### Basic Data Structures: Dynamic Arrays and Amortized Analysis

Neil Rhodes

Department of Computer Science and Engineering University of California, San Diego

## 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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```

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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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Abstract data type with the following operations (at a minimum):

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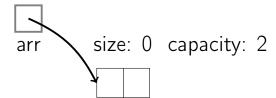
- Get(i): returns element at location  $i^*$
- Set(i, val): Sets element i to  $val^*$
- PushBack(val): Adds val to the end
- Remove(i): Removes element at location i
- Size(): the number of elements

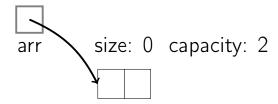
<sup>\*</sup>must be constant time

#### **Implementation**

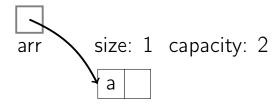
#### Store:

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

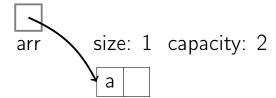


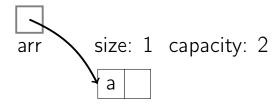


PushBack(a)

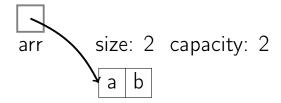


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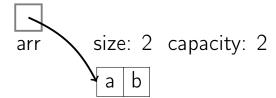


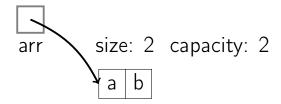


PushBack(b)

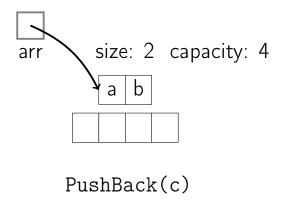


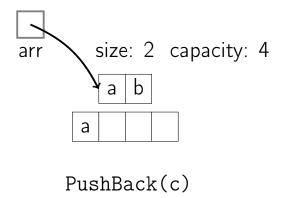
PushBack(b)

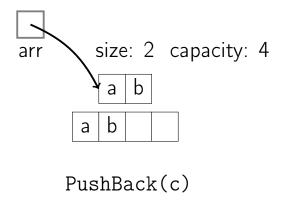


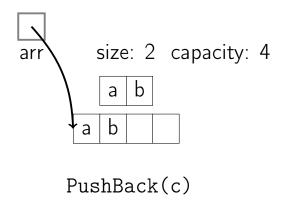


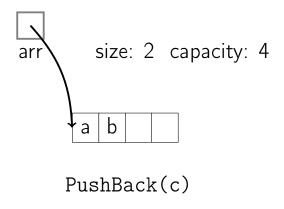
PushBack(c)

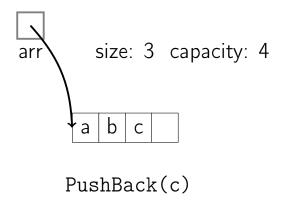


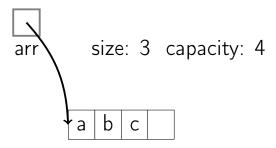


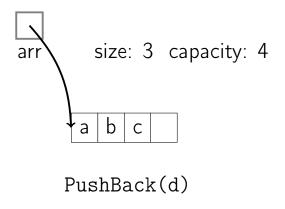


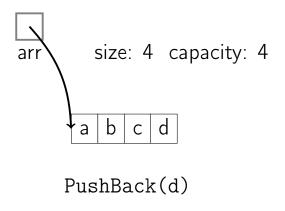


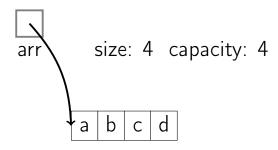


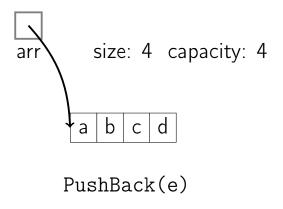


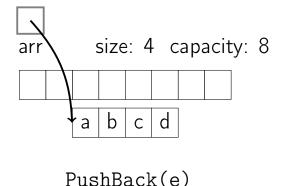


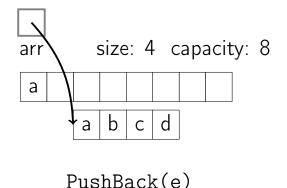


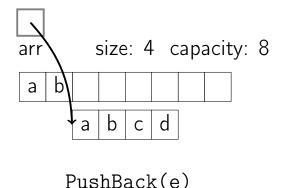


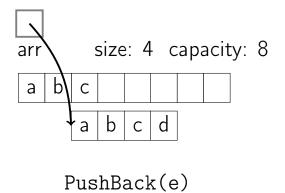


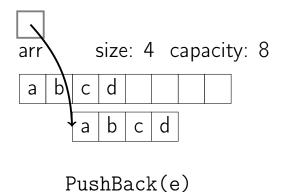


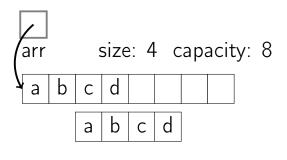












PushBack(e)

```
arr size: 4 capacity: 8
```

PushBack(e)

```
arr size: 5 capacity: 8
```

PushBack(e)

