Basic Data Structures: Dynamic Arrays and Amortized Analysis

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Data Structures Data Structures and Algorithms

Outline

1 Dynamic Arrays

2 Amortized Analysis—Aggregate Method

3 Amortized Analysis—Banker's Method

4 Amortized Analysis—Physicist's Method

Problem: static arrays are static!

```
int my_array[100];
```

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Semi-solution: dynamically-allocated arrays:

int *my_array = new int[size];

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Solution: *dynamic arrays* (also known as *resizable arrays*)

Idea: store a pointer to a dynamically allocated array, and replace it with a newly-allocated array as needed.

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Dynamic Array:

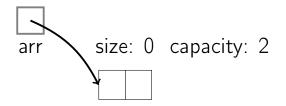
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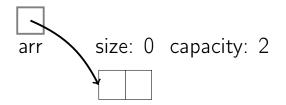
- Get(*i*): returns element at location i^*
- Set(i, val): Sets element i to val*
- PushBack(val): Adds val to the end
- Remove(i): Removes element at location i
- Size(): the number of elements

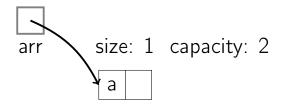
Implementation

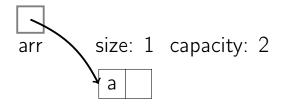
Store:

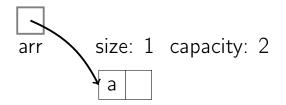
- arr: dynamically-allocated array
- capacity: size of the dynamically-allocated array
- size: number of elements currently in the array

