

# 07. Amortized Analysis and Fibonacci Heaps

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## Amortized Efficiency

### **amortized** performance

- ▶ *amortized efficiency*: average of worst-case time over any sequence of operations
- ▶ different from (probabalistic) expected efficiency, which is averaged over random choices made by a PRNG
- ▶ amortized time bound is a weaker math condition than the same worst-case bound
- ▶ e.g.  $\Theta(1)$  amortized vs.  $\Theta(1)$  worst-case
- ▶  $\implies$  downgrading to an amortized bound may admit upgrades to other aspects of a data structure

## Big Idea in Algorithm Design: Work Smart not Hard

Lifecycle of an implemented algorithm:

### Phase of Life

design and analysis

learned by students

implemented

run and maintained

### Frequency of Creation

once, by discoverer and peer reviewers

annually, by thousands

once per programming language lifetime (decade)

millions-billions of times indefinitely

- ▶ given the choice, we'd prefer for the ongoing tasks to be easy
- ▶ even if that means the one-time phases are complicated
- ▶ holy grail: algorithm is tough to conceive and analyze, but easy to understand and implement
- ▶ example: universal hashing, open addressing

## Amortized Analysis — What to Prove

A *splay tree* is a kind of binary search tree.

Lemma: the *INSERT*, *SEARCH*, and *DELETE* operations each take  $\Theta(\log n)$  amortized time on a splay tree.

Would need to prove:

- ▶ average time/operation =  $\Theta(\log n)$
- ▶ or, any sequence of  $n$  of these operations takes a grand total of  $n \cdot \Theta(\log n) = \Theta(n \log n)$  worst-case time
- ▶ any sequence: includes the worst-case for the data structure
- ▶ three conventional proof techniques

## Aggregate Analysis

- ▶ count total time  $T(n)$  for any sequence of  $n$  operations
- ▶  $\implies$  each operation takes  $T(n)/n$  amortized time
- ▶ pro: simple logic; from first principles
- ▶ con: only works when the goal is to prove all operations take the same time (e.g. *INSERT*, *SEARCH*, and *DELETE* are each  $\Theta(\log n)$ ); wouldn't work if one of them is  $\Theta(1)$
- ▶ con: not much inspiration for difficult analyses

## The Accounting Method

- ▶ inspired by financial amortization, balanced budgets
- ▶ analyst picks a **cost** for each operation (ex.: *INSERT* costs  $4 \cdot \lceil \log_2 n \rceil$  units)
- ▶ some operations **over-charge** and turn a **profit** that is deposited somewhere in the structure (e.g. in a node)
- ▶ other operations **under-charge** and incur a **loss** that is withdrawn from prior deposits
- ▶ show: never go bankrupt, i.e. withdrawn units always exist
- ▶ show: for every operation,

$$(\text{charge}) + (\text{withdrawal}) \leq \text{actual time spent}$$

- ▶  $\implies$  amortized time = charged cost

## Accounting Method Pros/Cons

- ▶ pro: possible to prove different amortized efficiency classes for each operation (e.g. one is  $\Theta(1)$ , another  $\Theta(\log n)$ )
- ▶ pro: cost story helps us reason through analysis
- ▶ loss operation = procrastinating, deferring work to later
- ▶ profit operation = catching up on deferred work
- ▶ con: cost story may overcomplicate things
- ▶ con: sometimes awkward to store profits in specific data structure locations

## Potential Method

- ▶ premise: for data structure  $D$ , **potential function**  $\Phi(D)$
- ▶ ( $\Phi$  pronounced like “fee”)
- ▶ potential is stored energy, like a battery
- ▶ amortized time of operation = actual time + change to  $\Phi(D)$
- ▶ procrastinating/loss operations decrease  $\Phi(D)$
- ▶ catch-up/profit operations increase  $\Phi(D)$
- ▶ show:  $\Phi(D) \geq 0$  always
- ▶ ultimately a less structured way of thinking of the accounting method



## Fibonacci Heap Overview

- ▶ alternative to a binary heap (as in heapsort)
- ▶ pro: faster INSERT and DECREASE-KEY operations (optimal in fact)
- ▶ pro: supports additional operations
- ▶ con: more complicated to describe, implement, *especially* analyze
- ▶ con: operations have amortized time bounds
- ▶ con: much worse constant factors, not in-place
- ▶ not currently practical

