

Amortized Analysis

Heejin Park

Hanyang University

Contents

- **Aggregate analysis**
- **Accounting method**
- **Potential method**
- **Dynamic Table**

Aggregate analysis

- Example of stack operation
 - Stack operations
 - $\text{PUSH}(x)$
 - $\text{POP}()$
 - PUSH and POP run in $O(1)$ time.
 - Thus the cost of each is 1.
 - Actual running time for n operations is $\Theta(n)$.

Aggregate analysis

- Example of stack operation

- **MULTIPOP(k)**

- Actual running time is linear in the number of POP operations actually executed.

MULTIPOP(S, k)

1 while not STACK-EMPTY(S) and $k > 0$

2 POP(S)

3 $k = k - 1$

← $T(1) * k$

- So, cost of MULTIPOP(S, k) is $O(k)$.

Aggregate analysis

- Example of stack operation

- MULTIPOP(k)

- Remove 2 top objects

- Remove 4 top objects

index	1	2	3	4	5	6
S						

MULTIPOP($S, 2$)

↑
top

index	1	2	3	4	5	6
S						

MULTIPOP($S, 6$)

↑
top

index	1	2	3	4	5	6
S						

Aggregate analysis

- Example of stack operation

- Analysis of a sequence of n PUSH, POP and MULTIPOP operations
 - on an initially empty stack
- Intuitive analysis of time complexity (wrong way)
 - The worst-case cost of one MULTIPOP: $O(n)$
 - Stack size: at most n

➔ Total cost : $O(n^2)$

- This cost isn't tight

Aggregate analysis

- Example of stack operation
 - Using Aggregate analysis
 - Can obtain a better upper bound the entire sequence of n operations
 - Any sequence of n PUSH, POP and MULTIPOP operations
 - on an initially empty stack
 - [Push, push, pop, push, push, push, multipop(2), ...]
= [Push, push, pop, push, push, push, {pop, pop}, ...]
$$n \geq \#(\text{push}) \geq \#(\text{pop})$$
$$2n \geq \#(\text{push}) + \#(\text{pop})$$

➔ Total cost : $O(n)$
 - Amortized cost is $O(n) / n = O(1)$

Aggregate analysis

- Example of incrementing binary counter
 - Consider the problem of implementing a k -bit binary counter that counts upward from 0
 - Use an array $A[0..k-1]$ of bits

$A[k-1]$...	$A[2]$	$A[1]$	$A[0]$
----------	-----	--------	--------	--------

Aggregate analysis

- Example of incrementing binary counter
 - Cost of INCREMENT operation is proportional to the number of bits flip

INCREMENT (A)

1 $i = 0$

2 **while** $i < A.length$ and $A[i] == 1$

3 $A[i] = 0$

4 $i = i + 1$

5 **if** $i < A.length$

6 $A[i] = 1$

← Bit flip, $T(1)$

← Bit flip, $T(1)$

Aggregate analysis

- Example of incrementing binary counter
 - Cost of INCREMENT operation is proportional to the number of bits flip

Ex. INCREMENT(*A*)

Counter value	<i>A</i> [4]	<i>A</i> [3]	<i>A</i> [2]	<i>A</i> [1]	<i>A</i> [0]	cost	Total cost
0	0	0	0	0	0	0	0
1	0	0	0	0	1	1	1

Aggregate analysis

- Example of incrementing binary counter
 - Cost of INCREMENT operation is proportional to the number of bits flip

Ex. INCREMENT(A)

Counter value	A[4]	A[3]	A[2]	A[1]	A[0]	cost	Total cost
0	0	0	0	0	0	0	0
1	0	0	0	0	1	1	1
2	0	0	0	1	0	2	3

Aggregate analysis

- Example of incrementing binary counter
 - Cost of INCREMENT operation is proportional to the number of bits flip

Ex. INCREMENT(A)

Counter value	A[4]	A[3]	A[2]	A[1]	A[0]	cost	Total cost
0	0	0	0	0	0	0	0
1	0	0	0	0	1	1	1
2	0	0	0	1	0	2	3
3	0	0	0	1	1	1	4

Aggregate analysis

- Example of incrementing binary counter
 - Cost of INCREMENT operation is proportional to the number of bits flip

Ex. INCREMENT(*A*)

Counter value	<i>A</i> [4]	<i>A</i> [3]	<i>A</i> [2]	<i>A</i> [1]	<i>A</i> [0]	cost	Total cost
0	0	0	0	0	0	0	0
1	0	0	0	0	1	1	1
2	0	0	0	1	0	2	3
3	0	0	0	1	1	1	4
4	0	0	1	0	0	3	7

Aggregate analysis

- Example of incrementing binary counter
 - A single execution of INCREMENT takes time $\Theta(k)$ in the worst case
 - In which array A contains all 1s.

$A[k-1]$...	$A[2]$	$A[1]$	$A[0]$	cost
1	...	1	1	1	-
0	...	0	0	0	k

- Thus, a sequence of n INCREMENT operations on an initially zero counter takes time $O(nk)$ in the worst case.

Aggregate analysis

- Example of incrementing binary counter
 - Aggregate Analysis
 - can tighten our analysis to yield a worst-case cost of $O(n)$ for a sequence of n INCREMENT operations
 - by observing that not all bits flip each time INCREMENT is called

Aggregate analysis

● Example of incrementing binary counter

● Compute bit flip of Array A

- Time of flip of $A[0]$: n
- Time of flip of $A[1]$: $\lfloor n/2 \rfloor$
- Time of flip of $A[2]$: $\lfloor n/4 \rfloor$

● The total number of flip in the sequence

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

➔ Total cost $O(n)$

● Amortized cost = $O(n)/n = O(1)$

	A[3]	A[2]	A[1]	A[0]
1	0	0	0	1
2	0	0	1	0
3	0	0	1	1
4	0	1	0	0
5	0	1	0	1
6	0	1	1	0
7	0	1	1	1
8	1	0	0	0
	↑ $n/8$	↑ $n/4$	↑ $n/2$	↑ n

Contents

- *Aggregate analysis*
- **Accounting method**
- **Potential method**
- **Dynamic Table**

