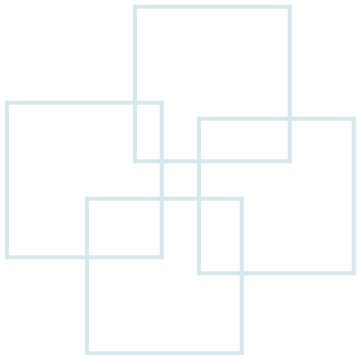


Topic 6: Amortized Analysis





Amortized (分期償還) Analysis

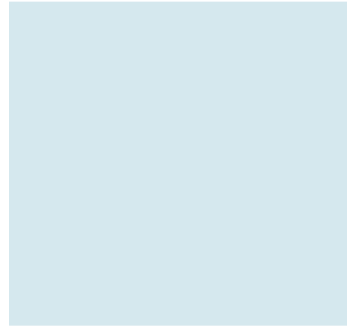
- Analyze a *sequence* of operations on a data structure.
- **Goal:**
 - Show that although some individual operations may be expensive, *on average* the cost per operation is small.
- *Average* in this context does not mean that we're averaging over a distribution of inputs. Instead,
 - ***No probability is involved.***
 - We're talking about
 - ***Average performance of each operation in the worst case.***
 - *The time required to perform a sequence of data structure operations in average over all the operations performed.*

For all n , a sequence of n operations takes worst time $T(n)$ in total. The amortize cost of each operation is $T(n)/n$.

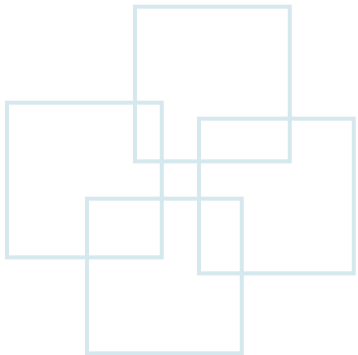


Outline

- Aggregate analysis
- The accounting method
- The potential method
- Dynamic tables



Aggregate Analysis



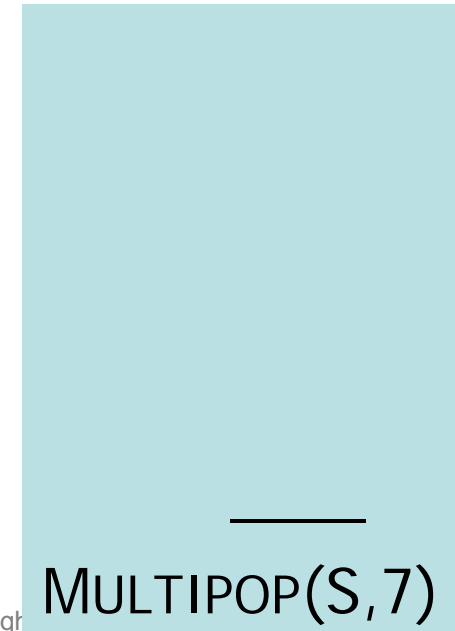
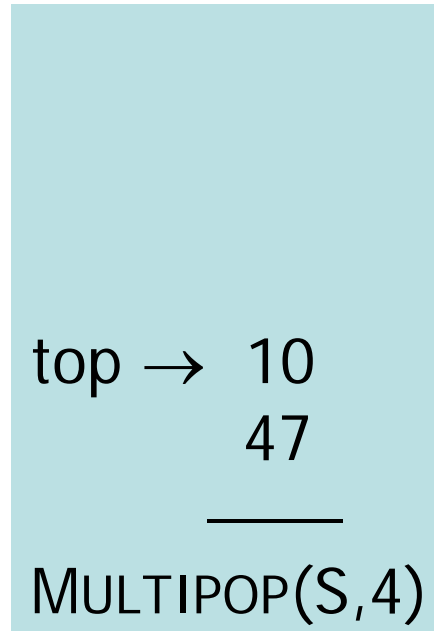
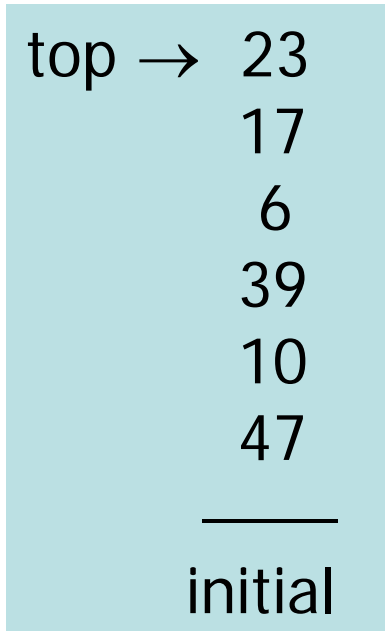