## Mergeable Heap API

CREATE-HEAP(): initialize empty heap

INSERT( $H, x$ ): insert entry  $x$  into  $H$

MINIMUM( $H$ ): return minimum-key entry

EXTRACT-MIN( $H$ ): remove and return minimum-key entry

UNION( $H_1, H_2$ ): consume heaps  $H_1, H_2$  and return a new heap with their elements

DECREASE-KEY( $H, x, k$ ): lower key of  $x$  to  $k$

DELETE( $H, x$ ): remove entry  $x$  from  $H$

## Heap Efficiency Comparison

| Operation    | Binary Heap (worst-case) | Fib. Heap (amort.) |
|--------------|--------------------------|--------------------|
| CREATE-HEAP  | $\Theta(1)$              | $\Theta(1)$        |
| INSERT       | $\Theta(\log n)$         | $\Theta(1)$        |
| MINIMUM      | $\Theta(1)$              | $\Theta(1)$        |
| EXTRACT-MIN  | $\Theta(\log n)$         | $O(\log n)$        |
| UNION        | $\Theta(n)$              | $\Theta(1)$        |
| DECREASE-KEY | $\Theta(\log n)$         | $\Theta(1)$        |
| DELETE       | $\Theta(\log n)$         | $\Theta(\log n)$   |

## Fibonacci Heap Structure

Overall heap  $H$  has

- ▶ a *forest* of individual  $k$ -ary trees (nonbinary); each tree in min-heap-order
- ▶  $H.min$  = pointer to the root with the global minimum key
- ▶  $H.n$  = number of entries in heap

Each node  $x$  has

- ▶  $x.parent$  = parent node (NIL if root)
- ▶  $x.child$  = pointer to an arbitrary child
- ▶  $x.left, x.right$  = pointers to siblings
- ▶ (siblings are in circular, unsorted, doubly-linked lists)
- ▶  $x.degree$  = # children
- ▶  $x.mark$  = boolean, true iff  $x$  has lost a child since the last time  $x$  was assigned a new parent (manages procrastination)

## Potential Function and Degree Bound

For analysis purposes, we define the potential function

$$\Phi(H) = t(H) + 2 \cdot m(H)$$

where

- ▶  $t(H) = \#$  trees in root list
- ▶  $m(H) = \#$  marked nodes

Claim: If Fibonacci heap  $H$  has  $n$  nodes, then the maximum degree ( $\#$  children) of any node, written  $D(n)$ , is

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

## CREATE-HEAP

```
1: function CREATE-HEAP
2:    $H.min = \text{NIL}$ 
3:    $H.n = 0$ 
4:   return  $H$ 
5: end function
```

Takes  $\Theta(1)$  time;  $t(H) = m(H) = 0$ , so  $\Phi(H) = 0$ ; no profit or loss.

## INSERT Pseudocode

Just create node, initialize node, insert into root list, update  $H.min$

```
1: function INSERT( $H, x$ )
2:    $x.degree = 0$ 
3:    $x.parent = x.child = \text{NIL}$ 
4:    $x.mark = \text{false}$ 
5:   if  $H.min == \text{NIL}$  then
6:      $H.min = x$ 
7:      $x.left = x.right = \text{NIL}$ 
8:   else
9:     insert  $x$  into root list as  $H.min$ 's right sibling
10:    if  $x.key < H.min.key$  then
11:       $H.min = x$ 
12:    end if
13:  end if
14:   $H.n++$ 
15: end function
```

## INSERT Analysis

Actual time steps are  $\Theta(1)$ .

Let  $H$  be heap before INSERT, and  $H'$  after; then

$$t(H') = t(H) + 1 \text{ (one tree created)}$$

$$m(H') = m(H) \text{ (no marking)}$$

$$\implies \Phi(H') = \Phi(H) + 1$$

$\therefore$  INSERT takes  $O(1) + 1 = O(1)$  amortized time.



## MINIMUM

Just follow the  $H.min$  pointer.

$O(1)$  actual time.

No new trees, no new marked nodes, so  $O(1)$  amortized time.

## UNION

```
1: function UNION( $H_1, H_2$ )
2:   if  $H_1.n == 0$  then
3:     return  $H_2$ 
4:   else if  $H_2.n == 0$  then
5:     return  $H_1$ 
6:   else
7:      $H =$  new heap object
8:      $H.min = \min(H_1.min, H_2.min)$ 
9:      $H.n = H_1.n + H_2.n$ 
10:    concatenate root lists of  $H_1, H_2$  into one linked list
11:    return  $H$ 
12:  end if
13: end function
```

$O(1)$  actual time; trees and marked nodes move around, but their number is unchanged, so no change to  $\Phi$ ;  $O(1)$  amortized time.