#### 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:
```

arr[i] = val

ERROR: index out of range

# 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* 

 $arr[size] \leftarrow val$ 

 $size \leftarrow size + 1$ 

 $arr \leftarrow new\_arr$ ; capacity  $\leftarrow 2 \times capacity$ 

#### Remove(i)

ERROR: index out of range

for j from i to size - 2:

 $arr[j] \leftarrow arr[j+1]$ 

 $size \leftarrow size - 1$ 

if i < 0 or  $i \ge size$ :

# Size()

return size

#### Common Implementations

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

 $Get(i) \mid O(1)$ 

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

```
egin{array}{c|c} \operatorname{Get}(i) & O(1) \ \operatorname{Set}(i,\mathit{val}) & O(1) \ \operatorname{PushBack}(\mathit{val}) & O(n) \ \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:

 $\frac{\mathsf{Cost}(n \text{ operations})}{n}$ 

Dynamic array: n calls to PushBack

$$c_i = 1 + \left\{ \right.$$

$$c_i = 1 + \begin{cases} i-1 & \text{if } i-1 \text{ is a power of 2} \end{cases}$$

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$$\frac{\sum_{i=1}^{n} c_i}{n}$$

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#### Banker's Method

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Like an amortizing loan.

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Charge 3 for each insertion: 1 token is the raw cost for insertion.

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- Place one token on the newly-inserted element, and one token  $\frac{capacity}{2}$  elements prior.

Z arr

size: 0 capacity: 0

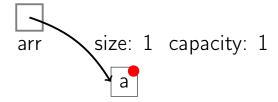
arr

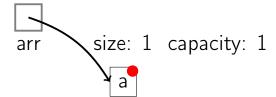
size: 0 capacity: 0

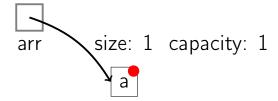
arr size: 0 capacity: 1

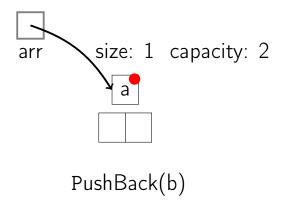
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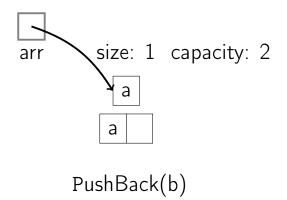
```
arr size: 1 capacity: 1
```

