Amortized Analysis: Dynamic Arrays

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Lecture 7

Today

- Amortized Analysis
 - Aggregate Method
 - Accounting Method
- Dynamic Arrays

Reading Assignments



Chapter 17

Running time

- Best case
- Average case
- Worst case
- Expected running time
- Amortized cost

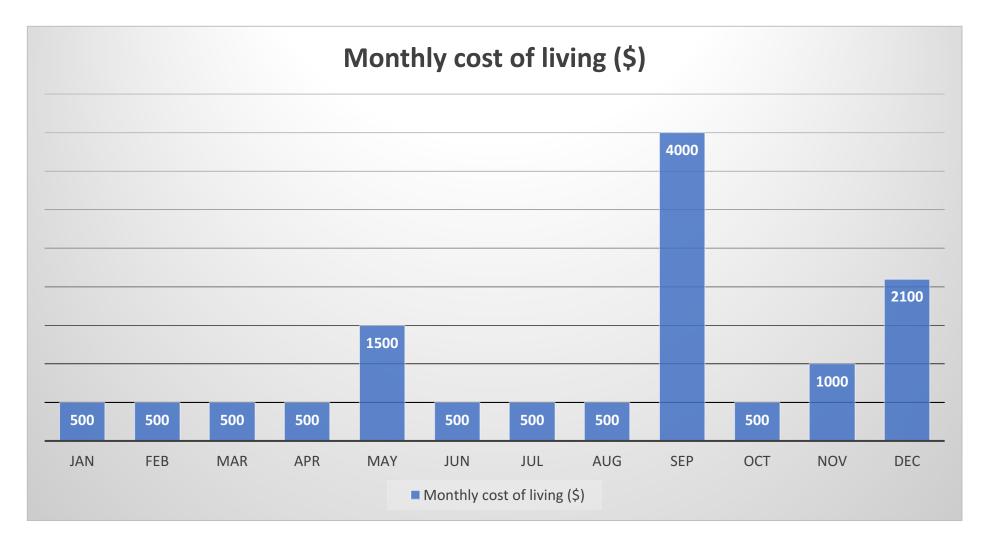


Amortized Analysis

- We do amortized analysis when we are interested in the total complexity of a sequence of operations.
- Unlike in average-case analysis where we are interested in a **single** operation.
- The *amortized sequence complexity* is the "average" cost per operation over the sequence. But unlike average-case analysis, there is NO probability or expectation involved.

Real life intuition

Monthly cost of living, a sequence of 12 operations



You are to maintain a collection of lists and support the following operations.

- insert(item, list): insert item into list (cost = 1).
- o sum(list): sum the items in list, and replace the list with a list containing one item that is the sum (cost = length of list).

Cost of an insert operation and the amortized cost of a sum operation.



Insert(4)

You are to maintain a collection of lists and support the following operations.

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Insert(7)

You are to maintain a collection of lists and support the following operations.

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Cost of an insert operation and the amortized cost of a sum operation.

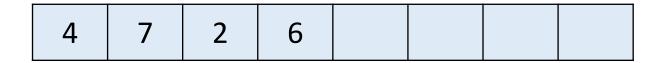


Insert(2)

You are to maintain a collection of lists and support the following operations.

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Cost of an insert operation and the amortized cost of a sum operation.



Insert(6)

Sum()

You are to maintain a collection of lists and support the following operations.

- insert(item, list): insert item into list (cost = 1).
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Cost of an insert operation and the amortized cost of a sum operation.

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Insert(6)

Sum()

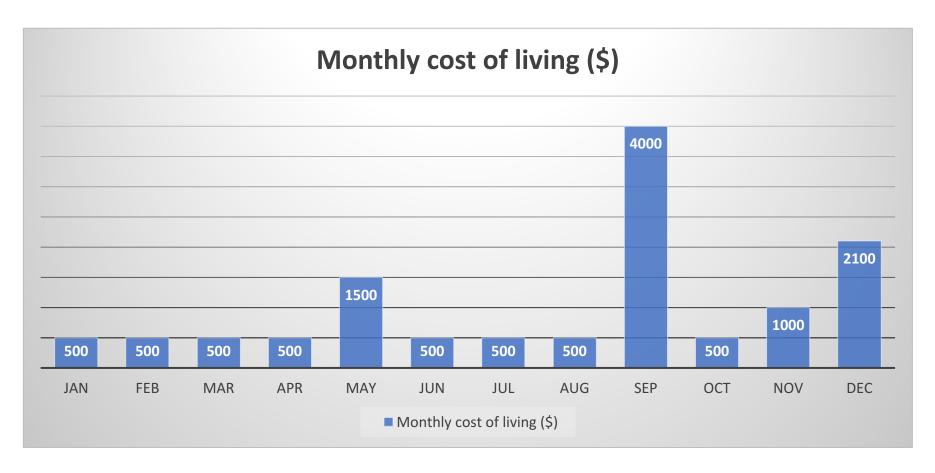
Sum()