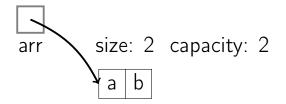


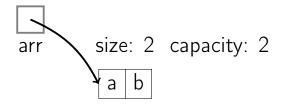


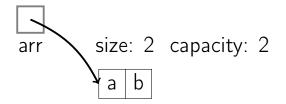


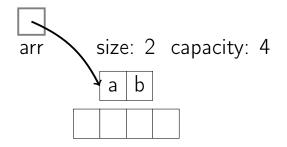


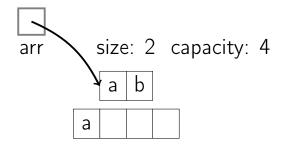


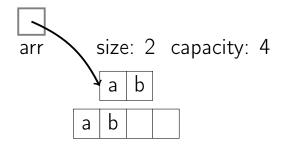


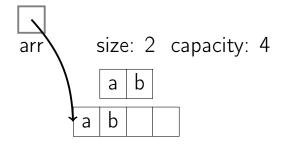


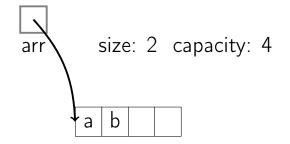


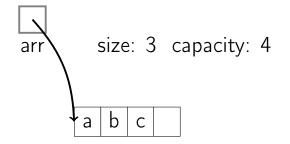


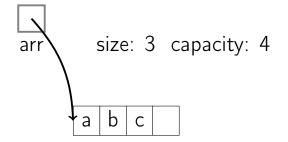


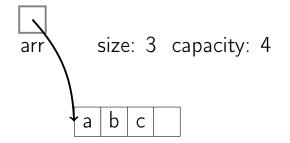


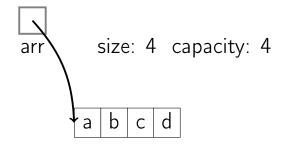


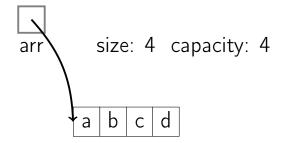


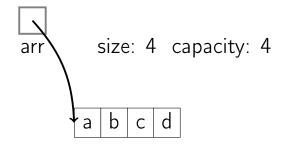


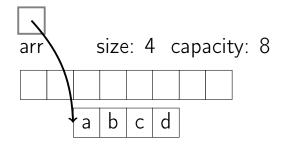


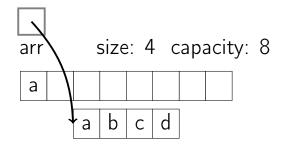


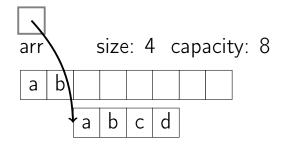


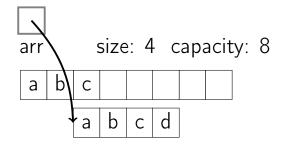


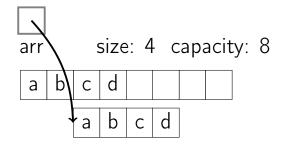


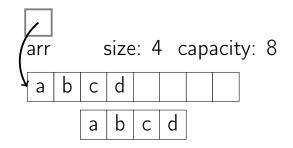


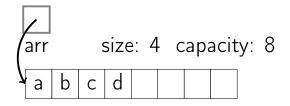


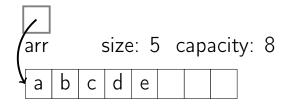












Get(i)

if i < 0 or $i \ge size$: ERROR: index out of range return arr[i]

Set(*i*, *val*)

if i < 0 or $i \ge size$: ERROR: index out of range arr[i] = val

PushBack(val)

if size = capacity: allocate *new_arr* $[2 \times capacity]$ for *i* from 0 to size -1: $new_arr[i] \leftarrow arr[i]$ free *arr* arr \leftarrow new_arr; capacity $\leftarrow 2 \times$ capacity $arr[size] \leftarrow val$ $size \leftarrow size + 1$

$\operatorname{Remove}(i)$

if i < 0 or $i \ge size$: ERROR: index out of range for j from i to size - 2: $arr[j] \leftarrow arr[j + 1]$ $size \leftarrow size - 1$

Size()

return *size*

Common Implementations

- C++: vector
- Java: ArrayList
- **Python**: list (the only kind of array)

 $\operatorname{Get}(i) \mid O(1)$

$\begin{array}{c|c} \operatorname{Get}(i) & O(1) \\ \operatorname{Set}(i, \mathit{val}) & O(1) \end{array}$

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- Appending a new element to a dynamic array is often constant time, but can take O(n).
- Some space is wasted—at most half.

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Sometimes, looking at the individual worst-case may be too severe. We may want to know the total worst-case cost for a sequence of operations.

Dynamic Array

We only resize every so often. Many O(1) operations are followed by an O(n) operations. What is the total cost of inserting many elements?

Definition

Amortized cost: Given a sequence of n operations, the amortized cost is:

n

Dynamic array: n calls to PushBack

$$c_i = 1 + \left\{ {
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Banker's Method

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- Use the tokens to pay for expensive operations.
- Like an amortizing loan.

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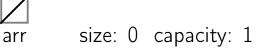
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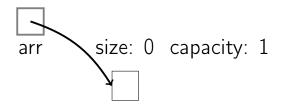
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- Place one token on the newly-inserted element, and one token ^{capacity}/₂ elements prior.

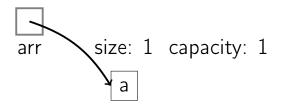
arr size: 0 capacity: 0

arr

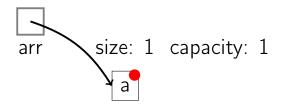
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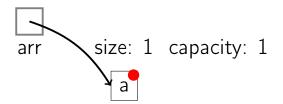