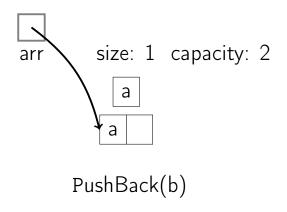


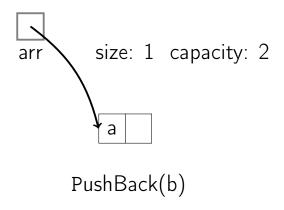


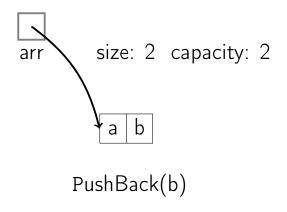


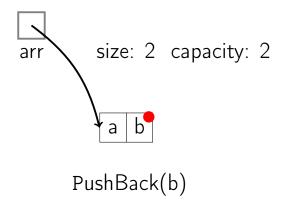


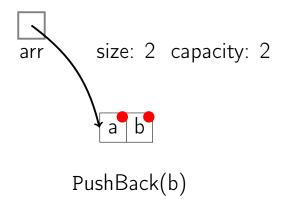


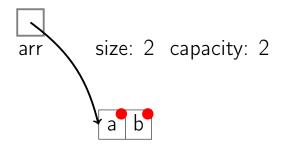


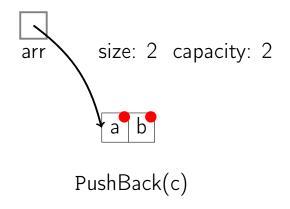


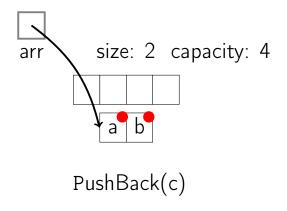


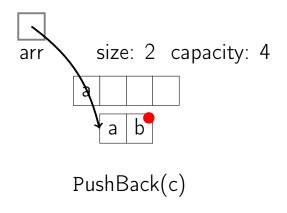


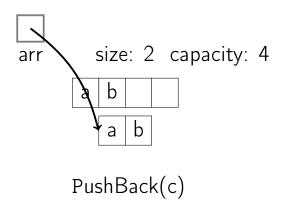


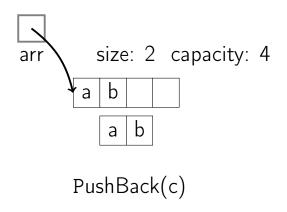


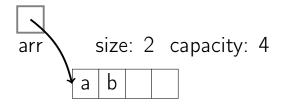


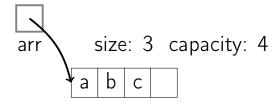


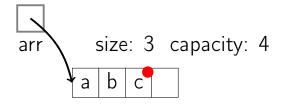


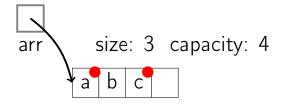


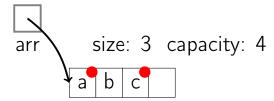


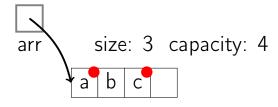


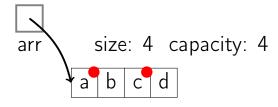


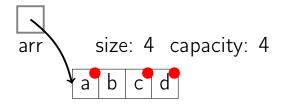


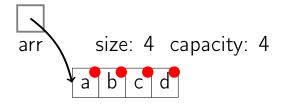




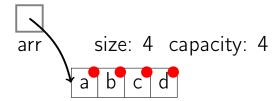


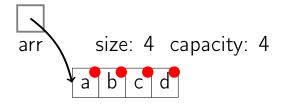




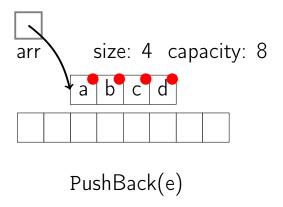


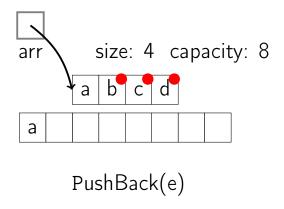
PushBack(d)

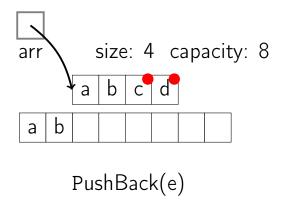


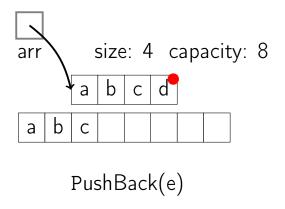


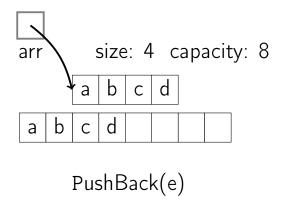
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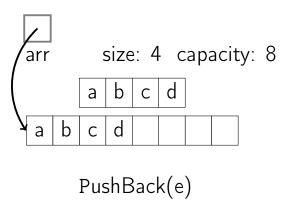


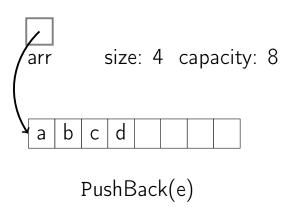


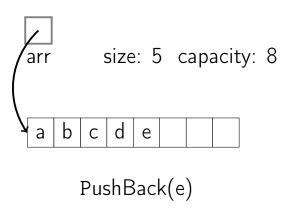


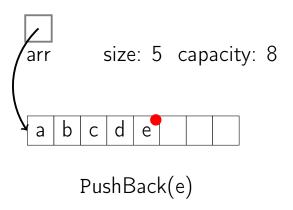


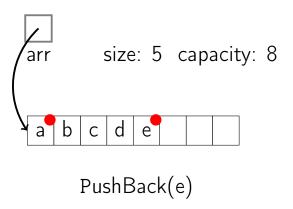


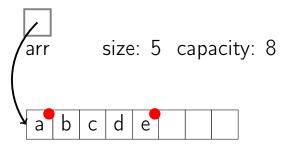


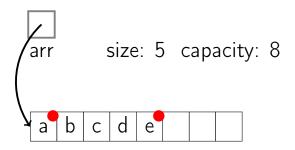












O(1) amortized cost for each PushBack

#### 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  $\frac{capacity}{2}$  elements prior.

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  - $\Phi(h_0) = 0$
  - $\Phi(h_t) > 0$
- amortized cost for operation t:

$$c_t + \Phi(h_t) - \Phi(h_{t-1})$$

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Choose  $\Phi$  so that:

- lacktriangle if  $c_t$  is small, the potential increases
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■ The cost of *n* operations is:  $\sum_{i=1}^{n} c_i$ 

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

- The cost of n operations is:  $\sum_{i=1}^{n} c_i$
- The sum of the amortized costs is:

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

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$$c_2 + \Phi(h_2) - \Phi(h_1) \cdot \cdot \cdot +$$

$$c_n + \Phi(h_n) - \Phi(h_{n-1})$$

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$$= \Phi(h_n) - \Phi(h_0) + \sum_{i=1}^n c_i \geq \sum_{i=1}^n c_i$$

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

Without resize when adding element i

Amortized cost of adding element i:

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=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?

If we expand by 10 each time, then: Let  $c_i = \cos t$  of i'th insertion.

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- Three ways to do analysis:
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- Nothing changes in the code: runtime analysis only.