

# Expanding Stacks & Queues

CS16: Introduction to Data Structures & Algorithms

Summer 2021

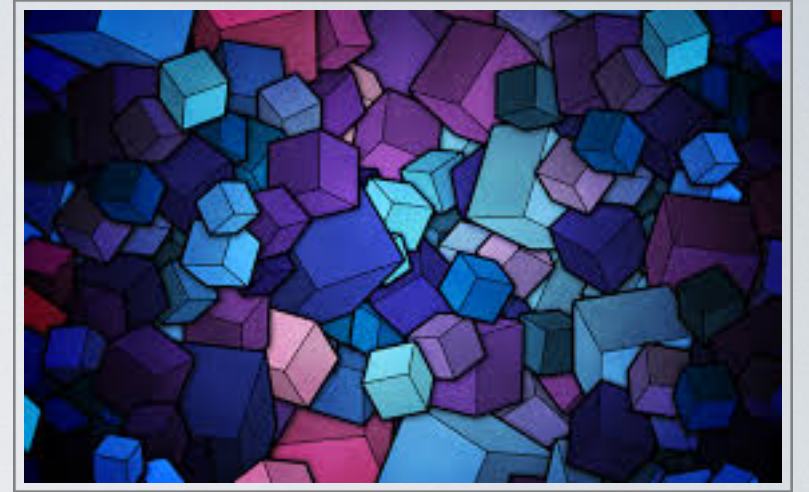
# Outline

- ▶ Abstract data types
- ▶ Stacks
  - ▶ Capped-capacity
  - ▶ Expandable
- ▶ Amortized analysis
- ▶ Queues
  - ▶ Expandable queues





# Abstract Data Types



- ▶ Abstraction of a data structure
- ▶ Specifies “functionality”
  - ▶ type of data stored
  - ▶ operations it can perform
- ▶ Like a Java interface
  - ▶ Specifies name & purpose of methods
  - ▶ But not implementations
- ▶ Think of lists: can have ArrayLists or LinkedLists

# Stacks

- ▶ Stores arbitrary objects
- ▶ Operations
  - ▶ **Push**: adds object
  - ▶ **Pop**: returns *last* object
  - ▶ LIFO: last-in first-out
- ▶ Can be implemented with
  - ▶ Linked lists, arrays, ...





# Stack ADT

- ▶ **push**(object)
  - ▶ inserts object
- ▶ object **pop**( )
  - ▶ returns and removes last inserted object
- ▶ **int size**( )
  - ▶ returns number objects in stack
- ▶ **boolean isEmpty**( )
  - ▶ returns **TRUE** if empty; **FALSE** otherwise



# Capped-Capacity Stack

- ▶ Array-based Stack
  - ▶ Stores objects in array
  - ▶ keeps pointer to last inserted object
- ▶ Problem?
  - ▶ Size of the stack is bounded by size of array :- (

# Capped-Capacity Stack

```
Stack( ):
    data = array of size 20
    count = 0
```

```
function size( ):
    return count
```

$O(1)$

```
function isEmpty( ):
    return count == 0
```

$O(1)$

```
function push(object):
    if count < 20:
        data[count] = object
        count++
    else:
        error("overfull")
```

$O(1)$

```
function pop( ):
    if count == 0:
        error("empty stack")
    else:
        count--
        return data[count]
```

$O(1)$



# Expandable Stack



- ▶ Capped-capacity stack is fast
  - ▶ but—what if we don't know how many items?
- ▶ How can we design an *uncapped* Stack?
- ▶ Remember—arrays can't be resized
  - ▶ “resize”—copy contents of size N array to size N' array



# Expandable Stack



- ▶ Strategy #1: **Incremental**
  - ▶ increase size of array by constant **e** when full
- ▶ Strategy #2: **Doubling**
  - ▶ double size of array when full

# Expandable Stack



```
Stack( ):
```

```
data = array of size 20
```

```
count = 0
```

```
capacity = 20
```

What is the runtime?

```
function push(object):
```

```
data[count] = object
```

```
count++
```

```
if count == capacity
```

```
    new_capacity = capacity + e /* incremental */
```

```
                = capacity * 2 /* doubling */
```

```
new_data = array of size new_capacity
```

```
for i = 0 to capacity - 1
```

```
    new_data[i] = data[i]
```

```
capacity = new_capacity
```

```
data = new_data
```



