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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```

int my_array[100];

Semi-solution: dynamically-allocated arrays:

```
int *my_array = new int[size];
```

Problem: might not know max size when allocating an array

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All problems in computer science can be solved by another level of indirection

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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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- Remove(i): Removes element at location i

*must be constant time

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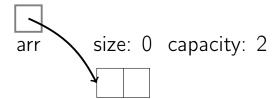
- \blacksquare Get(i): returns element at location i^*
- Set(i, val): Sets element i to val^*
- PushBack(val): Adds val to the end
- Remove(i): Removes element at location i
- Size(): the number of elements

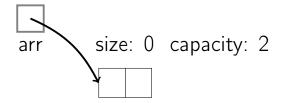
^{*}must be constant time

Implementation

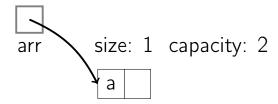
Store:

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

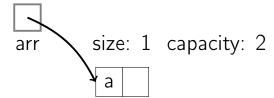


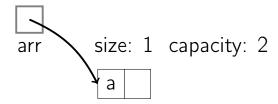


PushBack(a)

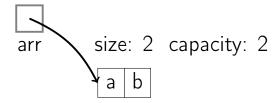


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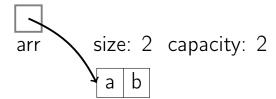


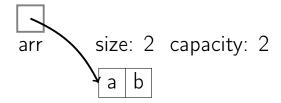


PushBack(b)

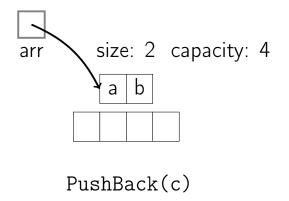


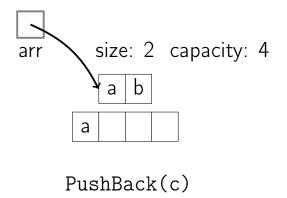
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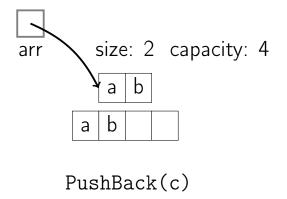


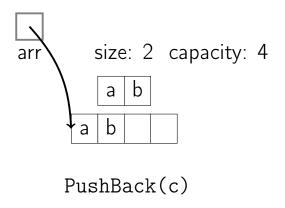


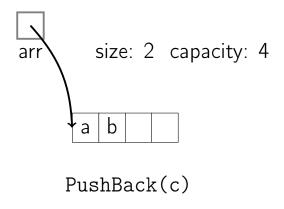
PushBack(c)

