

“The main purpose of deliberate practice is to develop effective mental representations, and, as we will discuss shortly, mental representations in turn play a key role in deliberate practice. The key change that occurs in our adaptable brains in response to deliberate practice is the development of better mental representations, which in turn open up new possibilities for improved performance.”

-- Peak: Secrets from the New Science of Expertise by Ericsson and Pool

CSCI 136:
Data Structures
and
Advanced Programming

Lecture 10

Recursion, part 2

Instructor: Kelly Shaw

[Williams](#)

Topics

- Recursion costs
- Mathematical Induction

Your to-dos

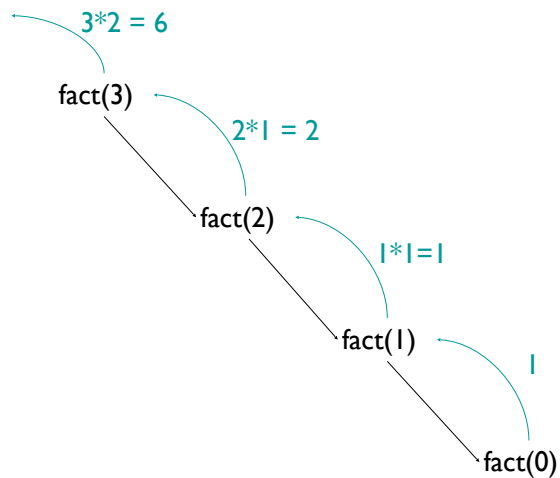
1. Lab 3, **due Tuesday 10/4 by 10pm**
2. Read **before Mon**: Bailey, Ch 9.4–9.5.
3. Weekly quiz, **due Saturday 10/1 at 12pm**

Recall: Factorial

- $n! = n \times (n-1) \times (n-2) \times \dots \times 1$

How much does a **recursive** solution **cost**?

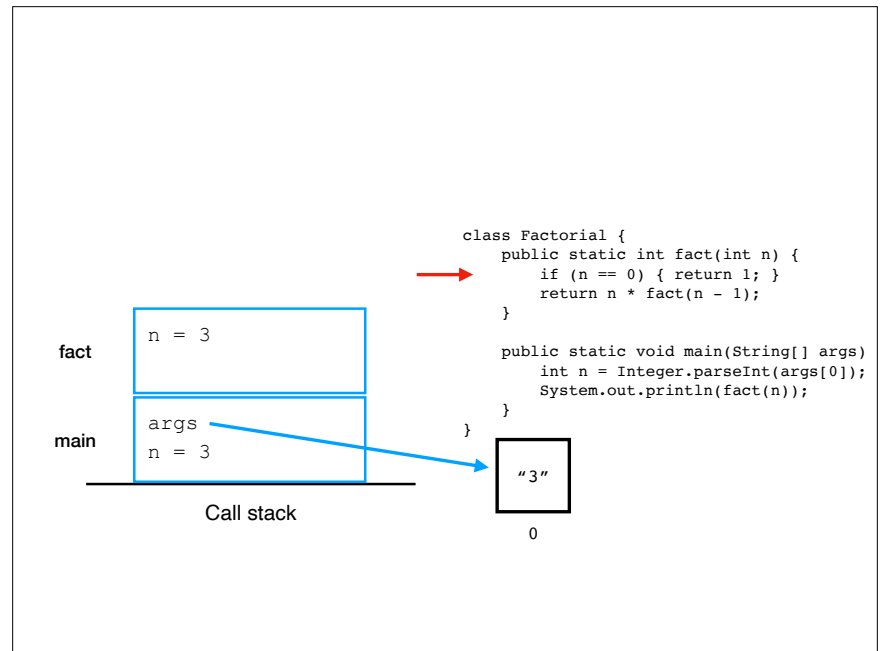
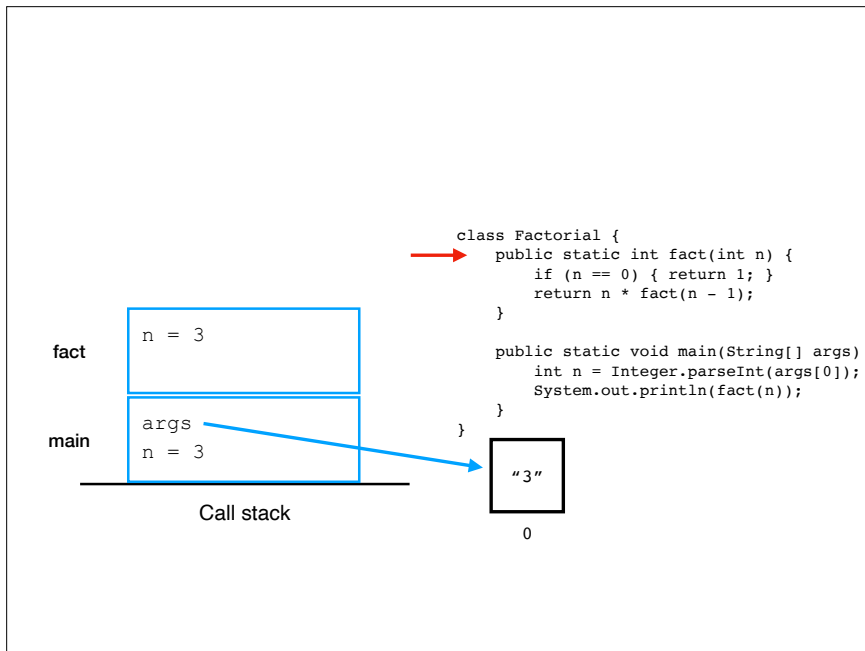
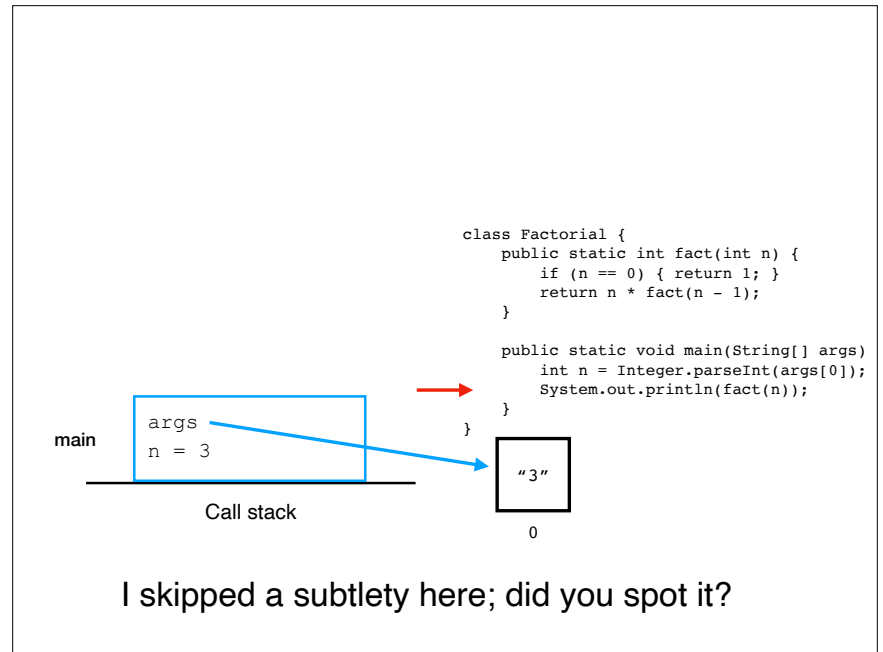
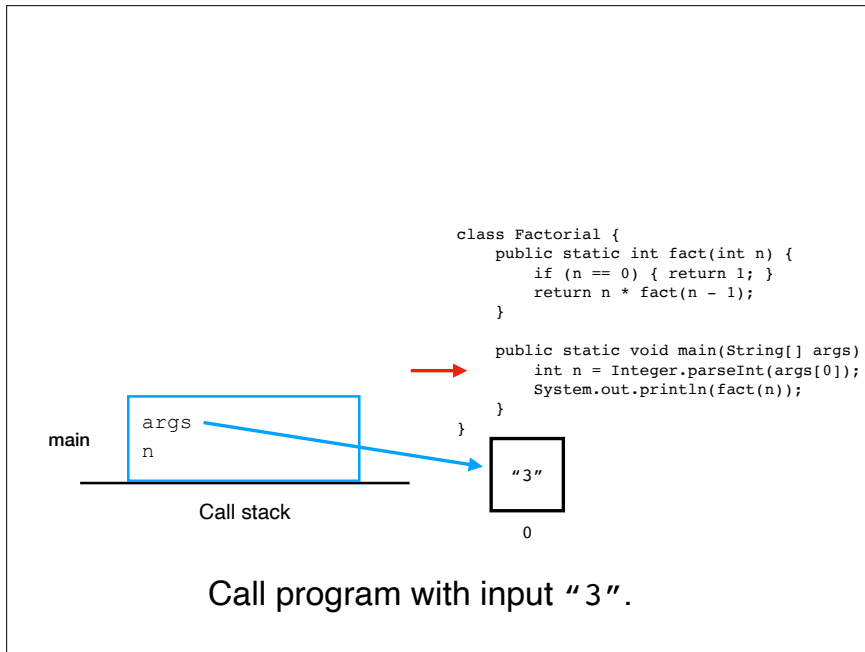
Graphically...

