## Notes on Amortized Analysis

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## Amortized analysis

- Analyze a sequence of operations on a data structure.
- ▶ Goal: Show that although some operations may be expensive, on average the cost per operation is small.
- Average is not over a distribution of inputs, but over a sequence of operations.
- ▶ No probability is involved: *Average* cost in the *worst* case.
- We look at three methods of calculating:
  - 1. aggregate analysis
  - 2. accounting method
  - 3. potential method
- And two simple examples:
  - 1. stack with multipop
  - 2. binary counter
- ▶ And a more interesting example:
  - dynamic tables

## Stack operations

```
Push(S, x): O(1)
Pop(S): O(1)
```

```
Multipop(S, k)
```

```
1 while S is not empty and k > 0
```

2 
$$Pop(S)$$

$$3 k = k - 1$$

```
top \rightarrow 23
17
6
39
10
47
47
(a)
(b)
MULTIPOP(S, 4)
MULTIPOP(S, 7)
```

# Running time of MULTIPOP:

- ▶ Linear in # of Pop operations.
- ▶ Let each PUSH/POP cost 1.
- # iterations of **while** loop is min(s, k)
  - where s = # of objects in stack.
- ▶ Total cost = min(s, k)

#### Worst-case analysis without amortization

- ightharpoonup Sequence of n PUSH, POP, and MULTIPOP operations.
- ▶ May have up to *n* PUSH operations.
- ▶ So worst-case there are *n* items on the stack.
- ▶ Therefore, worst-case cost of a MULTIPOP operation is O(n).
- ▶ Have n operations, each of which could be MULTIPOP.
- ▶ Therefore, worst-case cost of sequence of n operations is  $O(n^2)$ .

#### Something wrong with worst-case analysis

- ▶ There's clearly something wrong with this analysis.
- ▶ What is actual worst-case number of Pushs and Pops as a function of *n*?
- But how can we get a more accurate worst-case analysis?
- We need to consider how the operations interact with each other.
- ▶ We need to keep an account of how much time is spent in each one, because that affects the time spent in the others.

#### Aggregate analysis

#### Observations

- Each object can be popped only once per time that it's pushed.
- ▶ Have  $\leq n$  Pushs, therefore  $\leq n$  Pops, including those in MULTIPOP.
- ▶ Therefore, total cost = O(n).
- Average over n operations is = O(1) per operation on average, including those in MULTIPOP.
- ► This is called aggregate analysis.
  - No probability involved.
  - ▶ Showed worst-case O(n) for entire sequence.
  - ▶ Therefore, O(1) per operation on average.

#### Binary counter

Counter value	MINGHSHONSHONING	Total cost
0	$0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0$	0
1	$0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 1$	1
2	0 0 0 0 0 0 1 0	3
3	$0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 1$	4
4	0 0 0 0 0 1 0 0	7
5	$0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 0 \ 1$	8
6	0 0 0 0 0 1 1 0	10
7	$0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 1 \ 1$	11
8	0 0 0 0 1 0 0 0	15
9	0 0 0 0 1 0 0 1	16
10	0 0 0 0 1 0 1 0	18
11	$0 \ 0 \ 0 \ 0 \ 1 \ 0 \ 1 \ 1$	19
12	0 0 0 0 1 1 0 0	22
13	$0 \ 0 \ 0 \ 0 \ 1 \ 1 \ 0 \ 1$	23
14	0 0 0 0 1 1 1 0	25
15	$0 \ 0 \ 0 \ 0 \ 1 \ 1 \ 1 \ 1$	26
16	0 0 0 1 0 0 0	31

- Bits that flip upon increment shaded.
- ► Total cost of flipping bits at right.
- ► Total cost always less than twice number of increments.

#### Binary counter

- ▶ k-bit binary counter A[0..k-1] of bits.
- ► *A*[0] is the least significant bit.
- Counts upward from 0.
- ▶ Value of counter is

$$\sum_{i=0}^{k-1} A[i] \cdot 2^{i}$$

- ▶ Initially counter is 0, so A[0..k-1] = 0.
- ▶ To increment, add 1 (mod  $2^k$ ):

```
INCREMENT(A, k)

1  i = 0

2  while i < k and A[i] == 1

3  A[i] = 0

4  i = i + 1

5  if i < k
```

A[i] = 1

## Worst case analysis of binary counter

- ▶ Each call could flip *k* bits.
- ▶ n increments is O(nk).

