



# Exercise Session 04 – Amortized Analysis

## **Data Structures and Algorithms**

*These slides are based on those of the lecture, but were adapted and extended by the teaching assistant Adel Gavranović*

# Today's Schedule

Intro

Follow-up

Feedback for **code expert**

Learning Objectives

Entry Quiz

Amortized Analysis

Code-Example: Dynamically Sized

Array

Tips for **code expert**

Old Exam Question

Outro



`n.ethz.ch/~agavranovic`

▶ Exercise Session Material

▶ Adel's Webpage

▶ Mail to Adel

# Comic of the Week

I MET A TRAVELER FROM AN ANTIQUE LAND  
WHO SAID: "I MET A TRAVELER FROM AN AN-  
TIQUE LAND, WHO SAID: "I MET A TRAVELER FROM  
AN ANTIQUE LAND, WHO SAID: "I MET...



# 1. Intro

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# Intro

- You get the XP points to unlock the bonus tasks with way fewer than all points (i.e. 1/3 usually suffices)

## 2. Follow-up

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# Follow-up from last exercise session



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- If time allows, we'll have a look at an exam question

### 3. Feedback for **code** expert

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# Task "Some Proofs"

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# Task "Some Proofs"

- use counterexamples whenever you can – they're easy to prove and even easier to correct ;)
- the majority seems to grasp the concepts well, but the "mathy proofs" are lacking – make sure to study the master solutions

Questions regarding **code expert** from your side?



## 4. Learning Objectives

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# Learning Objectives

- Understand the basics of the three *Amortized Analysis* methods
  - Aggregate Analysis
  - Account Method
  - Potential Method
- Be prepared for Double Ended Queue exercise on **code expert**

## 5. Entry Quiz

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# Quiz

Among a huge number ( $n$ ) of students present, a prize will be awarded to the student with the median Legi number. There is an argument what kind of algorithm shall be used to find this student. Mark the correct statements.

(1) In order to have a worst case runtime of  $\mathcal{O}(n \log n)$ , we use

- BubbleSort
- Selection Sort
- Mergesort
- Quicksort

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(2) We use Quickselect with random pivot choice. Then we have

- a worst case running time of  $\mathcal{O}(n \log n)$
- a worst case running time of  $\mathcal{O}(n)$
- an expected running time of  $\mathcal{O}(\log n)$
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- an expected running time of  $\mathcal{O}(n)$  😊

## 6. Amortized Analysis

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# Amortized Analysis

## **Three Methods**

# Amortized Analysis

"Amortized constant runtime"  
→  $O(1)$

## Three Methods

- Aggregate analysis
- Account Method
- Potential Method

# Example: simple multi-set

Supports operations `Insert` and `Find`.

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Idea:

- Collection of arrays  $A_i$  with Length  $2^i$

$A_0$  [ $*$ ]

$A_1$  [ $*$   $*$ ]

$A_2$  [ $*$   $*$   $*$   $*$ ]

# Example: simple multi-set

Supports operations **Insert** and **Find**.

Idea:

- Collection of arrays  $A_i$  with Length  $2^i$
- Every array is either entirely empty or entirely full and stores items in a sorted order

$A_0$  [ $*$ ]

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Supports operations **Insert** and **Find**.

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- Collection of arrays  $A_i$  with Length  $2^i$
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- Between the arrays there is no further relationship

Data  $\{1, 8, 10, 18, 20, 24, 36, 48, 50, 75, 99\}$ ,  $n = 11$

$A_0$ : [50]

$A_1$ : [8, 99]

$A_2$ :  $\emptyset$

$A_3$ : [1, 10, 18, 20, 24, 36, 48, 75]

We use 0-indexing, such that for the lengths  $|A_i| = 2^i$ .

## Example: simple multi-set

For any  $n \in \mathbb{N}$ , we can store exactly  $n$  elements in our multi set, without partially-filled arrays. Intuition: binary representation of  $n$ .

$$\begin{aligned}\text{\#elements in multi-set} &= |A_k| + |A_{k-1}| + \dots + |A_0| \\ &= b_k 2^k + b_{k-1} 2^{k-1} + \dots + b_0 2^0 \\ &= (b_k \quad b_{k-1} \quad \dots \quad b_0)_2\end{aligned}$$

Where  $b_i = 0$  if  $|A_i| = 0$ , and 1 if  $|A_i| = 2^i$ .