Stack Operation

- Stack operation
 - PUSH(S, x)
 - POP(S)
 - MULTIPOP(S, k)

```

MULTIPOP( $S, k$ )
  while  $S$  is not empty and  $k > 0$ 
    POP( $S$ )
     $k = k - 1$ 
  
```





Stack Operation (Cont.)

- Running time of MULTIPOP
 - Linear in # of POP operations.
 - Let each PUSH/POP cost 1.
 - # of iterations of **while** loop is $\min(s, k)$, where $s = \#$ of objects on stack.
 - Therefore, total cost $\min(s, k)$.
- Sequence of n PUSH, POP, MULTIPOP operations:
 - Worst-case cost of MULTIPOP is $O(n)$.
 - Have n operations.
 - Therefore, worst-case cost of sequence is $O(n^2)$.
- **Observation**
 - Each object can be popped only once per time that it's pushed.
 - Have $\leq n$ PUSHes $\Rightarrow \leq n$ POPs, including those in MULTIPOP.
 - Therefore, total cost = $O(n)$.
 - Average over the n operations $\Rightarrow O(1)$ per operation on average.
- Again, notice no probability:
 - Showed *worst-case* $O(n)$ cost for sequence.
 - Therefore, $O(1)$ per operation on average. \rightarrow called **aggregate analysis**



Binary Counter

- k -bit binary counter $A[0..k-1]$ of bits, where $A[0]$ is the least significant bit and $A[k-1]$ is the most significant bit.
- Counts upward from 0.

- Value of counter is $\sum_{i=0}^{k-1} A[i] \cdot 2^i$.

- Initially, counter value is 0, so $A[0..k-1] = 0$.

- To increment, add 1 (mod 2^k):

```
INCREMENT( $A, k$ )
```

```
 $i = 0$ 
```

```
while  $i < k$  and  $A[i] == 1$ 
```

```
     $A[i] = 0$ 
```

```
     $i = i + 1$ 
```

```
if  $i < k$ 
```

```
     $A[i] = 1$ 
```



Binary Counter (Cont.)

- Each call could flip k bits, so n INCREMENTS takes $O(nk)$ time.
- **Observation**
 - 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

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



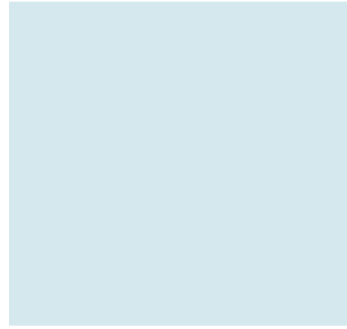
Binary Counter (Cont.)

- Analysis

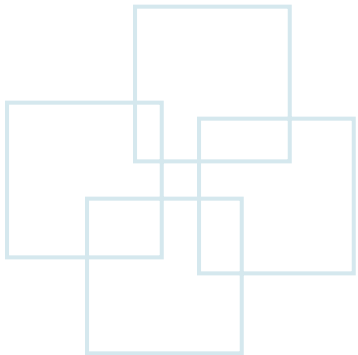
- n INCREMENTS costs $O(n)$.
- Average cost per operation = $O(1)$.