# Expandable Stack



```
function push(object):  
    data[count] = object  
    count++  
    if count == capacity  
        new_capacity = capacity + e /* incremental */  
                      = capacity * 2 /* doubling */  
        new_data = array of size new_capacity  
        for i = 0 to capacity - 1  
            new_data[i] = data[i]  
        capacity = new_capacity  
        data = new_data
```

- ▶ Runtime when not expanding is  $O(1)$  & runtime when expanding is  $O(n)$
- ▶ When does it expand?
  - ▶ after  $n$  pushes, where  $n$  is capacity of array

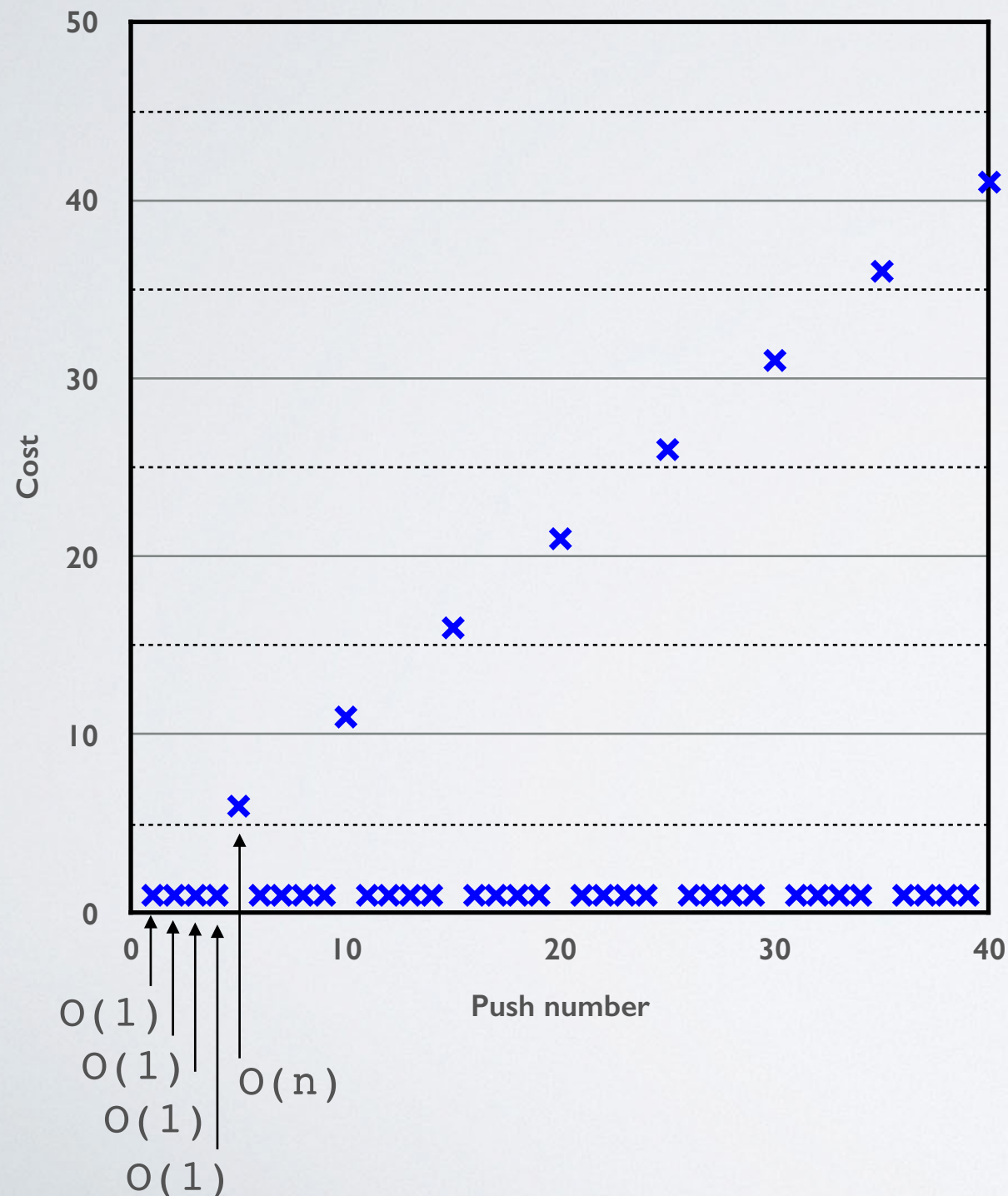
# Incremental & Doubling

- ▶ What are the *worst-case* runtimes?
  - ▶ incremental:  $O(n)$
  - ▶ doubling:  $O(n)$
- ▶ But are they really the same?

