

$$1] T(n) = 64 T(n/8) - n^2 \log_2 n$$

$$2] T(n) = \sqrt{2} T(n/2) + \log_2 n$$

$$1] \text{ ~~func~~ } a=64 \quad b=8 \quad f(n) = -n^2 \log_2 n$$

★ CANNOT BE SOLVED

$$= n^2 \log_2 \frac{1}{n}$$

$$2] T(n) = \sqrt{2} T(n/2) + \log_2 n$$

$$a = \sqrt{2} \quad b = 2 \quad f(n) = \log_2 n$$

$$\therefore n^{\log_b a} = n^{\log_2 \sqrt{2}} = n^{0.5}$$

$$\therefore f(n) < n^{\log_b a}$$

$$T(n) = O(n^{\log_b a}) = O(n^{\log_b a - \epsilon}) = O(n^{0.5 - \epsilon}) = O(n^{0.5})$$

Amortized ANALYSIS  $\rightarrow$  Related Sequence

$$k=1;$$

func(int n)

{ if (k is even)

$O(n)$ ; // expensive

else

$O(1)$ ; // cheap

$k = k + 1;$

}

For m calls Normally  $O(m \times n)$

For Amortized

$$O(m/2 \times n) + O(m/2)$$

$$\Rightarrow \frac{m}{2} O(1) + \frac{m}{2} O(n) = O(1) + O(n)$$

Amortized Analysis is - Spreading out big cost over a period of time

It is applied on Data structures that support many operations. Asymptotic analysis gives worst case analysis of each operation without taking effect of one operation on another, whereas amortized analysis focusses on sequence of operations & interplay between operations & thus it yields an analysis which is precise and micro level analysis.

This analysis is applied where an occasional operation is very slow (expensive) but other operations are faster (cheap).

Amortized Analysis is an upper bound & is average performance of each operation in the worst case. This analysis is concerned with overall cost of sequence of operations doesn't say anything about cost of specific operation in that sequence. This analysis is not <sup>asymptotic</sup> average time analysis because here probability distribution of input data is not assumed.

### Aggregate Analysis.

It is simplest way of analysis. Compute the total cost for sequence of  $n$  operations i.e.  $T(n)$ .

Avg cost of Amortized cost is given as  $\frac{T(n)}{n}$

$$\hat{C} = \frac{T(n)}{n}$$

The limitation of this method is that sr //



amortized cost is assigned for different of operations, whereas other two methods give precise analysis i.e. accounting method & potential method.

~~Stack~~

Multipop(stack s, int k)

{ while (not StackEmpty(s) & k > 0)

    u = pop(s);  
    // do something with u.  
    k--;

}

For Multipop Asymptotic Analysis  $\Rightarrow T(n) = O(n)$

for sequence of n operations Complexity turns out to be  $O(n^2)$

This is not a precise value because no. of times pop is called is equal to no. of times items are pushed into stack.

Let there be n operations in sequence.

Let l be no. of multipop operations.

$\therefore (n-l)$  no. of push & pop operations

$$\text{Avg cost} = \frac{2O(n)}{n} = 2 \approx O(1)$$

For n no. of operations total cost

$$O(n) + O(n) = 2O(n)$$

$\therefore$  (At most n elements can be popped)

Time reqd to perform sequence of data structure operation is avg over all the operations performed.

Avg cost of an operation is small even though that operation might be expensive.

Ex. 2. For Aggregate Analysis

Consider a binary counter of 8 bits

Analyse Amortized cost of sequence of  $n$  operations where counter is set to 0.

increment( $a, k$ )

{

$i = 0$ ;

while ( $i < k$  &&  $a[i] \neq 0$ )

{

$a[i] = 0$ ;

}  $i++$ ;

if ( $i < k$ )

$a[i] = 1$ ;

}

$T(n) = O(k)$

0010

010011

0100 ←



## Binary Counter

Analyse Amortised cost of  $n$  inc operations.

$a[0]$  flips  $n/2^0$  times

$a[1]$  flips  $n/2^1$  times

$a[2]$  flips  $n/2^2$  times

$a[i]$  flips  $n/2^i$  times

For Asymptotic Analysis,

$$T(n) = O(n \log n)$$

No of operations      No of bits

For Amortized Analysis

$$\begin{aligned} \text{Total Cost} &= \sum_{i=0}^{\lfloor \log_2 n - 1 \rfloor} n/2^i \\ &= n \sum_{i=0}^{\log_2 n - 1} 1/2^i \\ &= 2n \\ &\approx O(n) \end{aligned}$$

$$\begin{aligned} \text{Aggregate Cost} &= \frac{\text{Total Cost}}{n} \\ &= \frac{O(n)}{n} = O(1) \end{aligned}$$