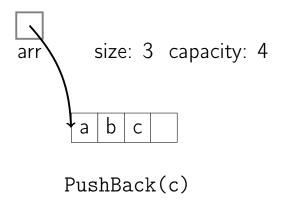


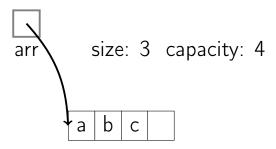


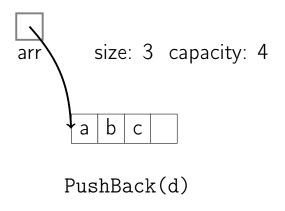


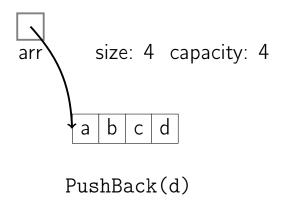


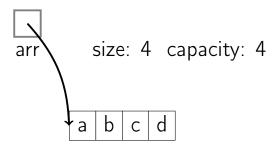


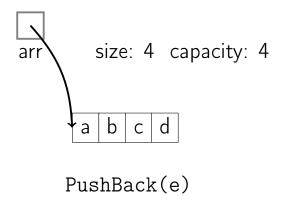


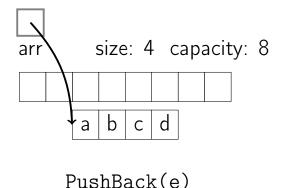


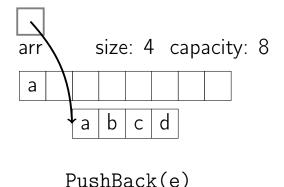


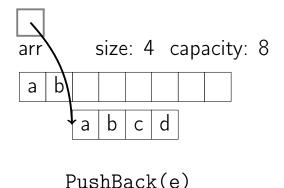


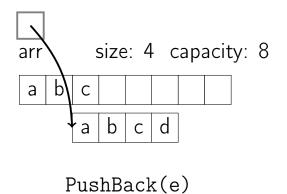


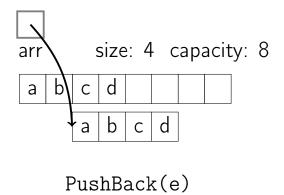


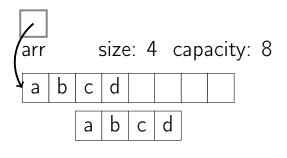












PushBack(e)

```
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)

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



 $size \leftarrow size - 1$

ERROR: index out of range

for j from i to size - 2:

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

Size()

return size

Common Implementations

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

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

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

```
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

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

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

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

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Banker's Method

■ Charge extra for each cheap operation.

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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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Resize needed: To pay for moving the elements, use the token that's present on each element that needs to move.

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

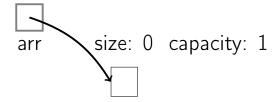
arr

size: 0 capacity: 0

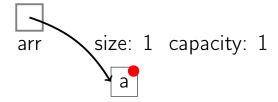
arr

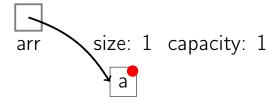
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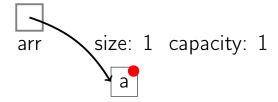
arr size: 0 capacity: 1

