
Amortized Analysis

Though the worst case performance of an algorithm may be $f(n)$, the average worst case over n runs may be asymptotically less than $n \cdot f(n)$.

We will study three methods for average worst case analysis:

1. **Aggregate Method.** Show all n runs take a total worst case time T , thus the average worst case is T/n .
2. **Accounting Method.** Performance costs assigned to some of the n runs are overestimates and are used as credit for underestimates of other runs.
3. **Potential Method.** The overestimates add to the “potential energy” of the method.

Example: Binary Counter $A[0, \dots, k-1]$

$A[0]$ is the low-order bit.

$A[k-1]$ is the high-order bit.

Increment(A)

```
i = 0
while i < length(A) and A[i] = 1
  A[i] = 0
  i = i+1
if i < length(A)
  then A[i] = 1
```

3	2	1	0
0	0	0	0
0	0	0	1
0	0	1	0
0	0	1	1
0	1	0	0
0	1	0	1
0	1	1	0
0	1	1	1
1	0	0	0

- The worst case run time is $\Theta(k)$,
- Over n calls, worst case is _____
- But actual worst-case run time for n calls is _____

Aggregate Method

Note: Not all k bits flip for each call.

Bit $A[0]$ flips $_$ times for n calls.

Bit $A[1]$ flips $\lfloor n/2 \rfloor$ times for n calls.

Bit $A[2]$ flips $\lfloor n/4 \rfloor$ times for n calls.

Bit $A[i]$ flips $\lfloor n/2^i \rfloor$ times for n calls, where $i = 0, 1, \dots, \lfloor \lg n \rfloor$.

For $i > \lfloor \lg n \rfloor$, $A[i]$ does not flip.

Aggregate Method

$$\begin{aligned} T(n) & \text{ is the worst case time for } n \text{ calls} \\ & = \sum_{i=0}^{\lfloor \lg n \rfloor} n/2^i \\ & = n \sum_{i=0}^{\lfloor \lg n \rfloor} 1/2^i < n \sum_{i=0}^{\infty} 1/2^i \\ & = n \left(\frac{1}{1-1/2} \right) \\ & = 2n \\ T(n) & = O(n) \end{aligned}$$

The amortized cost of each call is thus $O(n)/n = O(1)$.

Accounting Method

Different operations have different costs.

Cost overestimates fund cost underestimates.

Example

Operation	Cost	
bit to 1	2*	overestimate
bit to 0	0**	underestimate

Need to set some bits to 1 before set to 0.

* one for actual bit flip, one for credit

* use credit from the time when set to 1

Because resetting bits in the while loop are “paid for”, each call to Increment incurs a cost of 2 for the bit set to 1.

Each call is _____ amortized cost

n calls is _____ amortized cost

Constraints

1. The total amortized cost must be an upper bound on the actual cost.

2. The total credit in data structures must always be nonnegative.
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Potential Method

Credit adds to “potential” of whole data structure instead of to individual objects.

Definitions

- D_0 = initial data structure
- c_i = actual cost of i th operation resulting in data structure D_i after operating on D_{i-1} ($i = 1, \dots, n$).
- $\Phi(D_i)$ = real number potential associated with data structure D_i . Φ represents the potential function.
- \hat{c}_i = amortized cost of i th operation with respect to Φ
 \hat{c}_i = actual cost + potential increase
 $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$.
- The total amortized cost is

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

This is a telescoping sum $\sum_{i=1}^n (a_i - a_{i-1}) = a_n - a_0$

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

If $\Phi(D_n) \geq \Phi(D_0)$, then the amortized cost is an upper bound on the actual cost.

However, we do not know n .

If $\Phi(D_i) \geq \Phi(D_0)$ for all i , then we always pay in advance.

Let $\Phi(D_0) = 0$, thus we want $\Phi(D_i) \geq 0$.

This is similar to the accounting method since we credit potential when $\Phi(D_i) - \Phi(D_{i-1})$ is positive and debit potential when $\Phi(D_i) - \Phi(D_{i-1})$ is negative.

Example

In our example, $\Phi(D_i) = b_i$, the number of 1s in D_i (counter).

