.find-set():**
 - Return the set M_i containing e
 - **union(M_i, M_j):**
 - Unite the sets M_i and M_j into a new set

Union-Find Structures

- ▶ **Representation of set M_i :**
 - M_i is identified by a representative, which is some member of M_i .

Union-Find Structures

- ▶ **Operation using representative elements:**
- ▶ **e.make-set():**
 - Creates a new set whose only member is e. The representative is e.
- ▶ **e.find-set():**
 - Returns the name of representative of the set containing e.
- ▶ **e.union(f):**
 - Unites the sets M_e and M_f that e and f into a new set M and returns a member $M_e \cup M_f$ as the new representative of M. The sets M_e and M_f are erased.

Observations

- ▶ If n is the number of the **make-set** operations and m the total number of **makes-set**, **find-set** and **union** operations, then
 - $m \geq n$
 - after at most $(n - 1)$ **union** operations, only one set remains in the collection

Application: Connected Components

Input: Graph $G = (V, E)$

Output: collection of the connected components of G

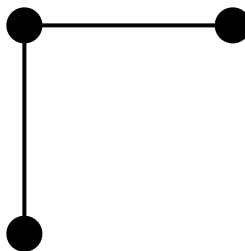
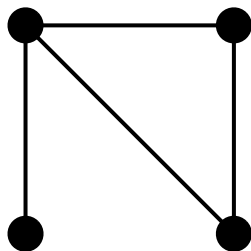
Algorithm: Connected-Components

for all v **in** V **do** $v.makeset()$

for all (u, v) **in** E **do**

if $u.findset() \neq v.findset()$

then $u.union(v)$

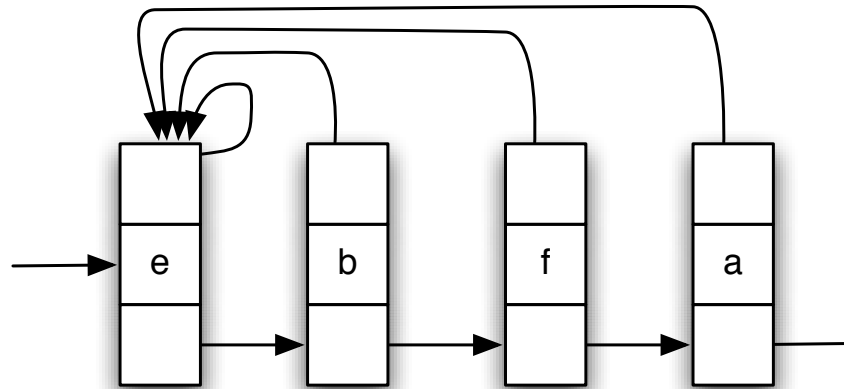


Same-Component (u, v) :