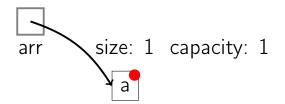


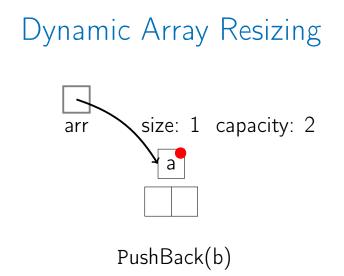


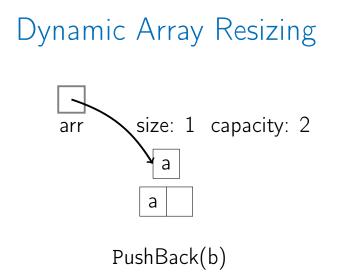


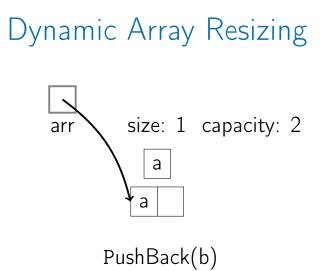


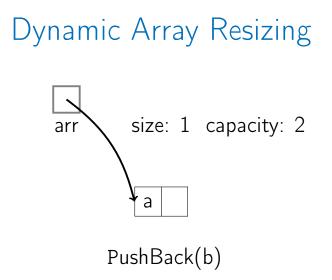


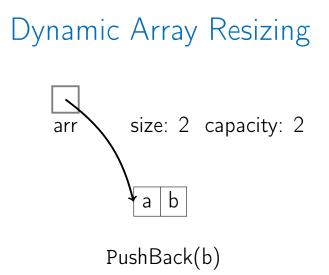


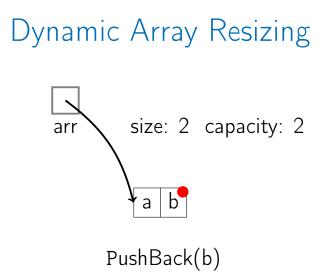


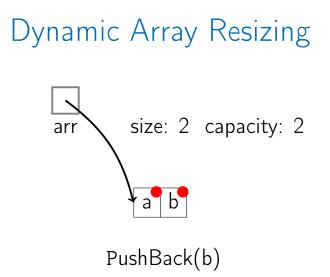




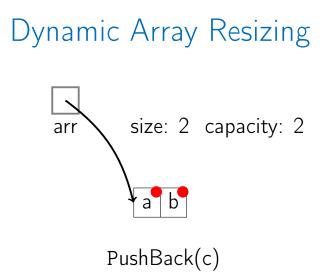


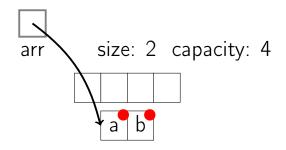


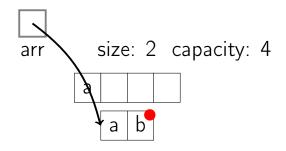


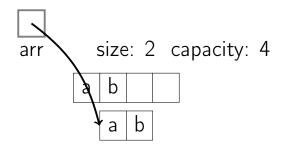


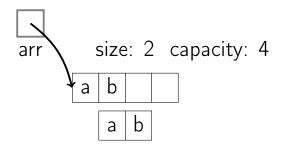
Dynamic Array Resizing size: 2 capacity: 2 arr а

