

# U-5 Amortized Analysis.

- measure performance of algo in avg. case.
- measures time/resource complexity in context of computer program.
- some method/resource  $\rightarrow$  cheaper
- says about avg cost of all operations in worst case.

## Methods of amortized analysis.

### 1) Aggregate Method

- assign  $n \rightarrow$  same cost all oper<sup>n</sup>
- find upper bound cost  $T(n)$  of oper<sup>n</sup> & calculate amortized cost as avg of that  $T(n)/n$ .

- ↓ Stack total
- Total no. of pops/multiple pops  $\leq$  no. of push
  - max no. of push  $\rightarrow n$
  - time for entire sequence  $\rightarrow O(n)$
  - amortized cost  $= \frac{O(n)}{n} = O(1)$

Binary ↓

- if flips  $\rightarrow$
- $n + n/2 + n/4 + \dots + n/2^k + r = 2n$
  - total cost of sequence  $= O(n)$
  - amortized cost  $= \frac{O(n)}{n} = O(1)$

### 2) Accounting Method

- diff cost for diff operation.
- less or greater actual cost
- amortized cost  $>$  actual cost  $\rightarrow$  credit
- diff of two  $\rightarrow$  credit
- credit  $\rightarrow$  pay debt for oper<sup>n</sup>

- ↓ Stack
- Push - 2
  - Pop - 0
  - MultiPop - 0
  - push  $\rightarrow$  cost 1
  - 1 as credit.

- enough credit  $\rightarrow$  each oper<sup>n</sup>
- amortized cost  $\rightarrow O(1)$
- cost of entire sequence  $\rightarrow O(n)$

Binary ↓

- all changes from 1 to 0 are paid with already stored credit.
- amortized.
- cost per op  $= O(1)$

### 3) Potential method.

- similar to accounting.
- not credit  $\rightarrow$  potential energy
- special ds  $\rightarrow$  store operation of object.
- potential diff  $> 0 \rightarrow$  overcharge
- PD  $< 0 \rightarrow$  undercharge.

- ↓ Stack
- push = 2
  - pop = 0.
  - multiPop = 0.

Binary ↓

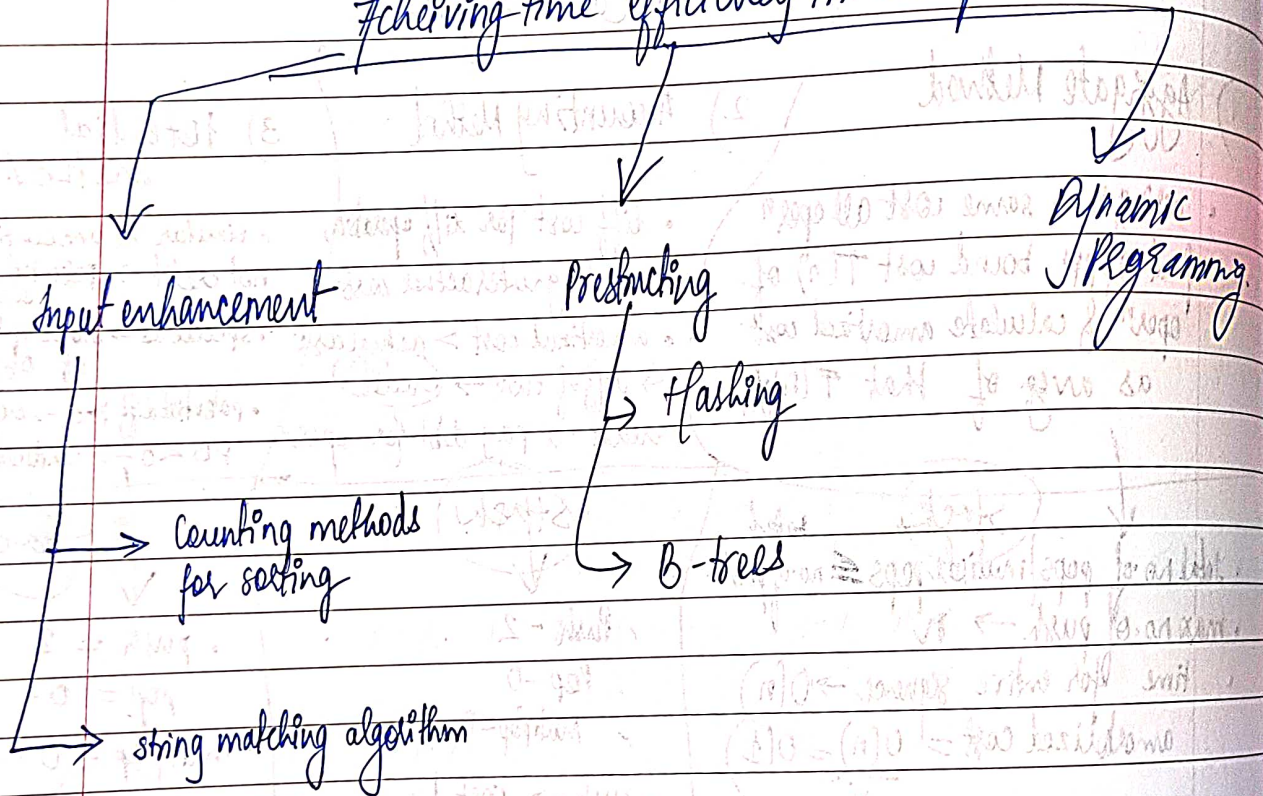
- Bit no. of 1 after its increment.
- $B_i \leq B_{i-1} - t_i + 1$ .
- diff in potential  $= B_i - B_{i-1} \leq B_{i-1} + t_i + 1 - B_{i-1} = t_i + 1$
- amortized cost  $C_i + B_i - B_{i-1} < t_i + 1 + 1 = 2$