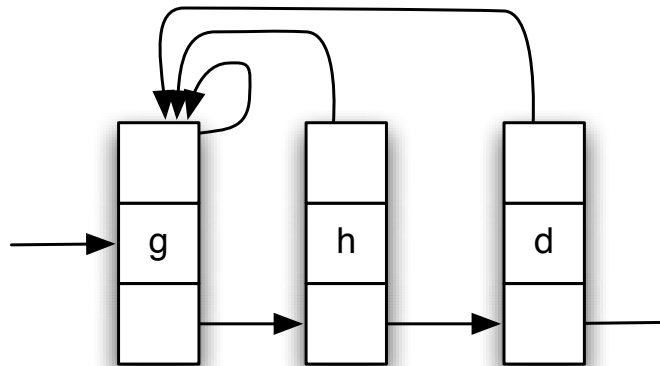
if $u.findset() = v.findset()$

then return $true$

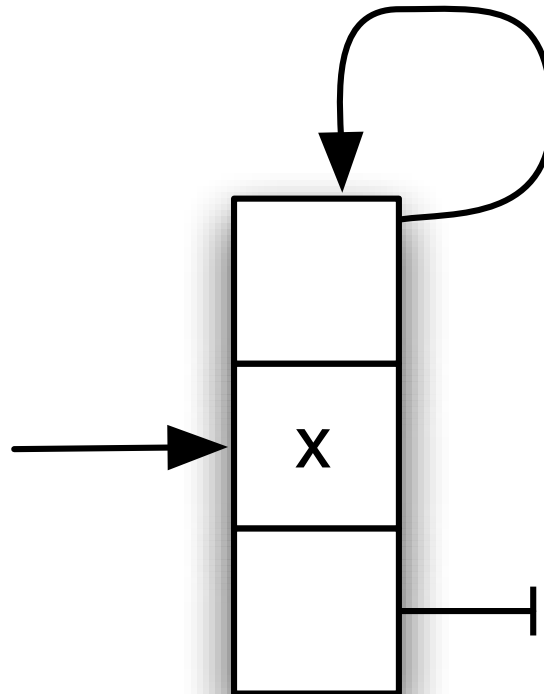
Linked-List Representation



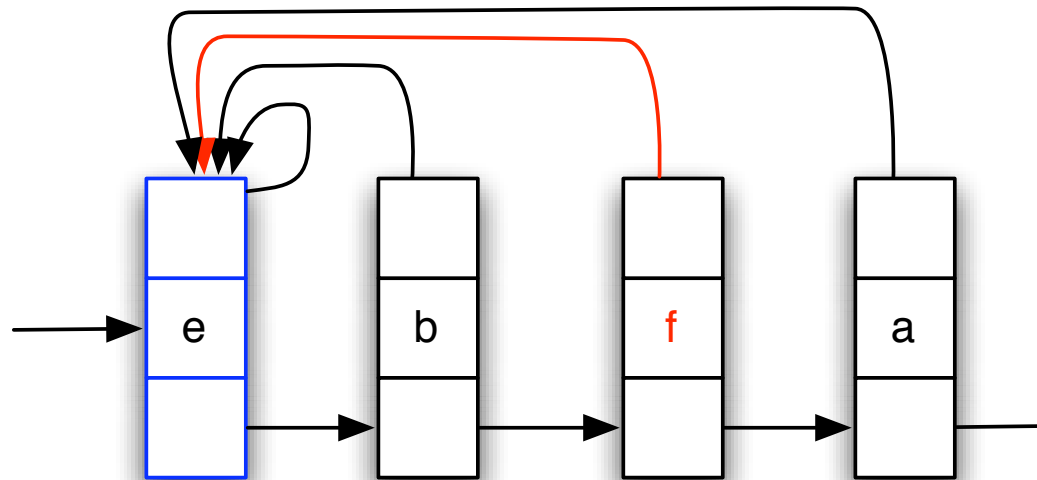
- *x.make-set()*
- *x.find-set()*
- *x.union(y)*



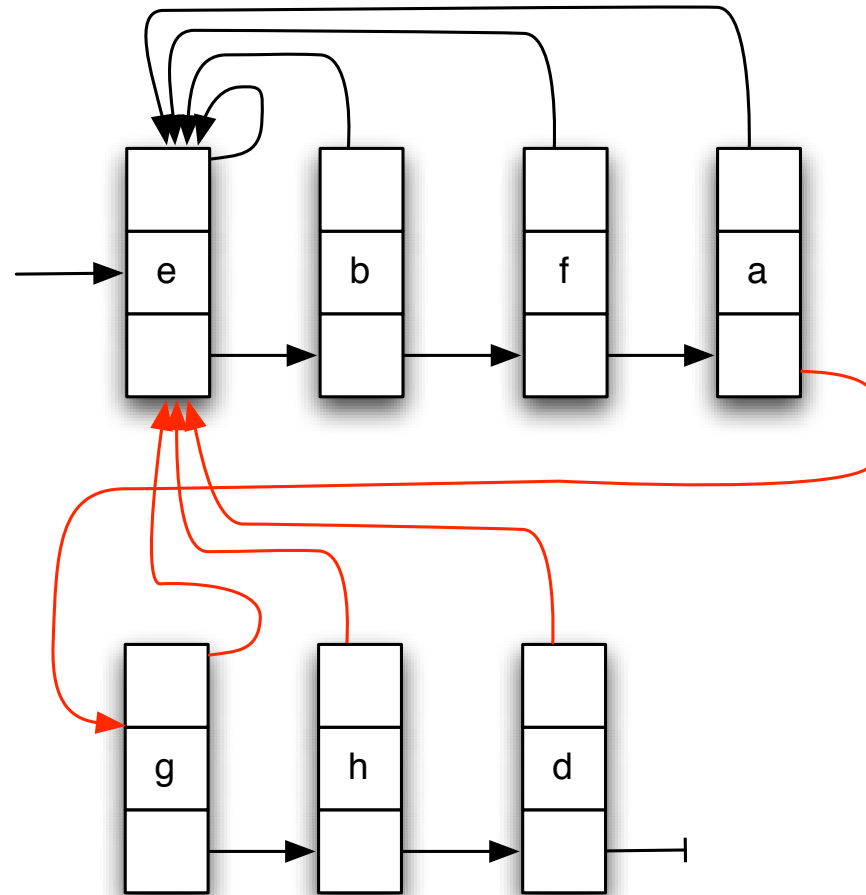
x.makeset()



