

Amortized Analysis

October 2020

The problem solved by amortized analysis

Some algorithms on data structures have operations with varying time complexities

long-run computation $C =$ sequence of operations o_1, o_2, \dots, o_n

- the operations are often fast, but from time to time they are slow
- ⇒ The worst-case runtime estimate $T(C) = \sum_{i=1}^n T(o_i)$ where $T(o_i)$ where $T(o_i)$ are the worst-case runtime estimates of o_i is **too inaccurate**.

The problem solved by amortized analysis

Some algorithms on data structures have operations with varying time complexities

long-run computation $C =$ sequence of operations o_1, o_2, \dots, o_n

- the operations are often fast, but from time to time they are slow
- ⇒ The worst-case runtime estimate $T(C) = \sum_{i=1}^n T(o_i)$ where $T(o_i)$ where $T(o_i)$ are the worst-case runtime estimates of o_i is **too inaccurate**.

Amortized analysis is a better method to estimate the time complexity of many operations at once.

Case study: Dynamic arrays

An array A with 2^n elements

`insert(A, x)`: inserts item x at the next free position in the array

- ▶ we double the size of A when it becomes full, and copy all elements from the old array to the newly created array
→ time consuming operation

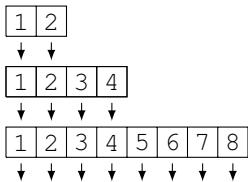
Case study: Dynamic arrays

An array A with 2^n elements

$\text{insert}(A, x)$: inserts item x at the next free position in the array

- ▶ we double the size of A when it becomes full, and copy all elements from the old array to the newly created array
→ time consuming operation

Analyzing the time complexity:



⇒ time complexity of i -th operation $\text{insert}(A, x)$:

- ▶ $O(2^n)$ if $i = 2^n$
- ▶ $O(1)$, otherwise.

Case study: Dynamic arrays

Analyzing time complexity

Number of elementary operations (assignment or copy) required by the i -th operation `insert(A, x)`:

item number	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
elem. ops.	1	2	1	4	1	1	1	8	1	1	1	1	1	1	1	16

Case study: Dynamic arrays

Analyzing time complexity

Number of elementary operations (assignment or copy) required by the i -th operation `insert(A, x)`:

item number	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
elem. ops.	1	2	1	4	1	1	1	8	1	1	1	1	1	1	1	16

- **Amortized analysis** with the **aggregate method**

- computes an upper bound of the total cost $T(n)$ of a sequence of n operations. In our example

$$\begin{aligned} T(n) &\leq \sum_{i=1}^n 1 + \sum_{i=1}^{\lfloor \log_2 n \rfloor} 2^i = n + (2^{\lfloor \log_2 n \rfloor + 1} - 1) \\ &\leq 2 + ((n + 1) - 1) = 2 \cdot n \end{aligned}$$

⇒ $T(n)/n$ is called the **amortized cost** per operation:

$$\frac{T(n)}{n} = \frac{2 \cdot n}{n} = 2 = O(1)$$

- ⇒ amortized cost is constant.
- ⇒ this cost applies to each operation, even if there are several types of operations in the sequence.

Other methods for amortized analysis

There are three frequently-used methods for amortized analysis:

- 1 aggregate method
- 2 accounting method
- 3 potential method

Accounting method

Assigns different amortized costs to different operations.

- The amortized cost depends on i (the i -th iteration) and may differ from the actual cost.
 - With aggregate method, all operations are assumed to have same amortized cost. With accounting method, they can have different amortized costs.
- When an operation's amortized cost exceeds its actual cost, the surplus goes into a "bank".
- Idea: Need to **overcharge** for simpler operations, to build up enough **savings** to afford a more expensive operation later
 - Bank balance must always be ≥ 0