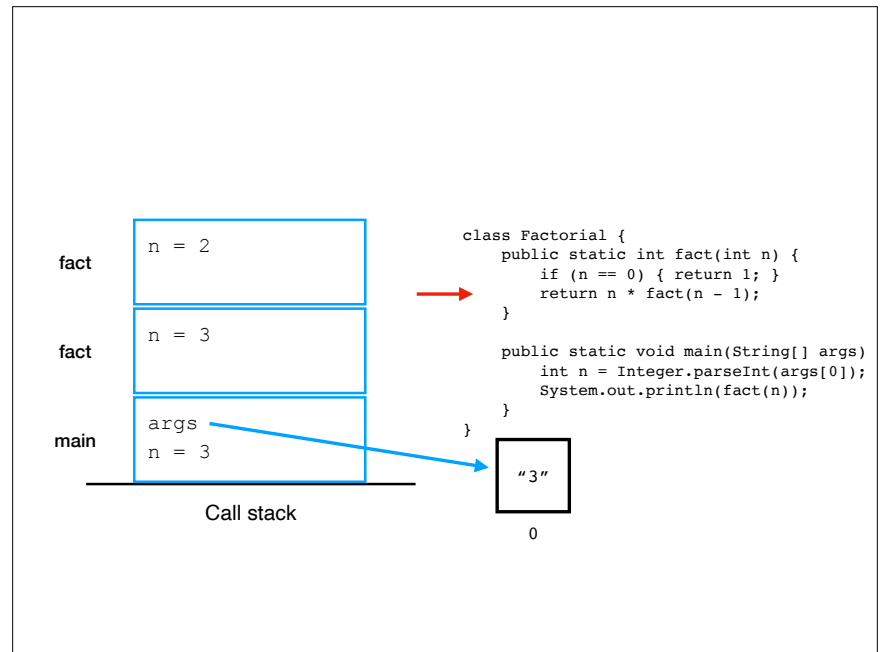
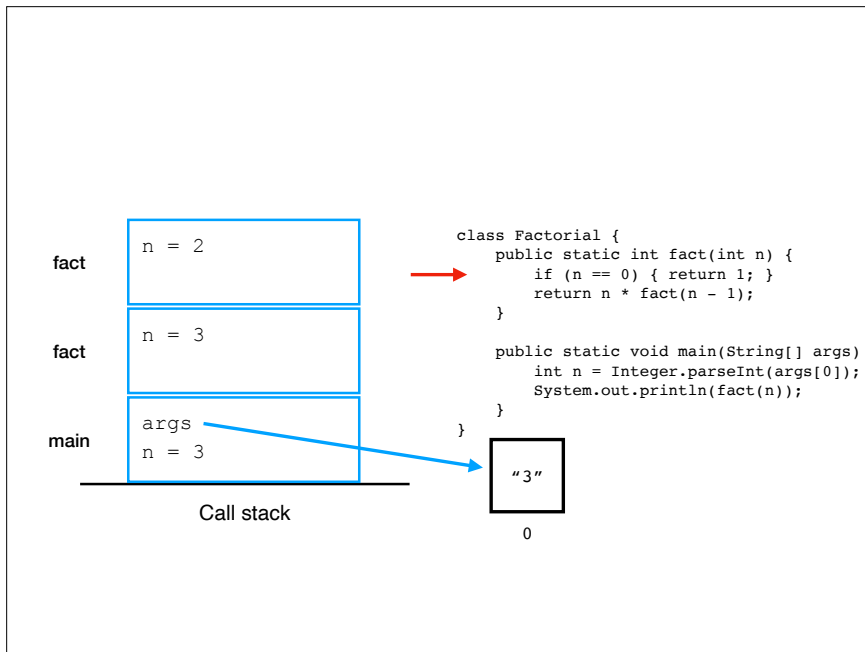
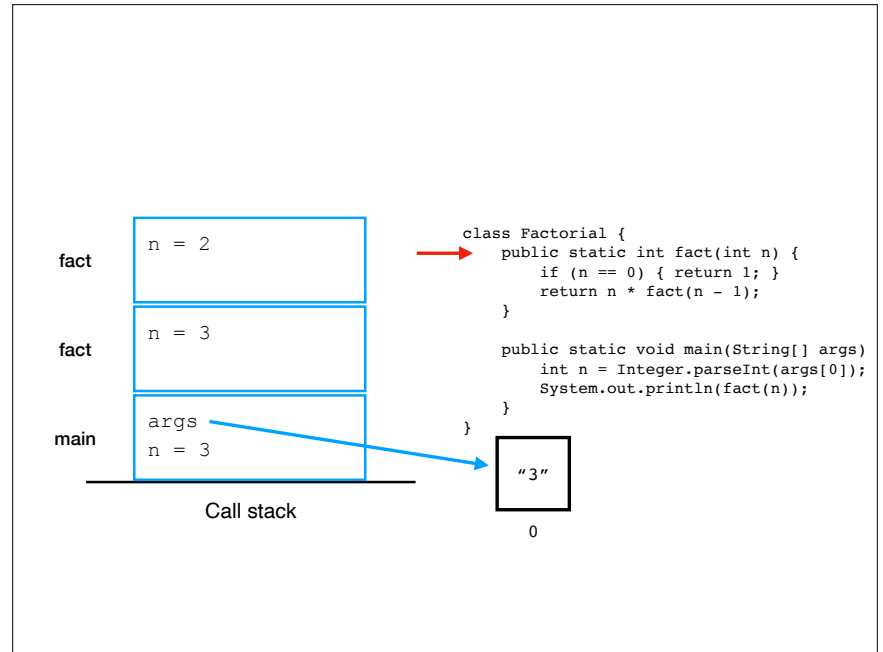
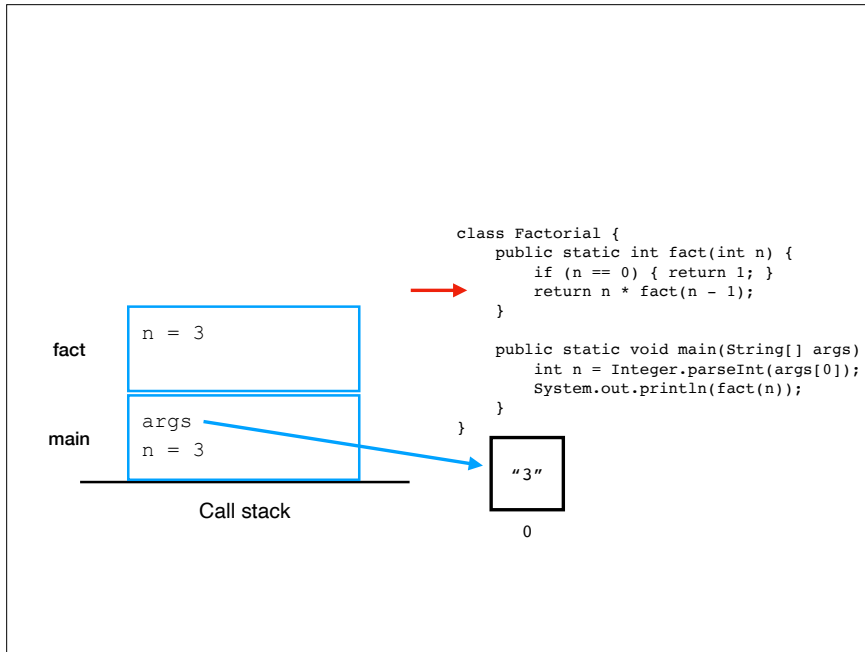