Accounting method

Accounting method

- We want to show that in the worst case the average cost per operation is small by analyzing with amortized costs,
 - c_i : actual cost of the i th operation
 - \hat{c}_i : amortized cost of the i th operation
 - $\sum_{i=0}^n \hat{c}_i \geq \sum_{i=0}^n c_i$: all sequences of n operations required
- The total credit
 - $\sum_{i=0}^n \hat{c}_i - \sum_{i=0}^n c_i$

Accounting method

- Example of stack operation
 - The actual costs of the operations
 - PUSH 1
 - POP 1
 - MULTIPOP(k) $\min(k, s)$
 - The amortized costs of the operations
 - PUSH 2
 - POP 0
 - MULTIPOP(k) 0

Accounting method

● Example of stack operation

index	1	2	3	4
<i>S</i>				

cost				
------	--	--	--	--

<i>S</i>	1	2	3	4
credit				

Accounting method

● Example of stack operation

● PUSH

index	1	2	3	4
S				

↑
top

cost	1			
------	---	--	--	--

S	1	2	3	4
credit	1			

- PUSH : actual cost 1 + prepaid credit 1
- Amortized cost : actual cost + credit = 2

Accounting method

● Example of stack operation

● PUSH

index	1	2	3	4
S				

↑
top

cost	1	1		
------	---	---	--	--

S	1	2	3	4
credit	1	1		

- PUSH : actual cost 1 + prepaid credit 1
- Amortized cost : actual cost + credit = 2

Accounting method

● Example of stack operation

● PUSH

index	1	2	3	4
S				

↑
top

cost	1	1	1	
------	---	---	---	--

S	1	2	3	4
credit	1	1	1	

- PUSH : actual cost 1 + prepaid credit 1
- Amortized cost : actual cost + credit = 2

Accounting method

● Example of stack operation

● POP

index	1	2	3	4
S				

↑
top

cost	1	1	1	
------	---	---	---	--

S	1	2	3	4
credit	1	1	0	

- POP and MULTIPOP : pay credit 1
- Amortized cost : $\text{actual cost} - \text{credit} = 0$

Accounting method

● Example of stack operation

● PUSH

index	1	2	3	4
S				

↑
top

cost	1	1	1	1
------	---	---	---	---

S	1	2	3	4
credit	1	1	1	

- PUSH : actual cost 1 + prepaid of credit 1
- Amortized cost : actual cost + credit = 2

Accounting method

- Example of stack operation
 - POP and MULTIPOP must execute after PUSH operation
 - Charging the PUSH operation a little bit more (= credit)
So, credit pay actual cost of POP and MULTIPOP operation
 - The amount of credit is always nonnegative
 - Because the stack always has nonnegative objects.
 - Thus, the total amortized cost is an upper bound on the total actual cost
 - Total amortized cost : $O(n)$
 - Total actual cost : $O(n)$

Accounting method

● Example of incrementing binary counter

● The actual costs

- Bit set ($0 \rightarrow 1$) : 1
- Bit reset ($1 \rightarrow 0$) : 1

● The amortized costs

- Bit set : 2
- Bit reset : 0

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1

cost	1				
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit	1				

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1
			0

cost	1				
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit					

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0

cost	1	1			
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit		1			

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0
0	0	1	1

cost	1	1	1		
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit	1	1			

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0
0	0	1	1
			0

cost	1	1	1		
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit		1			

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0
0	0	1	1
		0	0

cost	1	1	1		
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit					

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0
0	0	1	1
	1	0	0

cost	1	1	1	1	
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit			1		

Accounting method

● Example of incrementing binary counter

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0
0	0	1	1
0	1	0	0
0	1	0	1

cost	1	1	1	1	1
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit	1		1		

Accounting method

● Example of incrementing binary counter

- Bit reset must execute after bit set

- Charging the bit set in credit

So, credit pay for actual cost of reset operation

- The amount of credit is always nonnegative

Because the number of 1s in the counter never becomes negative

Thus, the total amortized cost is an upper bound on the total actual cost

- the total amortized cost : $O(n)$

- the total actual cost : $O(n)$

Accounting method

- Amortized cost
 - $O(n)$ time in total
- Running time
 - $O(n)$ time in total

Contents

- *Aggregate analysis*
- *Accounting method*
- **Potential method**
- **Dynamic Table**

Potential method

- Potential method
 - Similar to accounting method
 - Credit → “potential energy” or just “potential”
 - The potential with the data structure as a whole rather than with specific objects within the data structure.

Potential method

● Potential method

A[3]	A[2]	A[1]	A[0]
0	0	0	1

Credit

cost	1				
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit	1				

Potential

1	1
---	---

cost

Potential

Potential method

● Potential method

A[3]	A[2]	A[1]	A[0]
0	0	0	1
			0

Credit

cost	1				
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit					

Potential

1
cost Potential

Potential method

Potential method

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0

Credit

cost	1	1			
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit		1			

Potential

1	
1	1

cost

Potential

Potential method

● Potential method

A[3]	A[2]	A[1]	A[0]
0	0	0	1
0	0	1	0
0	0	1	1

Credit

cost	1	1	1		
A	A[0]	A[1]	A[2]	A[3]	A[4]
credit	1	1			

Potential

1	
1	1
1	1

cost

Potential

Potential method

● Potential method

- will perform n operations,

D_0 : an initial data structure

D_i : the data structure that results after applying the i th operation to data structure D_{i-1}

$\Phi(D_i)$: the potential associated with data structure D_i

- Potential difference ($\Phi(D_i) - \Phi(D_{i-1})$)

- positive

The potential of the data structure increases

- negative

The decrease in the potential pays for the actual cost of the operation

Potential method

- Potential method

- Amortized cost

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

- The total amortized cost of the n operations

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

- We require $\Phi(D_i) \geq \Phi(D_0)$ for all i

- So that $\sum_{i=1}^k \hat{c}_i \geq \sum_{i=1}^k c_i$ for all $1 \leq k \leq n$

Potential method

- Example of stack operation
 - Potential function Φ
 - the number of objects in the stack
 - $\Phi(D_0) = 0$
 - The stack D_i after the i th operation has nonnegative potential
 - $\Phi(D_i) \geq 0 = \Phi(D_0)$

