

Data Structures and Algorithms

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$$11 \times 2_m + 1 \times 5_m = 27_m \cancel{12}$$

Amortized Analysis

Data Structures and Algorithms
Undergraduate course



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Amortized Analysis

Amortized analysis refers to determining the time-averaged running time for a sequence of operations.

Example

Consider selecting the k^{th} smallest element in an array by sorting the array...

How to perform the amortized analysis?

- Aggregate method
- Accounting method
- Potential method

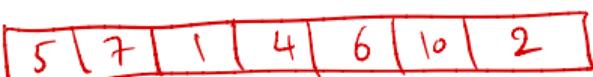
→ average cost per operation

k-Select

$O(n)$



Time Complexity for single execution = $O(n \log n)$
 n execs = $O(n \log n) + O(n)$



$k=3$

sort →



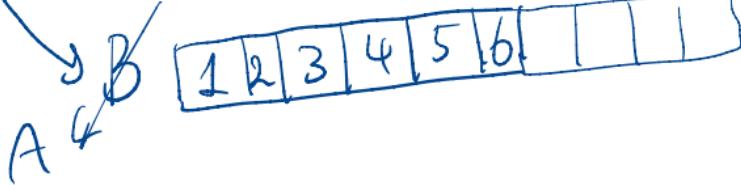
$O(n \log n)$

$O(\log n)$

Array :

A	1	2	3	4	5
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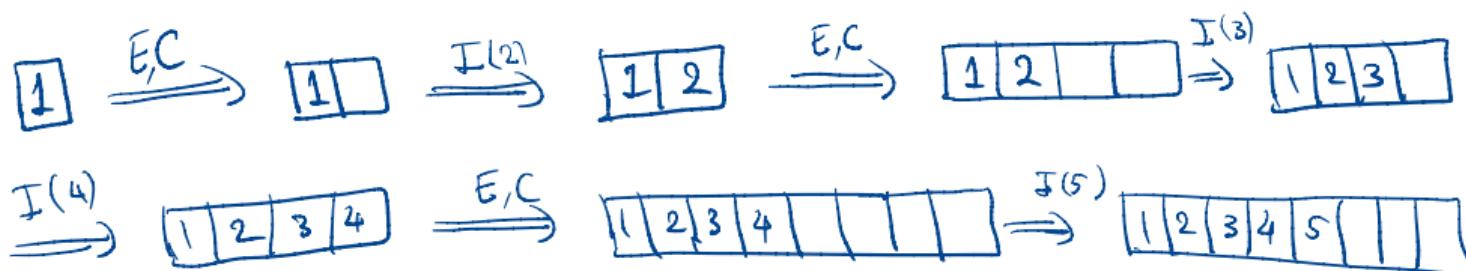
insert (6)



Example: Dynamic Array

Idea: Double the array size when it becomes full.

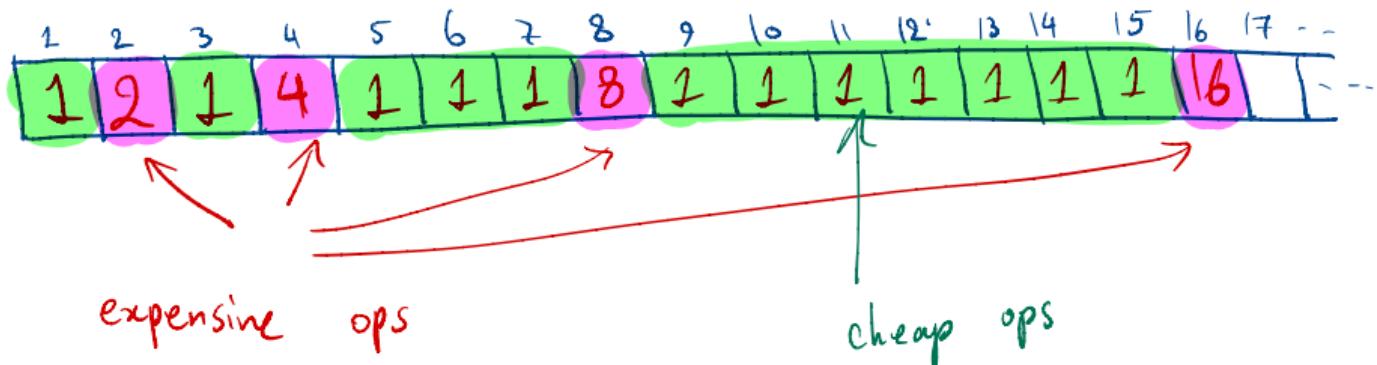
- Insertion Time Complexity: $O(n)$ (which is not tight!) \times
- Cheap operation: Insertion with $O(1)$
- Expensive operation: array expansion and copy with $O(n)$



Dynamic Array: Aggregate method

$$\frac{o(\text{total cost})}{n} = \frac{O(n)}{n} = O(1)$$

- * Simplest way.
- * $\frac{O(\text{total cost})}{\# \text{ operation}}$ should be calculated.