# Aggregate analysis of binary counter

▶ Not every bit flips every time.

bit	flips how often	times in $n$ Increments
0	every time	n
1	1/2 the time	$\lfloor n/2 \rfloor$
2	1/4 the time	$\lfloor n/4 \rfloor$
	:	
i	$1/2^i$ the time	$\lfloor n/2^i \rfloor$
	:	
$i \ge k$	never	0

## Total number of flips

$$\sum_{i=0}^{k-1} \lfloor n/2^i \rfloor < n \sum_{i=0}^{k-1} (1/2)^i$$

$$= n \frac{(1/2)^k - 1}{1/2 - 1}$$

$$= n \frac{1 - (1/2)^k}{1 - 1/2}$$

$$< n \left(\frac{1}{1/2}\right)$$

$$= 2n$$

- ▶ n Increments costs O(n).
- Average cost per operation O(1).

## Accounting Method and Potential Method

- Aggregate method works when we can add up all operations.
- ▶ More complex operations need a more sophisticated method.
- ▶ Two approaches:

#### Accounting method:

- assign charges to each operation
- some operations charged more than they cost
- others, charged less, can use accrued credit

#### Potential method:

- prepaid work is "potential energy"
- energy is assigned to data structures as a whole
- some operations increase potential energy
- some operations can release potential energy to reduce costs
- most flexible of the amortized analysis methods

#### Accounting method

- Amortized cost = amount we charge
- ▶ Amortized cost must always be ≥ actual cost
- When amortized cost > actual cost, store the difference on specific objects in the data structure as credit.
- When we have credit, we have accounted for expenses not yet accrued
- Use credit later to pay for operations whose actual cost > amortized cost.
- Differs from aggregate analysis:
  - In the accounting method, different operations can have different costs.
  - ▶ In aggregate analysis, all operations have the same cost.
- Credit must never go negative.
  - ▶ Otherwise we have a sequence of operations for which amortized cost is not an upper bound on actual cost.
  - Amortized cost would tell us nothing.



## Accounting method costs

 $c_i =$ actual cost of ith operation

 $\hat{c}_i = \text{amortized cost of } i \text{th operation}$ 

#### Require

$$\sum_{i=1}^n \hat{c}_i \geq \sum_{i=1}^n c_i$$

Total credit stored

$$\sum_{i=1}^n \hat{c}_i - \sum_{i=1}^n c_i$$

must never be negative.

## Accounting method amortized analysis of stack operations

operation	actual cost	amortized cost
Push	1	2
Рор	1	0
Multipop	$\min(k,s)$	0

#### Intuition:

- When pushing an object, pay \$2
- ▶ \$1 pays for the Push
- ▶ \$1 is prepayment for it being popped by POP or MULTIPOP
- Since each object has \$1 credit, the credit can never go negative.
- ▶ Total amortized cost, O(n), is an upper bound on total cost.
- ▶ Worst cast amortized cost is 2n = O(n).

## Accounting method amortized analysis of binary counter

- ► Charge \$0 to set a bit to 0
- Charge \$2 to set a bit to 1
  - \$1 pays for setting the bit to 1
  - ▶ \$1 prepayment for setting it back to 0
  - Have \$1 credit for every 1 in the counter
  - ► Therefore credit ≥ 0
- ► Amortized cost of INCREMENT:
  - Cost of resetting bits to 0 is paid by credit.
  - At most 1 bit is set to 1.
  - ▶ Amortized cost is always ≤ 2.
  - For *n* operations amortized cost is O(n).

#### The Potential Method

- Like the accounting method, but think of the credit as the *potential* stored with the entire data structure.
- Accounting method stores credit with specific objects.
- ▶ Potential method stores potential in the data structure as a whole.
- Can release potential to pay for future operations.
- Most flexible of the amortized analysis methods.

#### Potential function

 $D_i = data structure after the ith operation$ 

 $D_0 = initial data structure$ 

 $c_i$  = actual cost of *i*th operation

 $\hat{c}_i = \text{amortized cost of the } i \text{th operation}$ 

Potential function:  $\Phi: D_k \to \mathbb{R}$ 

 $\Phi(D_i)$  is the *potential* associated with the data structure  $D_i$ .

$$\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$$
  
=  $c_i + \Delta\Phi(D_i)$ 

The amortized cost is the *increase in potential* due to the *i*th operation.