Methods for amortized analysis

- Aggregate method
- Accounting method
- Potential method (skipped, read Chapter 17 if interested)

Aggregate method

What is the amortized cost per month (operation)? Just **sum up** the costs of all months (operations) and **divide** by the number of months (operations). Amortized cost = $\frac{\$ 12600}{12} = 1050$



Aggregate method

For a sequence of m operations, let T(m) = worst-case complexity of m operations

Amortized sequence complexity $=\frac{T(m)}{m}$

The MAXIMUM possible *total* cost of among all possible sequences of m operations

Instead of calculating the average spending, we think about the cost from a **different angle**, i.e.,

How much money do I need to **earn** each month in order to **keep living**? That is, be able to pay for the spending every month and **never become broke**.



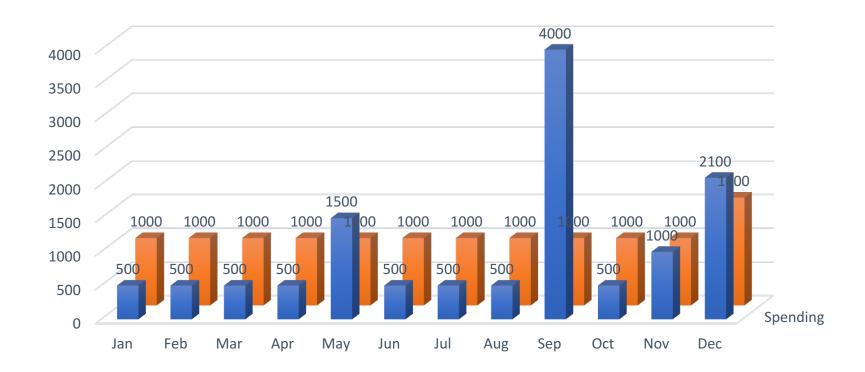
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How much money do I need to **earn** each month in order to **keep living**?

That is, be able to pay for the spending every month and **never become broke**.

Accounting method: if I **earn** \$1,000 per month from Jan to Nov and earn \$1,600 in December, I will never become broke (assuming earnings are paid at the beginning of month).

So the **amortized cost**: \$1,000 from Jan to Nov and \$1,600 in Dec.



- We assign differing charges to different operations.
- We call the amount we charge an operation its amortized cost.
- When an operation's amortized cost exceeds its actual cost, we assign the difference to specific objects in the data structure as credit.

- \circ the actual cost of the operation $i: c_i$
- o the amortized cost of the operation i: $\widehat{c_i}$.

$$\sum_{i=1}^{n} \widehat{c_i} \ge \sum_{i=1}^{n} c_i$$

Aggregate vs Accounting

- Aggregate method is easy to do when the cost of each operation in the sequence is concretely defined.
- Accounting method is more interesting since it works even when the sequence of operation is not concretely defined.
- Accounting method can obtain more refined amortized cost than aggregate method (different operations can have different amortized cost)

Example: Incrementing a binary counter

• Implementing a k-bit binary counter that counts upward from o.

```
A = [0,1,...n]

INCREMENT(A)

1   i = 0

2   while i < A. length and A[i] == 1

3   A[i] = 0

4   i = i + 1

5   if i < A. length

6   A[i] = 1
```

```
A 76543210
  0000000
  0000001
  0000010
  0000011
  00000100
5 00000101
  00000110
  00000111
8 00001000
  00001001
10 00001010
11 00001011
12 00001100
13 00001101
14 00001110
15 00001111
16 00010000
```

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  00001001
10 00001010
11 00001011
12 00001100
13 00001101
14 00001110
15 00001111
  0001000
```

