

# CMPT307: Amortized Analysis

Week 9-3

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# Amortized Analysis

consider a sequence of  $n$  PUSH, POP, MULTIPOP operations on an initial empty stack

- ▷ PUSH( $S, x$ ) pushes  $x$  onto stack  $S$   $O(1)$
- ▷ POP( $S$ ) pops the top of  $S$  and returns the popped object  $O(1)$
- ▷ MULTIPOP( $S, k$ ) removes  $k$  top objects from  $S$   $O(n)$

what is the total running time for the  $n$  operations?

- ▷ POP, MULTIPOP can pop at most the number of PUSH operations, thus  $O(n)$
- ▷ **amortized cost** of each operation is  $\frac{O(n)}{n} = O(1)$

# Incrementing Binary Counter

binary counter  $A[0..k-1]$

range:  $\{0, \dots, 2^k - 1\}$



INCREMENT( $A$ )

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```
1  $i = 0$ ;  
2 while  $i < k$  and  $A[i] == 1$  do  
3    $A[i] = 0$ ;  
4    $i = i + 1$ ;  
5 if  $i < k$  then  
6    $A[i] = 1$ ;
```

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each INCREMENT takes  $O(k)$  in worst-case

# Example

in initial  $A[i] = 0$  and consider  $n$  INCREMENT operations

A[3]	A[2]	A[1]	A[0]	value
0	0	0	0	0
0	0	0	1	1
0	0	1	0	2
0	0	1	1	3
0	1	0	0	4
0	1	0	1	5
0	1	1	0	6
0	1	1	1	7
1	0	0	0	8
1	0	0	1	9
1	0	1	0	10
1	0	1	1	11
1	1	0	0	12
1	1	0	1	13
1	1	1	0	14
1	1	1	1	15

# Aggregate Analysis

$A[0]$  flips  $n$  times

$A[1]$  flips  $\lfloor \frac{n}{2} \rfloor$  times

$A[2]$  flips  $\lfloor \frac{n}{2^2} \rfloor$  times

$\vdots$

$A[k-1]$  flips  $\lfloor \frac{n}{2^{k-1}} \rfloor$  times

$$T(n) = \sum_{i=0}^{k-1} \lfloor \frac{n}{2^i} \rfloor < n \sum_{i=0}^{\infty} \frac{1}{2^i} = 2n \Rightarrow \frac{T(n)}{n} = O(1)$$

- ▷ amortized analysis present a worst-case bound for  $n$  operations
- ▷ average-case analysis makes assumptions on input, and get “expected” costs
- ▷ amortized cost is not as exact as average-case cost  
amortized analysis makes no assumption about distribution
- ▷ when considering different operations, it is often easier to use amortized cost

## approaches for amortized analysis

- ▷ aggregate analysis (already seen)
- ▷ accounting method
- ▷ potential method

- ▷ assign different **charges**  $\hat{c}_i$  to different operations
- ▷ **amortized cost** of operation  $i$  is  $\hat{c}_i$
- ▷ let  $c_i$  be the real cost and require that

$$\sum_{i=1}^n \hat{c}_i \geq \sum_{i=1}^n c_i = T(n)$$

asymptotic

we may ignore constant factors

# Incrementing Binary Counter

$T(n)$  is proportional to the total number of flips

- ▷ setting to one:  $A[i] = 1$ , charge 2 dollars
  - \* one dollar for cost and one dollar for credit
- ▷ recover:  $A[i] = 0$ , charge 0 dollars
  - \* pay cost using credit