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

$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x}$$



The Accounting Method





Accounting Method

- Assign different charges to different operations.
 - Some are charged more than actual cost.
 - Some are charged less.
- **Amortized cost** = amount we charge.
- When **amortized cost > actual cost**, store the difference *on specific objects* in the data structure as **credit**.
- 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 same cost.
- **Need credit to never go negative**. Otherwise,
 - We have a sequence of operations for which the amortized cost is not an upper bound on actual cost.
 - Amortized cost would tell us *nothing*.



Accounting Method (Cont.)

Let c_i = actual cost of i th operation ,

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

Then require $\sum_{i=1}^n \hat{c}_i \geq \sum_{i=1}^n c_i$ for *all* sequences of n operations.

Total credit stored = $\sum_{i=1}^n \hat{c}_i - \underbrace{\sum_{i=1}^n c_i}_{\text{had better be}} \geq 0 .$



Stack Operation

• *Intuition*

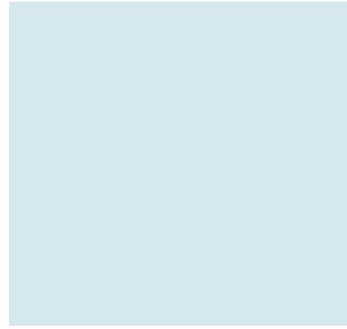
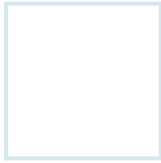
- When pushing an object, pay \$2.
 - \$1 pays for the PUSH.
 - \$1 is prepayment for it being popped by either POP or MULTIPOP.
 - Since each object has \$1, which is credit, the credit can never go negative.
 - **Total amortized cost = $O(n)$** is an upper bound on total actual cost.

operation	actual cost	amortized cost
PUSH	1	2
POP	1	0
MULTIPOP	$\min(k, s)$	0

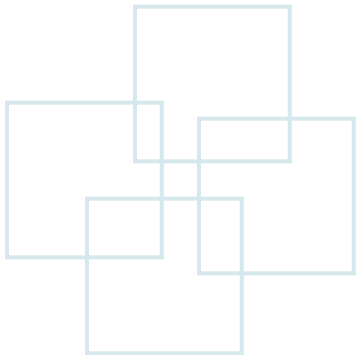


Binary Counter

- Charge \$2 to set a bit to 1.
 - \$1 pays for setting a bit to 1.
 - \$1 is prepayment for flipping it back to 0.
 - We have \$1 of credit for every 1 in the counter.
 - Therefore, $\text{credit} \geq 0$.
- Amortized cost of INCREMENT:
 - Cost of resetting bits to 0 is paid by credit.
 - At most 1 bit is set to 1.
 - Therefore, amortized cost \$2.
 - For n operations, amortized cost = $O(n)$.



The Potential Method





Potential Method

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



Potential Method (Cont.)

Let D_i = data structure after i th operation ,

D_0 = initial data structure ,

c_i = actual cost of i th operation ,

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

Potential function $\Phi : D_i \rightarrow \mathbb{R}$

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

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

$$= c_i + \underbrace{\Delta\Phi(D_i)} .$$

increase in potential due to i th operation



Potential Method (Cont.)

- Total amortized cost:

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

$$= \sum_{i=1}^n (c_i + \Phi(D_i) - \Phi(D_{i-1}))$$

(telescoping sum: every term other than D_0 and D_n is added once and subtracted once)

$$= \sum_{i=1}^n c_i + \Phi(D_n) - \Phi(D_0) .$$

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

In practice: $\Phi(D_0) = 0$, $\Phi(D_i) \geq 0$ for all i .



Stack Operation

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

Φ = # of objects in stack

(= # of \$1 bills in accounting method)

D_0 = empty stack $\Rightarrow \Phi(D_0) = 0$.