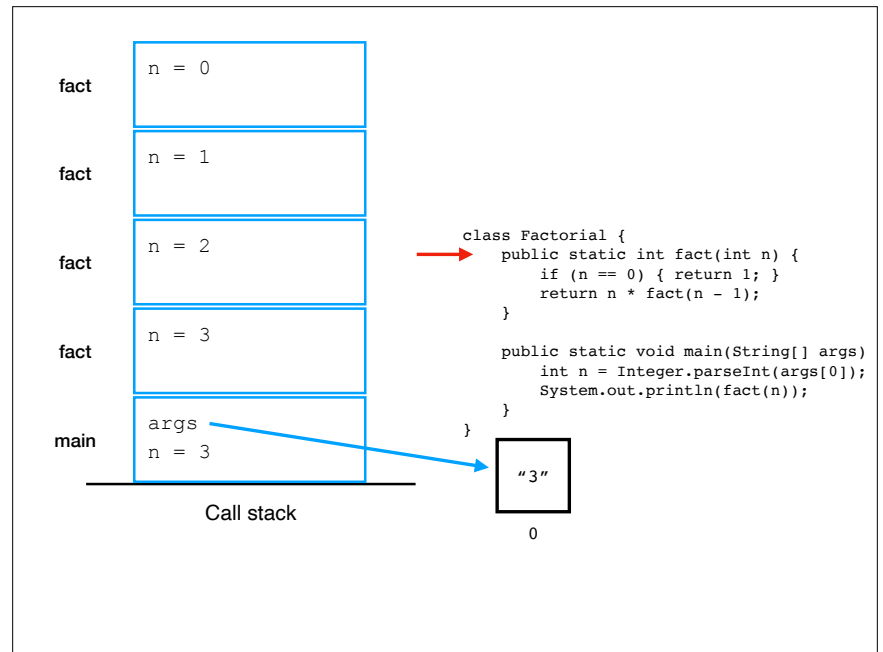
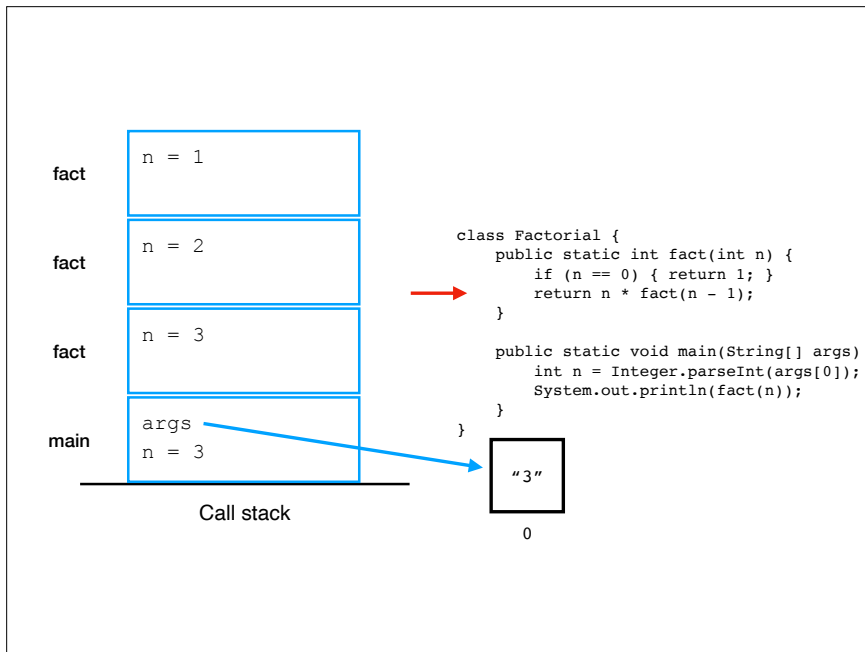
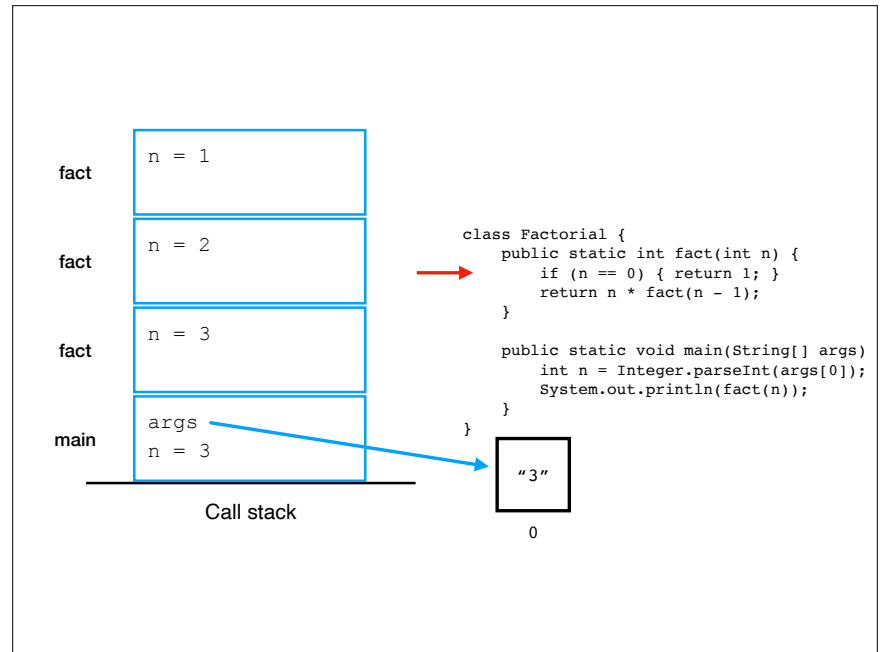
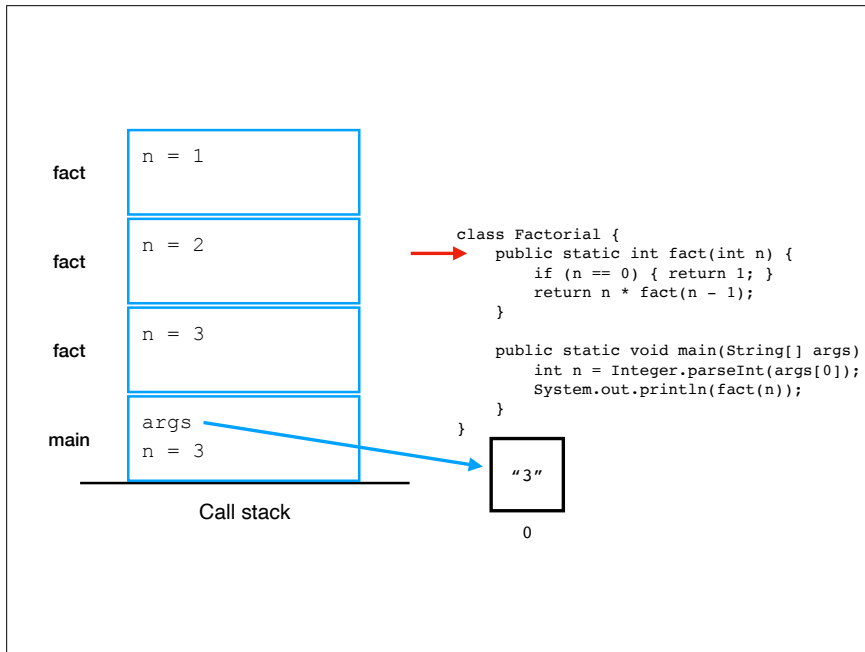
```
class Factorial {  
    public static int fact(int n) {  
        if (n == 0) { return 1; }  
        return n * fact(n - 1);  
    }  
    → public static void main(String[] args)  
        int n = Integer.parseInt(args[0]);  
        System.out.println(fact(n));  
    }  
}
```