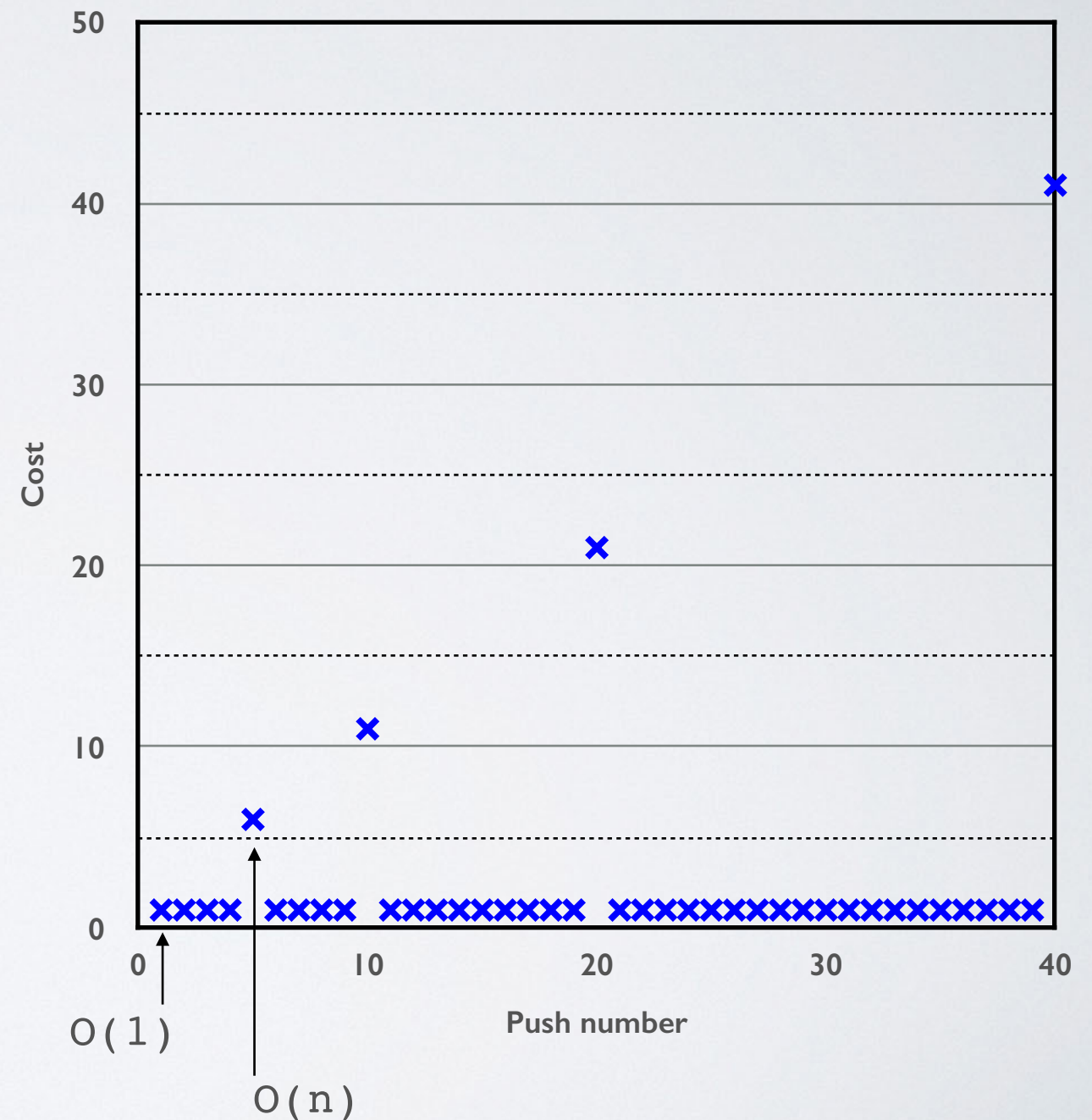


# Incremental & Doubling

## Incremental (5)



## Doubling



# Incremental & Doubling

- ▶ Worst-case analysis overestimates runtime
  - ▶ for algorithms that are fast *most* of time...
  - ▶ ...and slow *some* of the time
- ▶ For these algorithms we need an alternative
  - ▶ Amortized analysis!



