

# **Amortized Analysis**

Data Structures and Algorithms

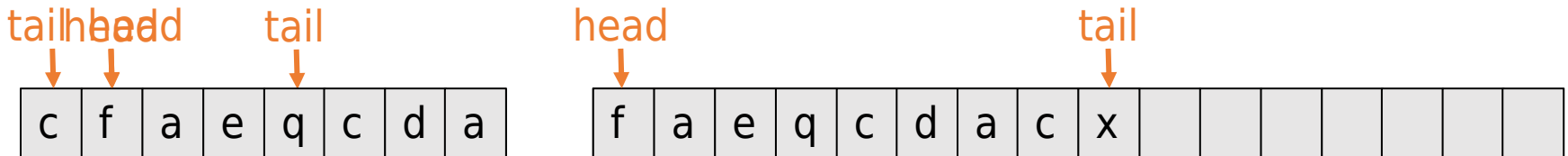
Nanjing University, Fall 2021

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# Implement **Queue** with **CircularArray**

- **CircularArray** supports **Queue** operations in  $O(1)$  time.
- But what to do when the array is full?!
- Allocate a new array of double size.  $\Theta(n)$
- Copy existing items  $\Theta(1)$  to the new array, and insert new element.
- Delete old array.
- But now the **Insert** operation may take  $\Theta(n)$  time!
- So a sequence of  $n$  operations can take  $O(n^2)$  time?!

**Correct but not tight!!**



# Amortized Analysis

- Technique for analyzing “average cost”:
  - Often used in data structure analysis
  - (Expensive Op. and Cheap Op.) + (Expensive Op. *can't be frequent*)  
=> Average cost of Op. for *any* sequence of Op. must be low.
- In some sense, like “pay in installments”.
  - Is using iPhone expensive?
  - Sure, average monthly salary in Jiangsu  $\approx 8635 / 53$
  - But you don't buy a new iPhone everyday!  
Pay < 550 per month if new iPhone every other year



RMB 12999  
1TB

# Amortized Analysis

- Technique for analyzing “average cost”:
  - Often used in data structure analysis
  - (Expensive Op. and Cheap Op.) + (Expensive Op. *not frequent*)  
=> Average cost of Op. for *any* sequence of Op. must be low.
- **Definition:** Operation has amortized cost  $\hat{c}(n)$ , if for *every*  $k \in \mathbb{N}^+$ , the total cost of *any*  $k$  operations is  $\leq \sum_{i=1}^k \hat{c}(n_i)$ .  
( $n_i$  is the size of the data structure when applying the  $i^{\text{th}}$  op.)
- Different operations may have different amortized costs.

# Amortized Analysis

- Consider a sequence operations:  
 $c_i$  = actual cost of the  $i^{\text{th}}$  op.;  $\widehat{c}_i$  = amortized cost of the  $i^{\text{th}}$  op.
- For the amortized cost to be valid:  
 $\sum_{i=1}^k c_i \leq \sum_{i=1}^k \widehat{c}_i$  for any  $k \in \mathbb{N}^+$
- Total cost of  $k$  operations is  $\leq \sum_{i=1}^k \widehat{c}_i$ , not  $\leq k \cdot \max \{c_i\}$ .
- Average cost of  $k$  operations is  $\leq \frac{\sum_{i=1}^k \widehat{c}_i}{k}$ , not  $\leq \max \{c_i\}$ .

# Amortized Analysis

- **Definition:** Operation has amortized cost  $\hat{c}(n)$ , if for *every*  $k \in \mathbb{N}^+$ , the total cost of *any*  $k$  operations is  $\leq \sum_{i=1}^k \hat{c}(n_i)$ .

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- Different operations may have different amortized costs.

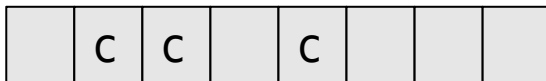
- Consider **CircularArray** implementation of **Queue**.

Ignore cost of array alloc and free for now.

- **Insert** have amortized cost 2? ( $\hat{c}(n) = 2$  if op. is **Insert**.)

- **Remove** has amortized cost 1? ( $\hat{c}(n) = 1$  if op. is **Remove**.)

<b>Insert</b> (c)	1+(1+1)=3	2+2=4
<b>Insert</b> (c)	3+(2+1)=6	4+2=6
<b>Insert</b> (c)	6+1=7	6+2=8
<b>Insert</b> (c)	7+(4+1)=12	> 8+2=10



# Amortized Analysis

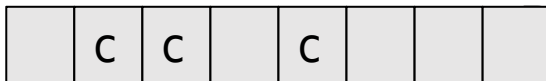
- **Definition:** Operation has amortized cost  $\hat{c}(n)$ , if for *every*  $k \in \mathbb{N}^+$ , the total cost of *any*  $k$  operations is  $\leq \sum_{i=1}^k \hat{c}(n_i)$ .  
( $n_i$  is the size of the data structure after the  $i$ -th operation.)

So **CircularArray** operations has  $O(1)$  amortized cost?  
(Even though some op. can cost  $\Theta(n)$ .)