Since # of objects in stack is always ≥ 0 , $\Phi(D_i) \geq 0 = \Phi(D_0)$ for all i .

operation	actual cost c_i	$\Delta\Phi = \Phi(D_i) - \Phi(D_{i-1})$	\hat{c}_i amortized cost
PUSH	1	$(s + 1) - s = 1$ where $s = \#$ of objects initially	$1 + 1 = 2$
POP	1	$(s - 1) - s = -1$	$1 - 1 = 0$
MULTIPOP	$k' = \min(k, s)$	$(s - k') - s = -k'$	$k' - k' = 0$

The amortized cost of a sequence of n operations = $O(n)$.



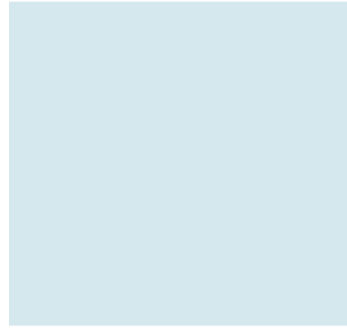
Binary Counter

- $\Phi = b_i = \#$ of 1's after i th INCREMENT
- Suppose i th operation resets t_i bits to 0.
- $c_i \leq t_i + 1$ (resets t_i bits, sets 1 bit to 1)
 - If $b_i = 0$, the i th operation reset all k bits and didn't set one, so $b_{i-1} = t_i = k \Rightarrow b_i = b_{i-1} - t_i$.
 - 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,

$$\begin{aligned} \Delta\Phi(D_i) &\leq (b_{i-1} - t_i + 1) - b_{i-1} \\ &= 1 - t_i . \end{aligned}$$

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

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



Dynamic Tables





Dynamic Tables

- A nice use of amortized analysis.
- **Scenario**
 - Have a table - maybe a hash table.
 - Don't know in advance how many objects will be stored in it.
 - When it fills, we must reallocate with a larger size, copying all objects into the new, larger table.
 - When it gets sufficiently small, *might* want to reallocate with a smaller size.
 - Details of table organization not important.



Dynamic Tables (Cont.)

- **Goals**

- $O(1)$ amortized time per operation.
- Unused space always \leq constant fraction of allocated space.

- **Load factor $\alpha = num/size$, where**

- num = # items stored,
- $size$ = allocated size.
- If $size = 0$, then $num = 0$. Call $\alpha = 1$.
- Never allow $\alpha > 1$
- Keep $\alpha >$ a constant fraction \Rightarrow goal (2).



Table Expansion

- Consider only insertion.
 - When the table becomes full, double its size and reinsert all existing items.
 - Guarantees that $\alpha \geq 1/2$.
 - Each time we actually insert an item into the table, it's an *elementary insertion*.



Table Expansion (Cont.)

Initially, $T.num = T.size = 0$.

TABLE-INSERT(T, x)

if $T.size == 0$

 allocate $T.table$ with 1 slot

$T.size = 1$

if $T.num == T.size$ // expand?

 allocate $new-table$ with $2 \cdot T.size$ slots

 insert all items in $T.table$ into $new-table$ // $T.num$ elem insertions

 free $T.table$

$T.table = new-table$

$T.size = 2 \cdot T.size$

insert x into $T.table$ // 1 elem insertion

$T.num = T.num + 1$



Running Time – Aggregate Analysis

- Charge 1 per elementary insertion.
- Count only elementary insertions, since all other costs together are constant per call.
- c_i = actual cost of i th operation.
 - If not full, $c_i = 1$.
 - If full, we have $i-1$ items in the table at the start of the i th operation.
 - We have to copy all $i-1$ existing items, then insert i th item $\Rightarrow c_i = i$.

n operations $\Rightarrow c_i = O(n) \Rightarrow O(n^2)$ time for n operations.

Not tight



Running Time - Aggregate Analysis (Cont.)

- Actual cost of i th operation (c_i):

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

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

aggregate analysis
says amortized cost
per operation = 3.



Accounting Method

- Charge \$3 per insertion of x .
 - \$1 pays for x 's insertion.
 - \$1 pays for x to be moved in the future.
 - \$1 pays for some other item to be moved.
- Suppose we've just expanded, $size = m$ before next expansion, $size = 2m$ after next expansion.
 - Assume that the expansion used up all the credit, so that there's no credit stored after the expansion.
 - It 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.
 - It will have \$2m of credit by next expansion, when there are $2m$ items to move.
 - Just enough to pay for the expansion, with no credit left over!