**Measure cost on sequence of calls not a single call!**



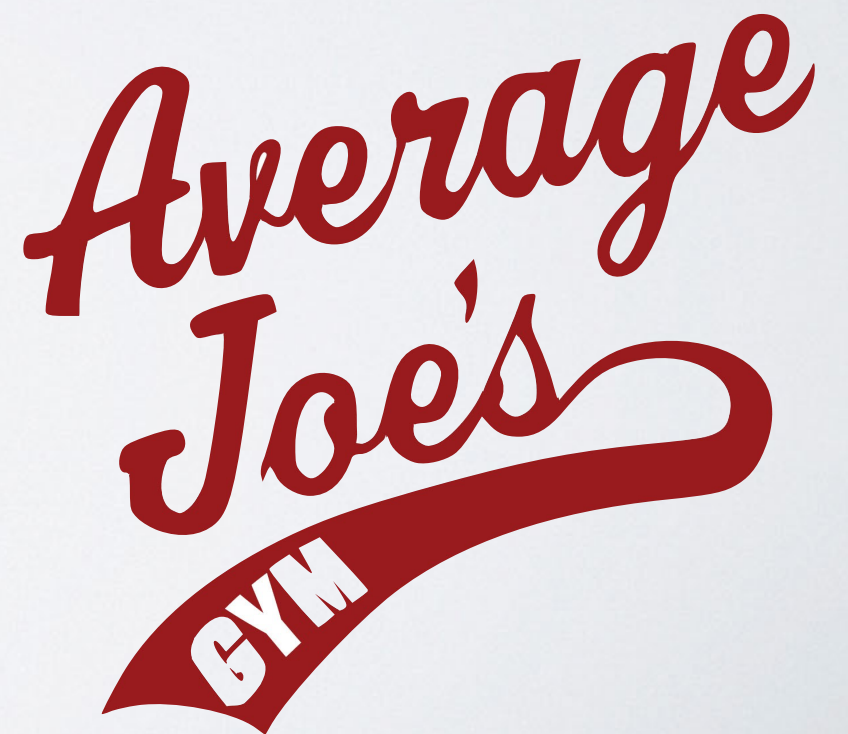
# Towards Amortized Analysis

- ▶ For certain algorithms it's better to measure
  - ▶ *total* running time on *sequence* of calls
  - ▶ instead of measuring on a single call
  - ▶  $S(n)$ : total #calls on *sequence* of  $n$  calls
  - ▶ **Not runtime on a single input of size  $n$**
- ▶ For a stack
  - ▶  $S(n)$ : cost push #1 + cost push #2 + ... + cost push # $n$

# Amortized Analysis

- ▶ Instead of reporting *total* cost of sequence
  - ▶ report cost of sequence *per call*

$$\frac{S(n)}{n}$$





# Amortized Analysis of Incremental

# Expandable Stack



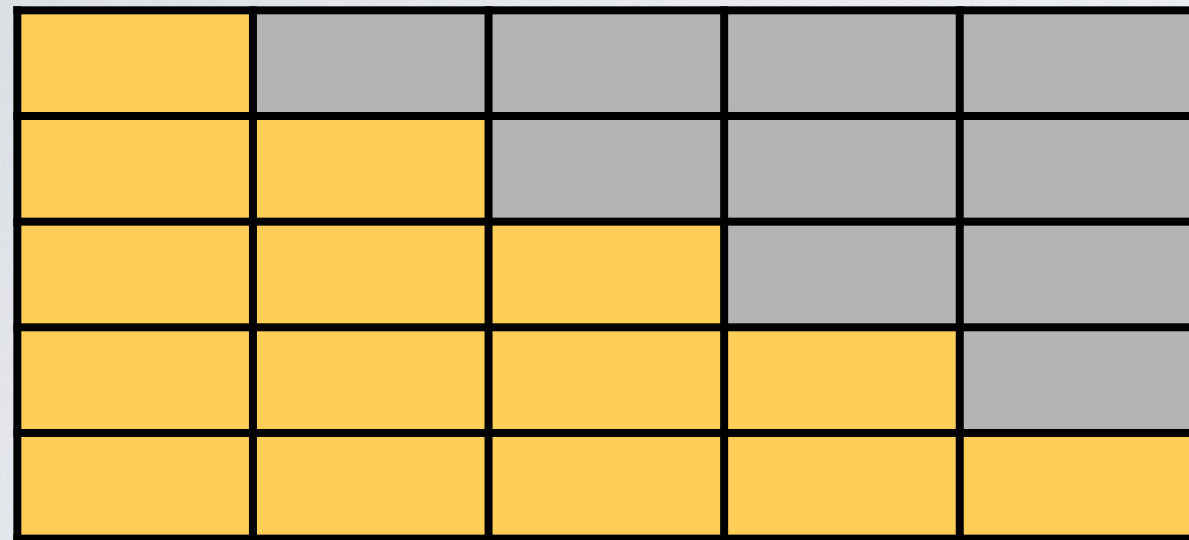
```
Stack( ) :  
    data = array of size 20  
    count = 0  
    capacity = 20
```