# Example: simple multi-set

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Algorithm Find:

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Algorithm **Find**: Perform a binary search on each array

Worst-case Runtime:

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Algorithm **Find**: Perform a binary search on each array

Worst-case Runtime:  $\Theta(\log^2 n)$ ,

$$\log 1 + \log 2 + \log 4 + \cdots + \log 2^k = \sum_{i=0}^k \log_2 2^i = \frac{k \cdot (k + 1)}{2} \in \Theta(\log^2 n).$$

$(k = \lfloor \log_2 n \rfloor)$

# Example: simple multi-set

Algorithm `Insert(x)`:

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Algorithm Insert(x):

- New array  $A'_0$   $\leftarrow$  [ $x$ ],  $i \leftarrow 0$

$A'_0$     [ $x$ ]

# Example: simple multi-set


Algorithm Insert(x):

- New array  $A'_0 \leftarrow [x], i \leftarrow 0$
- while  $A_i \neq \emptyset$ , set  $A'_{i+1} = \text{Merge}(A_i, A'_i), A_i \leftarrow \emptyset, i \leftarrow i + 1$

```
while (  $A_i \neq \emptyset$  ) {  
    set  $A'_{i+1} = \text{Merge}(A_i, A'_i)$ ;  
     $A_i \leftarrow \emptyset$ ;  
     $i++$ ;  
}
```

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Insert(11)

Pre-insert

$A_0$ : ~~[50]~~  $\emptyset$   
 $A_1$ : ~~[8, 99]~~  $\emptyset$   
 $A_2$ :  $\emptyset$   
 $A_3$ : [1, 10, 18, ..., 75]

$A'_0$ : [11]  
 $A'_1$ : [11, 50]  
 $A'_2$ : [8, 11, 50, 99]

Post-insert

$A_0$ :  $\emptyset$   
 $A_1$ :  $\emptyset$   
 $A_2$ : [8 | 11 | 50 | 99]  
 $A_3$ : [1, 10, 18, ...]

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`Insert(11)`

	Pre-insert		Temporary		Post-insert
$A_0$ :	[50]		$A'_0$ : [11]		$A_0$ : $\emptyset$
$A_1$ :	[8, 99]		$A'_1$ : [11, 50]	$\implies$	$A_1$ : $\emptyset$
$A_2$ :	$\emptyset$		$A'_2$ : [8, 11, 50, 99]		$A_2$ : [8, 11, 50, 99]
$A_3$ :	[1, 10, 18, ..., 75]				$A_3$ : [1, 10, 18, ..., 75]

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**Assumption:** creating new array  $A'_i$  with length  $2^i$  (and, for  $i > 0$  subsequent merge of  $A'_{i-1}$  and  $A_{i-1}$ ) has costs  $\Theta(2^i)$



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In the worst case, inserting an element into the data structure provides  $\log_2 n$  such operations.

⇒ **Worst-case Costs Insert:**

$$\sum_{i=0}^k 2^i = 2^{k+1} - 1 \in \Theta(n).$$

# Aggregate analysis

Level	Costs	Example Array
0	1	[*]
1	2	[*,*]
2	4	[*,*,*,*]
3	8	$\emptyset$
4	16	[*,*,*,*,*,*,*,*,*,*,*,*,*,*,*]

# Aggregate analysis

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**Observation:** Starting with an empty container, an insertion sequence reaches level 0 each time, level 1 (with costs 2) every second time, level 2 (with costs 4) every fourth time, etc.

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■ Total costs:  $1 \cdot \frac{n}{1} + 2 \cdot \frac{n}{2} + 4 \cdot \frac{n}{4} + \dots + 2^k \cdot \frac{n}{2^k} =$

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**Observation:** Starting with an empty container, an insertion sequence reaches level 0 each time, level 1 (with costs 2) every second time, level 2 (with costs 4) every fourth time, etc.

- Total costs:  $1 \cdot \frac{n}{1} + 2 \cdot \frac{n}{2} + 4 \cdot \frac{n}{4} + \dots + 2^k \cdot \frac{n}{2^k} = (k + 1)n$   
This is in  $\Theta(n \log n)$  because  $k = \log_2 n$ .
- **Amortized cost per operation:**  $\Theta((n \log n)/n) = \Theta(\log n)$ .

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Ideas?

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- Every element  $i$  ( $1 \leq i \leq n$ ) pays  $a_i = \log_2 n$  coins when it is inserted into the data structure.



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- The account provides enough credit to pay for all Merge operations of the  $n$  elements.

# Account method

*"guess a cost" and proof that account never runs out of credit*

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  - The element pays the allocation of the first array and every subsequent merge-step that can occur until the element has reached array  $A_{k+1}$  ( $k = \lfloor \log_2 n \rfloor$ ).
  - The account provides enough credit to pay for all Merge operations of the  $n$  elements.
- ⇒ **Amortized costs** for insertion  $\mathcal{O}(\log n)$

# Potential method

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We know from the account method that **each element on the way to higher levels requires  $\log n$  coins**, i.e. that an element on level  $i$  still needs to possess  $k - i$  coins.