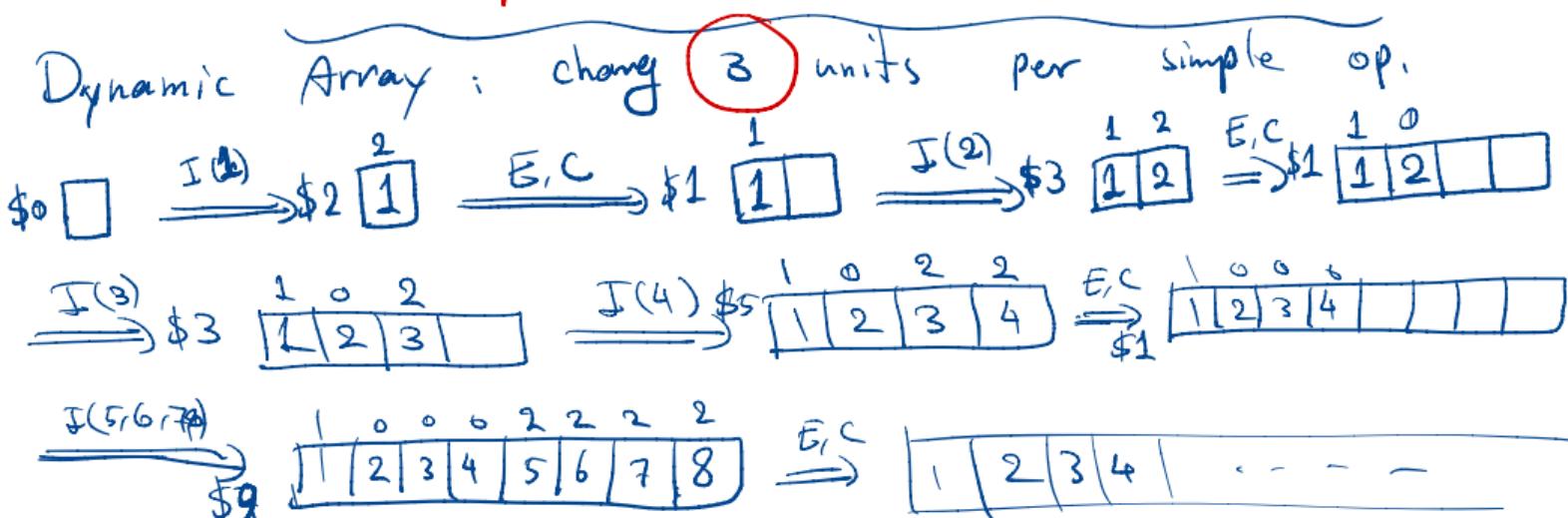


$$O(\text{total cost}) = O\left(\sum \text{cost of } n \text{ ops}\right) = O\left(\sum \text{cost of cheap ops} + \sum \text{cost of expensive ops}\right)$$

$$\Rightarrow \begin{cases} O\left(\sum \text{cost of } n \text{ insertions}\right) = O(n) \\ O\left(\sum \text{cost of expansions}\right) = 1 + 2 + 4 + 8 + \dots < 2n = O(n) \end{cases}$$

Dynamic Array: Accounting method (Banker's method)

- ★ Assign a cost to each operation (relative to i)
- ★ any surplus goes into a bank.
- ★ Overcharge the cheap ops to save enough money for more expensive ops.



Time Complexity for each operation = 3 = $O(1)$

Dynamic Array: Potential method

- * Similar to the accounting method.
- * The potential (energy) is analogous to the bank.
- * Potential must be ≥ 0 .
- * Potential method involves potential function $\varphi(h)$:
 - if the cost of an operation is low, the potential increases,
 - can be used to independently derive otherwise it decreases.
 - " " " " calculate the potential difference
(change in cost between two operations).
- * Finding the appropriate potential function is the challenging part!

Dynamic Array: Potential method

Example : Dynamic Array .

$$\varphi(h) = 2n - \text{size of the array}$$



$$\varphi(h) = 2 - 1 = 1$$

$$\varphi(h) = 2 - 2 = 0$$

$$\varphi(h) = 4 - 2 = 2$$

$$\varphi(h) = 4 - 4 = 0$$



$$\varphi(h) = 6 - 4 = 2$$

$$\varphi(h) = 8 - 4 = 4$$

$$\varphi(h) = 8 - 8 = 0$$

In general : $\varphi(h) = 2k - 2$ $\lceil \log_2 k \rceil$

Amortized time of the i^{th} operation h_i :

$c_i + \underbrace{\varphi(h_i) - \varphi(h_{i-1})}_{\text{cost of op.}}$ $\underbrace{\varphi(h_i) - \varphi(h_{i-1})}_{\text{potential difference}}$

Dynamic Array: Potential method

Cont.

{- cheap / simple operation :

$$\Rightarrow \{ c_i + (\cancel{2i - \text{size}}) - (\cancel{2(i-1) - \text{size}}) = 1 + 2 = 3 \}$$

after the operation before the operation

-expensive / complex operation:

$$(i+1) + \varphi(h_i) - \varphi(h_{i-1}) = (i+1) + (\cancel{2i} - \cancel{2i}) - (\cancel{2(i-1)} - \cancel{i})$$

↓
 i copy + I insert

$$= \cancel{i+1} - \cancel{2i} + 2 + \cancel{i}$$

the size after
expansion

the size
before
expansion

$\Rightarrow \boxed{\text{Amortized Time} = 3 = O(1)} \quad (= 3)$