

Unit - 4 - Amortized Analysis



→ If there are ~~some~~ ^{Series} of operations where the cost of a single operation is very large and rest of operation cost less, ~~then~~ worst case running asymptotic time complexity may not give us a tighter bound.

→ It is used for algorithm have a operation is very slow and but most of operation are faster. And we analysis algorithm and guarantee of worst case average time is lower than the ~~worst~~ ^{average} case time complexity.

eg Hash Table, Disjoint Set.

→ The amortized analysis does not say anything about the cost of individual operation. rather it average the cost of all the operation in the worst case.

1) Aggregate Method

→ Simple Method.

→ Amortized cost per operation = $\frac{\text{Total cost of all operations}}{\text{No. of operations}}$

$$\frac{T(n)}{n}$$

→ count the complexity of each operation.

Stack operation (Augmented stack)

Asymptotic Analysis

$$\text{pop}() = O(1)$$

$$\text{push}(x) = O(1)$$

$$\text{MultiPop}(k) = O(n) \quad \min(n, k)$$

for n operation.

$$\text{complexity will be } O(n) \times n = O(n^2)$$

Aggregate method

Total cost of pop, or multiPop \leq no. of pushes which is at most n .

So, let have n operation of push and pop

$$\text{Amortized Analysis} = O(n)$$

$$\text{for single operation} \quad \frac{O(n)}{n} = O(1)$$

Binary Counter

let we have array $B[K]$ - K bit

$$\text{MSB} = B[K-1]$$

$$\text{LSB} = B[0]$$

$$\text{It stores: no.} = \sum_{i=0}^{K-1} B[i] \cdot 2^i \quad \left[\text{eg, } B[0] \times 2^0 + B[1] \times 2^1 + B[2] \times 2^2 \right]$$

cost = no. of flip

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Counter value		cost	cumulative cost
0	0 0 0 0	0	0
1	0 0 0 1	1	1
2	0 0 1 0	2	3
3	0 0 1 1	1	4
4	0 1 0 0	3	7
5	0 1 0 1	1	8
6	0 1 1 0	2	10
7	0 1 1 1	1	11
8	1 0 0 0	4	15
9	1 0 0 1	1	16
10	1 0 1 0	2	18
11	1 0 1 1	1	19
12	1 1 0 0	3	22
13	1 1 0 1	1	23
14	1 1 1 0	2	25
15	1 1 1 1	1	26

$i=0$

while ($i < \text{length}$ and $BC[i] == 1$)

$BC[i] = 0$

$i = i + 1$

if ($i < \text{length}$)

$BC[i] = 1$

total no. of flip: $n + \frac{n}{2} + \frac{n}{4} + \frac{n}{8} \dots + \frac{n}{2^{k-1}}$
 $\leq 2n = O(n)$

$$= n \left(1 + \frac{1}{2} + \frac{1}{4} + \dots + \frac{1}{2^{k-1}} \right)$$

$$= n \left(\frac{1 - (\frac{1}{2})^k}{1 - \frac{1}{2}} \right) = 2n$$

So, for each operation $\frac{O(n)}{n} = O(1)$

Accounting method

→ Assign diff. charge to diff. operation. called amortized cost, which may less or more than its actual cost

→ If the amortized cost is more than its actual cost the diff. is called credit

$$\text{Total credit} = \sum_{i=1}^n \hat{C}_i - \sum_{i=1}^n C_i$$

\hat{C} = Amortized cost

C_i = actual cost

always non negative

Stack operation

let assign amortized cost

push = 2

pop = 0

Multi pop = 0

On every push we will

get 1 as credit and

for pop and multi pop

least is paid from credit.

So, $O(n)/n = O(1)$

Binary Counter

let for setting cost $(0 \rightarrow 1) = 2$ unit
resetting cost $(1 \rightarrow 0) = 0$ unit

Here also

added to credit and paid when resetting

Potential method

Simpler accounting method

But it uses potential or energy to pay cost of op.

- At any given point, there is potential in data structure.

→ Initial potential of DS = Φ_0

→ Φ_i will get after i operation on Φ_{i-1}

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

$\Phi(D_0) = 0$ and $\Phi(D_i) \geq 0$ for all i

→ Amortized cost = Actual cost + Change in potential

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

$$= C_i + \Delta\Phi$$

if $\Delta\Phi > 0$ overcharged
 $\Delta\Phi < 0$ undercharged.

Augmented Stack

let $\Phi(D_i) =$ no. of element in stack

push

push =

Amortized cost = Actual + potential diff

$$= 1 + [(n+1) - (n)]$$

$$= 1 + 1 = 2$$

pop

$$= 1 + [(n-1) - n]$$

$$= 0$$

multi pop

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

$$= 0$$

14,

$O(n)$ for n operation & $O(1)$ for each operation.

