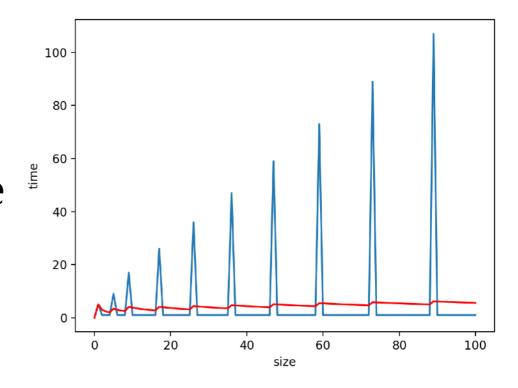
# CSE202 Design and Analysis of Algorithms

Week 9 — Amortization

# Various Kinds of Complexity Analysis

Worst-case: bound the worst-case scenario.

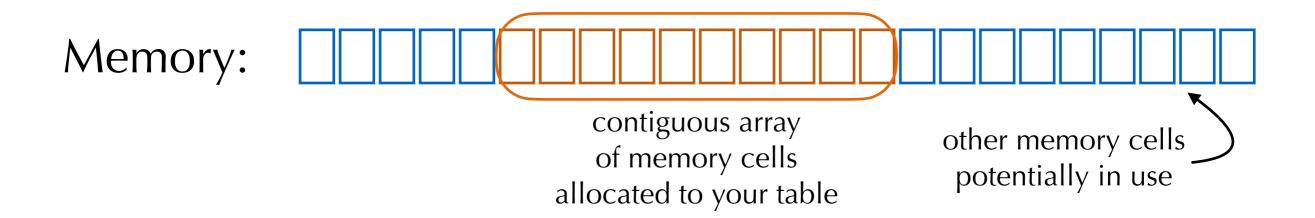
Amortized: average the worst-case over a sequence of operations.



Average-case: average complexity over random inputs or random executions.

# I. Dynamic Tables

# Tables in Low-Level Languages



Increasing the size of the table requires:

allocating a new array of memory; copying the old array to the new one.

Complexity linear in the size of the array

## **Dynamic Tables**

Use three fields: size, capacity, pointer to the array.

This is how lists are implemented in Python

```
def init (self):
    self.size = 0
    self.capacity = 0
    self.table = []
def __getitem__(self,i):
    if i>=self.size: raise IndexError
    return self.table[i]
def __setitem__(self,i,v):
    if i>=self.size: raise IndexError
    self.table[i] = v
def append(self,v):
    n = self.size
    self.resize(n+1)
    self.table[n] = v
def resize(self,newsize):
    if newsize>self.capacity:
      self.realloc((int)(α*newsize))
    self.size=newsize
```

Capacity is increased faster than size

Choice of  $\alpha > 1$ : after the analysis

In Python  $\alpha \approx 9/8$ 

Worst-Case cost of append: O(size)

Simplified & Pythonized C-code

## Amortized Cost of a Sequence of Append

Sequence of capacities:

$$t_{k+1} = [\alpha(t_k + 1)], \quad t_0 = 0.$$

A=[]
for i in range(N):
 A.append(1)

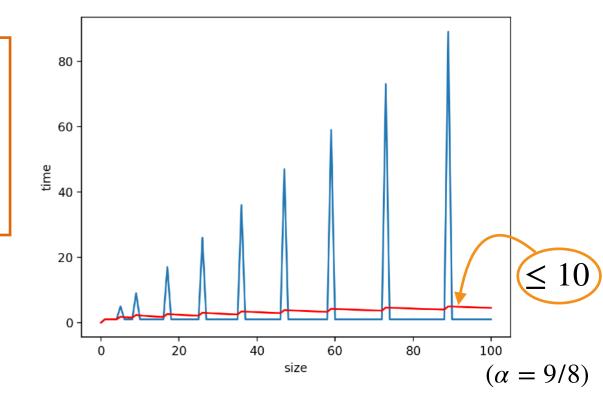
Total cost for N append:  $C_N \le N + \sum_{t_k \le N} t_k$ .