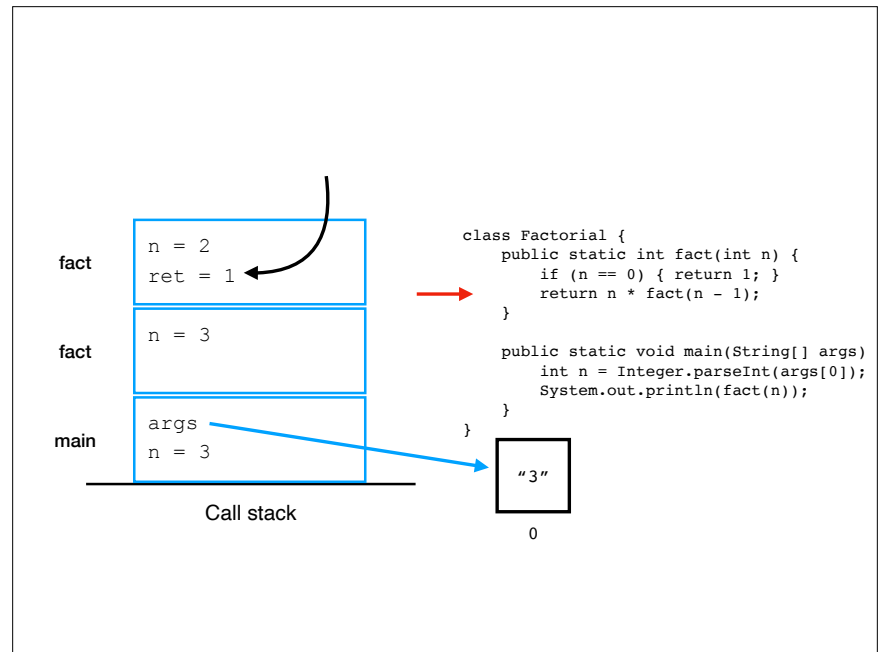
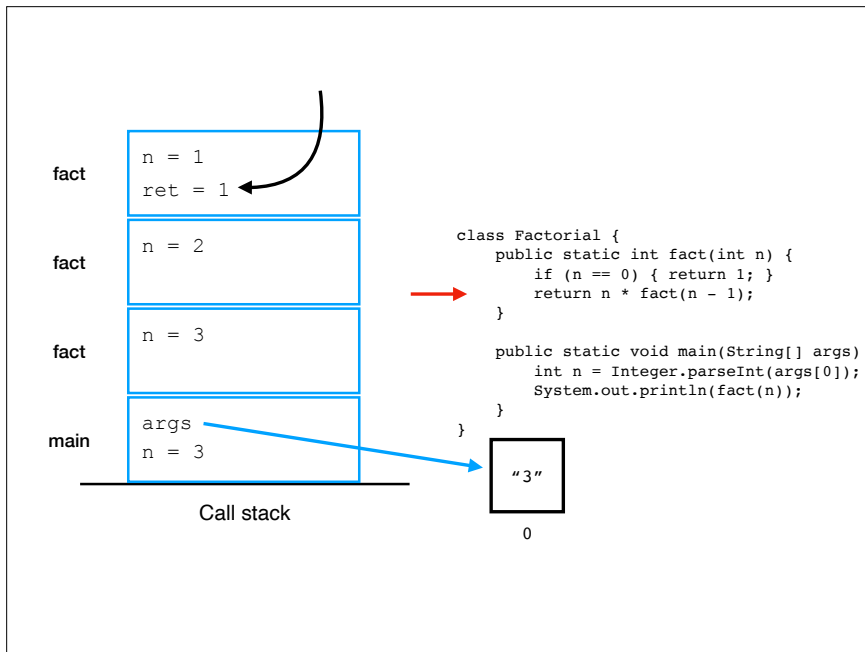
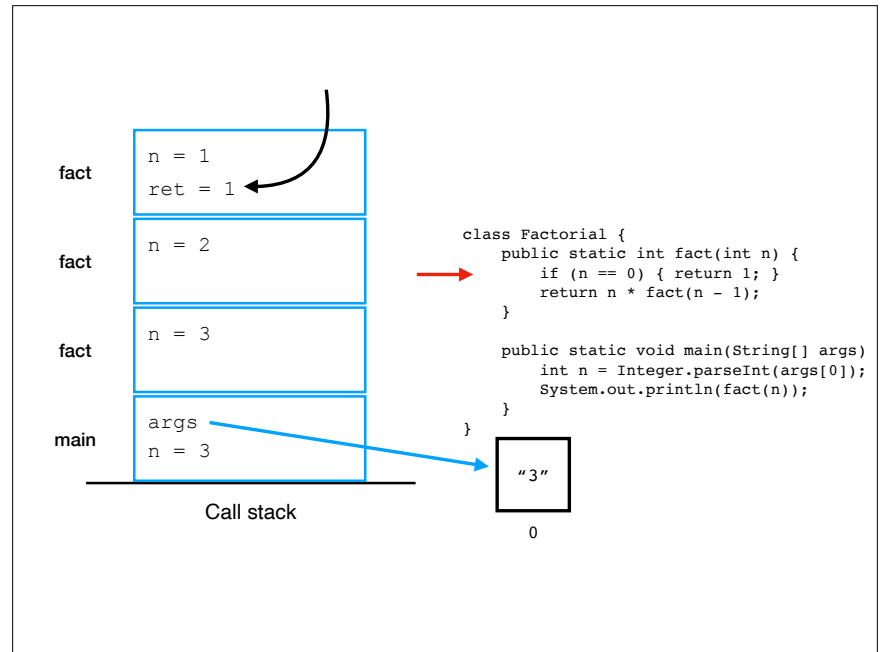
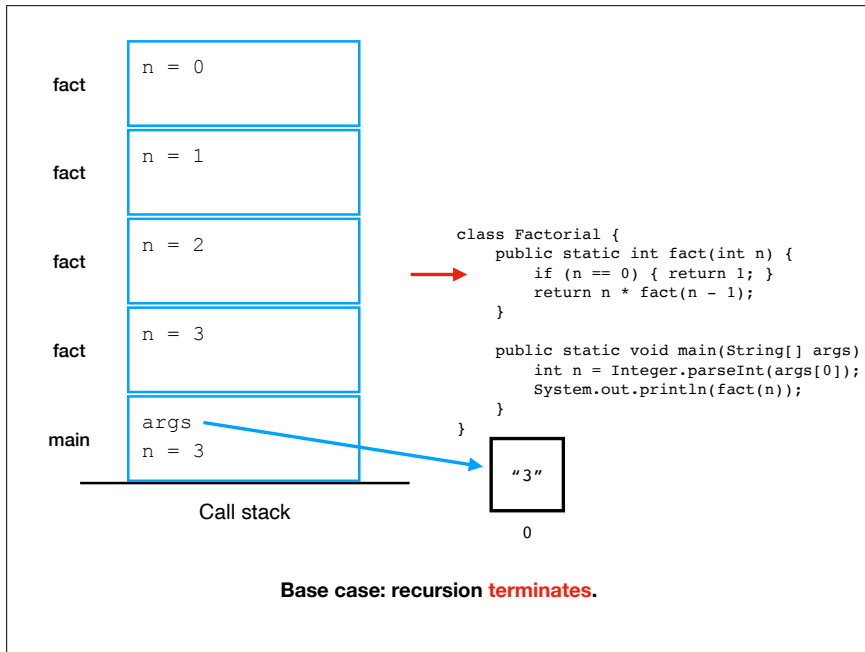
Call stack