## So Far So Good

So far all operations have been simple, and either potential-neutral (CREATE-HEAP, MINIMUM), or increased potential (INSERT)

Indeed,  $n$  INSERTs creates a glorified  $n$ -element linked list.

∴ need to expect remaining operations to be more complicated, decrease potential, decrease length of root list.

## EXTRACT-MIN

Follow  $H.min$  pointer; promote all children to roots; remove from root list; CONSOLIDATE to shorten root list and find new  $H.min$ .

```
1: function EXTRACT-MIN( $H$ )
2:    $z = H.min$ 
3:   for each child  $c$  of  $z$  do
4:     insert  $c$  into  $H$ 's root list
5:      $c.parent = NIL$ 
6:   end for
7:   delete  $z$  from root list
8:   if  $H.n == 1$  then
9:      $H.min = NIL$                                      ▷ heap just became empty
10:  else
11:    CONSOLIDATE( $H$ )   ▷ compact root list, recompute  $H.min$ 
12:  end if
13:   $H.n \leftarrow H.n - 1$ 
14:  return  $z$ 
15: end function
```

## Zooming in to CONSOLIDATE

Before moving on, observe EXTRACT-MIN takes, aside from CONSOLIDATE,  $\Theta(\text{degree}(H.min))$  time; claimed this is  $\Theta(\log n)$

Contract for CONSOLIDATE:

1: **function** CONSOLIDATE( $H$ )

**Require:**  $H.min$  invalid and  $H.n == \#nodes + 1$

**Ensure:**  $H.min$  valid and  $\#trees \in O(\log n)$

2: **end function**

## CONSOLIDATE Pseudocode

```
1: function CONSOLIDATE( $H$ )
2:    $A = \text{UNIQUE-DEGREE-ARRAY}(H)$ 
3:    $\text{ARRAY-TO-ROOT-LIST}(H, A)$ 
4:   free  $A$ 
5:    $\text{RECOMPUTE-MIN}(H)$ 
6: end function
```

Clearly  $\Theta(1)$  time except for the three subroutines.

## UNIQUE-DEGREE-ARRAY Pseudocode

```

1: function UNIQUE-DEGREE-ARRAY( $H$ )
Ensure: returns  $A[0..D(H.n)]$  where  $A[d]$  = only root w/ degree  $d$ 
2:    $A[0..D(H.n)]$  = new array of node pointers, all NIL
3:   for each root node  $r$  in  $H$  do                                ▷ move  $r$  into  $A$ 
4:      $p = r$                                                         ▷ parent node that needs to move into  $A$ 
5:     while  $A[p.degree] \neq \text{NIL}$  do                                ▷ another node in the way
6:        $c = A[p.degree]$                                             ▷  $p, c$  have same degree
7:        $A[p.degree] = \text{NIL}$                                         ▷ we will link them into one tree
8:       if  $p.key > c.key$  then
9:          $\text{swap}(p, c)$                                             ▷ ensure  $c$  should be  $p$ 's child
10:      end if
11:      make  $c$  a child of  $p$ , incrementing  $p.degree$ 
12:       $c.mark = \text{false}$ 
13:    end while
14:  end for
15:  return  $A$ 
16: end function

```

## UNIQUE-DEGREE-ARRAY Analysis

Create  $A$ :  $O(D(n))$  time

**for** loop — aggregate analysis

- ▶ # iterations = length of root list before CONSOLIDATE =  $t(H) + D(n) - 1$
- ▶ each iteration of inner **while** loop links two roots into one, decrementing # roots
- ▶ # roots at end of CONSOLIDATE  $\leq$  size of  $A = D(n) + 1$
- ▶  $\implies$  total time in all **while** iterations is

$$(t(H) + D(n) - 1) - (D(n) + 1) = t(H) - 2$$

- ▶  $\implies$  total time in **for** loop is

$$(t(H) + D(n) - 1) + (t(H) - 2) \leq 2 \cdot t(H) + D(n)$$

total for UNIQUE-DEGREE-ARRAY is  $O(t(H) + D(n))$



## ARRAY-TO-ROOT-LIST

```
1: function ARRAY-TO-ROOT-LIST( $H, A$ )
2:   clear  $H$ 's root list to empty
3:   for index  $i$  of  $A$  do
4:     if  $A[i] \neq \text{NIL}$  then
5:       insert  $A[i]$  into  $H$ 's root list
6:     end if
7:   end for
8: end function
```