#### Total amortized cost

$$egin{aligned} \sum_{i=1}^n \hat{c_i} &= \sum_{i=1}^n (c_i + \Phi(D_i) - \Phi(D_{i-1})) \ &= \sum_{i=1}^n c_i + \Phi(D_n) - \Phi(D_0) \end{aligned}$$

- ▶ If we require that  $\Phi(D_i) \ge \Phi(D_0)$  for all i, then the amortized cost is always an upper bound on the actual cost.
- ▶ In practice:

$$\Phi(D_0) = 0$$

$$\Phi(D_i) > 0 for all i$$

# Amortized analysis of stack operations using the potential method

$$\Phi=\#$$
 of objects in the stack 
$$=\# \text{ of $\$1$ bills in the accounting method}$$
  $\Phi({\it D}_0)=0$ 

Since # of objects in stack is always  $\geq 0$ ,

$$\Phi(D_i) \ge 0 = \Phi(D_0) \qquad \qquad \text{for all } i$$

operation	actual cost	ΔΦ
Push	1	(s+1)-s=1
Рор	1	(s-1)-s=-1
Multipop	$k' = \min(k, s)$	(s-k')-s=k'

Therefore the amortized cost of a sequence of n operations is O(n).



## Amortized analysis of binary counter: potential method

- lacktriangledown  $\Phi=b_i=\#$  of 1's after *i*th INCREMENT
- ▶ Suppose *i*th operation resets  $t_i$  bits to 0.
- ▶  $c_i \le t_i + 1$ , since it resets  $t_i$  bits and sets  $\le 1$  bit to 1.
- ▶ If  $b_i = 0$ , the *i*th operation reset all *k* bits and didn't set one, so

$$b_{i-1} = t_i = k \Rightarrow b_i = b_{i-1} - t_i = 0$$

▶ If  $b_i > 0$  the *i*th operation reset  $t_i$  bits, set one, so

$$b_i = b_{i-1} - t_i + 1$$

Either way

$$b_i \leq b_{i-1} - t_i + 1$$

Therefore

$$\Delta\Phi(D_i) \le (b_{i-1} - t_i + 1) - b_{i-1} = 1 - t_i$$
$$\hat{c}_i = c_i + \Delta\Phi(D_i) \le (t_i + 1) + (1 - t_i) = 2$$

- ▶ If counter starts at 0,  $\Phi(D_0) = 0$ .
- ▶ Therefore, amortized cost of n operations is O(n).



#### Dynamic Tables

- Nice application of amortized analysis.
- ▶ Suppose you have a table, maybe a hash table, maybe a heap.
- Details of table organization not important.
- We will assume insertion and deletion take O(1).
- You don't know in advance how many items will be stored in it.
- ▶ When it fills, you must reallocate a larger table and copy all the items into the new table.
- When it gets sufficiently small, you might want to reallocate with a smaller size.
- ► How can you do this so it doesn't mess up the efficiency of your table?
- ▶ Does it turn O(1) (hash) or  $O(\lg n)$  (heap) into O(n), since in worst case we have to copy all n elements into new array?



## Dynamic Table Goals

- 1. O(1) amortized time per operation.
- 2. Unused space always  $\leq$  constant fraction of allocated space.
- Load factor α = num/size where num = # items stored, size = allocated size.
- ▶ Never allow  $\alpha > 1$
- Keep  $\alpha$  > constant fraction (goal 2).

#### Table expansion

- First we consider only expansion.
- When table becomes full, double its size and reinsert all existing items.
- ► Each time we actually insert an item, it's an **elementary insertion**.

```
Table-Insert (T, x)
 1 if T. size == 0
                                                                 // empty?
          allocate T. table with 1 slot.
          T. size = 1
    if T. num == T. size
                                                                // expand?
 5
          allocate new-table with 2 · T. size slots
 6
         insert all items in T. table into new-table
         free T. table
 8
          T. table = new-table
          T. size = 2 \cdot T. size
    insert x into T. table
10
11
     T.num = T.num + 1
```

## Running time

- ► Charge 1 per elementary insertion.
- Count only elementary insertions.
  - All other costs are constant per cell.
- $ightharpoonup c_i = actual cost of ith operation$
- ▶ If not full,  $c_i = 1$
- ▶ If full, insert i 1 items plus one more,  $c_i = i$ .
- n operations, worst case:

$$c_i = O(n)$$
  
 $n ext{ operations} = O(n^2)$ 

#### Aggregate analysis

Of course, we don't always expand:

$$c_i =$$
  $\begin{cases} i & \text{if } i-1 \text{ is exact power of 2.} \\ 1 & \text{otherwise.} \end{cases}$ 

Total cost 
$$=\sum_{i=1}^{n} c_i$$
  
 $\leq n + \sum_{j=0}^{\lfloor \lg n \rfloor} 2^j$   
 $= n + \frac{2^{\lfloor \lg n \rfloor + 1} - 1}{2 - 1}$   
 $< n + 2n$   
 $= 3n$ 

► Aggregate analysis: the amortized cost per operation is 3.