Potential method

- Example of stack operation

- Amortized cost analysis of each operation

- PUSH operation

- If the i th operation on a stack containing s objects is a PUSH operation,

- $\Phi(D_i) - \Phi(D_{i-1}) = (s + 1) - s = 1$

- So, the amortized cost is $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$
 $= 1 + (s + 1) - s$
 $= 2$

- POP operation

- If the i th operation on a stack containing s objects is a POP operation,

- $\Phi(D_i) - \Phi(D_{i-1}) = (s - 1) - s = -1$

- So, the amortized cost is $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$
 $= 1 + (s - 1) - s$
 $= 0$

Potential method

- Example of stack operation

- Amortized cost analysis of each operation

- MULTIPOP(S, k) operation

- If the i th operation on a stack containing s objects is a MULTIPOP operation,
- $k' = \min(k, s)$: The number of objects to be popped off the stack
- $\Phi(D_i) - \Phi(D_{i-1}) = -\min(k, s) = -k'$

$$\begin{aligned}\text{The amortized cost is } \hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &= k' - k' \\ &= 0\end{aligned}$$

Potential method

- Example of stack operation

- Amortized cost : $O(1)$
- Total amortized cost : $O(n)$
- Total actual cost : $O(n)$

3. Potential method

- Example of incrementing binary counter
 - Potential function Φ
 - The number of 1s in the array
 - b_i : The number of 1s in the counter after the i th INCREMENT operation
 - t_i : The number of bits reset in the i th INCREMENT operation
 - Actual cost of the operation
 - $c_i \leq t_i + 1$
 - since in addition to resetting t_i bits, it sets at most one bit to 1

INCREMENT (A)

```
1   $i = 0$ 
2  while  $i < A.length$  and  $A[i] == 1$ 
3       $A[i] = 0$ 
4       $i = i + 1$ 
5  if  $i < A.length$ 
6       $A[i] = 1$ 
```


3. Potential method

Example of incrementing binary counter

- Case of $b_i = 0$
 - the i th operation resets all k bits
 - $b_{i-1} = t_i = k$
- Case of $b_i > 0$
 - $b_i = b_{i-1} - t_i + 1$
- In either case
 - $b_i \leq b_{i-1} - t_i + 1$

Ex) 1111 \rightarrow 0000

Counter value	$A[k]$...	$A[2]$	$A[1]$	$A[0]$	b_i
$i-1$	1	...	1	1	1	k
i	0	...	0	0	0	0

Ex) 0111 \rightarrow 1000

Counter value	$A[k]$...	$A[2]$	$A[1]$	$A[0]$	b_i
$i-1$	0	...	1	1	1	$k-1$
i	1	...	0	0	0	1

3. Potential method

- Example of incrementing binary counter

- Potential difference

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

- Amortized cost

- $$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &\leq (t_i + 1) + (1 - t_i) = 2 \\ &\rightarrow O(1)\end{aligned}$$

3. Potential method

- Example of incrementing binary counter
 - If the counter starts at zero, $\Phi(D_0) = 0$ and since $\Phi(D_i) \geq 0$ for all i
 - The total amortized cost of a sequence of n INCREMENT operations is an upper bound on the total actual cost
 - The worst-case cost of n INCREMENT operations is $O(n)$

3. Potential method

- Example of incrementing binary counter
 - If does not start at zero
 - $b_0 \geq 0, b_n \leq k$ (k : the number of bits in the counter)
 - $\sum_{i=1}^n \hat{c}_i = \sum_{i=1}^n c_i + \Phi(D_n) - \Phi(D_0)$
 - $\sum_{i=1}^n c_i = \sum_{i=1}^n \hat{c}_i - \Phi(D_n) + \Phi(D_0)$ ($\hat{c}_i \leq 2$ for all $1 \leq i \leq n$)
$$\leq \sum_{i=1}^n 2 - b_n + b_0 \quad (\Phi(D_n) = b_n, \Phi(D_0) = b_0)$$
$$= 2n - b_n + b_0$$
 - The total actual cost is $O(n)$ ($b_0 \leq k, k = O(n)$)

Contents

- *Aggregate analysis*
- *Accounting method*
- *Potential method*
- **Dynamic Table**

Dynamic tables

- Table allocation problem
- We do not always know in advance how many objects some applications will store in a table
 - insertion
 - So allocate space for a table and reallocate the table when new item is added.
 - deletion
 - Similarly, if many objects have been deleted from the table, it may be worthwhile to reallocate the table with a smaller size
- Using amortized analysis, we shall show that the amortized cost of insertion and deletion is only $O(1)$

Aggregate analysis

● INSERT

- When inserting an item into a full table, we can expand the table by allocating a new table with more slots than the old table had.
- A common heuristic allocates a new table with **twice** as many slots as the old one.

Aggregate analysis

● INSERT

- *T.table* : a pointer to the block of storage representing the table.
- *T.num* : the number of items in the table
- *T.size* : the total number of slots in the table.

Aggregate analysis

TABLE-INSERT(T, x)

1 **if** $T.size == 0$

2 allocate $T.table$ with 1 slot

3 $T.size = 1$

4 **if** $T.num == T.size$

5 allocate *new-table* with $2 * T.size$ slots

6 insert all items in $T.table$ into *new-table*

7 free $T.table$

8 $T.table = new-table$

9 $T.size = 2 * T.size$

10 insert x into $T.table$

11 $T.num = T.num + 1$

elementary insertion

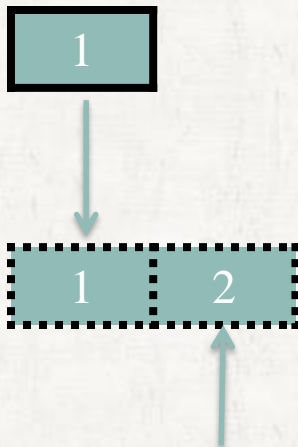
expansion

Aggregate analysis

- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - If the current table has room for the new item, then cost $c_i = 1$.
 - If the current table is full, an expansion occurs, then $c_i = i$.
 - 1 for insert new item, $i-1$ for move for extend

