

# Dynamic Arrays and Amortized Analysis

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To create an array, you need to specify a size, i.e., how many elements you can store in the array. Increasing the size is expensive because it means creating a new array and moving all the elements over.

This lecture will discuss clever tricks to change the array size efficiently! Our discussion will introduce **amortized analysis**, a new method for measuring the performance of a data structure.

## Dynamic Array Problem

Let  $S$  be a collection of integers (not necessarily distinct).  $S$  is empty in the beginning. Integers are then added to  $S$  one by one with **insertions**.

Let  $n$  be the number of elements in  $S$  currently. We want to maintain an array  $A$  satisfying:

- 1  $A$  has length  $O(n)$ .
- 2 For each  $i \in [1, n]$ ,  $A[i] = x$  if  $x$  is the  $i$ -th integer added to  $S$ .

The above requirements need to be satisfied after every insertion.

## Naive Algorithm

Perform  $\text{insert}(e)$  (which inserts an integer  $e$  to  $S$ ) as follows:

- If  $n = 0$ , set  $n$  to 1 and initialize  $A$  to have length 1 to store  $e$ .
- Otherwise ( $n \geq 1$ ):
  - Increase  $n$  by 1.
  - Initialize an array  $A'$  of length  $n$ .
  - Copy all the  $n - 1$  elements of  $A$  to  $A'$ .
  - Set  $A'[n] = e$ .
  - Destroy  $A$  and replace it with  $A'$ .

This algorithm spends  $O(n)$  time on the  $n$ -th insertion. Altogether, it takes  $O(n^2)$  time to do  $n$  insertions.

We will reduce the time of inserting  $n$  elements dramatically to  $O(n)$ . Our array  $A$  may have a length up to  $2n$ .

## A Better Algorithm

$A$  is **full** if its cells are all filled.

Perform  $\text{insert}(e)$  as follows:

- If  $n = 0$ , set  $n$  to 1 and initialize  $A$  of length 2 to store just  $e$  itself.
- Otherwise (i.e.,  $n \geq 1$ ), append  $e$  to  $A$  and increase  $n$  by 1. If  $A$  is full:
  - Initialize an array  $A'$  of length  $2n$ .
  - Copy all the elements of  $A$  to  $A'$ .
  - Destroy  $A$  and replace it with  $A'$ .

## Example

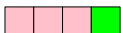
$n = 1$



$n = 2$



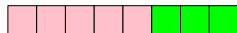
$n = 3$



$n = 4$

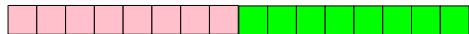


$n = 5$



...

$n = 8$



## Analysis

Cost of inserting the  $n$ -th element:

- if  $A$  is not full after the insertion,  $O(1)$ ;
- otherwise,  $O(n)$ , i.e., the time of **expanding**  $A$ .



## Analysis

Array expansions are infrequent:

- Initially, size 2.
- 1st expansion: size from 2 to 4.
- 2nd expansion: from 4 to 8.
- ...
- $i$ -th expansion: from  $2^i$  to  $2^{i+1}$ .

After  $n$  insertions, the size of  $A$  is at most  $2n$ . Hence:

$$2^{i+1} \leq 2n \quad \Rightarrow \quad i \leq \log_2 n$$

that is, at most  $\log_2 n$  expansions.

## Analysis

The total cost of  $n$  insertions is bounded by:

$$\left( \sum_{i=1}^n O(1) \right) + \sum_{i=1}^{\log_2 n} O(2^i) \quad (1)$$

where

- the first term captures the  $O(1)$  time compulsory for each insertion;
- the second term captures all the expansion cost.

(1) evaluates to  $O(n)$ .

We have shown that the total cost of  $n$  insertions is  $O(n)$ . In other words, each insertion entails  $O(1)$  cost “on average”. This does not mean that every insertion can be performed in  $O(1)$  time. The cost of some insertions can reach  $\Omega(n)$ .

In general, if a data structure can process any  $n$  operations in  $f(n)$  time, we say that it guarantees an **amortized cost** of  $\frac{f(n)}{n}$  per operation.

The dynamic array guarantees  $O(1)$  amortized cost per insertion.