#### Accounting method

- ► Charge \$3 per elementary insertion of *x*:
  - ▶ \$1 pays for x's insertion.
  - ▶ \$1 pays for *x*'s move in the future.
  - \$1 pays for some other item to be moved.
- ▶ Suppose we've just expanded, size = m.
- size = 2m after next expansion.
- ▶ Assume that the expansion used up all the credit, so that there's no credit stored after the expansion.
- ▶ Will expand again after another *m* insertions.
- ▶ Each insertion will put \$1 on one of the *m* items that were in the table just after expansion, and will put \$1 on the item inserted.
- ► Have \$2*m* of credit by next expansion, when there are 2*m* items to move.
- Just enough to pay for expansion, with no credit left over!
- ▶ Credit always ≥ 0.



#### Potential method

$$\Phi(T) = 2 \cdot T. num - T. size$$

▶ Initially, num = size = 0.

$$\Phi = 0$$

▶ Just after expansion,  $size = 2 \cdot num$ 

$$\Phi = 0$$

▶ Just before expansion, *size* = *num* 

$$\Phi = num$$

we have enough potential to pay for moving all items.

▶ Need  $\Phi \ge 0$  always.

## Amortized cost of ith operation

$$num_i = num$$
 after *i*th operation  $size_i = size$  after *i*th operation  $\Phi_i = \Phi$  after *i*th operation

If no expansion:

$$size_i = size_{i-1}$$
 $num_i = num_{i-1} + 1$ 
 $c_i = 1$ 

Then we have

$$\hat{c}_i = c_i + \Phi_i - \Phi_{i-1}$$
  
= 1 + (2 · num<sub>i</sub> - size<sub>i</sub>) - (2 · num<sub>i-1</sub> - size<sub>i-1</sub>)  
= 1 + (2 · num<sub>i</sub> - size<sub>i</sub>) - (2(num<sub>i</sub> - 1) - size<sub>i</sub>)  
= 1 + 2 = 3



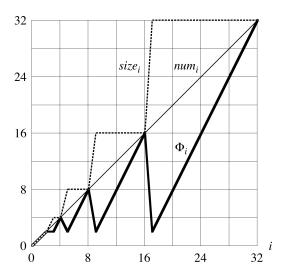
# Amortized cost of ith operation

► If expansion:

$$size_i = 2 \cdot size_{i-1}$$
  
 $size_{i-1} = num_{i-1} = num_i - 1$   
 $c_i = num_{i-1} + 1 = num_i$ 

Then we have

$$\begin{split} \hat{c_i} &= c_i + \Phi_i - \Phi_{i-1} \\ &= num_i + (2 \cdot num_i - size_i) - (2 \cdot num_{i-1} - size_{i-1}) \\ &= num_i + (2 \cdot num_i - 2(num_i - 1)) - (2(num_i - 1) - (num_i - 1)) \\ &= num_i + 2 - (num_i - 1) \\ &= 3 \end{split}$$



As we insert items, the potential builds up until we have enough to pay for moving all items, when the potential drops back to zero.

## Expansion and contraction

When  $\alpha$  drops too low, contract the table.

- Allocate a new, smaller one.
- Copy all items.

#### Still want:

- ightharpoonup  $\alpha$  bounded from below by a constant
- ightharpoonup amortized cost of O(1)

#### "Obvious strategy"

- ▶ Double size when inserting into a full table ( $\alpha = 1$ ).
- ▶ Halve size when deletion would make table less than half full  $(\alpha < 1/2)$ .
- ▶ Then would always have  $1/2 \le \alpha \le 1$ .
- Unfortunately, suppose we fill the table, then:

insert	$\Rightarrow$	double
two deletes	$\Rightarrow$	halve
two inserts	$\Rightarrow$	double
two deletes	$\Rightarrow$	halve
two inserts	$\Rightarrow$	double

Not performing enough operations in between expansion and contraction to pay for the next one.