Potential Method

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

- Initially, $num = size = 0 \Rightarrow \Phi = 0$.
- Just after expansion, $size = 2 \cdot num \Rightarrow \Phi = 0$.
- Just before expansion, $size = num \Rightarrow \Phi = num \Rightarrow$ have enough potential to pay for moving all items.
- Need $\Phi \geq 0$, always.

Always have

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



Potential Method (Cont.)

- Amortized Cost of i th Operation \hat{c}_i

num_i = num after i th operation ,
 $size_i$ = $size$ after i th operation ,
 Φ_i = Φ after i th operation .

- If no expansion:

$$size_i = size_{i-1} ,$$

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

$$c_i = 1 .$$

Then we have

$$\hat{c}_i = c_i + \Phi_i - \Phi_{i-1}$$

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

$$= 1 + (2 \cdot num_i - size_i) - (2(num_i - 1) - size_i)$$

$$= 1 + 2$$

$$= 3 .$$



Potential Method (Cont.)

num_i = num after i th operation ,
 $size_i$ = $size$ after i th operation ,
 Φ_i = Φ after 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

$$\hat{c}_i = c_i + \Phi_i + \Phi_{i-1}$$

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

$$= num_i + (2 \cdot num_i - 2(num_i - 1)) - (2(num_i - 1) - (num_i - 1))$$

$$= num_i + 2 - (num_i - 1)$$

$$= 3 .$$

Φ_i



Potential Method (Cont.)