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$$\Phi_j = \sum_{0 \leq i \leq k: A_i \neq \emptyset} (k - i) \cdot 2^i$$

# Potential method

For the **change of the potential**  $\Phi_j - \Phi_{j-1}$  we only have to consider the lower  $l$  levels that are occupied at time point  $j - 1$  (in analogy to the binary counter). Let  $l$  be the smallest index such that array  $A_l$  is empty.



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After merging arrays  $A_0 \dots A_{l-1}$ , array  $A_l$  is full and arrays  $A_i (0 \leq i < l)$  are now empty. Therefore:

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$$\Phi_j - \Phi_{j-1} = (k - l) \cdot 2^l - \sum_{i=0}^{l-1} (k - i) \cdot 2^i$$

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Real costs:

$$t_j = \sum_{i=0}^l 2^i = 2^{l+1} - 1$$

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$$\begin{aligned}\Phi_j - \Phi_{j-1} &= (k - l) \cdot 2^l - \sum_{i=0}^{l-1} (k - i) \cdot 2^i \\ &= (k - l) \cdot 2^l - k \cdot (2^l - 1) + \sum_{i=0}^{l-1} i \cdot 2^i \\ &= (k - l) \cdot 2^l - k \cdot (2^l - 1) + l \cdot 2^l - 2^{l+1} + 2 \\ &= k - 2^{l+1} + 2\end{aligned}$$

$$\implies \Phi_j - \Phi_{j-1} + t_j = k - 2^{l+1} + 2 + 2^{l+1} - 1 = k + 1 \in \Theta(\log n)$$

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$$\implies \Phi_j - \Phi_{j-1} + t_j = k - 2^{l+1} + 2 + 2^{l+1} - 1 = k + 1 \in \Theta(\log n)$$

See CLRS Chapter 16.

$$\sum i \cdot \lambda^i$$

Always the same trick:

$$\sum i \cdot \lambda^i$$

Always the same trick:

$$\begin{aligned} \lambda \cdot \sum_{i=0}^n i \cdot \lambda^i - \sum_{i=0}^n i \cdot \lambda^i &= \sum_{i=0}^n i \cdot \lambda^{i+1} - \sum_{i=0}^n i \cdot \lambda^i = \sum_{i=1}^{n+1} (i-1) \cdot \lambda^i - \sum_{i=0}^n i \cdot \lambda^i \\ &= n \cdot \lambda^{n+1} + \sum_{i=1}^n (i-1) \cdot \lambda^i - i \cdot \lambda = n \cdot \lambda^{n+1} - \sum_{i=1}^n \lambda^i \\ &= n \cdot \lambda^{n+1} - \frac{\lambda^{n+1} - 1}{\lambda - 1} + 1 \\ \implies (\lambda - 1) \cdot \sum_{i=0}^n i \cdot \lambda^i &= n \cdot \lambda^{n+1} - \frac{\lambda^{n+1} - 1}{\lambda - 1} + 1 \end{aligned}$$

For  $\lambda = 2$ :

$$\sum_{i=0}^n i \cdot 2^i = n \cdot 2^{n+1} - 2^{n+1} + 1 + 1 = (n-1) \cdot 2^{n+1} + 2$$



# Quiz

```
void g(unsigned n){
  for (unsigned k = 1; k != n ; ++k){
    // what does the following code do?
    unsigned prev = k-1;
    for (unsigned num = k; num != 0; num /= 2){
      if (num % 2 != prev % 2)
        f();
      prev /= 2;
    }
  }
}
```

Q: Asymptotic number of calls of  $f$ ?

# Quiz

```
void g(unsigned n){
    for (unsigned k = 1; k != n ; ++k){
        // call f for all bits that toggle from k-1 to k
        unsigned prev = k-1;
        for (unsigned num = k; num != 0; num /= 2){
            if (num % 2 != prev % 2)
                f();
            prev /= 2;
        }
    }
}
```