```
A[0] flips n times
A[1] flips \lfloor n/2 \rfloor times
A[1] flips \lfloor n/4 \rfloor times
A[i] flips |n/2^i| times
The total number of flips
\sum_{i=0}^{k-1} |n/2^i| < n \sum_{i=0}^{\infty} |1/2^i| =
2n
Total amortized cost
       = T(n)/n = O(1)
```

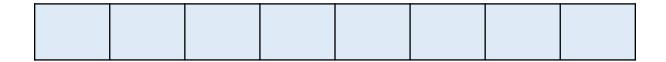
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Amortized Analysis on Dynamic Arrays

Problem description

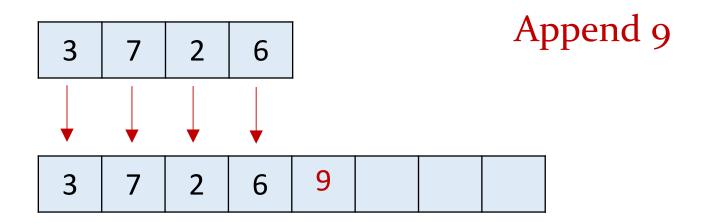
 Think of an array initialized with a fixed number of slots, and supports APPEND and DELETE operations.



- When we APPEND too many elements, the array would be **full** and we need to **expand** the array (make the size larger).
- When we DELETE too many elements, we want to **shrink** to the array (make the size smaller).
- Requirement: the array must be using one contiguous block of memory all the time.

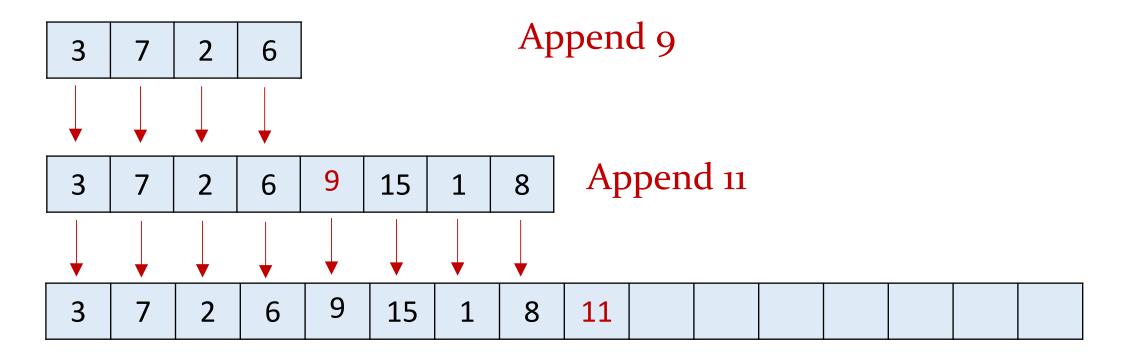
One way to expand

- If the array is full when APPEND is called
 - o Create a new array of **twice** the size
 - Copy the all the elements from old array to new array
 - Append the element



One way to expand

- If the array is full when APPEND is called
 - Create a new array of twice the size
 - Copy the all the elements from old array to new array
 - Append the element



Amortized analysis of expand

Now consider a dynamic array initialized with size 1 and a sequence of MPPEND operations on it.

Analyze the amortized cost per operation

Assumption: only count array assignments, i.e., **append** an element and **copy** an element

Use the aggregate method

What is the cost for **copying** the elements when an element is appended? What is the cost for **appending** the elements when an element is inserted?

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	•••	31	32	33
Copy	-	1	2	-	4	-	-	-	8	-	-	-	-	-	-	-	16	-	-	-	•••	-	-	32
Append	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	•••	1	1	1
total	1	2	3	1	5	1	1	1	9	1	1	1	1	1	1	1	17	1	1	1		1	1	33

$$c_i = \begin{cases} i+1 & if \ i \ is \ power \ of \ 2 \\ 1 & otherwise \end{cases}$$

Use the aggregate method

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19		31	32	33
Copy	-	1	2	-	4	-	-	-	8	-	-	-	-	-	-	-	16	-	-	-		-	-	32
Append	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	•••	1	1	1
total	1	2	3	1	5	1	1	1	9	1	1	1	1	1	1	1	17	1	1	1	•••	1	1	33

n operation of append operation.

$$Let m = \log n \rightarrow n = 2^m$$

Cost for copies:
$$1 + 2 + \dots 2^m = \sum_{i=0}^m 2^i = 2^{m+1} = 2n$$

Cost for appends:
$$\sum_{i=0}^{n} 1 = n$$

Total cost =
$$3n$$
 Amortized cost = $3n/n = 3$

Use the accounting method

Cost sequence concretely defined, sum-and-divide can be done, but we want to do something more interesting...

How much money do we need to **earn** at each operation, so that all future costs can be paid for?

How much money to earn for **each APPEND'ed element**? \$1 ?

\$2?

\$3?

\$log m ?

\$m?

Amortized cost 1\$?



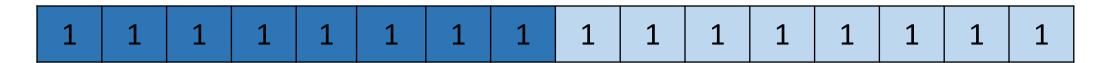


Earn \$1 for each appended element