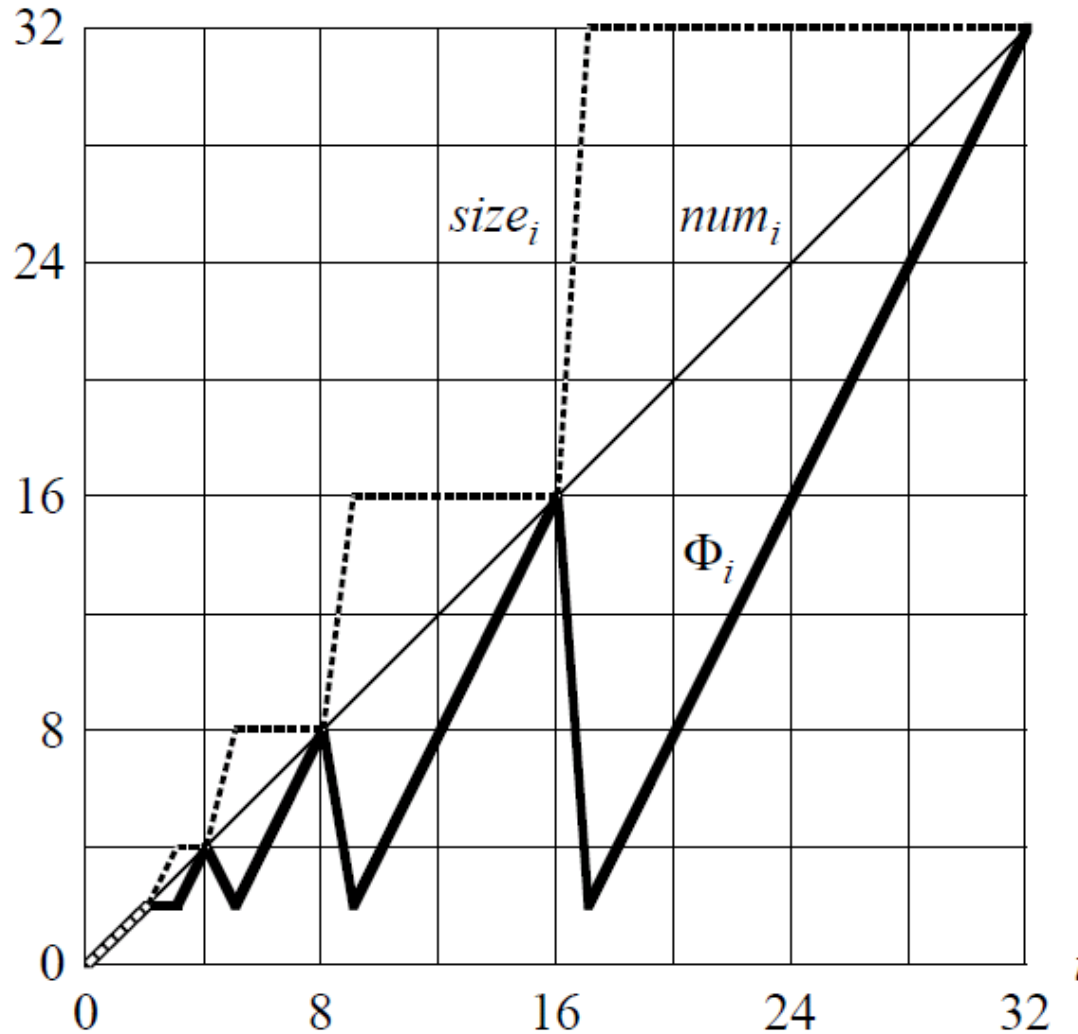




Table Expansion and Contraction

- When α drops too low, contract the table.
 - Allocate a new, smaller one.
 - Copy all items.
- Still want
 - α bounded from below by a constant,
 - Amortized cost per operation = $O(1)$.
- Measure cost in terms of elementary insertions and deletions.



Obvious Strategy

- Double size when inserting into a full table.
 - When $\alpha = 1$, so that after insertion α would become > 1 .
- Halve size when deletion would make table less than half full
 - When $\alpha = 1/2$, so that after deletion α would become $< 1/2$.
- Then always have $1/2 \leq \alpha \leq 1$.

Suppose we fill table.

Then insert \Rightarrow double

2 deletes \Rightarrow halve

2 inserts \Rightarrow double

2 deletes \Rightarrow halve

...

Not performing enough operations after expansion or contraction to pay for the next one.

The cost of each expansion and contraction is $\Theta(n)$ and there are $Q(n)$ operations. \rightarrow The total cost of the n operations is $\Theta(n^2)$.



Simple Solution

- Double as before: when inserting with $\alpha = 1 \Rightarrow$ after doubling, $\alpha = 1/2$.
- Halve size when deleting with $\alpha = 1/4 \Rightarrow$ after halving, $\alpha = 1/2$.
- Thus, immediately after either expansion or contraction, have $\alpha = 1/2$.
- Always have $1/4 \leq \alpha \leq 1$.

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

T empty $\Rightarrow \Phi = 0$.

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

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

Intuition

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



Further Intuition

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

Φ 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, have enough potential to pay for moving all num items.
- Potential increases linearly between $\alpha = 1/2$ and $\alpha = 1$, and it also increases linearly between $\alpha = 1/2$ and $\alpha = 1/4$.
- Since α has different distances to go to get to 1 or 1/4, starting from 1/2, rate of increase of Φ differs.
 - For α to go from 1/2 to 1, num increases from $size/2$ to $size$, for a total increase of $size/2$. Φ increases from 0 to $size$. Thus, Φ needs to increase by $\textcircled{2}$ for each item inserted. That's why there's a coefficient of 2 on the $T.num$ term in the formula for Φ when $\alpha \geq 1/2$.
 - For α to go from 1/2 to 1/4, num decreases from $size/2$ to $size/4$, for a total decrease of $size/4$. Φ increases from 0 to $size/4$. Thus, Φ needs to increase by $\textcircled{1}$ for each item deleted. That's why there's a coefficient of -1 on the $T.num$ term in the formula for Φ when $\alpha < 1/2$.