Accounting method

Illustrated example: Dynamic array

Charge 3 units per operation

\$0

Accounting method

Illustrated example: Dynamic array

Charge 3 units per operation

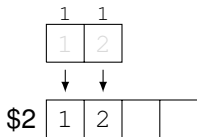
\$4

	²	²
1	2	

Accounting method

Illustrated example: Dynamic array

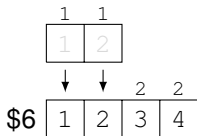
Charge 3 units per operation



Accounting method

Illustrated example: Dynamic array

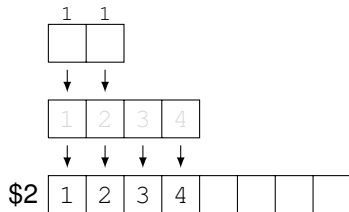
Charge 3 units per operation



Accounting method

Illustrated example: Dynamic array

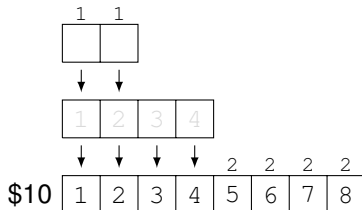
Charge 3 units per operation



Accounting method

Illustrated example: Dynamic array

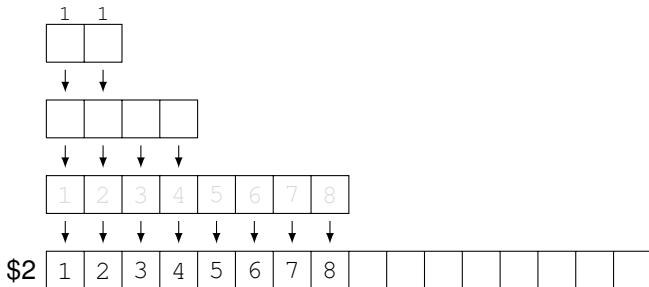
Charge 3 units per operation



Accounting method

Illustrated example: Dynamic array

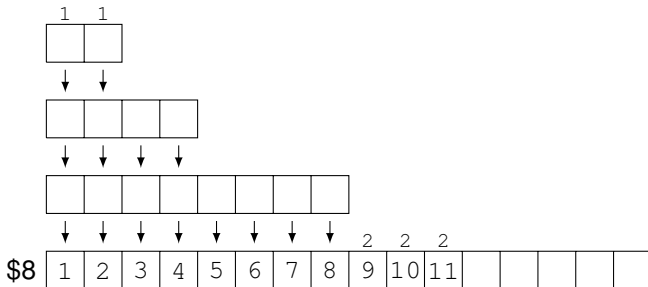
Charge 3 units per operation



Accounting method

Illustrated example: Dynamic array

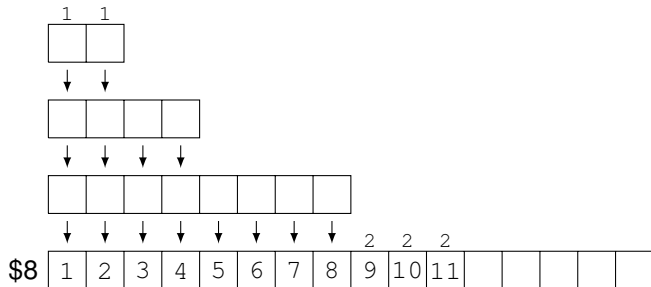
Charge 3 units per operation



Accounting method

Illustrated example: Dynamic array

Charge 3 units per operation



- Bank balance is always ≥ 2
- ⇒ each operation has constant amortized cost 3.

Potential method

Conceptually similar to accounting method

- Same idea of using stored surplus to pay for more expensive operations
- The **potential (energy)** is analogous to the **bank**
- Potential must always ≥ 0

Key differences:

- Accounting method: bank balance of a particular state is **dependent** on previous state
- Potential method involves a **potential function** $\Phi(A)$
 - can be used to derive the potential at any state
 - can also be used to compute a **potential difference**, which shows the change in cost between two operations

Potential method

Conceptually similar to accounting method

- Same idea of using stored surplus to pay for more expensive operations
- The **potential (energy)** is analogous to the **bank**
- Potential must always ≥ 0

Key differences:

- Accounting method: bank balance of a particular state is **dependent** on previous state
- Potential method involves a **potential function** $\Phi(A)$
 - can be used to derive the potential at any state
 - can also be used to compute a **potential difference**, which shows the change in cost between two operations

Finding a potential function is the challenging part!

Potential method

Illustrated example: Dynamic array

$\Phi(A) = 2(i + 1) - \text{size}(A)$ where i is the number of elements in A

- Easy to check that $\Phi(A)$ is always ≥ 0 .

The amortized cost \hat{c}_i of the i -th operation is defined to be

$\hat{c}_i = c_i + \Phi(A_i) - \Phi(A_{i-1})$ where

- c_i is the actual cost of the i -th operation
- A_i is the dynamic array after the i -th operation

We distinguish 2 cases:

- 1 i -th step is normal step. Then $c_i = 1$, $\text{size}(A_i) = \text{size}(A_{i+1})$ and $\hat{c}_i = 1 + (2(i + 1) - \text{size}(A_i)) - (2i - \text{size}(A_i)) = 3$
- 2 i -th step is an expansion step. This happens when $\text{size}(A_i) = i$ and $\text{size}(A_{i+1}) = 2i$. Then $c_i = i$ and $\hat{c}_i = i + (2(i + 1) - 2i) - (2i - i) = 3$

\Rightarrow every operation has constant amortized cost 3.

- Aggregate analysis defines amortized cost of any operation as an average:

$$\frac{\text{total cost of a sequence of operations}}{\text{number of operations}}$$

- Potential and Accounting methods involve assigning amortized costs per operation
 - Cheap operations are **overcharged** \Rightarrow we acquire a surplus
 - Expensive operations are paid for by the **surplus**