- Different **CircularArray** implementation of **Queue**.
- **insert** have amortized cost 1? ( $\hat{c}(n) = 1$  if op. is **insert**.)
- **Remove** has amortized cost 1? ( $\hat{c}(n) = 1$  if op. is **Remove**.)

Ignore cost of array alloc and free for now.

	Actual Total cost	Amortized total cost
<b>Insert(c)</b>	$1 + (1 + 1) = 3$	$3 + 3 = 6$
<b>Insert(c)</b>	$3 + (2 + 1) = 6$	$6 + 3 = 9$
<b>Insert(c)</b>	$6 + 1 = 7$	$9 + 3 = 12$
<b>Insert(c)</b>	$7 + (4 + 1) = 12$	$12 + 3 = 15$
<b>Remove()</b>	$12 + 1 = 13$	$15 + 1 = 16$



## Amortized Analysis

# The Accounting Method

- Consider a sequence operations:  
 $c_i$  = actual cost of the the  $i^{\text{th}}$  op.;  $\widehat{c}_i$  = amortized cost of the  $i^{\text{th}}$  op.
- For the amortized cost to be valid:  
 $\sum_{i=1}^k c_i \leq \sum_{i=1}^k \widehat{c}_i$  for any  $k \in \mathbb{N}^+$
- Imagine you have a bank account  $B$ .
- For the  $i^{\text{th}}$  op., you spend  $\widehat{c}_i$  money:
  - Recall the actual cost of the  $i^{\text{th}}$  op. is  $c_i$ .
  - If  $\widehat{c}_i \geq c_i$ , pay  $c_i$  for the op., and deposit  $\widehat{c}_i - c_i$  into  $B$ .
  - If  $\widehat{c}_i < c_i$ , pay  $c_i$  for the op., and withdraw  $c_i - \widehat{c}_i$  from  $B$ .
- Amortized analysis valid if  $B = \sum_{i=1}^k (\widehat{c}_i - c_i)$  always  $\geq 0$ .



## The Accounting Method

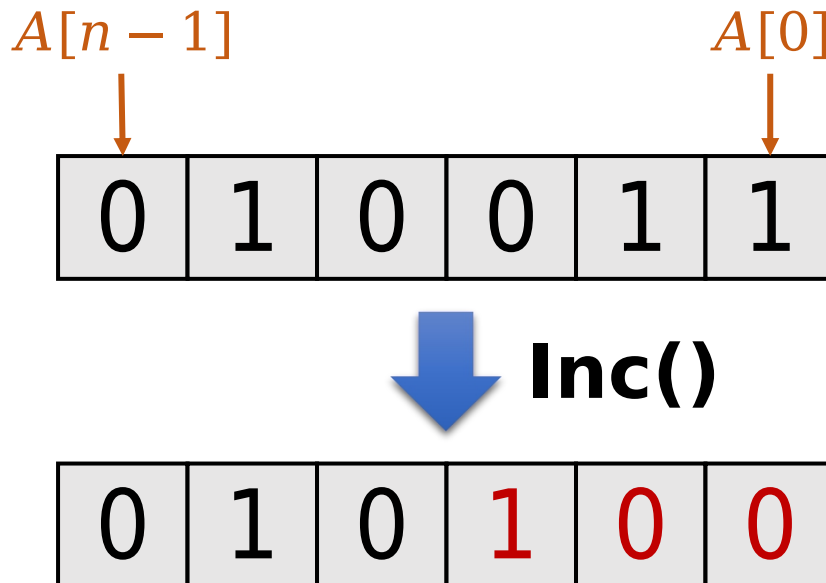
# Example: **CircularArray** based Queue

- $\widehat{c}_i = 3$  if the  $i^{\text{th}}$  op is **Insert**,  $\widehat{c}_i = 1$  if the  $i^{\text{th}}$  op is **Remove**.
- **Goal:** Prove  $\sum_{i=1}^k c_i \leq \sum_{i=1}^k \widehat{c}_i$  for any  $k \in \mathbb{N}^+$  operations.
- **Strategy:** account always non-negative via induction on  $k$ .
- [Basis] Prior to 1<sup>st</sup> op., account value is 0.
- [Hypothesis] Prior to  $i^{\text{th}}$  op., account value is always non-negative.
- [Inductive Step] Consider the  $i^{\text{th}}$  op.
  - If it's **Remove**, then we make no change to account value.
  - If it's **Insert** without expansion, we add 2 to account value.
  - If it's **Insert** with expansion. Assume expand from  $n$  to  $2n$ .  
Last expand must be from  $n/2$  to  $n$ .  
Since last expand, each **Insert** adds 2, each **Remove** makes no change.

## The Accounting Method

# Example: Binary Counter

- Use length  $n$  binary array  $A$  to represent a number.
- The number is 0 initially, and **Inc** op. adds 1 to this number.
- Cost of **Inc**: number of bits it flipped.
- Average cost of  $k$  **Inc** operations?
  - Easy answer:  $O(n)$
  - More careful analysis... (Amortized analysis...)