Let $t_i =$ number of bits reset to 0 on the i th call to Increment.

Thus, the actual cost c_i is at most (cost of reset) + (cost to set one) = $t_i + 1$.

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

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

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

If $\Phi(D_0) = 0$, then $\Phi(D_i) \geq 0$ for all i , and the total amortized cost is an upper bound to the actual cost.

The counter starts at zero, $\Phi(D_0) = 0$.

$$\begin{aligned}\hat{c}_i &= O(2) \\ \text{n calls is } &\underline{\hspace{2cm}}\end{aligned}$$

Dynamic Tables

TableInsert(T, x)

```
1   if size( $T$ ) = 0
2   then new table[ $T$ ] with 1 slot
3     size[ $T$ ] = 1
4   if num( $T$ ) = size( $T$ )
5   then create new table with  $2 \cdot \text{size}[T]$  slots           ;  $\alpha \geq 1/2$ 
6     copy items from table[ $T$ ] to new table
7     free table[ $T$ ]
8     table[ $T$ ] = new table
9     size[ $T$ ] =  $2 \cdot \text{size}[T]$ 
10  insert  $x$  into table[ $T$ ]
11  num[ $T$ ] = num[ $T$ ] + 1
```

Aggregate Method

Let n = number of items in table.

The worst-case running time of this algorithm is $O(n)$.

For n calls the worst-case running time is $O(n^2)$.

Double the table size when full. This expansion is performed once every power of 2 steps in $1 \dots n$.

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

$$\sum_{i=1}^n c_i \leq n + \sum_{j=0}^{\lfloor \lg n \rfloor} 2^j$$

This is a geometric series $\sum_{k=0}^n x^k = \frac{x^{n+1} - 1}{x - 1}$

$$< n + 2n$$

$$= 3n$$

The total amortized cost of a single call to TableInsert is 3.

Accounting Method

TableInsert cost should be 3. This cost pays for:

- Insert in existing table
- Copy to new table
- Copy one item already in table

If $m = \text{size}[T]$ after expanding, then $\text{num}[T] = \underline{\hspace{2cm}}$ and $\underline{\hspace{4cm}}$

Charge \$3 per Insert.

Insert costs \$1 (\$2 left).

For $m/2$ items, $m/2$ credit for new items + $m/2$ credit for existing items.

Potential Method

Define potential function Φ .

- $\Phi(T) = 2 * \text{num}[T] - \text{size}[T]$
- $\Phi(T) = 0$ immediately after expansion
- $\Phi(T)$ approaches $\text{size}[T]$ when T gets full
- $\text{num}[T] \geq \text{size}[T]/2$, so $\Phi(T) \geq 0$

Thus, the sum of the amortized costs of n TableInsert operations is an upper bound on the sum of the actual costs.

Analysis

To analyze the amortized cost of the i th TableInsert operation, let

num_i denote the number of items stored in the table after the i th operation

$size_i$ denote the total size of the table after the i th operation

Φ_i denote the potential after the i th operation

Initially, $num_0 = 0$, $size_0 = 0$, and $\Phi_0 = 0$.

Consider two cases based on whether a table expansion is done.

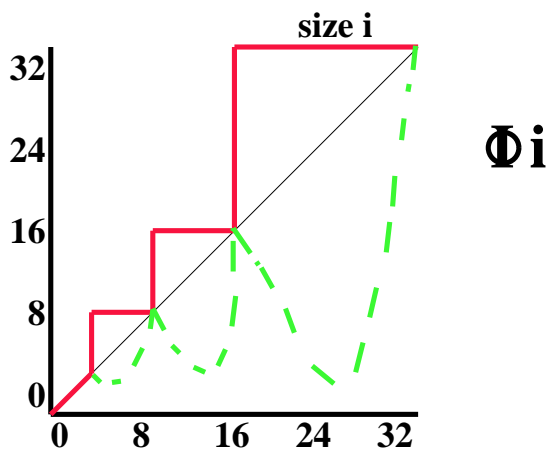
No expansion:

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

Analysis

Expansion: ($size_i/2 = size_{i-1} = num_i - 1$)

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



Note how potential builds up to number of elements just before expansion.

Applications