

# 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

(a)

MULTIPOP( $S, 4$ )

top  $\rightarrow$  10

47

(b)

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

- ▶ 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	A[7]	A[6]	A[5]	A[4]	A[3]	A[2]	A[1]	A[0]	Total cost
0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	1	1
2	0	0	0	0	0	0	1	0	3
3	0	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	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       $A[i] = 0$                                 // reset to 0
4       $i = i + 1$ 
5  if  $i < k$ 
6       $A[i] = 1$                                 // set to 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	$\lfloor n/4 \rfloor$
	$\vdots$	
$i$	$1/2^i$ the time	$\lfloor n/2^i \rfloor$
	$\vdots$	
$i \geq k$	never	0

## Total number of flips

$$\begin{aligned}\sum_{i=0}^{k-1} \lfloor n/2^i \rfloor &< n \sum_{i=0}^{\infty} 1/2^i \\ &= n \left( \frac{1}{1 - 1/2} \right) \\ &= 2n\end{aligned}$$

- ▶  $n$  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  $\geq$  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  $i$ th operation

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

Require, for *all* sequences of  $n$  operations:

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

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
POP	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 case 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  $\geq 0$
- ▶ 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  $\leq 2$ .
  - ▶ 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  $i$ th operation

$D_0$  = initial data structure

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

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

**Potential function:**  $\Phi : D_k \rightarrow \mathbb{R}$

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

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

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

## Total amortized cost

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

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

$$\Phi(D_0) = 0$$

$$\Phi(D_i) \geq 0 \quad \text{for all } i$$

# Amortized analysis of stack operations using the potential method

$\Phi = \#$  of objects in the stack

$= \#$  of \$1 bills in the accounting method

$$\Phi(D_0) = 0$$

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

$$\Phi(D_i) \geq 0 = \Phi(D_0) \quad \text{for all } i$$

Let  $s = \#$  objects initially

operation	actual cost	$\Delta\Phi$	amortized cost
PUSH	1	$(s + 1) - s = 1$	$1 + 1 = 2$
POP	1	$(s - 1) - s = -1$	$1 - 1 = 0$
MULTIPOP	$k' = \min(k, s)$	$(s - k') - s = -k'$	$k' - k' = 0$

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

# Amortized analysis of binary counter: potential method

- ▶  $\Phi = b_i = \#$  of 1's after  $i$ th INCREMENT
- ▶ Suppose  $i$ th operation resets  $t_i$  bits to 0.
- ▶  $c_i \leq t_i + 1$ , since it resets  $t_i$  bits and sets  $\leq 1$  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

$$\Delta\Phi(D_i) \leq (b_{i-1} - t_i + 1) - b_{i-1} = 1 - t_i$$

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

- ▶ If counter starts at 0,  $\Phi(D_0) = 0$ .
- ▶ Therefore, amortized cost of  $n$  operations is  $O(n)$ .
- ▶ Method works even if counter does not start at 0.