$O(D(n))$  time

## RECOMPUTE-MIN

```
1: function RECOMPUTE-MIN( $H$ )
2:    $H.min = \text{NIL}$ 
3:   for node  $r$  in  $H$ 's root list do
4:     if  $H.min == \text{NIL}$  then
5:        $H.min = A[i]$ 
6:     else if  $H.min.key > r.key$  then
7:        $H.min = r$ 
8:     end if
9:   end for
10: end function
```

$O(D(n))$  time

## CONSOLIDATE Worst-Case Analysis

```

1: function CONSOLIDATE( $H$ )
2:    $A = \text{UNIQUE-DEGREE-ARRAY}(H)$        $\triangleright O(D(n) + t(H))$ 
3:    $\text{ARRAY-TO-ROOT-LIST}(H, A)$            $\triangleright O(D(n))$ 
4:   free  $A$                               $\triangleright O(1)$ 
5:    $\text{RECOMPUTE-MIN}(H)$                    $\triangleright O(D(n))$ 
6: end function

```

Time spent is

$$O(D(n) + t(H) + D(n) + 1 + D(n)) = O(3 \cdot D(n) + t(H)) = O(\log n + t(H)).$$

Want this to be  $O(\log n)$ ; the  $t(H)$  overage can be withdrawn from the potential function.

## CONSOLIDATE Amortized Analysis

Recall potential function

$$\Phi(H) = t(H) + 2 \cdot m(H)$$

Let  $H'$  be  $H$  after EXTRACT-MIN

EXTRACT-MIN does not mark any nodes, so  $m(H') = m(H)$ .

# roots decreases:  $t(H') = D(n) + 1 \leq t(H)$

$$\Phi(H') = t(H') + 2 \cdot m(H') = (D(n) + 1) + 2 \cdot m(H)$$

$$\Phi(H) - \Phi(H') = t(H) - (D(n) + 1)$$

Amortized cost of EXTRACT-MIN

$$= \Delta\Phi = O(t(H) + D(n)) - O(t(H) - D(n) - 1) = O(2 \cdot D(n)) = O(\log n)$$

## Where did that time come from? Accounting Perspective

- ▶ potential function's  $t(H)$  term over-charges INSERT and under-charges EXTRACT-MIN
- ▶ when  $t(H) = O(\log n)$ , EXTRACT-MIN takes  $O(\log n)$  worst-case time
- ▶ first  $\log n$  roots are “clean”, remaining  $t(H) - \log n$  are “mess”
- ▶ deposit  $O(1)$  time when INSERTing a messy root
- ▶ withdraw those deposits in CONSOLIDATE
- ▶ 1 deposit pays for making 1 messy root a child of a clean parent
- ▶ (recall that link operation takes  $O(1)$  time)
- ▶ hard to use accounting method directly because the identity of the clean nodes changes during CONSOLIDATE based on heap-order

## DELETE

```
1: function DELETE( $H, x$ )  
2:   DECREASE-KEY( $H, x, -\infty$ )  
3:   EXTRACT-MIN( $H$ )  
4: end function
```

Know EXTRACT-MIN is  $O(\log n)$  amortized time.

Claim: DECREASE-KEY is  $O(1)$  amortized time.

$\implies$  DELETE is  $O(\log n)$  amortized time.

## DECREASE-KEY Sketch

DECREASE-KEY( $H, x, k$ )

- ▶ update  $x.key = k$
- ▶ if  $x$  is a root, or heap-order maintained, done
- ▶ else **cut**  $x$ 
  - ▶ make  $x$  a root, no longer a child of parent  $p$
  - ▶ update  $H.min$  if  $x$  is new minimum
  - ▶ mark  $p$ ; but if  $p$  was already marked, recursively cut  $p$
- ▶  $O(1)$  time not counting recursive cuts
- ▶ recursive cuts are paid for by withdrawing potential

## Heap Efficiency Comparison

| Operation    | Binary Heap (worst-case) | Fib. Heap (amort.) |
|--------------|--------------------------|--------------------|
| CREATE-HEAP  | $\Theta(1)$              | $\Theta(1)$        |
| INSERT       | $\Theta(\log n)$         | $\Theta(1)$        |
| MINIMUM      | $\Theta(1)$              | $\Theta(1)$        |
| EXTRACT-MIN  | $\Theta(\log n)$         | $O(\log n)$        |
| UNION        | $\Theta(n)$              | $\Theta(1)$        |
| DECREASE-KEY | $\Theta(\log n)$         | $\Theta(1)$        |
| DELETE       | $\Theta(\log n)$         | $\Theta(\log n)$   |