### **Inc(A):**

```
i=0
while (i<n and A[i]==1)
  A[i]=0
  i=i+1
if (i<n)
  A[i]=1
```

## The Accounting Method

# Example: Binary Counter

- The number is 0 initially, and **Inc** op. adds 1 to this number.
- Cost of **Inc**: number of bits it flipped.
- In each **Inc**:  $0 \rightarrow 1$ : at most 1 bit;  $1 \rightarrow 0$ : many bits.
- But a bit has to be set to 1 before it resets to 0!
- If we deposit 1 whenever we  $0 \rightarrow 1$ , later  $1 \rightarrow 0$  are “**free**”!

• 

0	1	0	0	1	1
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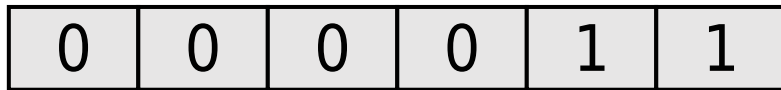
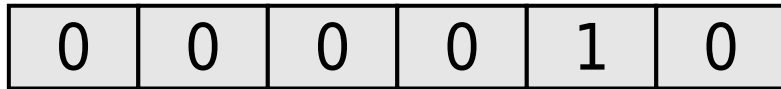
 **Inc()**

0	1	0	1	0	0
---	---	---	---	---	---

cost

**Inc(A):**

```
i=0
while (i<n and A[i]==1)
  A[i]=0
  i=i+1
if (i<n)
  A[i]=1
```



Actual  
Total Cost

1

$$1 + 2 = 3$$

$$3 + 1 = 4$$

$$4 + 3 = 7$$

Amortized  
Total Cost



× 2



× 4



× 6



× 8

## Amortized Analysis

# The Potential Method

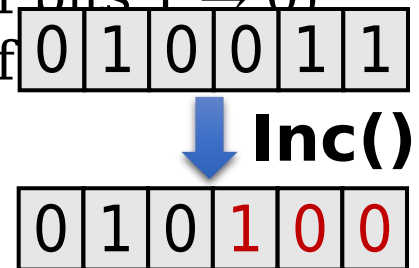
- Consider a sequence operations:  
 $c_i$  = actual cost of  $i^{\text{th}}$  op.;  $\widehat{c}_i$  = amortized cost of  $i^{\text{th}}$  op.
- For the amortized cost to be valid:  
$$\sum_{i=1}^k c_i \leq \sum_{i=1}^k \widehat{c}_i \text{ for any } k \in \mathbb{N}^+$$
- Design a **potential function**  $\Phi$  that maps D.S. status to real values.
  - $\Phi(D_0)$ : initial potential of D.S., usually set to 0.
  - $\Phi(D_i)$ : potential of D.S. after  $i^{\text{th}}$  operation.
- Define  $\widehat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$
- For amortized cost to be valid, need  $\Phi(D_k) \geq \Phi(D_0)$  for all  $k$ .
- “Potential” is like the **balance in account** in “Counting Method”.
  - Potential slowly accumulates during “cheap” operations (deposit).
  - Potential drops a lot after an “expensive” operation (withdraw).
- But the Potential Method could be more powerful in general...

## The Potential Method

# Example: Binary Counter

- Design a **potential function**  $\Phi$  that maps D.S. status to real values.
  - $\Phi(D_0)$ : initial potential of D.S., usually set to 0.
  - $\Phi(D_i)$ : potential of D.S. after  $i^{\text{th}}$  operation.
- Define  $\widehat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$ , need  $\Phi(D_k) \geq \Phi(D_0)$  for all  $k$ .
- How to define  $\Phi(D_i)$  for Binary Counter? (Recall potential is like “balance”.)
- $\Phi(D_i)$  = number of 1s in the array after the  $i^{\text{th}}$  **Inc** operation.
- Clearly “ $\Phi(D_k) \geq \Phi(D_0)$  for all  $k$ ” is satisfied, how large is  $\widehat{c}_i$ ?
- $$c_i = (\# \text{ of bits } 0 \rightarrow 1) + (\# \text{ of bits } 1 \rightarrow 0)$$

$$\Phi(D_i) - \Phi(D_{i-1}) = (\# \text{ of bits } 0 \rightarrow 1) - (\# \text{ of bits } 1 \rightarrow 0)$$
- $\widehat{c}_i = 2 \cdot (\# \text{ of bits } 0 \rightarrow 1) \leq 2$



# Back to **CircularArray** based Queue

- **Problem:** Array has limited size, what to do when it's full?
- **Solution:** Double the size when array is full and **Insert** comes. (Copy items to new array, insert new item, and delete old array.)
- **Solution is Good:** amortized cost of **Insert** and **Remove** both  $O(1)$ .
- **New Problem:** Lots of **Insert**, then lots of **Remove**. A lot of space wasted!
- **Solution:** Reduce array size to half when array only half loaded after **Remove**. (Allocate new array of half size, copy items to new array, and delete old array.)
- Does the above solution achieves  $O(1)$  amortized cost?
- **No!** Consider a full array and following ops: **Insert, Remove, Insert, Remove, ...**
- Better solutions? How to prove new solutions indeed "better"?

# Reading

- [CLRS] Ch.17 (including 17.4)