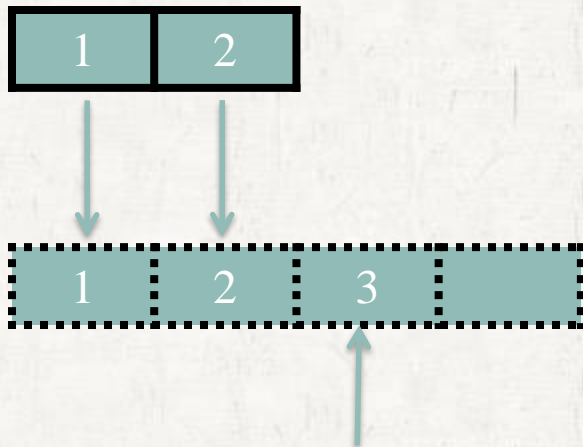
Aggregate analysis

- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - If the current table has room for the new item, then cost $c_i = 1$.
 - If the current table is full, an expansion occurs, then $c_i = i$.
 - 1 for insert new item, $i-1$ for move for extend



Aggregate analysis

- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - If the current table has room for the new item, then cost $c_i = 1$.
 - If the current table is full, an expansion occurs, then $c_i = i$.
 - 1 for insert new item, $i-1$ for move for extend



Aggregate analysis

- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - If the current table has room for the new item, then cost $c_i = 1$.
 - If the current table is full, an expansion occurs, then $c_i = i$.
 - 1 for insert new item, $i-1$ for move for extend



Aggregate analysis

- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.

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

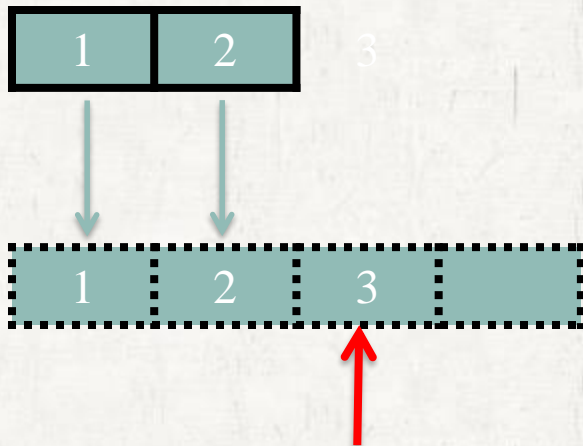
Aggregate analysis

- The total cost of n TABLE-INSERT operations is therefore

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

Aggregate analysis

- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - For 1 to n , when item inserted in table, it's cost is 1.
 - It requires $1 * n = n$ cost.
 - It is expressed by the red arrow.



$$\boxed{n} + \sum_{j=0}^{\lfloor \lg n \rfloor} 2^j$$

Aggregate analysis

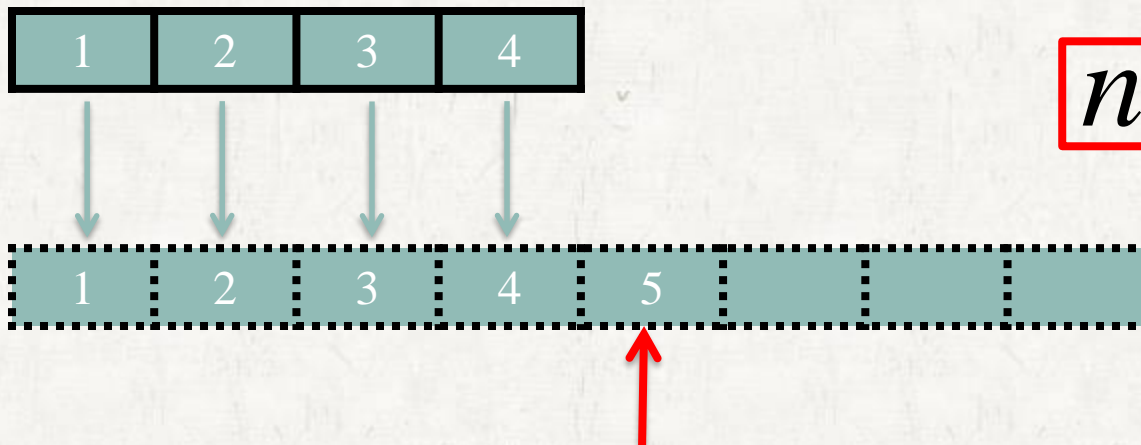
- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - For 1 to n , when item inserted in table, it's cost is 1.
 - It requires $1 * n = n$ cost.
 - It is expressed by the red arrow.



$$\boxed{n} + \sum_{j=0}^{\lfloor \lg n \rfloor} 2^j$$

Aggregate analysis

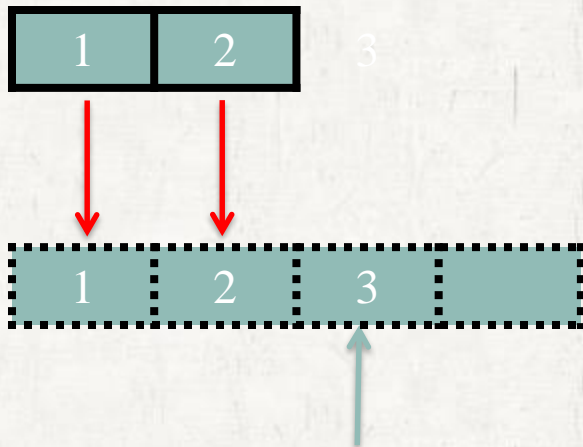
- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - For 1 to n , when item inserted in table, it's cost is 1.
 - It requires $1 * n = n$ cost.
 - It is expressed by the red arrow.