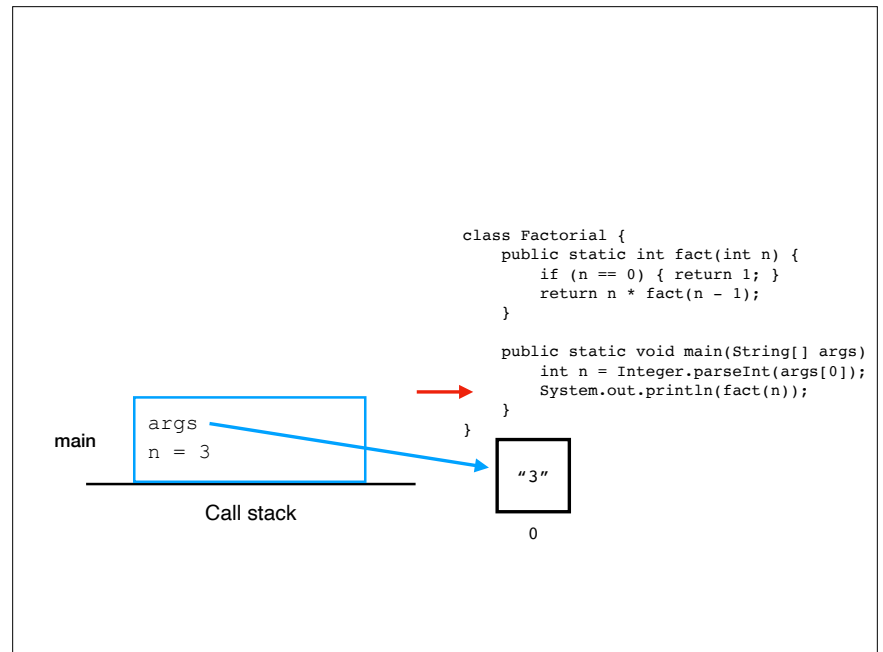
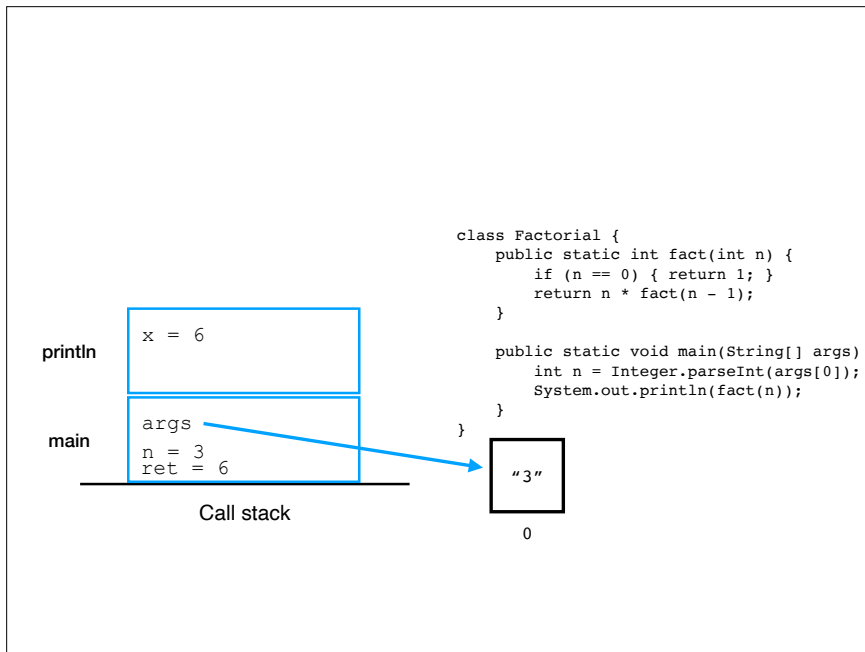
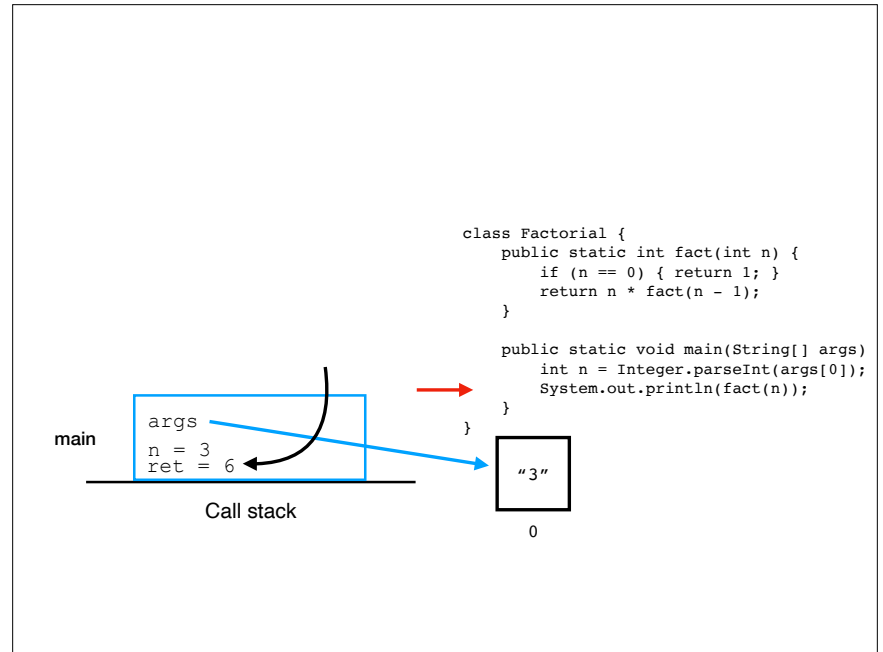
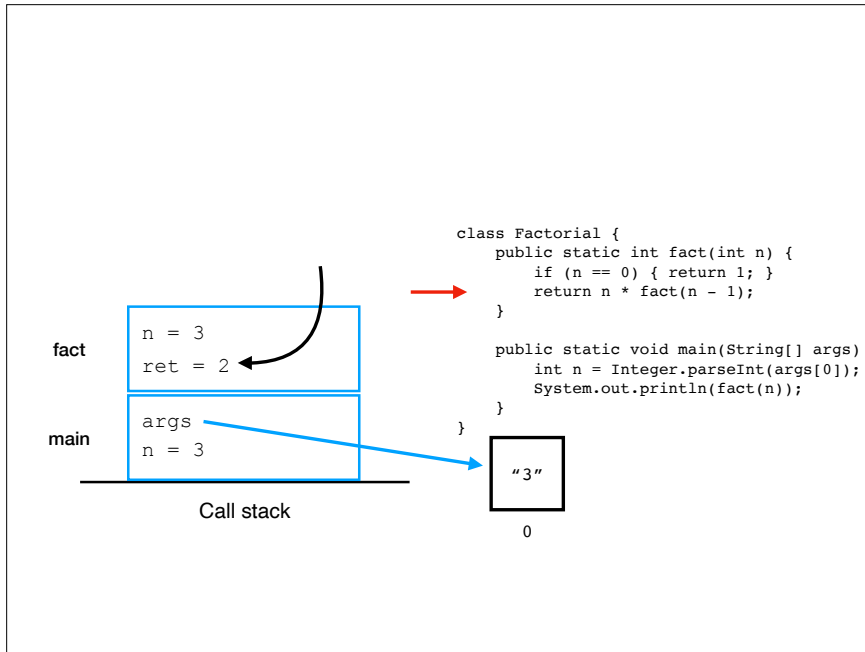
Call program with input "3".