## \* Time Space Trade Off -

- situation where either a space efficiency (memory utilization) can be achieved at cost of time or time efficiency / performance efficiency can be achieved at cost of memory.

Achieving time efficiency in computation -



Issues → 1) time & space compete → efficient algo → space efficient  
 2) Data compression techniques -

#

## Tractable

- solved in polynomial time.
- verified in " " " "
- easy to solve
- principle of func<sup>n</sup> known
- Independence to other systems
- eg → sorting list  
multiplication of integers  
search unordered list.  
search ordered list

## Non-Tractable

- solved in exponential time.
- verified in polynomial time
- not easy to solve.
- principles of func<sup>n</sup> are partly known.
- interdependence.
- eg → traveling Salesman Problem  
→ Knapsack problem

#

## Randomized Algos-

- NP may be → randomized or approximation algo in polynomial time.
- random number generator or randomizers
- decision → based current output.
- goal → quickly give a sol<sup>n</sup> nearest to optimal sol<sup>n</sup>

Categories → Las Vegas Algo • (produce same outcome on each execution)  
(Always right answers) for same input

↓ Monte Carlo Algo • (produce diff outcomes for same input on each execution)-  
• (may produce wrong answers)

Adv → simple to implement  
→ many times efficient than trad<sup>n</sup> algo

Disadv → small degree of error (dangerous)  
→ not possible to obtain better results.



## \* Approximate Algo -

- Quick sol<sup>n</sup>  $\rightarrow$  no guarantee of optimal sol<sup>n</sup>
- apply heuristics to cut computation
- feasible sol<sup>n</sup>  $\rightarrow$  near to optimal sol<sup>n</sup>  $\rightarrow$  called approximate sol<sup>n</sup>

characteristics

$I \rightarrow$  Instance

$P \rightarrow$  problem.

$FO(I)$  &  $FA(I) \rightarrow$  value of optimal & feasible sol<sup>n</sup>

based on their diff  $\rightarrow$  categorize approximation.

- 1)  $|FO(I) - FA(I)| \leq K$  [absolute Approx.]
- 2)  $|FO(I) - FA(I)| \leq f(n)$  [ $f(n)$ -Approx]
- 3)  $|FO(I) - FA(I)| / FO(I) \leq \epsilon$  [ $\epsilon$  approx]

Performance. sation

$C^* \rightarrow$  cost of optimal sol<sup>n</sup>

$C \rightarrow$  cost of sol<sup>n</sup> returned.

$\epsilon \rightarrow$  approx algo

Ratio  $P(n)$  for input size  $n \rightarrow$

$$\max \left[ \frac{C}{C^*}, \frac{C^*}{C} \right] \leq P(n)$$



## \* Embedded Algorithms.

↳ computer system with dedicated func<sup>n</sup> with target mechanical or electrical system.

characteristics → single functions.

- tightly ~~performance~~ constrained
- part of large systems
- reactive & real time
- programmable.

- embedded system scheduling → power optimized scheduling algo.  
→ fixed priority algo.

algo → 1) Fixed priority algo

- priority task → assigned → design time
- Task → exit → blocked or waiting state.
- simple lifecycle is followed here.

2) Dynamic priority algo →

- priority of task → changes dynamically
- rest is same as fixed priority.

• sorting algo for embedded system

- characteristics →
  - ↳ should be in place
  - ↳ not be recursive
  - ↳ code size → as per need of problem.
  - ↳ running time → increase linearly or logarithmically with no. of elements sorted.

Insertion sort used here →

Best time complex →  $O(n)$   
 Avg →  $O(n^2)$   
 worst →  $O(n^2)$