**Thm**. Amortized cost bounded by

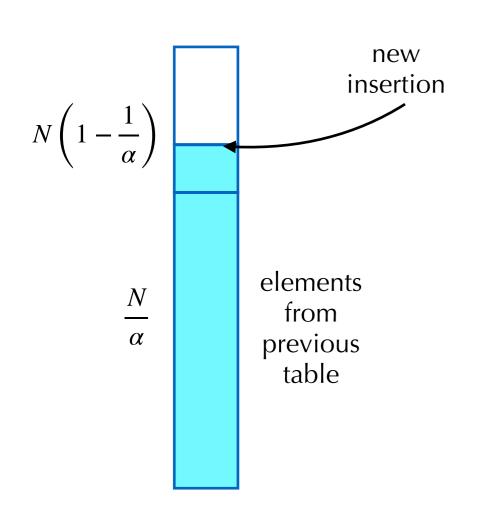
$$C_N/N \le 1 + \frac{\alpha}{\alpha - 1}$$
.

Proof on the blackboard.

A larger  $\alpha$  lowers the constant, but penalizes small tables.



# Interpretation by Accounting Method



When a new element is inserted, it is charged:

1 for its own insert

1 for its future copy when the table next grows

 $1/(\alpha - 1)$  for its share of the future copy of the previous table

Total: 
$$1 + 1 + \frac{1}{\alpha - 1} = 1 + \frac{\alpha}{\alpha - 1}$$
. Exercise: make this rigorous

using the actual  $(t_k)$ .

The cost of future copies is prepaid.

#### **Deletion**

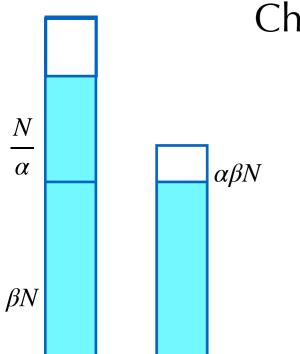
Retrieve memory when the size of the table decreases

#### Dangerous scenario:

increase by a factor  $\alpha$  when full; decrease by a factor  $1/\alpha$  when possible.

Append  $t_m$  times, then ADDAADD.. copies too often.

Solution: leave space to prepay for the next growth.



Charge for Insert unchanged:

$$1 + \frac{\alpha}{\alpha - 1}$$

Charge for Delete:

$$1 + \frac{\alpha\beta}{1 - \alpha\beta}$$

```
def pop(self):
    if self.size==0: raise IndexError
    res = self.table[self.size]
    self.resize(self.size-1)
    return res

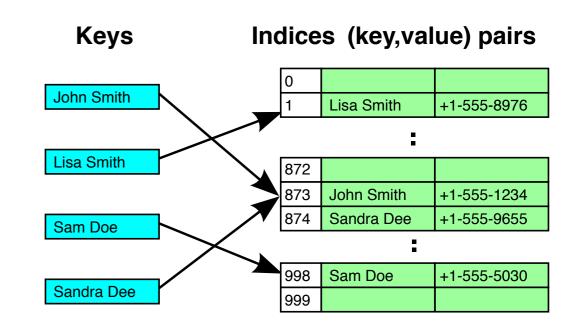
def resize(self,newsize):
    if newsize > self.capacity or \
        newsize < self.capacity/2>
        self.realloc((int)(α*newsize))
    self.size=newsize
```

Amortized cost O(1) per operation.

## **Application to Hash Tables**

Hash tables with linear probing require a filling ratio bounded away from 1.

Implemented with dynamic tables.



Resizing the table requires to rehash all the entries.

In Python, the hash function is computed once as a 64-bit integer, and stored with the object. Only its value mod the new size is recomputed.