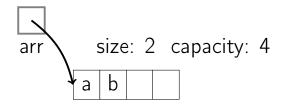


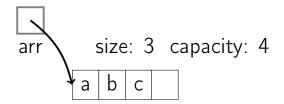


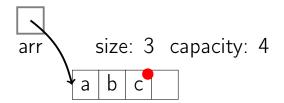


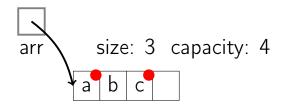


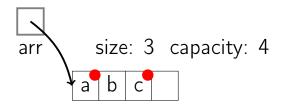


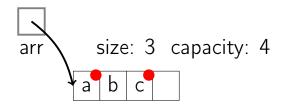


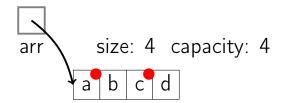


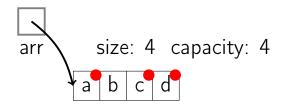


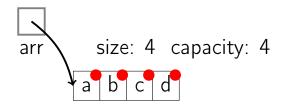


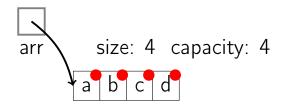


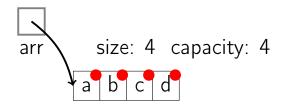


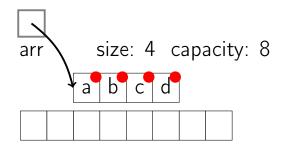


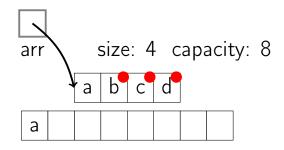


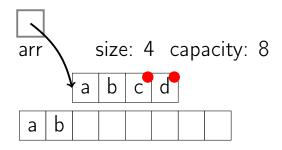


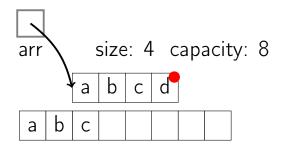


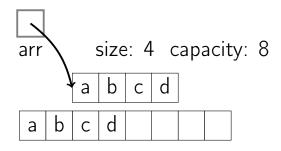


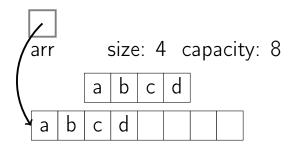


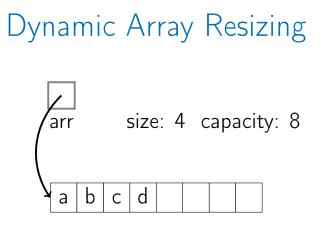


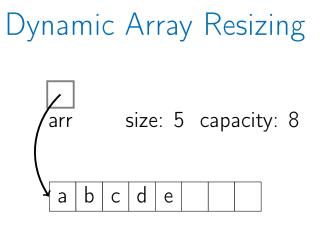


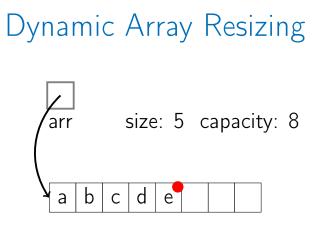


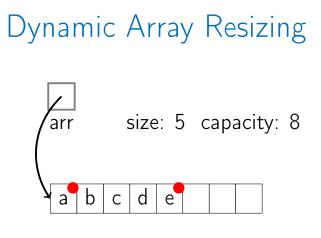


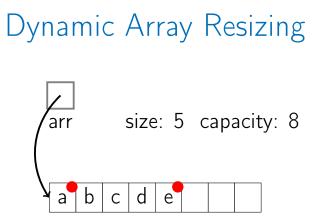


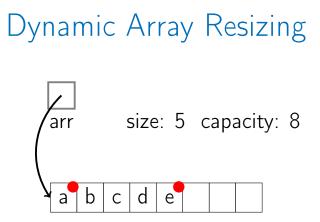












O(1) amortized cost for each PushBack

Banker's Method

Dynamic array: *n* calls to PushBack Charge 3 for each insertion. 1 coin is the raw cost for insertion.

- Resize needed: To pay for moving the elements, use the coin that's present on each element that needs to move.
- Place one coin on the newly-inserted element, and one coin ^{capacity}/₂ elements prior.

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Choose Φ so that:

- if c_t is small, the potential increases
- if *c*_t is large, the potential decreases by the same scale

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 - $\sum (c_i + \Phi(h_i) \Phi(h_{i-1}))$ i=1 $=c_1 + \Phi(h_1) - \Phi(h_0) +$ $c_2 + \Phi(h_2) - \Phi(h_1) \cdots +$ $c_n + \Phi(h_n) - \Phi(h_{n-1})$ $=\Phi(h_n)-\Phi(h_0)+\sum c_i$ i=1

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- The sum of the amortized costs is:

$$egin{aligned} &\sum_{i=1}^n (c_i + \Phi(h_i) - \Phi(h_{i-1})) \ = &c_1 + \Phi(h_1) - \Phi(h_0) + \ &c_2 + \Phi(h_2) - \Phi(h_1) \cdots + \ &c_n + \Phi(h_n) - \Phi(h_{n-1}) \ = &\Phi(h_n) - \Phi(h_0) + \sum_{i=1}^n c_i \geq \sum_{i=1}^n c_i \end{aligned}$$

