50001 - Algorithm Analysis and Design - Lecture  $9\,$ 

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## Lecture Recording

Lecture recording is available here

# Amortization

The complexity of **tail** is an example of **Amortized analysis**, where operation's wider context are considered when calculating the complexity.

$$xs_0 \stackrel{op_0}{\leadsto} xs_1 \stackrel{op_1}{\leadsto} xs_2 \stackrel{op_2}{\leadsto} xs_3 \stackrel{op_3}{\leadsto} \dots \stackrel{op_{n-1}}{\leadsto} xs_n$$

We defined 3 parts:

#### 1. Cost Function

 $C_{op_i}(xs_i)$  determines the cost of operation  $op_i$  on data  $xs_i$ . Estimating how many steps it takes for each operation to execute.

#### 2. Amortized Cost Function

 $A_{op_i}(xs_i)$  for each operation  $op_i$  on data  $xs_i$ .

### 3. Size Function

S(xs) that calculates the size of data xs

We define these functions with the goal to show that:

$$C_{op_i}(xs_i) \le A_{op_i}(xs_i) + S(xs_i) - S(xs_{i+1})$$

The cost of the operation is smaller than the amortized cost, plus the difference in size of the data structure before and after the operation.

Once this is shown, we can infer that:

$$\sum_{i=0}^{n-1} C_{op_i}(xs_i) \le \sum_{i=0}^{n-1} A_{op_i}(xs_i) + S(xs_0) - S(xs_n)$$

Furthermore when  $S(xs_0) = 0$  this implies:

$$\sum_{i=0}^{n-1} C_{op_i}(xs_i) \le \sum_{i=0}^{n-1} A_{op_i}(xs_i) - S(xs_n) \Rightarrow \sum_{i=0}^{n-1} C_{op_i}(xs_i) \le \sum_{i=0}^{n-1} A_{op_i}(xs_i)$$

This means the cost of the operations is less than the sum of the amortized costs.

For example, if  $A_{op_i}(xs) = 1$  then the total cost will be bounded by O(n).

#### Tail example

$$C_{cons}(xs) = 1$$
  $C_{snoc}(xs) = 1$   $C_{head}(xs) = 1$   $C_{last}(xs) = 1$ 

For tail we can do the following:

```
C_{tail}(Dequeue\ us\ sv) = length\ sv
                                                       (Create a cost function of tail.)
(2)
      A_{op}(xs) = 2
                                                       (Create an arbitrary cost function.)
(3)
      S(Dequeue\ us\ sv) = |length\ us - length\ sv|
                                                       (Create a size function for dequeue.)
     Dequeue us sv where length sv = k
                                                       (Worst case where us is a singleton)
      S(Dequeue\ us\ sv) = k-1
                                                       (Size of the next data structure can be at most 1.)
       S(Dequeue\ us'\ sv') = 1
(6)
      C_{tail}(Dequeue\ us\ sv) = k
                                                       (Calculate the worst case cost of tail.)
      k \le 2 + (k-1) - 1 = k+2
                                                       (As this inequality holds, the time complexity of all n instructi
(7)
```

As the time complexity of all n instructions together is O(n), the amortized cost of a single instruction is O(1).

#### About the size function

We want to balance the size function such that:

- The size function is 0 to start with.
- The size between operations is large enough to prove the inequality.

The size function is arbitrary, if you cannot choose a size function that satisfied the goal inequality, then you're probably making a mistake

#### Peano Numbers

```
data Peano = Zero | Succ Peano
3
       analogous to (:) Cons
   incr :: Peano -> Peano
5
   incr = Succ
6
7
      analogous to tail
8
   decr :: Peano -> Peano
9
    decr (Succ n) = n
   decr Zero = error "Cannot decrement zero"
10
11
       analogous to (++) concatenate
   add :: Peano -> Peano -> Peano
13
14
   add \ a \ Zero = a
15
   add \ a \ (Succ \ b) = Succ \ (add \ a \ b)
16
17
       tail recursive version for extra goodness!
   add \ a \ b = add \ (incr \ a) \ (decr \ b)
```

This shows how similar operations of similarly structured data can be.

# Binary Numbers

```
6 | incr :: Binary -> Binary
7 | incr [] = [I]
8 | incr (O:bs) = I:bs
9 | incr (I:bs) = O:incr bs
```

we can do amortized analysis on incr:

```
C_{incr}(bs) = t + 1 where t = length (takeWhile (== I) bs)
                                                                             (Create a cost function.)
(1)
(2)
      A_{incr}(bs) = 2
                                                                             (Create Amortized Cost)
      S(bs) = length.filter (== I) $ bs
(3)
                                                                             (Create size function.)
      Given bs' = incr\ bs we show C_{incr}(bs) \le A_{incr}bs + S(bs) - S(bs')
                                                                             (Setup inequality)
(5)
     t + 1 \le 2 + S(bs) - (S(bs) - t + 1)
                                                                             (Substitute in inequality)
      t+1 \le 1+t
(6)
                                                                             (Hence inequality holds)
      S(start) = 0
                                                                             (Start size is zero.)
```

Hence The sum of C is smaller than the sum of A, as this is over n operations and  $\sum A = 2n$ ,  $C_{incr}(bs) \in O(1)$ .