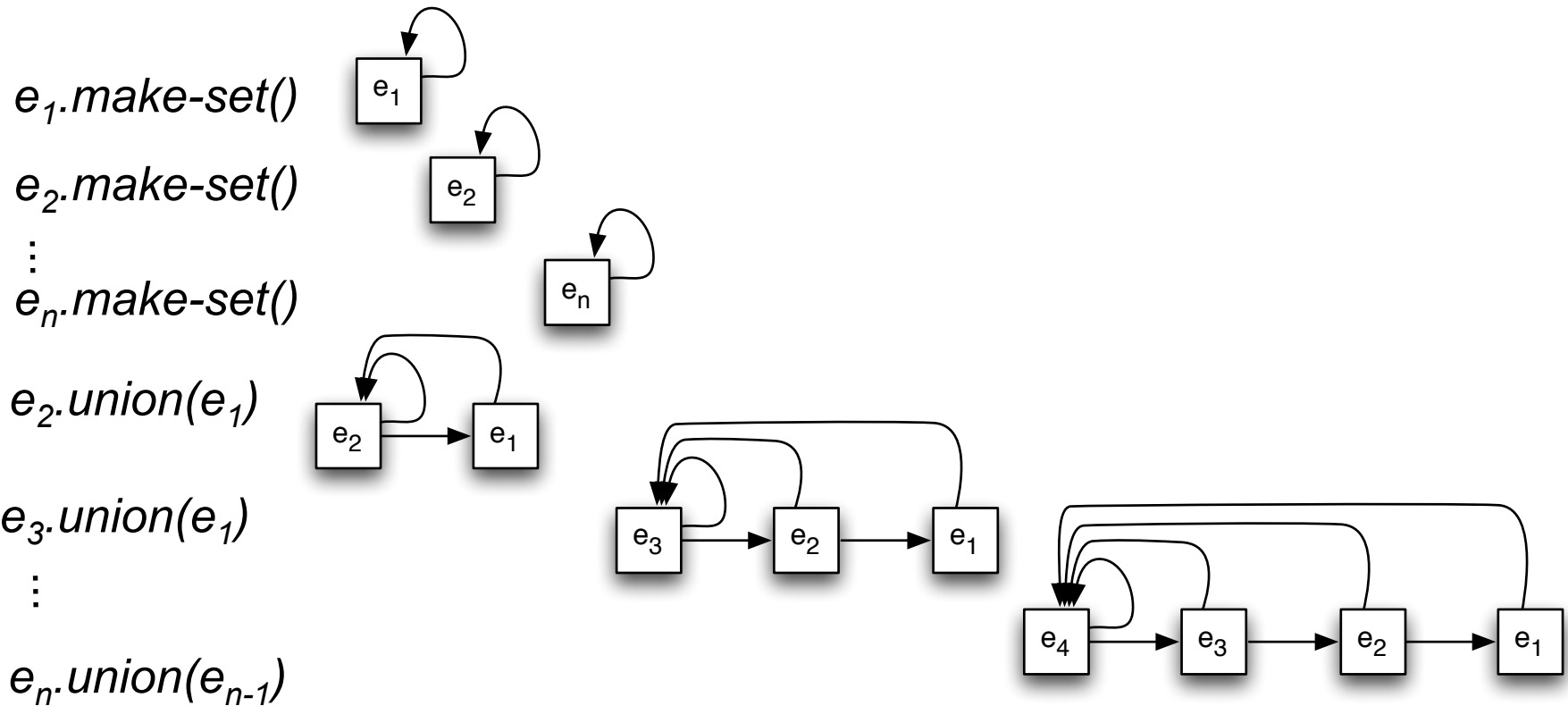
f.find-set()



b.union(d)



Expensive Sequence of Operations



The longer list is always is appended to the shorter list!