Binary Counter

$\Phi(D_i) =$ no. of 1's in counter after i th operation
 $\Phi(D_i) \geq 0$

		Actual cost	$\Phi(D_i)$	$\Phi(D_{i-1})$	$\Delta\Phi$	Amortized
0	0 0 0 0	0	0	0	0	0
1	0 0 0 1	1	1	0	1	2
2	0 0 1 0	2	1	1	0	2
3	0 0 1 1	1	2	1	1	2
4	0 1 0 0	3	1	2	-1	2
5	0 1 0 1	1	2	1	1	2

We are getting 2 for every operation. So,
 $2+2+2+\dots+2 = O(2n) = O(n)$ total. each $O(1)$

* Tractable problem

→ Can be solved in polynomial time

→ Can be verified in polynomial time

→ easy to solve

→ Bubble sort $O(n^2)$

Class - P

Non tractable problem

→ Can't solve in polynomial time

→ Can verify in polynomial time

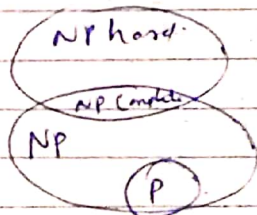
→ not easy to solve

→ TSP - Traveling Salesman Problem

to $O(2^n)$

Class NP

→ Solved by guessing and verification



Randomized Algorithm

→ An algorithm that uses a random number at least once during the computation to make a decision

→ follow probabilistic Analysis

→ Uses Random no. generator

eg

① Choosing pivot in Quicksort

② Pick edge from graph randomly

Deterministic Algo

I/P → [Algo] → O/P

O/P and time is same for same I/P

Types

(a) - Las Vegas.

→ Run within a specified amount of time

→ If solution is found within time, the solution will be exactly correct

→ If algorithm run out of time it does not find any solution.

(b) Monte Carlo (probabilistic)

→ Depending on the I/P, there is slight probability

of

(a) producing an incorrect result or

(b) failing to produce a result altogether

Adv. → Simple, fast, require less Resource

disadv. → not reach global optimum solution

→ Some time. Suboptimal is better than optimal solution

Random algo.

I/P → [Algo] → O/P

Random no.

O/P and exe. time varies for same input.

Approximation Algorithm.

- Heuristic Algorithm
- finding an optimal solution for NP-complete.
- Goal → To design an algorithm and come as close as possible to the optimum value in a reasonable amount of time
- It find feasible solution.

eg. TSP, optimization problem is find shortest cycle and approximation problem find short cycle.

Characteristic

- (1) It guarantee to run in polynomial time. But don't guarantee for effective solution
- (2) It used to get the answer near to optimal.
- (3) High accuracy and top quality.

Performance Ratio

Let C be the cost of solution by approximation and C^* is cost of ~~sub~~ optimal solution

$$\max\left(\frac{C}{C^*}, \frac{C^*}{C}\right) \leq P(n) \quad | \quad P(n) \geq 1$$

High value of $P(n)$ indicate bad solution.

Vertex cover problem

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Set  $C \leftarrow \{ \}$ 
 $E' \leftarrow E$ 
while  $|E'| \neq 0$  do
    Select an edge  $(u, v) \in E'$ 
    Add  $u$  and  $v$  to  $C$ 
     $E' \leftarrow E' - (\text{Adjacent to } u \text{ or } v)$ 
end
return  $C$ 

```

Embedded system.

- Capable of doing computation and handling inputs.
- But made for specific functionality
- eg washing machine, AC, Bluetooth, etc.

Characteristic

- ① Single function
- ② Tightly constrained
- ③ time bounded.
- ④ Microprocessor based.

* Algorithm implemented on the embedded system.
They have tight constraints.

Problem for Embedded Algorithm

- ① Limited Memory
- ② deadline based
- ③ low power consumption

Embedded System Scheduling

Scheduler - The software that decide which operation should be executed next.

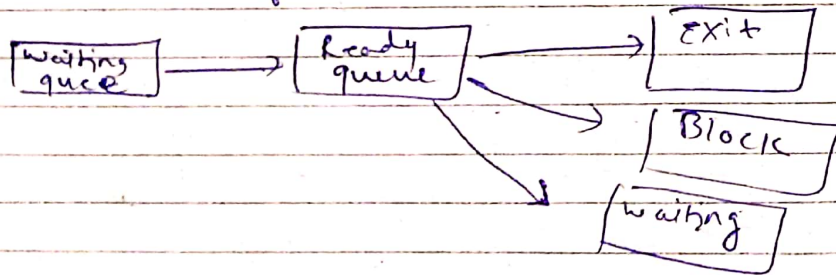
The algorithm describe the logic and the mechanism of the scheduler is called the "Scheduling Algorithm".

Each task has priority, based on that.

2 types

(1) Fixed Priority Algorithm.

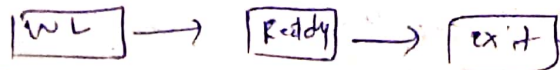
priority assign at design time and not change during lifetime.



(2) Dynamic priority algorithm.

→ priority change dynamically

→ check deadline of task in the highest priority job and schedule with the nearest deadline



Sorting Algorithm for Embedded System

It must have. (1) Sort in place - No dynamic Allocation

(2) Iterative

(3) Invariable running time

(4) I/P size dependence

(5) Reasonable code size

(6) Clean implementation.

Insertion sort is best for this

→ Stable

→ In place

Algorithm

for $j=2$ to $n-1$

key = $A[j]$

$i = j-1$

while $i > 0$ and $A[i] > \text{key}$

$A[i+1] = A[i]$

$i = i-1$

~~$A[i+1] = \text{key}$~~

$A[i+1] = \text{key}$