## II. Union-Find

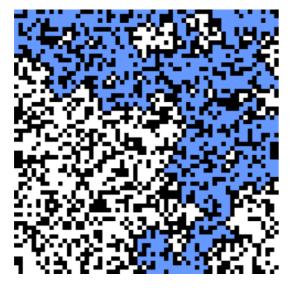
### Recall Union-Find (CSE103)

Abstract Data Type for Equivalence Classes

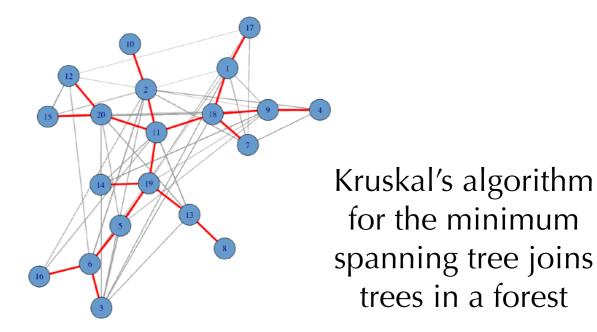
Main operations:

Find(p): identifier for the equivalence class of p

Union(p, q): add the relation  $p \sim q$ 



Connected components in a graph as equivalence classes



## **Forests in Arrays**

#### 2 3 2 3 106 6 6 106 2 11

```
p[i] := parent(i)
(init with p[i] := i)
```

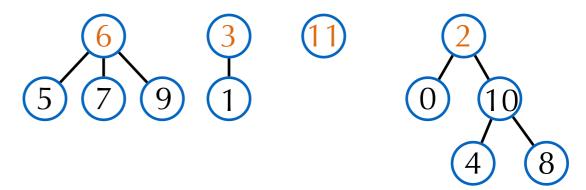
#### First version

```
def find(p,a):
    while p[a]!=a: a=p[a]
    return a

def union(p,a,b):
    link(p,find(p,a),find(p,b))

def link(p,a,b):
    p[a]=b
```

Only find uses more than O(1) array accesses



current equivalence classes

Worst-case:

```
for i in range(N):
    union(p,0,i)
```

uses  $O(N^2)$  array accesses

# Union by Rank

Maintain rank (=height). Link short trees to higher ones.

```
def link(p,a,b):
    if a == b: return
    if rk[b]>rk[a]: p[a]=b
    else: p[b]=a
    if rk[a]==rk[b]: rk[a]+=1
```

Starting from









```
0 \sim 1

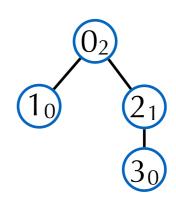
2 \sim 3 produce

0 \sim 3 successively
```







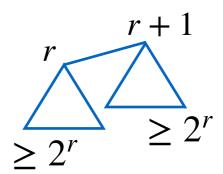


Exercise: join this tree to another one of its shape

#### Properties.

- . rank increases from leaf to root;
- . size of tree  $\geq 2^{\text{rank(root)}}$ ;
- . num nodes of rank  $r \le n/2^r$  .

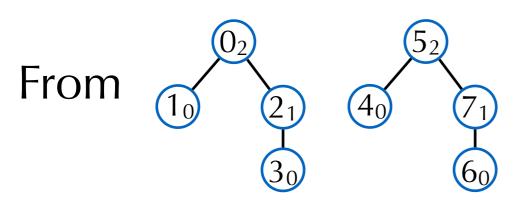
Proof by induction.



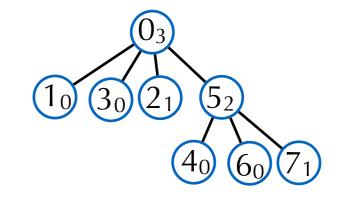
 $\Longrightarrow$  Worst case for find:  $O(\log n)$ .

# **Path Compression**

Every find branches all the nodes it visits to their root.



 $3 \sim 6$  gives



```
def find(p,a):
   if p[a]!=a: p[a]=find(p,p[a])
   return p[a]
```

Preserves the properties of rank (becomes an upper bound on height)

Worst-case for find unchanged.