Pointer updates for the i -th operation $e_i.union(e_{i-1})$:

Running time of $2n - 1$ operations:

Improvement

▶ Union by Size

- Always append the smaller list to the longer list. (Maintain the length of a list as a parameter)

▶ Theorem

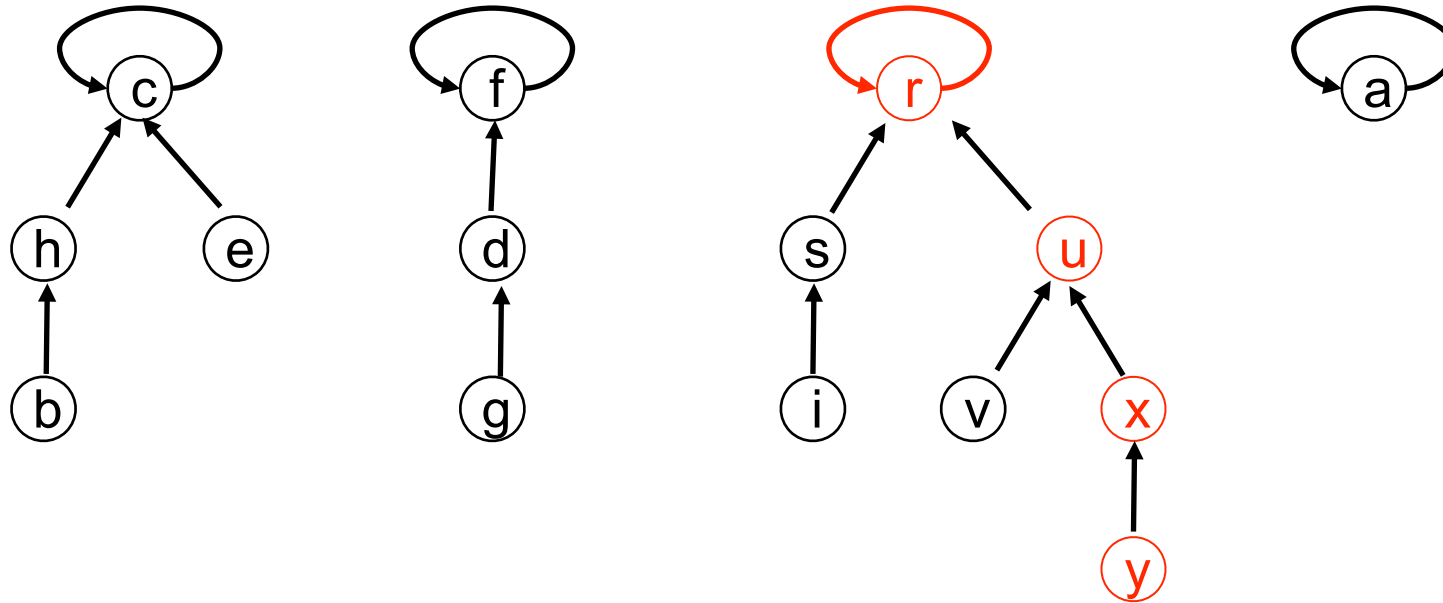
- Using the weighted union heuristic, the running time of a sequence of m make-set, find-set, and union operations, where n are make-set() operations, is $O(m+n \log n)$

▶ Proof

- Consider element e .
- Number of times e 's pointer to the represented is updated: $\log n$

- Whenever an element has to update its pointer it becomes the element of a set with a size which is at least doubled
- because of the weighted union heuristic
- Every element experiences at most $\log n$ union pointer changes