## Imp Dynamic Table

This table grows as we insert the element into the array. If there are no empty cells left at the end of table then new table of double size is created & all the data from old table is copied to new table & newly inserted element is appended at end.

```

insertTable (T, x)
{
    if (T.size == 0)
    {
        Create Table of size 1 &
        T[0] = x
        exit
    }
    else if (Load factor > 1)
    Load factor = T.num / T.size
    if (Load factor == 100%)
    {
        Allocate newTable of size 2 * T.size
        Copy all elements of Table T
        to new Table
        T = newTable
        T.size = newTable.size
        T[T.num] = x
        T.num++
    }
}
    
```

Asymptotic Time Complexity =  $O(n^2)$

	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17
Size	1	2	4	4	8	8	8	8	16	16	16	16	16	16	16	16	32
Cost	1	2	3	1	5	1	1	1	9	1	1	1	1	1	1	1	17
Insert Cost	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
Instruction	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1



$$c_i = i \quad \text{if } (i-1) \text{ is perfect power} \\ = 1 \quad \text{else}$$

$$d_i = i-1 \quad \text{if } (i-1) \text{ is perfect power} \\ = 0 \quad \text{else}$$

$$\sum_{i=1}^n c_i = \sum_{i=1}^n c_i + \sum_{i=0}^{\log_2 n - 1} 2^i$$

$$= n + 2n$$

$$= 3n$$

$$\approx O(n)$$

$$\text{Aggregate Cost} = \frac{3n}{n} = 3 \approx O(1)$$

### Accounting Method

Here we maintain an account w/ undetermined data structure. Initially it will have zero balance (zero credits). We assign different charges different operation with some operation charge more or less than their actual cost. The cost assigned to a by us is called amortized operation<sup>(cost)</sup>. When the amortized cost is greater than actual cost of operation then there will be a balance (credit). That credit we can use later to pay for operations having amortized cost lesser than actual cost.

Let  $\hat{C}_i$  be the actual time of operation

Let  $\hat{C}_i$  be amortized cost of  $i^{\text{th}}$  op

If  $\hat{C}_i > C_i$  then there will be a credit

$$\sum_{i=1}^n C_i \geq \sum_{i=1}^n C_i$$

∴ Credit is always positive

Low cost operations are charged little more than their actual cost & surplus is deposited in bank account whereas high cost operations are charged less than their true cost & the deficit is paid by savings in the bank account.

Note: The charge through each operation must be safe large enough that the balance in the bank account always remains positive but small enough so that no operation can charge significantly more than its actual cost.

- In stack push operation takes 2 units in amortized analysis so out of which 1 unit is used for actual push operation & 1 unit goes as credit & is used in pop & multipop later.

- For Binary counter we are assigning 2 unit cost to each bit flipping when it is flipping from 0 to 1.

$$C_0 \rightarrow 1 = 2 \Rightarrow 0(1)$$

$$C_1 \rightarrow 0 = 0 \Rightarrow 0(1)$$

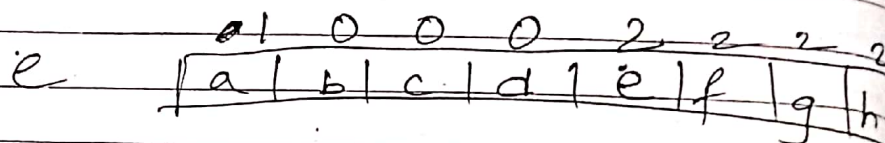
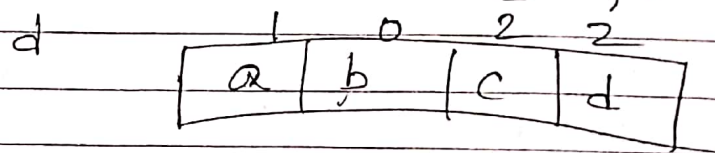
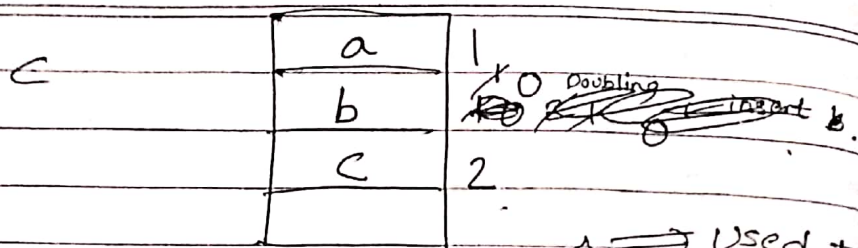
- ~~Insert~~ Dynamic Table :-