**Thm**. A sequence of  $m \ge n$  union or find operations uses  $O(m \log^* n)$  array accesses.

S log Proof next

4 slides.

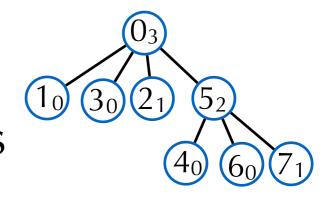
log\*n: number of iterations of log₂ before reaching ≤1. log\*2 = 1, log\*4 = 2, log\*16 = 3, log\*65536 = 4, log\*10<sup>19000</sup> = 5.

Constant in practice.

# Strategy for the Amortized Analysis

We analyse a sequence of  $m \ge n$  union or find.

Difficulty in the analysis: a node can change parents several times



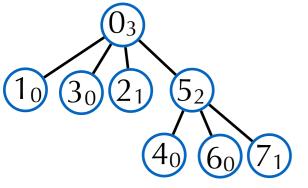
Idea 1: analyze another algorithm with the same cost, easier to handle.

Idea 2: treat high-ranking elements separately, recursively.

#array accesses = O(m + #parent changes)

## Link & Compress

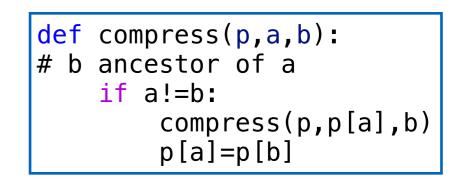
1. Rewrite the sequence of m union or find as a sequence of O(m) link or compress



```
l(0,1), l(2,3), l(0,2), l(5,4),

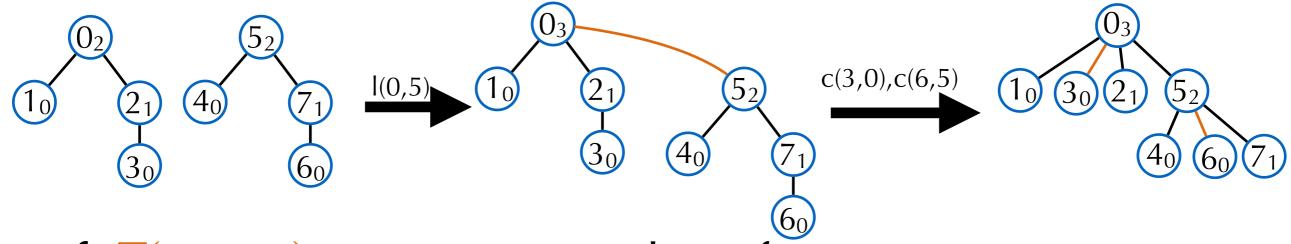
l(7,6), l(5,7), c(3,0), c(6,5), l(0,5)

union(3,6)
```



Links determine the ranks

2. Perform the links first (each in O(1) operations)

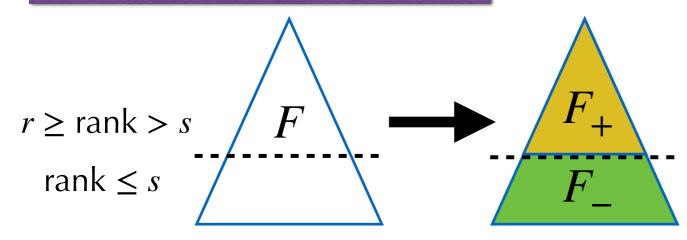


**Def.** T(m, n, r) worst-case number of parent changes in  $\leq m$  compress in a forest of  $\leq n$  nodes, each of rank  $\leq r$ .

Simple bound  $T(m, n, r) \leq nr$ .