$$\boxed{n} + \sum_{j=0}^{\lfloor \lg n \rfloor} 2^j$$

Aggregate analysis

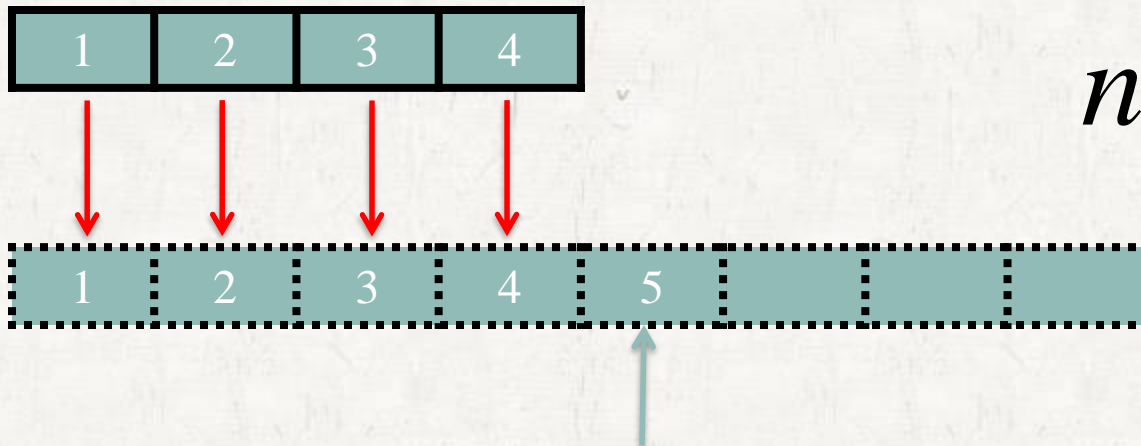
- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - When table size is exact power of 2, table expansion occur
 - 2^j insert is occurred.
 - And it occurred $\lfloor \lg n \rfloor$ times.
 - It is expressed by the red arrow



$$n + \sum_{j=0}^{\lfloor \lg n \rfloor} 2^j$$

Aggregate analysis

- Let us analyze a sequence of n TABLE-INSERT operations on an initially empty table.
 - When table size is exact power of 2, table expansion occur
 - 2^j insert is occurred.
 - And it occurred $\lfloor \lg n \rfloor$ times.
 - It is expressed by the red arrow



$$n + \sum_{j=0}^{\lfloor \lg n \rfloor} 2^j$$

Aggregate analysis

- The total cost of n TABLE-INSERT operations is therefore

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

- Since the total cost of n TABLE-INSERT operations is bounded by $3n$, the amortized cost of single operation is at most 3 ($3n / n$)

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.

Accounting method

TABLE-INSERT(T, x)

```
1      if  $T.size == 0$ 
2          allocate  $T.table$  with 1 slot
3           $T.size = 1$ 
4      if  $T.num == T.size$ 
5          allocate  $new-table$  with  $2 * T.size$  slots
6          insert all items in  $T.table$  into  $new-table$ 
7          free  $T.table$ 
8           $T.table = new-table$ 
9           $T.size = 2 * T.size$ 
10     insert  $x$  into  $T.table$ 
11      $T.num = T.num + 1$ 
```

elementary insertion

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - There are two types of elementary insertion:
 - 6 insert all items in $T.table$ into new-table
 - 10 insert x into $T.table$

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - 1 **cost** for line 10,
 - 2 **credit** for line 6.
 - Credit is used to move items when expansion occurs.

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands

1	2		
0	0		

Credit for move

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands

1	2	3	
0	0	0	

Credit for move

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands

1	2	3	
0	0	1	

Credit for move

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands

1	2	3	
1	0	1	

Credit for move

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands

1	2	3	4
1	1	1	1


Credit for move

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands

1	2	3	4
1	1	1	1

Credit for move




1							
0							

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands

1	2	3	4
1	1	1	1

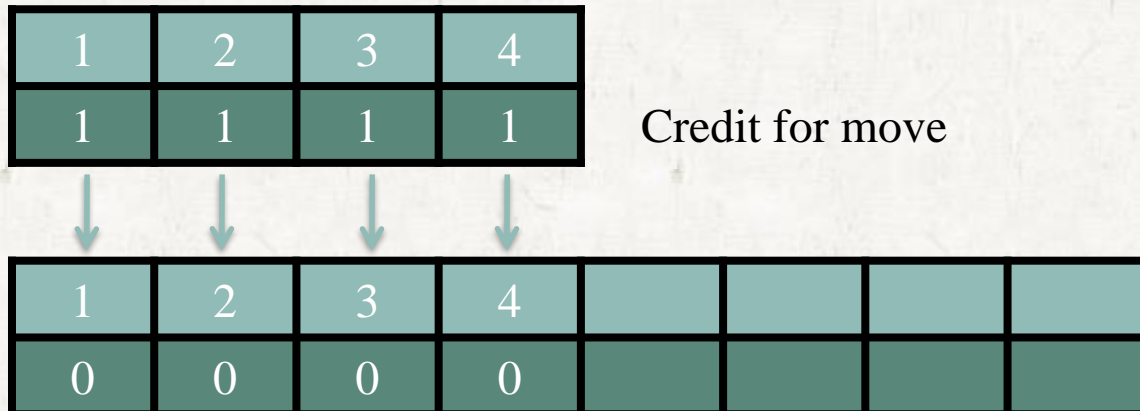
Credit for move



1	2						
0	0						

Accounting method

- By using the accounting method, we can gain some feeling for why the amortized cost of a TABLE-INSERT operation should be 3.
 - each item pays for 3 elementary insertions:
 - inserting itself into the current table
 - moving itself when the table expands
 - moving another item that has already been moved once when the table expands



Potential method

- We can use the potential method to analyze a sequence of n TABLE-INSERT operations.
 - and we shall use it in Section 17.4.2 to design a TABLE-DELETE operation that has an $O(1)$ amortized cost as well