Disjoint Set Forests



- **a.make-set()**: as before
- **y.find-set()**: Follow path upwards
- **d.union(e)**: Make the representative of one set (e.g. f) the parent of the representative of the other set

Example

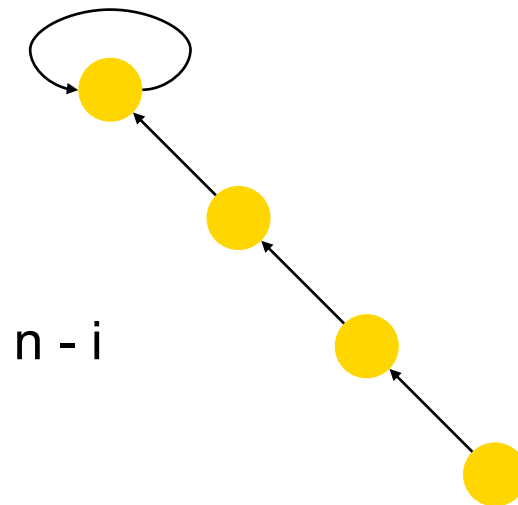
m = total number of operations ($\geq 2n$)

for $i = 1$ to n do e_i .make-set()

for $i = n$ to 2 do e_{i-1} .union(e_i)

for $i = 1$ to f do e_n .find-set()

i - th step



running time of f find-set operations: $O(f n)$

Union by Size

Additional variable:

e.size = (#nodes in the subtree rooted *e*)

e.make-set()

1 *e.parent* = *e*

2 *e.size* = 1

e.union(f)

1 *Link(e.find-set(), f.find-set())*

Union by Size

Link(e,f)

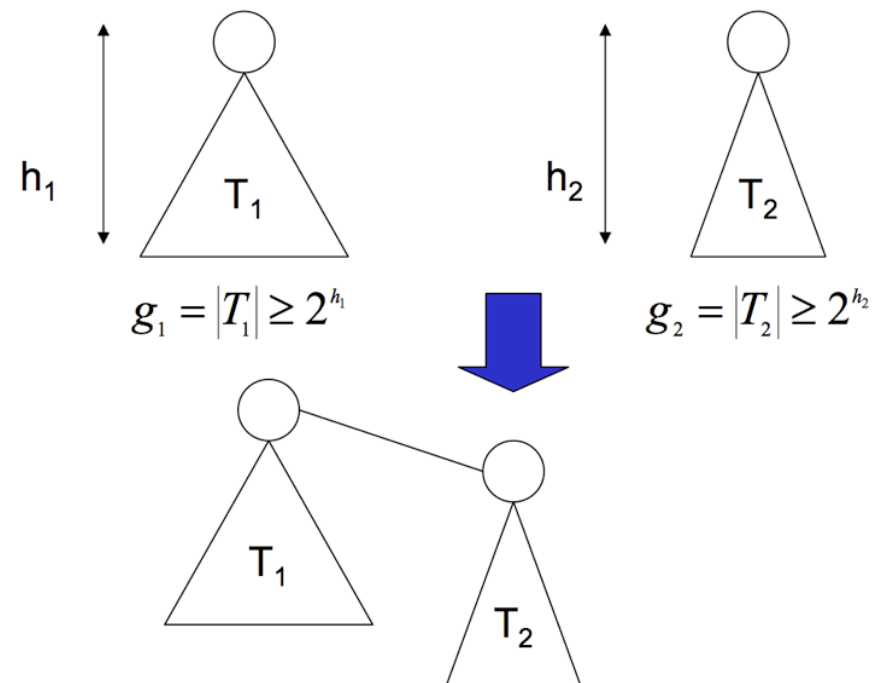
```
1 if e.size ≥ f.size
2   then f.parent = e
3       e.size = e.size + f.size
4   else /* e.size < f.size */
5       e.parent = f
6       f.size = e.size + f.size
```


Union by Size

▶ Theorem

- The method union-by-size maintains the following invariant: A tree of height h contains at least 2^h