Amortized Costs: More Cases

Insert

- $\alpha_{i-1} \geq 1/2$, same analysis as before $\Rightarrow \hat{c}_i = 3$.
- $\alpha_{i-1} < 1/2 \Rightarrow$ no expansion (only occurs when $\alpha_{i-1} = 1$).
- If $\alpha_{i-1} < 1/2$ and $\alpha_i < 1/2$:

$$\begin{aligned}\hat{c}_i &= c_i + \Phi_i + \Phi_{i-1} \\ &= 1 + (size_i/2 - num_i) - (size_{i-1}/2 - num_{i-1}) \\ &= 1 + (size_i/2 - num_i) - (size_i/2 - (num_i - 1)) \\ &= 0.\end{aligned}$$

- If $\alpha_{i-1} < 1/2$ and $\alpha_i \geq 1/2$:

$$\begin{aligned}\hat{c}_i &= 1 + (2 \cdot num_i - size_i) - (size_{i-1}/2 - num_{i-1}) \\ &= 1 + (2(num_{i-1} + 1) - size_{i-1}) - (size_{i-1}/2 - num_{i-1}) \\ &= 3 \cdot num_{i-1} - \frac{3}{2} \cdot size_{i-1} + 3 \\ &= 3 \cdot \alpha_{i-1} size_{i-1} - \frac{3}{2} \cdot size_{i-1} + 3 \\ &< \frac{3}{2} \cdot size_{i-1} - \frac{3}{2} \cdot size_{i-1} + 3 \\ &= 3.\end{aligned}$$

amortized cost of insert is < 3 .



Amortized Costs: More Cases (Cont.)

Delete

- If $\alpha_{i-1} < 1/2$, then $\alpha_i < 1/2$.

- If no contraction:

$$\begin{aligned}\hat{c}_i &= 1 + (size_i/2 - num_i) - (size_{i-1}/2 - num_{i-1}) \\ &= 1 + (size_i/2 - num_i) - (size_i/2 - (num_i + 1)) \\ &= 2.\end{aligned}$$

- If contraction:

$$\begin{aligned}\hat{c}_i &= \underbrace{(num_i + 1)}_{\text{move + delete}} + (size_i/2 - num_i) - (size_{i-1}/2 - num_{i-1}) \\ &\quad [size_i/2 = size_{i-1}/4 = num_{i-1} = num_i + 1] \\ &= (num_i + 1) + ((num_i + 1) - num_i) - ((2 \cdot num_i + 2) - (num_i + 1)) \\ &= 1.\end{aligned}$$



Amortized Costs: More Cases (Cont.)

- If $\alpha_{i-1} \geq 1/2$, then no contraction.

- If $\alpha_i \geq 1/2$:

$$\begin{aligned}\hat{c}_i &= 1 + (2 \cdot \text{num}_i - \text{size}_i) - (2 \cdot \text{num}_{i-1} - \text{size}_{i-1}) \\ &= 1 + (2 \cdot \text{num}_i - \text{size}_i) - (2 \cdot \text{num}_i + 2 - \text{size}_i) \\ &= -1.\end{aligned}$$

- If $\alpha_i < 1/2$, since $\alpha_{i-1} \geq 1/2$, have

$$\text{num}_i = \text{num}_{i-1} - 1 \geq \frac{1}{2} \cdot \text{size}_{i-1} - 1 = \frac{1}{2} \cdot \text{size}_i - 1.$$