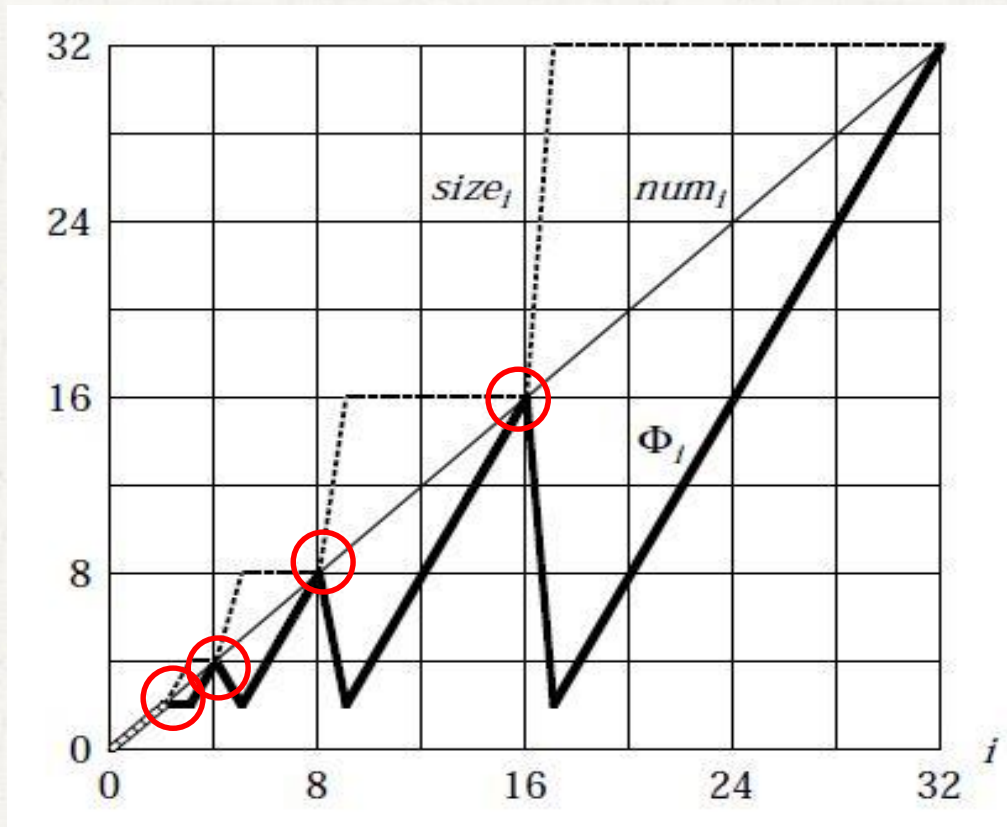
Potential method

- We can use the potential method to analyze a sequence of n TABLE-INSERT operations.
 - and we shall use it in Section 17.4.2 to design a TABLE-DELETE operation that has an $O(1)$ amortized cost as well
- We start by defining a potential function Φ
 - 0 immediately after an expansion
 - table size by the time the table is full

Potential method

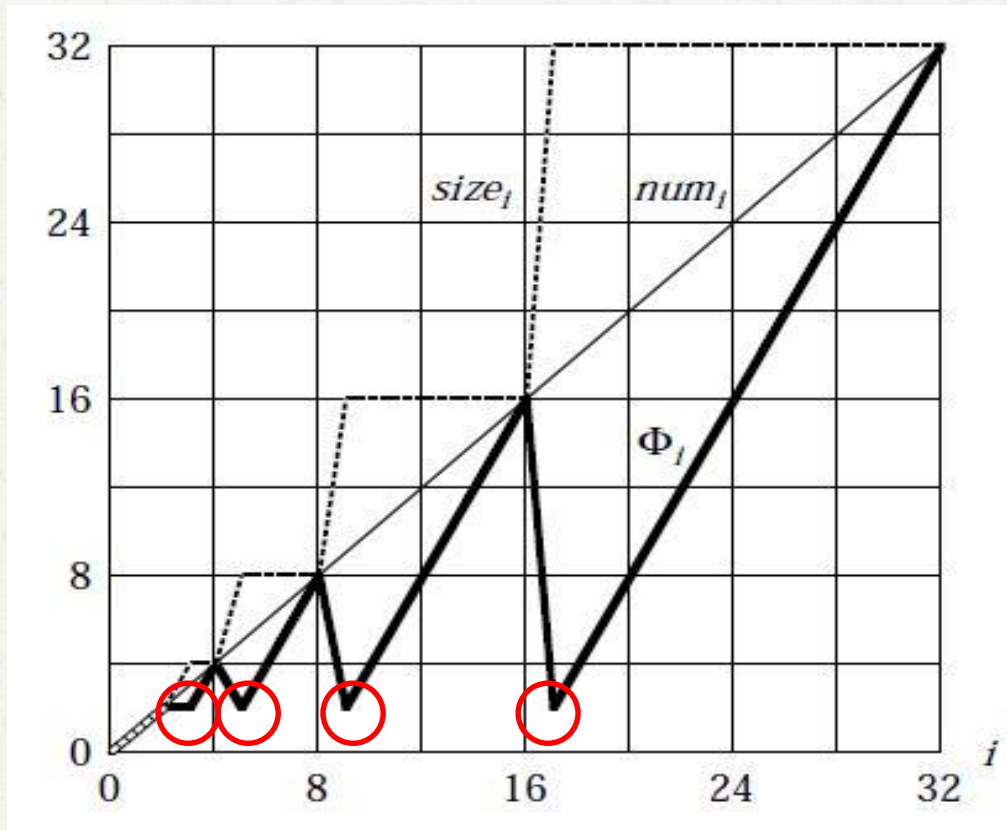
- $\Phi(T) = 2 * T.num - T.size$ (17.5)
- Immediately before an expansion, we have $T.num = T.size$ and thus $\Phi(T) = T.num$
- $\Phi(T)$ is always nonnegative
 - The initial value of the potential is 0
 - and since the table is always at least half full, $T.num \geq T.size/2$

Potential method



- Before expansion, $\Phi_i = num_i$

Potential method



- After expansion, $\Phi_i = 0$ but immediately increased by 2

Potential method

- The amortized cost of the i th TABLE-INSERT operation
 - num_i : the number of items in the table after the i th operation
 - $size_i$: the total size of the table after the i th operation
 - Φ_i : the potential after the i th operation
 - \hat{c}_i : its amortized cost with respect to Φ
- Initially, we have $num_0 = 0$, $size_0 = 0$, and $\Phi_0 = 0$.

Potential method

- The amortized cost of the i th TABLE-INSERT operation
 - If the i th TABLE-INSERT operation does not trigger an expansion, then we have $size_i = size_{i-1}$ and the amortized cost of the operation is
 - $\Phi(T) = 2 * T.num - T.size$

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

Potential method

- The amortized cost of the i th TABLE-INSERT operation

- If the i th operation does trigger an expansion, then we have

$$size_i = 2 * size_{i-1}$$

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

$$size_i = 2 * (num_i - 1).$$

Thus, the amortized cost of the operation is

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

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

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

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

$$= 3$$

Table expansion and contraction

- TABLE-DELETE operation.