**feasibility:** total charge is (asymptotically) no less than  $T(n)$

**amortized cost:** INCREMENT runs “setting to one” at most once

# Potential Method

- ▷ perform  $n$  operations, starting with initial data structure  $D_0$
- ▷ apply the  $i$ th operation to  $D_{i-1}$  yields  $D_i$
- ▷  $c_i$  = actual cost of the  $i$ th operation
- ▷ define **potential function**  $\Phi(D_i) \in \mathbb{R}$  s.t.

$$\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1}) \quad \text{amortized cost}$$

$$\begin{aligned} \sum_{i=1}^n \hat{c}_i &= \sum_{i=1}^n [c_i + \Phi(D_i) - \Phi(D_{i-1})] \\ &= \sum_{i=1}^n c_i + \Phi(D_n) - \Phi(D_0) \geq \sum_{i=1}^n c_i \end{aligned}$$

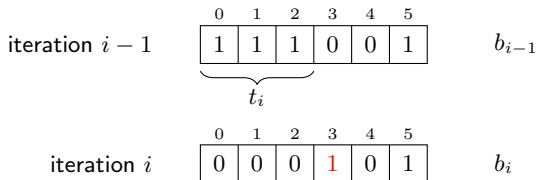
- ▷ may set  $\Phi(D_0) = 0$ , hence require that  $\Phi(D_n) \geq 0$   
often show  $\Phi(D_i) \geq 0$  for all  $i$ , as  $n$  is not known
- ▷  $\Phi(D_i) - \Phi(D_{i-1})$  indicates **overcharge** or **undercharge** of  $c_i$

# Incrementing Binary Counter

$b_i = \#1\text{s in the counter after the } i\text{th INCREMENT} = \Phi(D_i)$

$t_i = \# \text{ bits to be reset to zero at the } i\text{th INCREMENT}$

▷ if  $b_i > 0$ , then  $b_i = b_{i-1} - t_i + 1$



# Incrementing Binary Counter

▷ if  $b_i = 0$ , reset  $k$  bits, hence  $b_{i-1} = t_i = k$

	0	1	2	3	4	5
iteration $i - 1$	1	1	1	1	1	1
iteration $i$	0	0	0	0	0	0

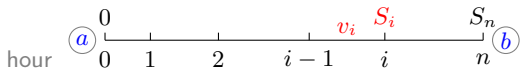
in any case:  $\Phi(D_i) - \Phi(D_{i-1}) \leq b_{i-1} - t_i + 1 - b_{i-1} = 1 - t_i$

amortized cost  $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1}) \leq t_i + 1 + (1 - t_i) = 2$

since  $\Phi(D_i) \geq 0$ , it holds that  $\sum_{i=1}^n \hat{c}_i \geq \sum_{i=1}^n c_i$

# Intuition

a vehicle takes  $n$  hours from  $a$  to  $b$



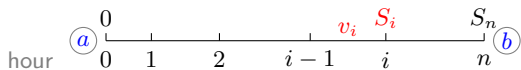
- ▷  $v_i$  = average speed during  $i - 1$  and  $i$
- ▷  $S_i$  = distance from  $a$  to the position at hour  $i$
- ▷ what is the average speed  $\bar{v}$ ?

aggregate analysis:  $\bar{v} = \frac{\text{dist}(a,b)}{n} = \frac{S_n}{n}$

accounting method

- ▷ if  $v_i > \bar{v}$ , assign speed  $v_i - \bar{v}$  to “slow hours”
- ▷ if  $v_i < \bar{v}$ , get speed  $\bar{v} - v_i$  from “fast hours”

# Intuition



potential method:  $\Phi_i - \Phi_{i-1} = \text{overcharge or undercharge}$

$$\begin{aligned}\Phi_i &= \text{expected distance} - \text{actual distance at hour } i \\ &= \bar{v} \cdot i - S_i\end{aligned}$$

$$\Phi_i - \Phi_{i-1} = \bar{v} - (S_i - S_{i-1}) = \bar{v} - v_i$$

$$\hat{v} = v_i + \Phi_i - \Phi_{i-1} = \bar{v}$$

but we do not know  $\bar{v}$  and  $S_i$

- ▷ guess  $\bar{v}$  and estimate  $S_i$ , then define potential function
- ▷ show that  $\Phi_i$  works well