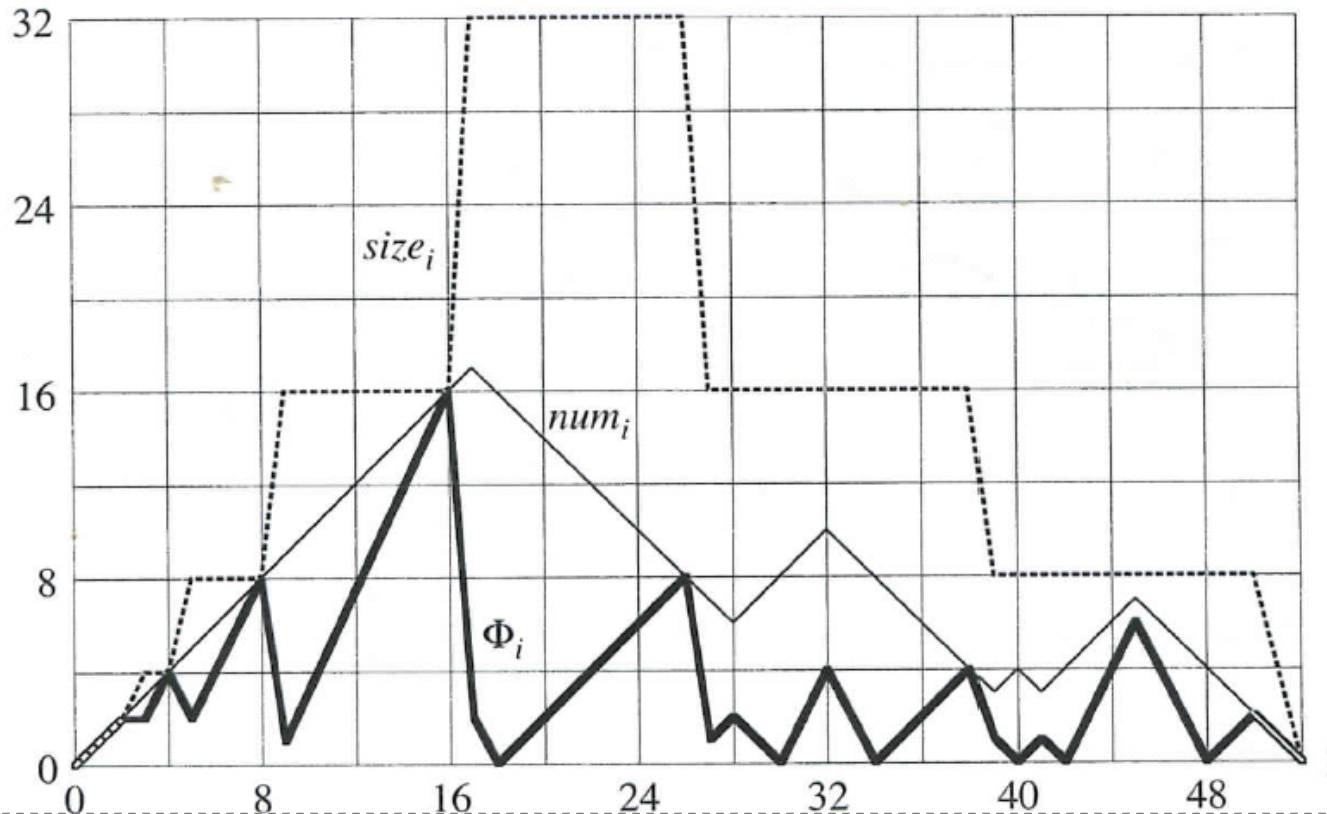
Thus,

$$\begin{aligned}\hat{c}_i &= 1 + (\text{size}_i/2 - \text{num}_i) - (2 \cdot \text{num}_{i-1} - \text{size}_{i-1}) \\ &= 1 + (\text{size}_i/2 - \text{num}_i) - (2 \cdot \text{num}_i + 2 - \text{size}_i) \\ &= -1 + \frac{3}{2} \cdot \text{size}_i - 3 \cdot \text{num}_i \\ &\leq -1 + \frac{3}{2} \cdot \text{size}_i - 3 \left(\frac{1}{2} \cdot \text{size}_i - 1 \right) \\ &= 2.\end{aligned}$$

amortized cost of delete is ≤ 2 .



Example



The effect of a sequence of n TABLE-INSERT and TABLE-DELETE operations on the number num_i of items in the table, the number $size_i$ of slots in the table, and the potential

$$\Phi_i = \begin{cases} 2 \cdot num_i - size_i & \text{if } \alpha_i \geq 1/2, \\ size_i/2 - num_i & \text{if } \alpha_i < 1/2, \end{cases}$$