• This \$1 is spent when appending the element.

But, when we need to copy this element to a new array (when expanding the array), we don't any money to pay for it.

Amortized cost 2\$?





Earn \$2 for each appended element

• \$1 (the "append-dollar") will be spent when appending the element.

Amortized cost 2\$?



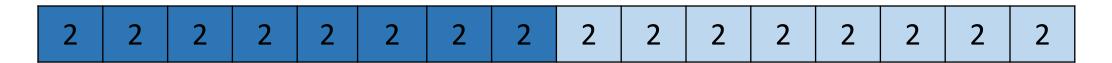




Earn \$2 for each appended element

\$1 (the "copy-dollar") will be spent when copying the element to a new array. How about the elements that have been copied more than one time?

Amortized cost 3\$?





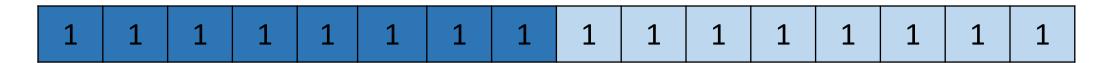




Earn \$3 for each appended element

• \$1 (the "append-dollar") will be spent when appending the element.

Amortized cost 3\$?





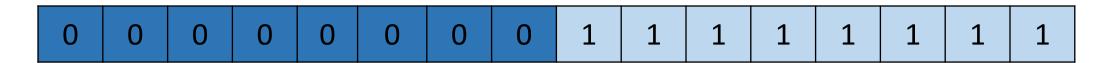




Earn \$3 for each appended element

\$1 (the "copy-dollar") will be spent when copying the element to a new array.

Use the accounting method!









Earn \$3 for each appended element

• The elements in the first half have been copied twice.

 1
 1
 1
 1
 1
 1
 0
 0
 0
 0
 0
 0
 0

• The elements in the first quarter have been copied three times.



Use the accounting method!

If we earn \$3 upon each APPEND it is enough money to pay for all costs in the sequence of APPEND operations.

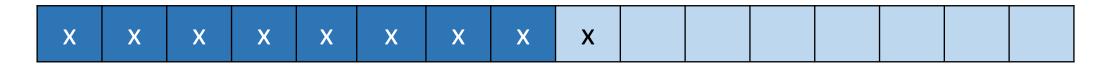
In other words, for a sequence of m APPEND operations, the amortized cost per operations is 3, which is in O(1).

In a regular worst-case analysis (non-amortized), what is the worst-case runtime of an APPEND operation on an array with m elements?

Amortized Analysis on Shrinking Dynamic Arrays

First idea

When the array is ½ full after DELETE, create a new array of half of the size, and copy all the elements.



Consider the following sequence of operations performed on a **full** array with **n**element...

• APPEND, DELETE, APPEND, DELETE, APPEND, ...

\(\theta(n)\) amortized cost per operation since every APPEND or DELETE causes allocation of new array.

NO GOOD!

Better solution

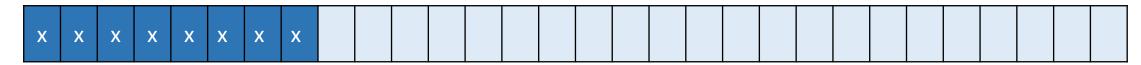
When the array is ¼ full after DELETE, create a new array of ½ of the size, and copy all the elements.



Earning \$3 per APPEND and \$3 per DELETE would be enough for paying all the cost.

- 1 append/delete-dollar
- 1 copy-dollar
- 1 recharge-dollar

Shrinking cost



The array, after shrinking...



Array is half-empty

Elements who just spent their copy-dollars

Before the **next expansion**, we need to **fill** the **empty half**, which will spare enough money for copying the green part.



Before the **next shrinking**, we need to **empty** half, which will spare enough money for copying what's left.

Summary

- In a dynamic array, if we expand and shrink the array as discussed (double on ½ full, halve on 1/4 full)...
- For any sequence of APPEND or DELETE operations, earning \$3 per operation is enough money to pay for all costs in the sequence,...
- Therefore the amortized cost per operation of any sequence is upper-bounded by 3, i.e., O(1).

Questions