$k-1$       10111  
 $k$             11000

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# Quiz

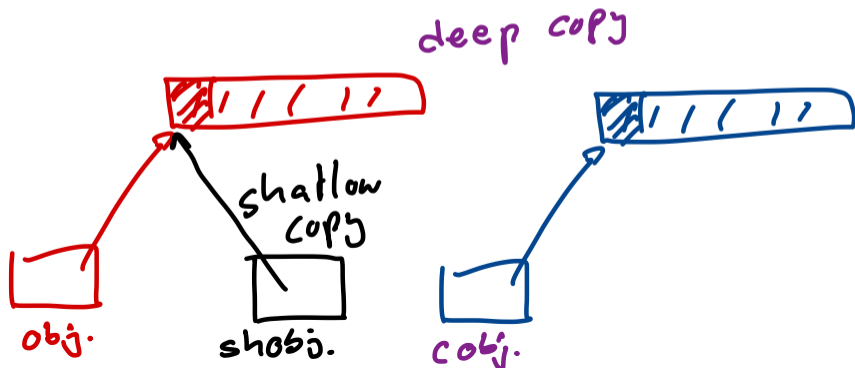
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        }
    }
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```

Q: Asymptotic number of calls of  $f$ ?

A:  $\Theta(n)$  (Counting example from class).

# Recap dynamically allocated memory

Important: Every `new` needs its `delete` and only one!



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Therefore “Rule of three”:

- constructor
- copy constructor
- destructor

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Therefore “Rule of three”:

- constructor
- copy constructor
- destructor

Being lazy “Rule of two”:

- never copy (unsafe)
- make copy constructor private (safe) or deleted

## 7. Code-Example: Dynamically Sized Array

Preparation for **code expert** exercise *Double Ended Queue*

## 8. Tips for **code** expert

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# Tips for next **code expert** exercises

## **Task "Stable and In-Situ Sorting"**

# Tips for next **code expert** exercises

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- "...in their unmodified form..."

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## **Task "Amortized Analysis: Dynamic Array"**

- Ottman/Widmayer, Chapter 3.3 (depending on version)
- Cormen et al, Chapter 17 (or 16 depending on version)

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## **Task "Double Ended Queue"**

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## Task "Double Ended Queue"

- Takes time – make sure to start early!
- Dynamic data types and memory management (fun!)
- By the way: *the name Double Ended Queue may be misleading because it suggests to be implemented with a linked list. This would make it hard, if not impossible, to fulfill the requirements stated above. Rather think of something like a vector and extend it with `push_front()`*

## 9. Old Exam Question

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# Recurrence Equation

Gegeben sei die folgende Rekursionsgleichung:

$$T(n) = \begin{cases} 4T(n/2) + 3n, & n > 1 \\ 3 & n = 1 \end{cases}$$

Leiten Sie eine geschlossene (nicht rekursive), einfache Formel für  $T(n)$  her. Nehmen Sie an, dass es ein  $k \in \mathbb{N}$  gibt mit  $2^k = n$ . Zeigen Sie mit vollständiger Induktion, dass Ihr Ergebnis stimmt.

Hinweis: Es gilt

$$4^{\log_2 n} = n^2 \text{ und } \sum_{i=0}^{k-1} 2^i = 2^k - 1$$

(D&A Exam 25.8.2022)

Consider the following recursion equation:

not just asymptotic!

Derive a closed (non-recursive), simple formula for  $T(n)$ . Assume that there is some  $k \in \mathbb{N}$  for which  $2^k = n$ . Prove by induction that your solution is correct.

Hint: it holds that

optional



# Recurrence Equation – Solution I

$$\begin{aligned}T(n) &= 4T(n/2) + 3n \\&= 4(4T(n/4) + 3n/2) + 3n \\&= 4(4(4T(n/8) + 3n/4) + 3n/2) + 3n \\&= \dots \\&= T(1) \cdot 4^k + 3n \cdot \sum_{i=0}^{k-1} 2^i \\&= 3n^2 + 3n(2^k - 1) \\&= 3n^2 + 3n(n - 1) \\&= 3n(2n - 1)\end{aligned}$$

(D&A Exam 25.8.2022)

# Recurrence Equation – Solution II

Let  $f(n) = 3n(2n - 1)$

We show that  $f(n) = T(n)$  for all  $n$  such that there is some  $k \in \mathbb{N}$  for which  $2^k = n$ .

Induction base: It holds that  $f(1) = 3 = T(1)$ .

Induction step: Assume that  $T(n) = f(n)$  (induction hypothesis). We now show that  $T(2n) = f(2n)$ .

$$\begin{aligned}T(2n) &= 4T(n) + 6n \\ &\stackrel{i.h.}{=} 12n(2n - 1) + 6n \\ &= 6n(2(2n - 1) + 1) \\ &= 3 \cdot 2n(2 \cdot 2n - 1) \\ &= f(2n).\end{aligned}$$

(D&A Exam 25.8.2022)

# 10. Outro

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# General Questions?

See you next time

Have a nice week!