**Run time depends on  
count which depends on  
# of *previous* pushes**

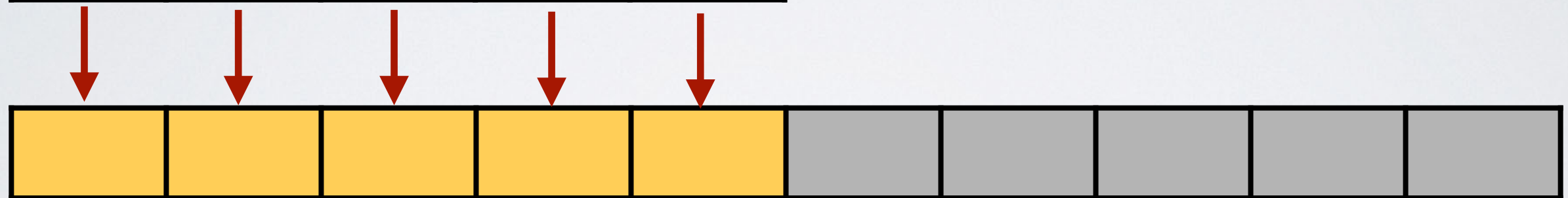
```
function push(object):  
    data[count] = object  
    count++  
    if count == capacity  
        new_capacity = capacity + e /* incremental */  
                       = capacity * 2 /* doubling */  
        new_data = array of size new_capacity  
        for i = 0 to capacity - 1  
            new_data[i] = data[i]  
        capacity = new_capacity  
        data = new_data
```

A black arrow originates from the text 'Run time depends on count which depends on # of previous pushes' and points to the condition 'count == capacity' in the push function code block.

# Amortized Analysis of Incremental



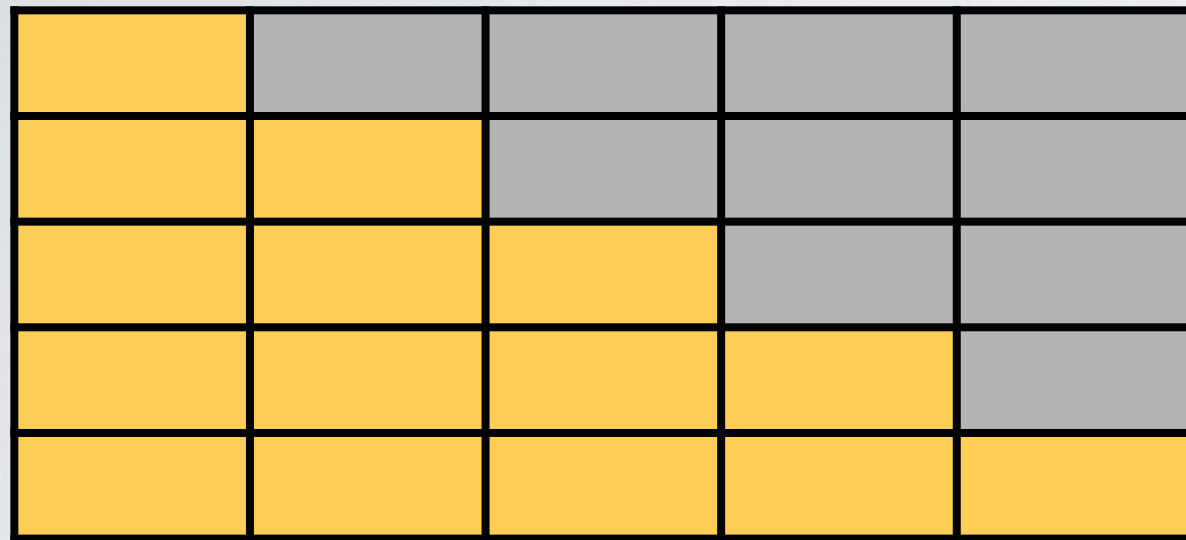
- ▶ Stack with start capacity  $c = 5$
- ▶ Expands by  $e = 5$
- ▶ push 5 items to stack