▶ Proof



Union by Size

Case 1: The height of the new tree is equal of the height of T_1

$$g_1 + g_2 \geq g_1 \geq 2^{h_1}$$

Case 2 : The new tree T has a greater height

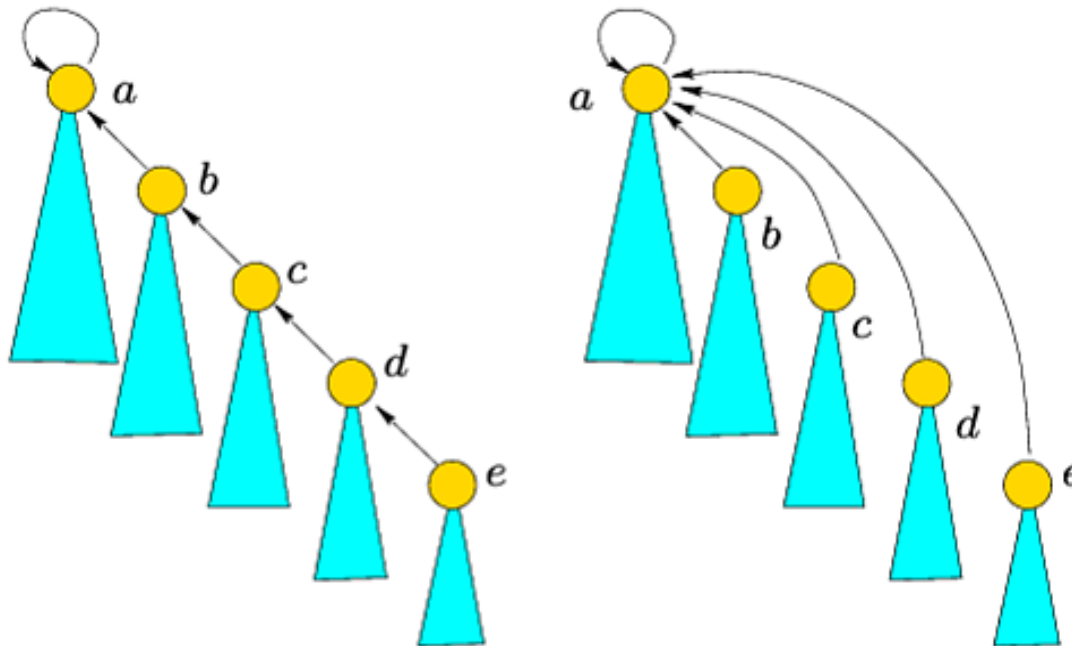
Height of T : $h_2 + 1$

$$g = g_1 + g_2 \geq 2^{h_2} + 2^{h_2} = 2^{h_2+1}$$

Consequence

The running time of find-set operation is $O(\log n)$, where n is the number of make-set operations.

Path Compression during Find-Set Operations



e.find-set()

- 1 **if** $e \neq e.parent$
- 2 **then** $e.parent = e.parent.find-set()$
- 3 **return**

Analysis of Running Time

m total number of operations,

f of which are *find-set* operations and

n of which are *make-set* operations

→ at most $n-1$ *union*-operations

Union by size:

$O(n + f \log n)$

Find-set operations with path compression:

if $f < n$, $\Theta(n + f \log n)$

if $f \geq n$, $\Theta(f \log_{1+f/n} n)$

Analysis of the Running Time

Theorem (Union by size with path compression)

Using the combined union-by-size and path-compression heuristic, the running time of m disjoint-set operation of n elements is $\Theta(m * \alpha(m,n))$

where $\alpha(m,n)$ is the inverse of the Ackermann's function

Ackermann Function und Inverse

Ackermann's function

$$A(1, j) = 2^j \quad \text{for } j \geq 1$$

$$A(i, 1) = A(i - 1, 2) \quad \text{for } i \geq 2$$

$$A(i, j) = A(i - 1, A(i, j - 1)) \quad \text{for } i, j \geq 2$$

Inverse Ackermann's function

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

Ackermann's function and its Inverse

$$A(i, \lfloor m/n \rfloor) \geq A(i, 1)$$

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

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

$$A(4, 1) = A(3, 2) = A(2, A(3, 1)) = A(2, 16) \\ \geq 2^{2^{2^2}} = 2^{65536}$$

$$\alpha(m, n) \leq 4, \text{ für } \log n < 2^{65536}$$



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