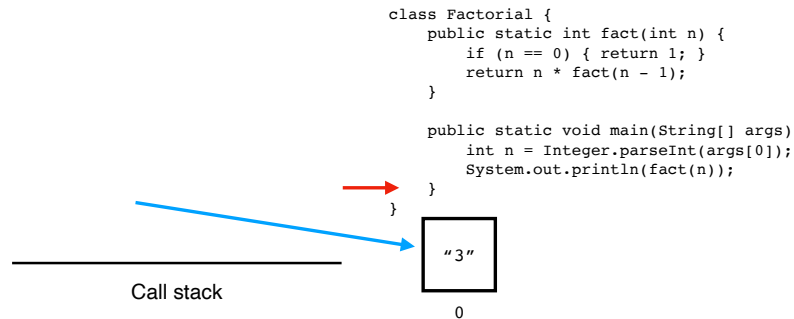






Recursion tradeoffs

- Advantages
 - Often **easier** to construct recursive solution than a loop
 - Code is usually **clearer**
 - Some problems do not have **obvious** non-recursive solutions
- Disadvantages
 - **Time cost** of recursive calls
 - **Memory cost** (need to store state for each recursive call until base case is reached)



Mathematical Induction



A note about “formal methods”



If the problem “fits” the mold, there is a procedure for determining truth.

Mathematical Induction

- The **mathematical cousin** of **recursion** is **induction**
- Induction is a **proof technique**
- Purpose: to **simultaneously prove** an **infinite number** of theorems!

Principle of Mathematical Induction

Let $P(n)$ be a **predicate** that is defined for **integers** n , and let a be a **fixed integer**.

If the following two statements are **true**:

1. $P(a)$ is **true**.
2. For all integers $k \geq a$, **if** $P(k)$ is **true** **then** $P(k + 1)$ is **true**.

then the statement

for all integers $n \geq a$, $P(n)$ is **true**

is **also true**.

Principle of Mathematical Induction (variant)

Let $P(n)$ be a **predicate** that is defined for **integers** n , and let a be a **fixed integer**.

If the following two statements are **true**:

1. $P(a)$ is **true**.
2. For all integers $k > a$, **if** $P(k-1)$ is **true** **then** $P(k)$ is **true**.

then the statement

for all integers $n \geq a$, $P(n)$ is **true**

is **also true**.

To be clear:

If you want to prove that $P(n)$ is **true** for all integers $n \geq a$,

1. You must first prove that $P(a)$ is **true**.
2. Then you must prove that:

For all integers $k \geq a$, **if** $P(k)$ is **true** **then** $P(k+1)$ is **true**.

Critically, when proving #2, **assume** that $P(k)$ is **true** and **show** that $P(k+1)$ **must also be true**.

Names for things and “form”

Hypothesis: $P(n)$ is true for all integers $n \geq a$,

1. Base case: $P(a)$ is true.

2. Inductive step:

For all integers $k \geq a$, if $P(k)$ is true then $P(k+1)$ is true.