## **High and Low Forests**

Idea: Most of the compressions take place in small rank

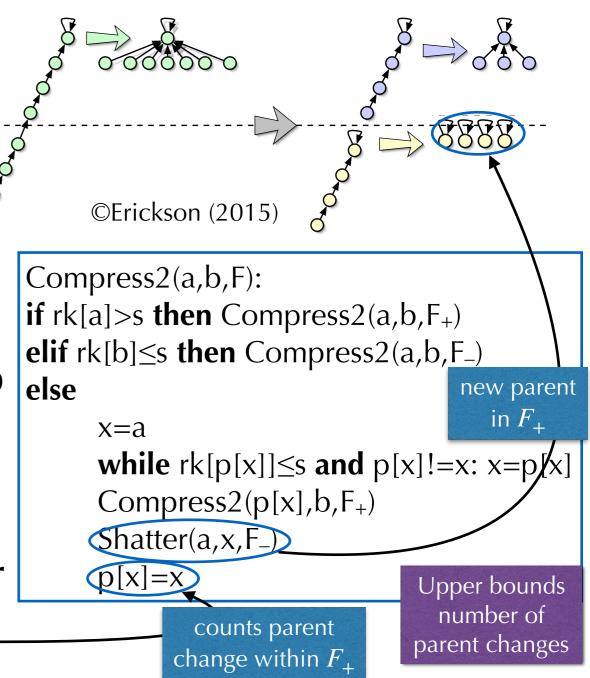


 $m_{-}$  compress purely inside  $F_{-}$ 

$$m_{+} := m - m_{-}$$

C sequence of m compress splits into

 $m_{-}$  compress in  $F_{-}$ , denoted  $C_{-}$   $m_{+}$  compress in  $F_{+}$ , denoted  $C_{+}$   $|F_{-}| \le n$  parent changes in Shatter  $\le m_{+}$  parent changes within  $F_{+}$  Split compress:



$$T(m, n, r) = T(F, C) \le T(F_+, C_+) + T(F_-, C_-) + m_+ + n$$

#### Conclusion

For any sequence C of length  $\leq m$  in a forest with n nodes of rank  $\leq r$ ,

$$T(F,C) - m \le T(F_-,C_-) - m_- + T(F_+,C_+) + n$$

$$rk \le r$$

$$rk \le s$$

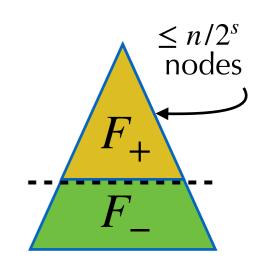
$$\leq rn/2^s$$
by the simple bound 2p. ago

Choose  $s = \log_2 r$ 

$$T(F,C) - m \le T(F_-,C_-) - m_- + 2n$$

$$rk \le r$$

$$rk \le \log_2 r$$



Iterating log\* r times yields

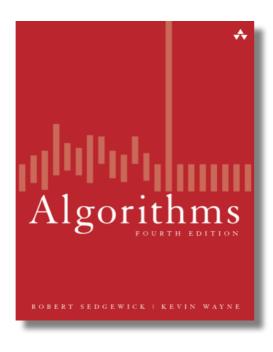
$$T(F,C) \le m + 2n\log^* r = O(m\log^* n) \qquad (m \ge n).$$

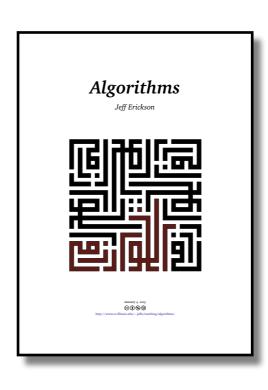
Can be improved further, see assignment.

#### References for this lecture

The slides are designed to be self-contained.

They were prepared using the following books that I recommend if you want to learn more:





#### Next

Assignment: improved amortized complexity of union-find

Next tutorial: experiments with union-find

Next week: Balancing against Worst-Case

#### **Feedback**

Moodle for the slides, TDs and exercises.

Questions or comments: <u>Bruno.Salvy@inria.fr</u>