- Table contraction is analogous to table expansion:
 - when the number of items in the table drops too low, we allocate a new, smaller table and then copy the items from the old table into the new one

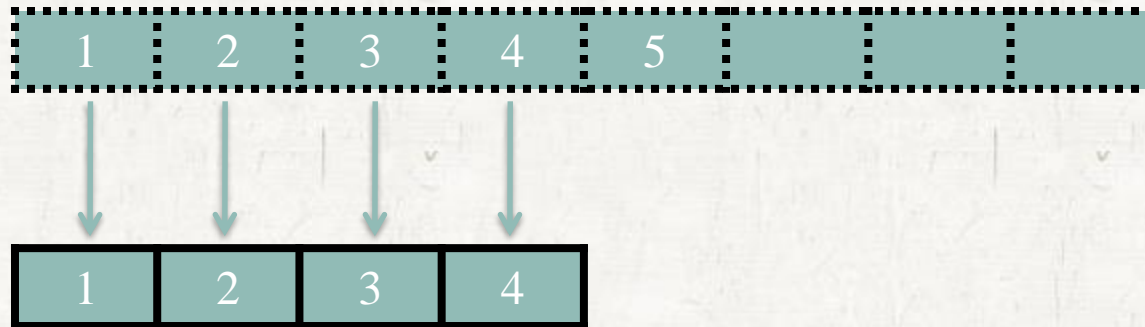


Table expansion and contraction

- TABLE-DELETE operation.
 - load factor : $\alpha(T) = T.num / T.size$



- we would like to preserve two properties:
 - the load factor of the dynamic table is bounded below by a positive constant
 - the amortized cost of a table operation is bounded above by a constant.

Table expansion and contraction

- Table expansion and contraction
 - double the table size upon inserting an item into a full table
 - halve the size when deleting an item would cause the table to become less than half full
 - This strategy would guarantee that the load factor of the table never drops below $1/2$, but have a **problem**

Table expansion and contraction

- Table expansion and contraction
 - We perform n operations on a table T , where n is an exact power of 2.
 - The first $n/2$ operations are insertions,
 - cost a total of $\Theta(n)$.
 - At the end of this sequence of insertions, $T.num = T.size = n/2$.
 - For the second $n/2$ operations, we perform the following sequence:
 - insert, delete, delete, insert, insert, delete, delete, insert, insert,

Table expansion and contraction

- Table expansion and contraction
 - First $n/2$ insertions

1	$n/2$
---	------	-------

Table expansion and contraction

- Table expansion and contraction
 - First $n/2$ insertions
 - **insert**, delete, delete, insert, insert, delete, delete, insert, insert, . . .

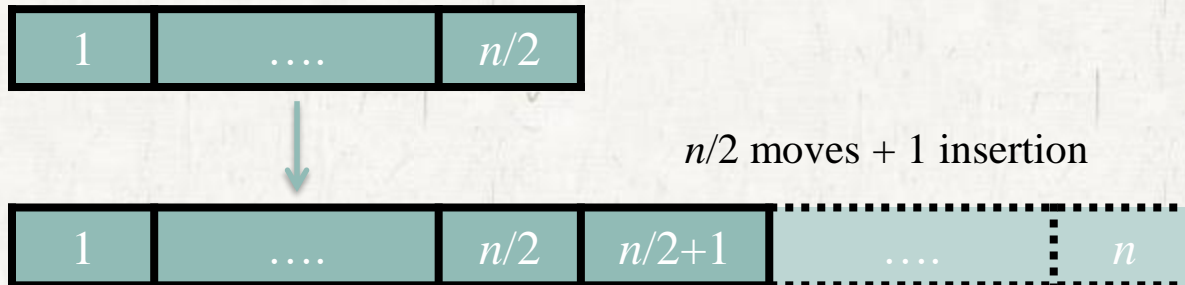


Table expansion and contraction

- Table expansion and contraction

- **insert, delete**, delete, insert, insert, delete, delete, insert, insert, . . .

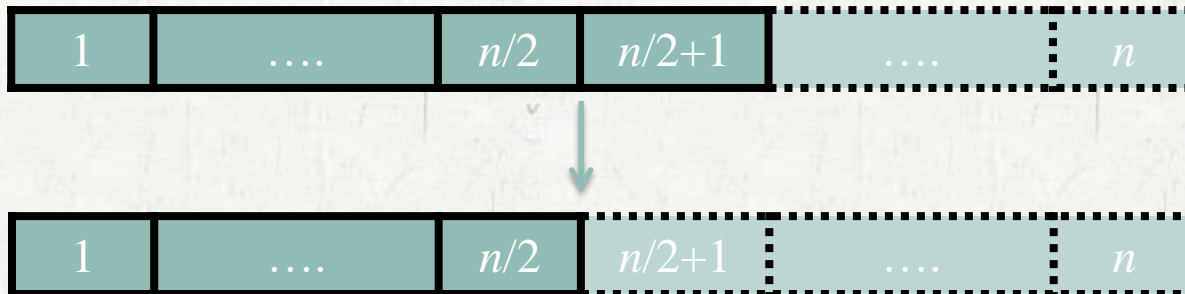


Table expansion and contraction

- Table expansion and contraction

- insert, delete, delete, insert, insert, delete, delete, insert, insert, . . .



Table expansion and contraction

- Table expansion and contraction

- insert, delete, delete, insert, insert, delete, delete, insert, insert, . . .

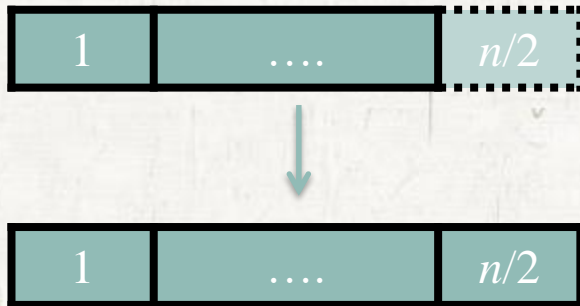


Table expansion and contraction

- Table expansion and contraction

- And **insert, delete, delete, insert, insert**, delete, delete, insert, insert, . . .