Dynamic array: n calls to PushBack

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Let $\Phi(h) = 2 \times size - capacity$

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Dynamic array: n calls to PushBack

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$$\Phi(h) = 2 \times size - capacity$$

• $\Phi(h_0) = 2 \times 0 - 0 = 0$
• $\Phi(h_i) = 2 \times size - capacity > 0$
(since $size > \frac{capacity}{2}$)

Without resize when adding element i

Amortized cost of adding element *i*:

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Amortized cost of adding element *i*: $c_i + \Phi(h_i) - \Phi(h_{i-1})$

Dynamic Array Resizing

Without resize when adding element i

Amortized cost of adding element *i*: $c_i + \Phi(h_i) - \Phi(h_{i-1})$ $=1 + 2 \times size_i - cap_i - (2 \times size_{i-1} - cap_{i-1})$

Dynamic Array Resizing

Without resize when adding element i

Amortized cost of adding element *i*: $c_i + \Phi(h_i) - \Phi(h_{i-1})$ $=1 + 2 \times size_i - cap_i - (2 \times size_{i-1} - cap_{i-1})$ $=1 + 2 \times (size_i - size_{i-1})$

Dynamic Array Resizing

Without resize when adding element i

Amortized cost of adding element *i*: $c_i + \Phi(h_i) - \Phi(h_{i-1})$ $=1 + 2 \times size_i - cap_i - (2 \times size_{i-1} - cap_{i-1})$ $=1 + 2 \times (size_i - size_{i-1})$ =3

Dynamic Array Resizing With resize when adding element *i*

Dynamic Array Resizing With resize when adding element *i* Let $k = size_{i-1} = cap_{i-1}$ Dynamic Array Resizing With resize when adding element *i* Let $k = size_{i-1} = cap_{i-1}$ Then: $\Phi(h_{i-1}) = 2size_{i-1} - cap_{i-1} = 2k - k = k$

$$c_i + \Phi(h_i) - \Phi(h_{i-1})$$

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=(size_i) + 2 - k

$$c_i + \Phi(h_i) - \Phi(h_{i-1})$$

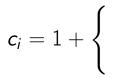
=(size_i) + 2 - k
=(k + 1) + 2 - k

$$c_i + \Phi(h_i) - \Phi(h_{i-1})$$

=(size_i) + 2 - k
=(k + 1) + 2 - k
=3

Alternatives to Doubling the Array Size

We could use some different growth factor (1.5, 2.5, etc.). Could we use a constant amount?



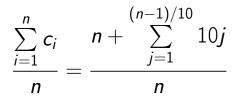
$$c_i = 1 + \left\{ egin{array}{cl} i-1 & ext{if } i-1 ext{ is a multiple of 10} \\ \end{array}
ight.$$

$$c_i = 1 + egin{cases} i-1 & ext{if } i-1 ext{ is a multiple of 10} \\ 0 & ext{otherwise} \end{cases}$$

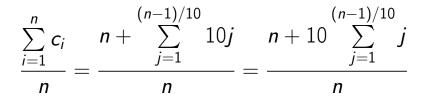
$$c_i = 1 + egin{cases} i-1 & ext{if } i-1 ext{ is a multiple of 10} \ 0 & ext{otherwise} \end{cases}$$



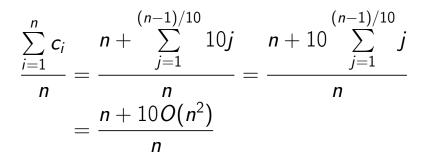
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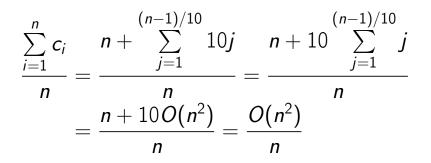
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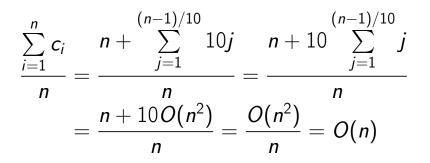
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- Three ways to do analysis:
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 - Physicist's method (potential function, Φ)
- Nothing changes in the code: runtime analysis only.