# 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**  $\alpha = num/size$  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?
2      allocate  $T.table$  with 1 slot
3       $T.size = 1$ 
4  if  $T.num == T.size$                                // expand?
5      allocate  $new-table$  with  $2 \cdot T.size$  slots
6      insert all items in  $T.table$  into  $new-table$ 
7      free  $T.table$ 
8       $T.table = new-table$ 
9       $T.size = 2 \cdot T.size$ 
10 insert  $x$  into  $T.table$ 
11  $T.num = T.num + 1$ 
```

# Running time

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

$$c_i = O(n)$$

$$n \text{ operations} = O(n^2)$$

# Aggregate analysis

- Of course, we don't *always* expand:

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

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

- 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 \$2m of credit by next expansion, when there are  $2m$  items to move.
- ▶ Just enough to pay for expansion, with no credit left over!
- ▶ Credit always  $\geq 0$ .

# 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  $\Phi \geq 0$  always.

$$\begin{aligned} size &\geq num &\geq size/2 &\Rightarrow \\ 2 \cdot num &\geq size &\Rightarrow \\ \Phi &\geq 0 \end{aligned}$$

## Amortized cost of $i$ th 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

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

## Amortized cost of $i$ th operation

- If expansion:

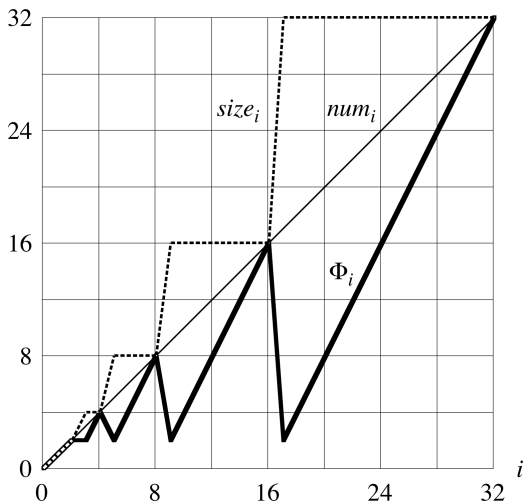
$$size_i = 2 \cdot size_{i-1}$$

$$size_{i-1} = num_{i-1} = num_i - 1$$

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

Then we have

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



- ▶ 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:

- ▶  $\alpha$  bounded from below by a constant
- ▶ 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  $1/2 \leq \alpha \leq 1$ .
- ▶ 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  $1/4 \leq \alpha \leq 1$

# 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.

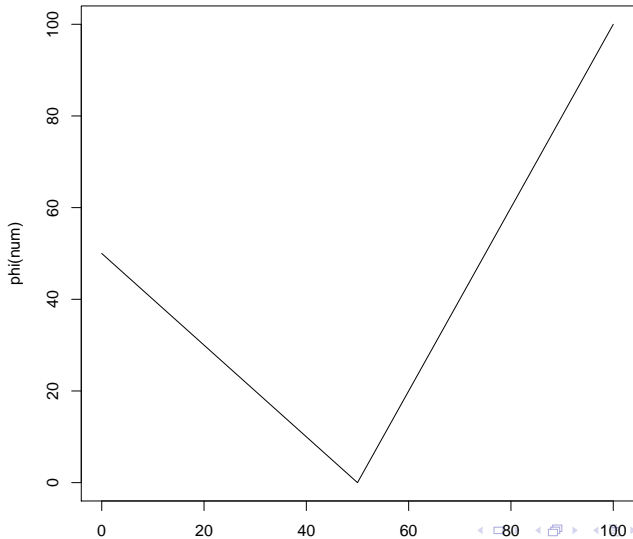
$$\Phi(T) = \begin{cases} 2 \cdot T.num - T.size & \text{if } \alpha \geq 1/2 \\ T.size/2 - T.num & \text{if } \alpha < 1/2 \end{cases}$$

$T$  empty  $\Rightarrow \Phi = 0$

$\alpha \geq 1/2 \Rightarrow num \geq size/2 \Rightarrow 2 \cdot num \geq size \Rightarrow \Phi \geq 0$

$\alpha \leq 1/2 \Rightarrow num < size/2 \Rightarrow \Phi \geq 0$

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



## Further intuition

- ▶  $\Phi$  measures how far from  $\alpha = 1/2$  we are.
- ▶  $\alpha = 1/2 \Rightarrow \Phi = 2 \cdot num - 2 \cdot num = 0$
- ▶  $\alpha = 1 \Rightarrow \Phi = 2 \cdot num - num = num$
- ▶  $\alpha = 1/4 \Rightarrow \Phi = size/2 - num = 4 \cdot num/2 - num = num$
- ▶ 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*
  - ▶  $\Phi$  increases from 0 to *size*
  - ▶  $\Phi$  needs to increase by 2 for each item inserted.
  - ▶ That's why the coefficient of 2 in the formula for  $\Phi$ .
- ▶ For  $\alpha$  to go from  $1/2$  to  $1/4$ :
  - ▶ *num* decreases from *size*/2 to *size*/4.
  - ▶  $\Phi$  increases from 0 to *size*/4
  - ▶ Thus,  $\Phi$  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
- ▶  $\alpha \geq 1/2$  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

- ▶  $\alpha_i < 1/2$

$$\begin{aligned}\hat{c}_i &= c_i + \Phi_i - \Phi_{i-1} \\ &= 1 + (\text{size}_i/2 - \text{num}_i) - (\text{size}_{i-1}/2 - \text{num}_{i-1}) \\ &= 1 + (\text{size}_i/2 - \text{num}_i) - (\text{size}_i/2 - (\text{num}_i - 1)) \\ &= 0\end{aligned}$$

- ▶  $\alpha_i \geq 1/2$

$$\begin{aligned}\hat{c}_i &= c_i + \Phi_i - \Phi_{i-1} \\ &= 1 + (2 \cdot \text{num}_i - \text{size}_i) - (\text{size}_{i-1}/2 - \text{num}_{i-1}) \\ &= 1 + (2(\text{num}_{i-1} + 1) - \text{size}_{i-1}) - (\text{size}_{i-1}/2 - \text{num}_{i-1}) \\ &= 3 \cdot \text{num}_{i-1} - \frac{3}{2} \cdot \text{size}_{i-1} + 3 \\ &= 3 \cdot \alpha_{i-1} \text{size}_{i-1} - \frac{3}{2} \cdot \text{size}_{i-1} + 3 \\ &< \frac{3}{2} \cdot \text{size}_{i-1} - \frac{3}{2} \cdot \text{size}_{i-1} + 3 \\ &= 3\end{aligned}$$

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

- ▶ Cannot happen.

# Insert

expansion	$\alpha \geq 1/2$	$\hat{c}_i = 3$
no expansion	$\alpha \geq 1/2$	$\hat{c}_i = 3$
expansion	$\alpha < 1/2$	impossible
no expansion	$\alpha_{i-1} < 1/2, \alpha_i < 1/2$	$\hat{c}_i = 0$
no expansion	$\alpha_{i-1} < 1/2, \alpha_i \geq 1/2$	$\hat{c}_i = 3$

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

# Delete

contraction	$\alpha < 1/2$	$\hat{c}_i = 1$
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	$\alpha_{i-1} \geq 1/2, \alpha_i < 1/2$	$\hat{c}_i = 2$

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