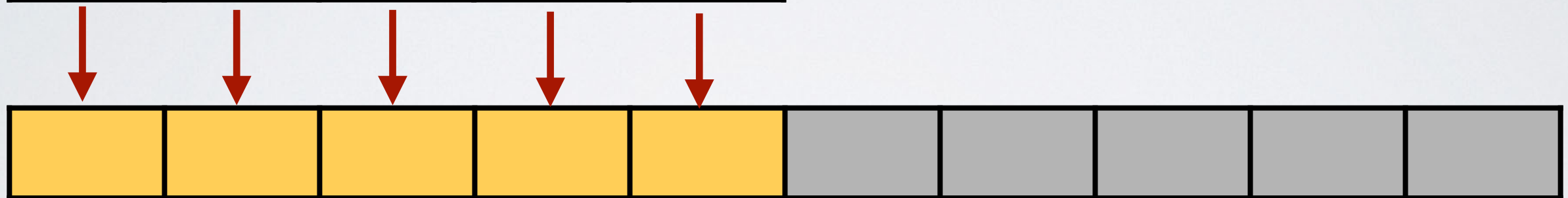
- ▶ 5th push brings to capacity
  - ▶ Objects copied to new array of size  $c+e = 5+5 = 10$
  - ▶ Cost per push over 5 pushes?



# Amortized Analysis of Incremental



- ▶ Stack with start capacity  $c = 5$
- ▶ Expands by  $e = 5$



$$\frac{S(n)}{n} = \frac{c + c}{c} = \frac{5 + 5}{5} = 2$$

**pushes** **expansion**

Is each push  
 $O(1)$ ?

# Amortized Analysis of Incremental

- ▶ What if we push 10 objects?

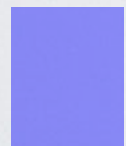
$$\frac{S(n)}{n} = \frac{\text{c} + \text{c} + \text{e} + (\text{c} + \text{e})}{n}$$

**1st batch of pushes**

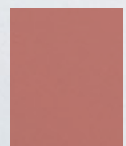
**1st expansion**

**2nd batch of pushes**

**2nd expansion**



pushes



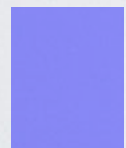
expansions

**c=5**  
**e=5**

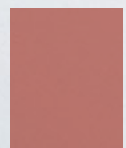
# Amortized Analysis of Incremental

- ▶ What if we push **10** objects?

$$\begin{aligned}\frac{S(n)}{n} &= \frac{\text{c} + \text{c} + \text{e} + (\text{c} + \text{e})}{10} \\ &= \frac{\text{c} + \text{e} + \text{c} + (\text{c} + \text{e})}{10} \\ &= \frac{10 + 5 + (5 + 5)}{10} \\ &= 2.5\end{aligned}$$



pushes



expansions

c=5  
e=5

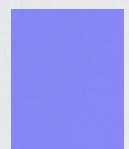


# Amortized Analysis of Incremental

$$\frac{S(10)}{10} = \frac{\boxed{c+e} + \boxed{c} + \boxed{(c+e)}}{10} = \frac{10 + 5 + 10}{10} = \frac{25}{10} = 2.5$$

$$\frac{S(15)}{15} = \frac{\boxed{c+e+e} + \boxed{c} + \boxed{(c+e)} + \boxed{(c+e+e)}}{15} = \frac{15 + 5 + 10 + 15}{15} = \frac{45}{15} = 3$$

$$\frac{S(20)}{20} = ?$$



pushes



expansions

$c=5$   
 $e=5$

# Amortized Analysis of Incremental

$$\begin{aligned} S(n) &= c + e + \dots + e + c + (c + e) + (c + 2e) + (c + 3e) + \dots \\ &= n + c + (c + e) + (c + 2e) + (c + 3e) + \dots \end{aligned}$$