Insert a  $\boxed{a} \quad 2(3-1(\text{for insert})) = 2$

Insert b  $\begin{array}{|c|} \hline a \\ \hline b \\ \hline \end{array} \begin{array}{l} 1 \\ 2 \end{array}$







i	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
Size <sub>i</sub>	1	2	4	4	8	8	8	8	16	16	16	16	16	16	16	16
Actual Cost	1	2	3	1	5	1	1	1	9	1	1	1	1	1	1	1
cost <sub>i</sub>	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
Balance	2	3	3	5	3	5	7	9	3	5	7	9	11	13	15	17

$$\sum_{i=1}^n C_i \geq \sum_{i=1}^n \hat{C}_i$$

$\hat{C}_i > C_i$  Surplus Amt - Credit

$\hat{C}_i < C_i$  Deficit paid by Credit

### Potential Method

$\Phi(D_i) \rightarrow$  Change in structural parameter of D after  $i$ th operation

$$C_i = \underset{\substack{\uparrow \\ \text{Actual} \\ \text{cost}}}{C_i} + \underset{\substack{\uparrow \\ \text{Change in potential}}}{\Phi(D_i) - \Phi(D_{i-1})} - (1)$$

$$\sum_{i=1}^n \hat{C}_i = \sum_{i=1}^n C_i + \sum_{i=1}^n [\phi(D_i) - \phi(D_{i-1})]$$

$$\therefore \sum_{i=1}^n \hat{C}_i = \sum_{i=1}^n C_i + \phi(D_n) - \phi(D_0)$$

Stack



Structural parameter. no of elements.

$$\hat{C}_{push} = C_{push} + [\phi(D_i) - \phi(D_{i-1})]$$

Assume before  $i^{th}$  push in  $x$

$$\therefore \hat{C}_{push} = 1 + [(x+1) - x]$$

$$= 1 + x + 1 - x = 2 = O(1)$$

$$\hat{C}_{push} = C_{pop} + [\phi(D_i) - \phi(D_{i-1})]$$

$$= C_{pop} + ((x-1) - x)$$

$$= 1 + x - 1 - x = 0 = O(1)$$

$$\hat{C}_{multipop} = C_{multipop} + [\phi(D_i) - \phi(D_{i-1})]$$

$$= k + [(n-k) - n]$$

$$= 0 = O(1)$$

Counter :-

$$\text{No. of bit changes} = (n - t + 1) \quad \begin{array}{l} \nearrow \text{To replace '0'} \\ \downarrow \text{Total no of 1's} \quad \downarrow \text{Total 1's before '0'} \end{array}$$

Structural parameter is no of 1's in pa



Let  $n$  be no of 1's before  $i^{th}$  op  
 let  $t$  be no of 1's before last  
 & after  $i^{th}$  operation,

$(n - t + 1)$  no of 1's in count

$$\begin{aligned} C_{increment} &= C_{increment} + [\phi(D_i) - \phi(D_{i-1})] \\ &= (t+1) + [(n-t+1) - n] \\ &= 2 = O(1) \end{aligned}$$

Dynamic Table

Amortized cost = Actual cost + Change in potential

$$\phi(T) = 2 * Num(T) - Size(T)$$

Initially No of Element = 0

Size of Table = 0

Therefore  $\phi(T) = 0$

There are 2 possible cases :-  
Case 1 :-  $i^{th}$  insertion does not trigger an expansion

$$C_{insertion} = C_{insertion} + (\phi_i - \phi_{i-1})$$

$$\phi_i = 2N_i - S_i$$

$$\phi_{i-1} = 2(N_{i-1}) - (S_{i-1})$$

$$= 2(N_i - 1) - (S_{i-1}) \rightarrow S_i$$

$$= 2(N_i - 1) - (S_i)$$

Case 2:-  $i^{\text{th}}$  instruction triggers an expansion  
 $C_{\text{insertion}} = C_{\text{insertion}} + (\phi_i - \phi_{i-1})$

$$\phi_i = 2N_i - S_i$$

$$\phi_{i-1} = 2(N_{i-1}) - S_{i-1}$$

$$= 2(N_i - 1) - \frac{S_i}{2}$$

~~No of elements in Table~~

$$C_{\text{insertion}} = N_i + (2N_i - S_i - 2N_{i-1} + 2 + \frac{S_i}{2})$$

$$= \frac{N_i}{2} - \frac{S_i}{2} + 2$$

But  $N_i = \frac{S_i}{2} + 1 \Rightarrow$  No of elements to be copied suppose on doubling size = 16

$$C_{\text{insertion}} = 3 = O(1) \quad \text{for } 9^{\text{th}} \text{ element insertion}$$

No of elements to insert =  $\frac{16}{2} + 1 = 8 + 1 = 9$  elements