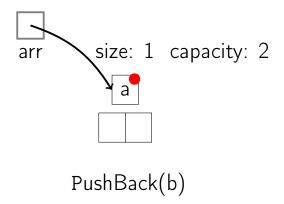


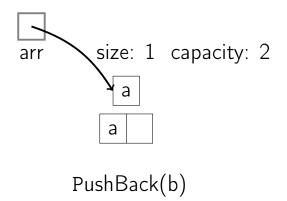
```
arr size: 1 capacity: 1
```

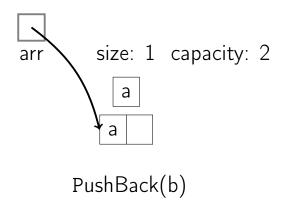


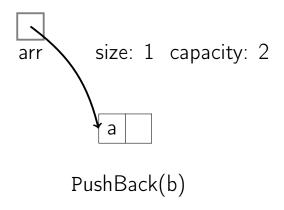


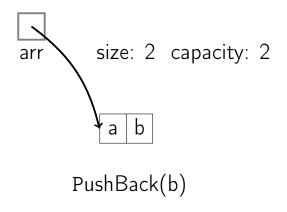


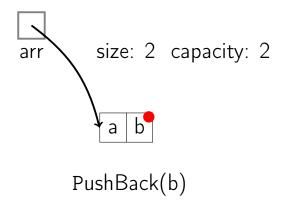


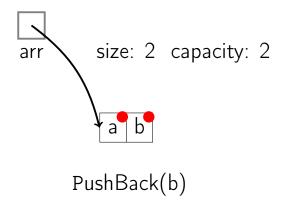


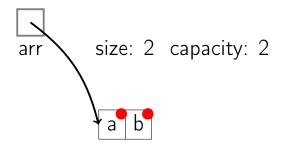


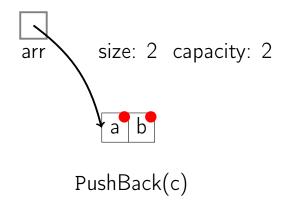


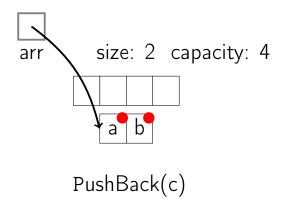


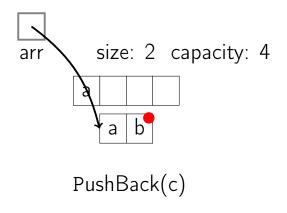


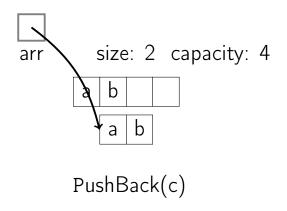


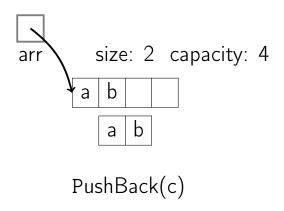


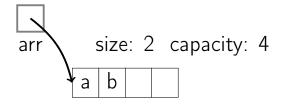


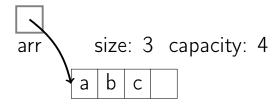


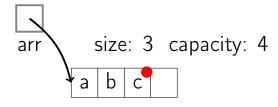


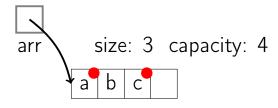


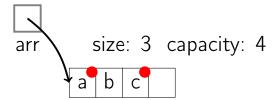


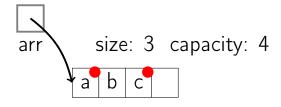


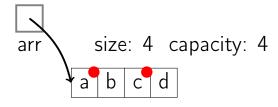


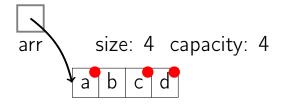


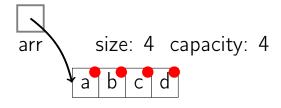




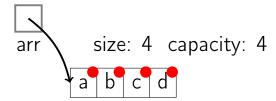


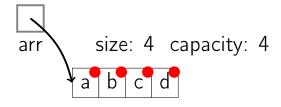




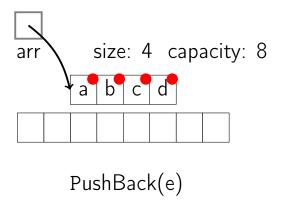


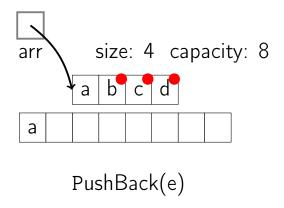
PushBack(d)

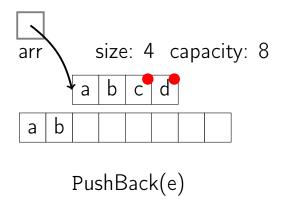


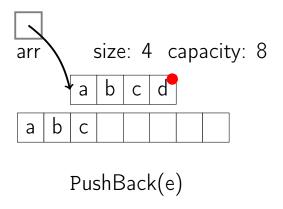


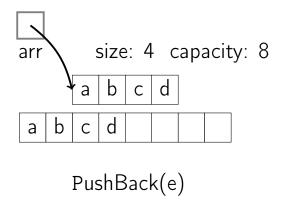
PushBack(e)

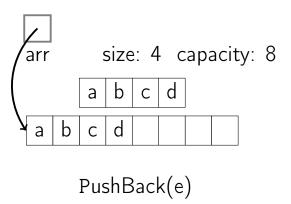


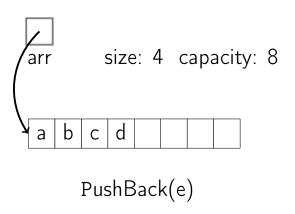


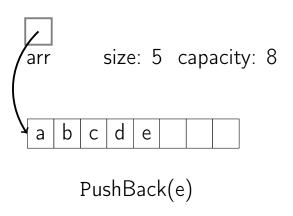


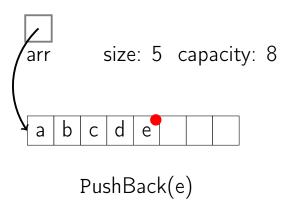


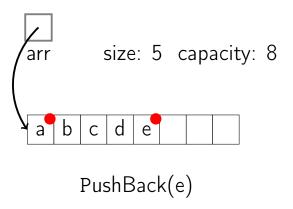


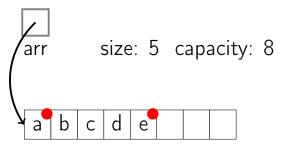


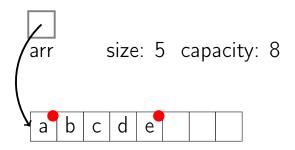












O(1) amortized cost for each PushBack

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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- lacktriangle if c_t is small, the potential increases
- if c_t is large, the potential decreases by the same scale

■ The cost of *n* operations is: $\sum_{i=1}^{n} c_i$

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

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

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

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

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

$$i=1$$

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

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 $f(h) = \frac{1}{h} \frac{1}{h}$

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

$$= \Phi(h_n) - \Phi(h_n) + \sum_{n=1}^{n} c_n > 0$$

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

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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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Without resize when adding element i

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$$= 3$$

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Amortized cost of adding element i:

=3

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

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

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.