To make things simpler, let's assume  
 $e = c$

$$= n + c + 2c + 3c + 4c + \dots + \frac{n}{c} \cdot c$$

**# of expansions**  
**(1 expansion per  $c$  (or  $e$ ) pushes)**



# Amortized Analysis of Incremental

**n pushes**

**cost of exp. # n/c**

$$S(n) = n + c + 2c + 3c + \dots + \frac{n}{c} \cdot c$$

$$= n + c \cdot \left(1 + 2 + \dots + \frac{n}{c}\right) \quad \leftarrow \text{--- factoring out } c$$

$$= n + c \cdot \frac{1}{2} \cdot \left(\frac{n}{c} \left(\frac{n}{c} + 1\right)\right) \quad \leftarrow \text{--- using:}$$

$$(1 + 2 + \dots + k) = \frac{k \cdot (k + 1)}{2}$$

$$= n + \frac{n^2/c + n}{2} \quad \leftarrow \text{--- distributing \& simplifying:}$$

$$= O(n^2)$$

$$\frac{S(n)}{n} = O(n)$$



# Amortized Analysis of Incremental

- ▶ Summary

- ▶ Total cost of  $n$  pushes:  $S(n) = O(n^2)$

- ▶ Amortized cost of  $n$  pushes:  $S(n)/n = O(n)$

# Amortized Analysis of Double

# Amortized Analysis of Doubling

- ▶ Doubling stack with initial capacity  $c=5$ ?

$$\begin{aligned} \frac{S(n)}{n} &= \frac{S(5)}{5} = \frac{5 + 5}{5} = 2 && \begin{array}{l} \text{cost of pushes} \\ \text{cost of exp} \end{array} \\ \frac{S(n)}{n} &= \frac{S(10)}{10} = \frac{10 + 5 + 10}{10} = 2.5 && \begin{array}{l} \text{cost of exp} \\ \text{\#2} \end{array} \\ \frac{S(n)}{n} &= \frac{S(20)}{20} = \frac{20 + 5 + 10 + 20}{20} = 2.75 && \begin{array}{l} \text{cost of exp} \\ \text{\#3} \end{array} \end{aligned}$$



# Amortized Analysis of Doubling

**cost of n pushes**      **cost of last exp**      **cost of second to last exp**

$$S(n) = n + n + \frac{n}{2} + \frac{n}{4} + \cdots + \frac{n}{2^{k-1}}$$

**cost of exp #1**

$$= n + n \cdot \left( 1 + \frac{1}{2} + \frac{1}{4} + \cdots + \frac{1}{2^{k-1}} \right)$$

$$< n + n \cdot 2$$

**using:**

$$= 3n$$

$\lim_{k \rightarrow \infty} \sum_{i=0}^k \frac{1}{2^i} = 2$

Assume:  
 $c=2$   
 $n=2^k$

$$\frac{S(n)}{n} = O(1)$$

# Amortized Analysis

- ▶ Summary for Incremental
  - ▶ Total cost of  $n$  pushes:  $S(n) = O(n^2)$
  - ▶ Amortized cost of  $n$  pushes:  $S(n)/n = O(n)$
- ▶ Summary for Doubling
  - ▶ Total cost of  $n$  pushes:  $S(n) = O(n)$
  - ▶ Amortized cost of  $n$  pushes:  $S(n)/n = O(1)$

# Amortized Analysis

- ▶ Summary for Incremental
  - ▶ Total cost of  $n$  pushes:  $S(n) = O(n^2)$
  - ▶ Amortized cost of  $n$  pushes:  $S(n)/n = O(n)$
- ▶ Summary for Doubling
  - ▶ Total cost of  $n$  pushes:  $S(n) = O(n)$
  - ▶ Amortized cost of  $n$  pushes:  $S(n)/n = O(1)$
- ▶ In practice: always use doubling



# How do we feel about amortized analysis?

- ▶ Situations where worst case is most important?

# Expandable Queue



# Queue ADT

- ▶ **enqueue(object):**
  - ▶ inserts object
- ▶ **object dequeue( )**
  - ▶ returns and removes first inserted object
- ▶ **int size( )**
  - ▶ returns number objects in queue
- ▶ **boolean isEmpty( )**
  - ▶ returns **TRUE** if empty; **FALSE** otherwise





# Expandable Queue



↑  
**head**  
**tail**

- ▶ Can be implemented with expandable array
  - ▶ need to keep track of head and tail