#### Simple solution

- ▶ Double when full  $(\alpha = 1)$ .
- ▶ Halve size when  $\alpha = 1/4$ .
- ▶ Immediately after expansion *or* contraction,  $\alpha = 1/2$ .
- ▶ Always have  $1/4 \le \alpha \le 1$

#### Intuition

- Want to make sure we perform enough operations in between consecutive expansions/contractions to pay for the change in table size.
- Need to delete half of the items before contraction.
- Need to double the number of items before expansion.
- Either way, the number of operations between expansions and contractions is at least a constant fraction of the number of items copied.

$$\Phi(T) = \begin{cases} 2 \cdot T. num - T. size & \text{if } \alpha \ge 1/2 \\ T. size/2 - T. num & \text{if } \alpha < 1/2 \end{cases}$$

$$\begin{array}{l} \textit{T} \; \mathsf{empty} \Rightarrow \Phi = 0 \\ \alpha \geq 1/2 \Rightarrow \textit{num} \geq \textit{size}/2 \Rightarrow 2 \cdot \textit{num} \geq \textit{size} \Rightarrow \Phi \geq 0 \\ \alpha \leq 1/2 \Rightarrow \textit{num} < \textit{size}/2 \Rightarrow \Phi \geq 0 \end{array}$$

#### Further intuition

- $\Phi$  measures how far from  $\alpha = 1/2$  we are.
- $ho \quad \alpha = 1 \Rightarrow \Phi = 2 \cdot num num = num$
- $ho \quad \alpha = 1/4 \Rightarrow \Phi = size/2 num = 4 \cdot num/2 num = num$
- Therefore, when we double or halve, we have enough potential to pay for moving all *num* items.

#### Further intuition

- ▶ Potential increases linearly between  $\alpha = 1/2$  and  $\alpha = 1$ .
- ▶ Potential increases linearly between  $\alpha = 1/2$  and  $\alpha = 1/4$ .
- ▶ Since  $\alpha$  has different distances to go to get to 1 or 1/4, starting from 1/2, rate of increase of  $\Phi$  differs.
- For  $\alpha$  to go from 1/2 to 1:
  - num increases from size/2 to size
  - Φ increases from 0 to size
  - Φ needs to increase by 2 for each item inserted.
  - That's why the coefficient of 2 in the formula for Φ.
- For  $\alpha$  to go from 1/2 to 1/4:
  - ▶ num decreases from size/2 to size/4.
  - Φ increases from 0 to size/4
  - ► Thus, Φ needs to increase by 1 for each item deleted.
  - ▶ That's why the coefficient of -1 in the formula for  $\Phi$ .

# Eight cases for calculating amortized costs

- ▶ insert *vs.* delete
- $\alpha \ge 1/2$  vs.  $\alpha < 1/2$
- ▶ size changes vs. size doesn't change

# Insert, $\alpha \geq 1/2$ , with or without expansion

- Same analysis as before.
- $\hat{c}_i = 3$

# Insert, $\alpha_{i-1} < 1/2$ , no expansion

▶ 
$$\alpha_i < 1/2$$

$$\begin{split} \hat{c_i} &= c_i + \Phi_i - \Phi_{i-1} \\ &= 1 + \left( size_i/2 - num_i \right) - \left( size_{i-1}/2 - num_{i-1} \right) \\ &= 1 + \left( size_i/2 - num_i \right) - \left( size_i/2 - \left( num_i - 1 \right) \right) \\ &= 0 \end{split}$$

$$\sim \alpha_i \geq 1/2$$

$$\begin{split} \hat{c_i} &= c_i + \Phi_i - \Phi_{i-1} \\ &= 1 + (2 \cdot num_i - size_i) - (size_{i-1}/2 - num_{i-1}) \\ &= 1 + (2(num_{i-1} + 1) - size_{i-1}) - (size_{i-1}/2 - num_{i-1}) \\ &= 3 \cdot num_{i-1} - \frac{3}{2} \cdot size_{i-1} + 3 \\ &= 3 \cdot \alpha_{i-1} size_{i-1} - \frac{3}{2} \cdot size_{i-1} + 3 \\ &< \frac{3}{2} \cdot size_{i-1} - \frac{3}{2} \cdot size_{i-1} + 3 \\ &= 3 \end{split}$$

# Insert, $\alpha < 1/2$ , expansion

Cannot happen.

#### Insert

=3
=3
ossible
= 0
=3

▶ Therefore, in all cases, the amortized cost of insertion is  $\leq 3$ .

#### Delete

contraction	lpha < 1/2	$\hat{c}_i = 1$
no contraction	$\alpha < 1/2$	$\hat{c}_i = 2$
contraction	$\alpha \geq 1/2$	impossible
no contraction	$\alpha_{i-1} \geq 1/2, \alpha_i \geq 1/2$	$\hat{c_i} = -1$
no contraction	$\alpha_{i-1} \ge 1/2, \alpha_i < 1/2$	$\hat{c}_i = 2$

▶ In all cases the amortized cost is  $\leq 2$ .

