

Design and Analysis of Algorithms

CS218M

Amortized Complexity

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Worst Case Complexity versus Amortized Complexity

Stack

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- Worst case complexity of **Multipop(S,k)** is $O(\min(s, k))$. Hence, **Aggregate cost** of n operations is $O(n^2)$. Pessimistic.
- Before you delete k elements you must do k push, each of $O(1)$. Hence **Aggregate cost** of k push and one **Muplipop(S,k)** is $O(k)$ and **amortized cost** per operation is $O(1)$.

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- Let, $CREDIT = \sum_{i=1}^n \hat{c}_i - \sum_{i=1}^n c_i$. We require that $CREDIT \geq 0$.
- **Potential** of a datatype configuration D is denoted $\Phi(D)$. It gives CREDIT retained by reaching D . It can define Credit

- Let op_i take datatype from config D_{i-1} to D_i .
Hence $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$.
Hence, $\sum_{i=1}^n \hat{c}_i = (\sum_{i=1}^n c_i) + \Phi(D_n) - \Phi(D_0)$
- If invariantly $\Phi(D_i) - \Phi(D_{i-1}) \geq 0$ then aggregate $\sum_{i=1}^n \hat{c}_i$ is upperbound on $\sum_{i=1}^n c_i$.
- Potential function determines amortized cost \hat{c}_i of each operation.
- We typically define $\Phi(D_0) = 0$ and require that $\Phi(D_i) \geq 0$.

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- Let $Op_i = \text{Push}(S, x)$: We have $\Delta\Phi = 1$.
Hence, $\hat{c}_i = c_i + \Delta\Phi(D) = 1 + 1$.
- Let $Op_i = \text{Pop}(S)$: We have $\Delta\Phi = -1$.
Hence, $\hat{c}_i = c_i + \Delta\Phi(D) = 1 + (-1)$.
- Let $Op_i = \text{Multipop}(s, k)$: We have $\Delta\Phi = -\min(s, k)$.
Hence, $\hat{c}_i = c_i + \Delta\Phi = \min(s, k) + (-\min(s, k)) = 0$.

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Hence, $\hat{c}_i = c_i + \Delta\Phi = \min(s, k) + (-\min(s, k)) = 0$.
- Hence amortized cost of each stack operation is $O(1)$.

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

Binary Counter of k bits

represented as bit-array $A[0..k - 1]$.

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       $A[i] = 1$ 
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- Worst case complexity of *INC* is $\theta(k)$.
- Amortized worst case complexity of *INC* is ???.

Amortized Complexity: Counter

Let $\Phi(D)$ be number of trailing 1 in config D . Let $t_i = \Phi(D_{i-1})$.

- $c_i = t_i + 1$, Also $\Delta\Phi = -t_i$.
- Hence $\hat{c}_i = c_i + \Delta\Phi = 1$. Hence $O(1)$.

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- TABLE-INSERT
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 - $T.size$ the size of table.
 - $T.num$ Number of occupying elements in table.
- If $T.size = T.num$ then before insert, allocate double sized table and copy.

Dynamic Table Insert

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  $2 \cdot T.size$  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  $T.num = T.num + 1$ 
```


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- Initially, $T.num = T.size = 0$. Hence $\Phi_0 = 0$.
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- Case 1: $T.num < T.size$. Then,

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

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- Case 2: $T.num = T.size$. Then,

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

TABLE-DELETE

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