# Expandable Queue



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# Expandable Queue



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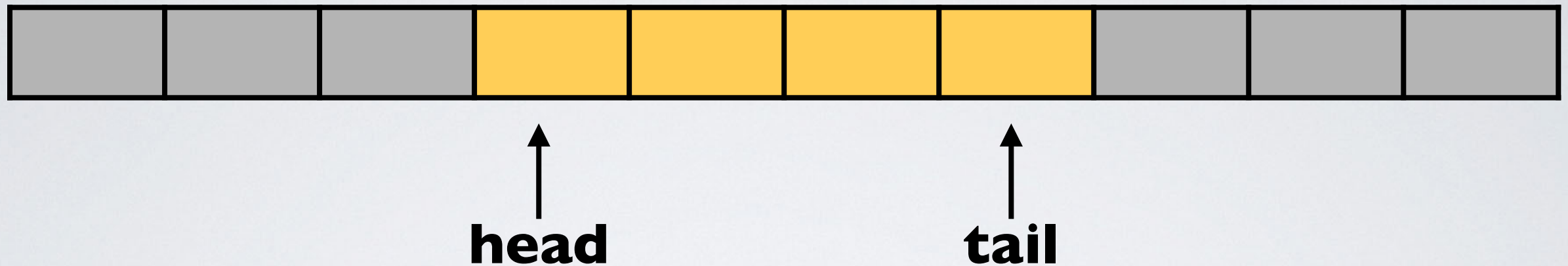


# Expandable Queue



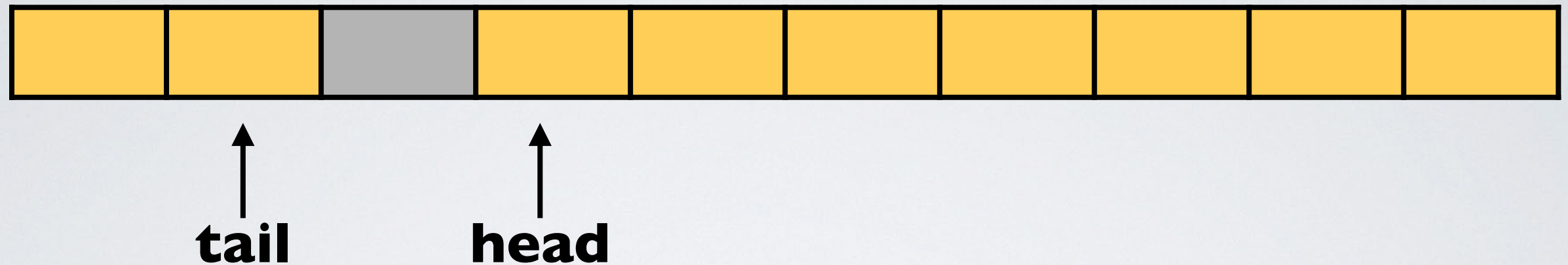
- ▶ Can be implemented with expandable array
  - ▶ need to keep track of head and tail

# Expandable Queue



- ▶ Can be implemented with expandable array
  - ▶ need to keep track of head and tail
- ▶ What happens when tail reaches end?
  - ▶ Is the queue full?
- ▶ So when should we expand array?

# Expandable Queue



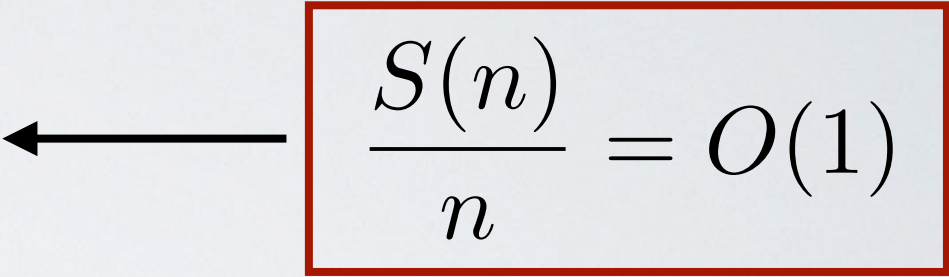
- ▶ Wrap around until array is completely full
- ▶ When expanding re-order objects properly



# Expandable Queue

```
function enqueue(object):  
    if size == capacity  
        double array and copy contents  
        reset head and tail pointers  
    data[tail] = object  
    tail = (tail + 1) % capacity  
    size++
```

```
function dequeue( ):  
    if size == 0  
        error("queue empty")  
    element = data[head]  
    head = (head + 1) % capacity  
    size--  
    return element
```


$$\frac{S'(n)}{n} = O(1)$$