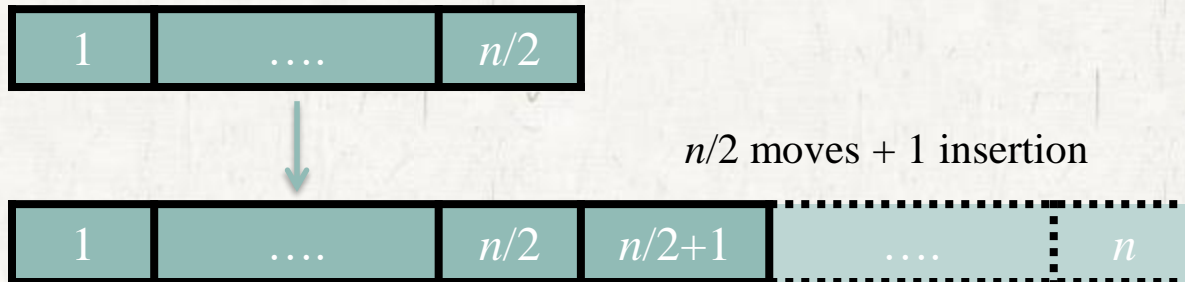


Table expansion and contraction

- Table expansion and contraction
 - And **insert, delete, delete, insert, insert, delete, delete, insert, insert, . . .**
 - about $n/2$ number of moves for $n/4$ operations
 - Thus, the total cost of the n operations is $\Theta(n^2)$.

Table expansion and contraction

- Improve upon this strategy
 - Specifically, we continue to double the table size upon inserting an item into a full table,
 - but we halve the table size when deleting an item causes the table to become less than $1/4$ full, rather than $1/2$ full as before.
 - The load factor of the table is therefore bounded below by the constant $1/4$.

Table expansion and contraction

- potential method to analyze the cost of a sequence of n TABLE-INSERT and TABLE-DELETE operations
 - Let us denote the load factor of a nonempty table T by $\alpha(T) = T.num / T.size$
 - Since for an empty table, $T.num = T.size = 0$ and $\alpha(T) = 1$
 - We shall use as our potential function

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

Table expansion and contraction

- potential method to analyze the cost of a sequence of n TABLE-INSERT and TABLE-DELETE operations
 - Let us denote the load factor of a nonempty table T by $\alpha(T) = T.num / T.size$
 - Since for an empty table, $T.num = T.size = 0$ and $\alpha(T) = 1$
 - We shall use as our potential function

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

Table expansion and contraction

● TABLE-INSERT and TABLE-DELETE

- c_i : the actual cost of the i th operation
- \hat{c}_i : its amortized cost with respect to Φ
- num_i : the number of items
stored in the table after the i th operation
- $size_i$: the total size of the table after the i th operation
- α_i : the load factor of the table after the i th operation
- Φ_i : the potential after the i th operation
- Initially, $num_0 = 0$, $size_0 = 0$, $\alpha_0 = 1$, and $\Phi_0 = 0$

Table expansion and contraction

• TABLE-INSERT

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

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

Table expansion and contraction

• TABLE-INSERT

- Case 2-1: $\alpha_{i-1} < 1/2$ and $\alpha_i < 1/2$.

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

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

Then the amortized cost of the i th operation is

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

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

Table expansion and contraction

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

● TABLE-INSERT

- Case 2-2: $\alpha_{i-1} < 1/2$ but $\alpha_i \geq 1/2$

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

$$= 1 + (2 * num_i - size_i) - (size_{i-1} / 2 - num_{i-1})$$

$$= 1 + (2 * (num_{i-1} + 1) - size_{i-1}) - (size_{i-1} / 2 - num_{i-1})$$

$$= 3 * num_{i-1} - 3size_{i-1}/2 + 3$$

$$= 3\alpha_{i-1}size_{i-1} - 3size_{i-1}/2 + 3$$

$$< 3size_{i-1}/2 - 3size_{i-1}/2 + 3$$

$$= 3$$

$$\bullet \quad num_i = num_{i-1} + 1 = \alpha_i * size_i$$

- Thus, the amortized cost of a TABLE-INSERT operation is at most 3.

Table expansion and contraction

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

● TABLE-INSERT

- Case 2-2: $\alpha_{i-1} < 1/2$ but $\alpha_i \geq 1/2$

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

$$= 1 + (2 * num_i - size_i) - (size_{i-1} / 2 - num_{i-1})$$

$$= 1 + (2 * (num_{i-1} + 1) - size_{i-1}) - (size_{i-1} / 2 - num_{i-1})$$

$$= 3 * num_{i-1} - 3size_{i-1}/2 + 3$$

$$= 3\alpha_{i-1}size_{i-1} - 3size_{i-1}/2 + 3$$

$$< 3size_{i-1}/2 - 3size_{i-1}/2 + 3$$

$$= 3$$

$$\bullet \quad num_i = num_{i-1} + 1 = \alpha_i * size_i$$

- Thus, the amortized cost of a TABLE-INSERT operation is at most 3.

Table expansion and contraction

● TABLE-DELETE

- $num_i = num_{i-1} - 1$
- If $\alpha_{i-1} < 1/2$, the i th operation causes
 - no contraction or
 - contraction

Table expansion and contraction

● TABLE-DELETE

- If $\alpha_{i-1} < 1/2$

- No contraction

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

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

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

Table expansion and contraction

● TABLE-DELETE

$$\Phi_i = \begin{cases} 2 * num_i - size_i & \text{if } \alpha(T) \geq 1/2 \\ size_i / 2 - num_i & \text{if } \alpha(T) < 1/2 \end{cases}$$

- If $\alpha_{i-1} < 1/2$

- contraction

- actual cost of the operation is $c_i = num_i + 1$
- $size_i / 2 = size_{i-1} / 4 = num_{i-1} = num_i + 1$

- the amortized cost of the operation is

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

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

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

$$= 1$$

Table expansion and contraction

• TABLE-DELETE

- If $\alpha_{i-1} \geq 1/2$
 - It's amortized cost is constant.

Table expansion and contraction

- In summary, since the amortized cost of each operation is bounded above by a constant, the actual time for any sequence of n operations on a dynamic table is $O(n)$.