Like recursion, there is an analogy



Like recursion, there is an analogy



Example

Prove that the sum of the first n integers is:

$$\frac{n(n+1)}{2}$$

Example

Put another way, prove

$$P(n) : 1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$$

for all $n \geq 1$.

We have an unbounded number of hypotheses (“for all $n \geq 1$ ”).

Use **mathematical induction**.

Remember the template!

Step 1: Prove **$P(a)$**

Step 2: Prove **$P(k) \Rightarrow P(k+1)$**

Therefore,

$$P(n) : 1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$$

For all $n \geq 1$.

Is **true**.

Example

Step 1: Prove **$P(a)$**

What would a good **a** be?

$$P(n) : 1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$$

The “simplest” instance is **$a = 1$** . Let’s start there.

Example

Step 1: Prove **$P(a)$**

$$P(a) : 1 = \frac{1(1+1)}{2}$$

Is this statement true? **Yes**.

$$\text{Proof: } \frac{1(1+1)}{2} = \frac{2}{2} = 1$$

Example

Step 2: Prove $P(k) \Rightarrow P(k+1)$

Assume the following is true:

$$P(k) : 1 + 2 + 3 + \dots + k = \frac{k(k+1)}{2}$$

Prove that $P(k)$ implies:

$$P(k+1) : 1 + 2 + 3 + \dots + (k + 1) = \frac{(k+1)((k+1)+1)}{2}$$

Example

Step 2: Prove $P(k) \Rightarrow P(k+1)$

$$P(k+1) : 1 + 2 + 3 + \dots + (k + 1) = \frac{(k+1)((k+1)+1)}{2}$$

Let's handle the left side first.

$$1 + 2 + 3 + \dots + (k + 1)$$

Looks familiar. Isn't it the same as:

$$(1 + 2 + 3 + \dots + k) + (k + 1)$$

Example

Step 2: Prove $P(k) \Rightarrow P(k+1)$

$$(1 + 2 + 3 + \dots + k) + (k + 1)$$

According to $P(k)$, which is true,
it must be equal to:

$$(1 + 2 + 3 + \dots + k) + (k + 1) = \frac{k(k+1)}{2} + (k + 1)$$

Example

Step 2: Prove $P(k) \Rightarrow P(k+1)$

$$\text{Simplify} \quad = \frac{k(k+1)}{2} + (k + 1)$$

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

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

Let's stop here.
The left side is

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

Example

Step 2: Prove $P(k) \Rightarrow P(k+1)$

$$P(k+1) : 1 + 2 + 3 + \dots + (k + 1) = \frac{(k+1)((k+1)+1)}{2}$$

Let's handle the right side now.

$$\frac{(k+1)((k+1)+1)}{2}$$

Simplify

$$\frac{(k+1)(k+2)}{2} \quad \text{Let's stop here.}$$

Example

Step 2: Prove $P(k) \Rightarrow P(k+1)$

$$P(k+1) : 1 + 2 + 3 + \dots + (k + 1) = \frac{(k+1)((k+1)+1)}{2}$$

We just showed that the left side

$$\frac{(k+1)(k+2)}{2}$$

equals the right side

$$\frac{(k+1)(k+2)}{2}$$

Example

Step 1: Prove $P(a)$ 

Step 2: Prove $P(k) \Rightarrow P(k+1)$ 

Therefore,

$$P(n) : 1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$$

For all $n \geq 1$.

Is true. 

Recap & Next Class

Today:

- Recursion costs
- Mathematical induction

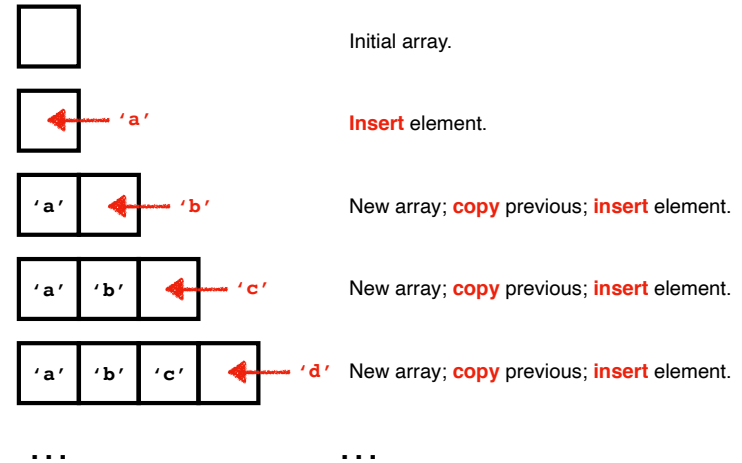
Next class:

- Vector doubling
- ADTs
- Lists

Expanding vectors: why double?

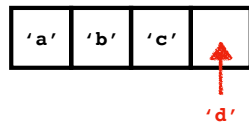
Why is the **array doubling** strategy for Vector **better** than expanding the array **one element at a time**?

One-at-a-time expansion



Insertion into an array

How much does **array insertion** cost?



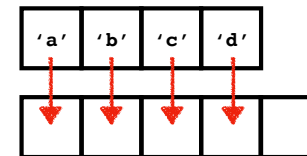
It costs **$O(1)$** .

In fact, lookup and insertion both cost **$O(1)$** .

Tradeoff: arrays are fixed size.

Copying an array

How much does an **array copy** cost?



It costs **$O(1) \times m$** , where **m** is the size of the original array.

$\approx O(m)$

One-at-a-time expansion costs?

(in the worst case, each time)



Initial array.



Insert element.



New array; **copy** previous; **insert** element.

$O(m) + O(1) \approx O(m)$, where **m** is the size of the original array.

Cost is **dominated by the size of the array** being copied.

How many copies?

of copies for one-at-a-time expansion:

	1	+	2	+	3	+	...	+	(n-1)
add()	2nd		3rd		4th				nth
	elem.		elem.		elem.				elem.

Recall theorem: $1 + 2 + 3 + \dots + k = k(k+1)/2$

Sub $n-1$ for k : $(n-1)((n-1)+1)/2 = n(n-1)/2$

$$= (n^2 - n) / 2$$

One-at-a-time expansion costs $\approx O(n^2)$