

**CSCI 136**  
**Data Structures &**  
**Advanced Programming**

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**Lecture 7**

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# Announcements

- Questions about Lab 2?
- Lab 3 will be handed out today
  - Lots of thinking...little typing
  - Problems can be done in any order!
  - Recursion can be frustrating...be patient!

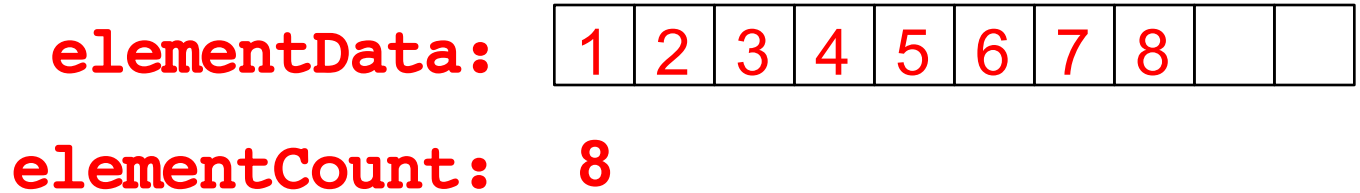
# Last Time

- Generics
- Started implementing vectors
  - `add(int index, E element)`, `remove(int index)`:
    - Require *shifting*
  - `add(int index, E element)`, `add(E element)`:
    - Vectors must *\*grow\** as we add more elements
    - How to expand a Vector's internal array?

# Today's Outline

- Wrap up Vectors
- Learn about Big-O analysis
- Briefly discuss recursion
  - More next class, but quick recursion review for lab this week...

# add(), remove()



Vector v

v.add(0,0);

v.remove(3);

Copy from  
right to left!

Copy from  
left to right!

# Growing Vectors

```
Vector v = new Vector(1);  
v.add(0);  
for (int i = 0; i < n; i++) {  
    v.add(i);  
}
```

**Grow to meet  
new capacity.**

**Double the current  
capacity.**

# Growing Vectors

- Two ways to grow when adding  $n$  new elements to Vector:
  - Additive increase (add some constant factor)
    - Requires  $\sim n^2/2$  operations (or copies)
  - Multiplicative increase (double)
    - Requires  $\sim n$  operations
  - Which is better?
  - Is there a tradeoff?

# Vectors

- These questions relate to the **time and space tradeoff**
  - We could just as easily avoid all copy operations by making a huge Vector/array initially...
  - ...but this wastes space and is inefficient



# Shrinking the Array

- When should we shrink the array in Vector implementation?
  - When 1/2 full?
  - When 1/4 full?
- We shrink when 1/4 full...
- Can get bad performance if array size changes too frequently

# Vector Constructors

```
protected Object elementData[]; // the data  
protected int elementCount; // number of elements in vector
```

```
public Vector() {  
    this(10);  
}
```

```
public Vector(int initialCapacity) {  
    elementData = new Object[initialCapacity];  
    elementCount = 0;  
  
}
```

# Vector Constructors

```
protected Object elementData[]; // the data
protected int elementCount; // number of elements in vector
protected int capacityIncrement; // the rate of growth for vector

public Vector() {
    this(10);
}

public Vector(int initialCapacity) {
    elementData = new Object[initialCapacity];
    elementCount = 0;
    capacityIncrement = 0;
}

// pre: initialCapacity >= 0, capacityIncr >= 0
// post: constructs an empty vector with initialCapacity capacity
// that extends capacity by capacityIncr, or doubles if 0
public Vector(int initialCapacity, int capacityIncr) {
    elementData = new Object[initialCapacity];
    elementCount = 0;
    capacityIncrement = capacityIncr;
}
```

# ensureCapacity()

```
public void add(E element) {  
    ensureCapacity(elementCount+1);  
    . . .  
}  
  
public void ensureCapacity(int minCapacity) {  
    if (elementData.length < minCapacity) {  
        if (capacityIncrement == 0) {  
            //double the array  
        } else {  
            //grow by capacityIncrement  
        }  
        //copy elements to new array  
    }  
}
```

# Growing the Array

- Vector.java
  - `ensureCapacity()`
- Chapter 3 of Bailey

# Observations about Vectors

- How long does it take to add an element?
  - Varies – sometimes takes a lot longer if we have to grow array before adding element
- How long does it take to insert/remove an element in the middle of the Vector?
  - Might take a long time if we have to move several other elements
- Key insight: The running time depends on the size of the Vector!

# Running Time Analysis

- We want *general tools* for understanding how running time and memory usage changes as the amount of data increases
- Example:
  - If I double my Vector's size, how much longer will it take to:
    - Find an element?
    - Insert an element at the front?
    - Remove an element from the middle?
    - Etc.

# Measuring Computational Cost

- How can we measure the cost of a computation?
  - Absolute clock time
    - Problems?
      - Different machines have different clocks
      - Lots of other stuff happening (network, OS, etc)
      - Not consistent. Need lots of tests to predict future behavior



# Measuring Computational Cost

- How can we measure the cost of a computation?
  - Count how many “expensive” operations were performed (i.e., array copies in Vector)
  - Count number of times “x” happens
    - For a specific event or action “x”
    - i.e., How many times a certain variable changes
  - Problems?
    - 64 vs 65? 100 vs 105? Does it really matter??

# Measuring Computational Costs

- Rather than keeping exact counts, we want to know the *order of magnitude* of occurrences
  - 60 vs 600 vs 6000, not 65 vs 68
- We want to make comparisons without looking at details and without running tests
- Avoid using specific numbers or values
- Look for overall trends

# Looking for Trends

- Rule of thumb: ignore constants (most of the time...)
- Examples:
  - Treat  $n$  and  $n/2$  as same order of magnitude
  - $n^2/1000$ ,  $2n^2$ , and  $1000n^2$  are “pretty much” just  $n^2$  (behave in same way)
  - $a_0n^k + a_1n^{k-1} + a_2n^{k-2} + \dots + a_k$  is roughly  $n^k$
- The key is to find the most significant or dominant term

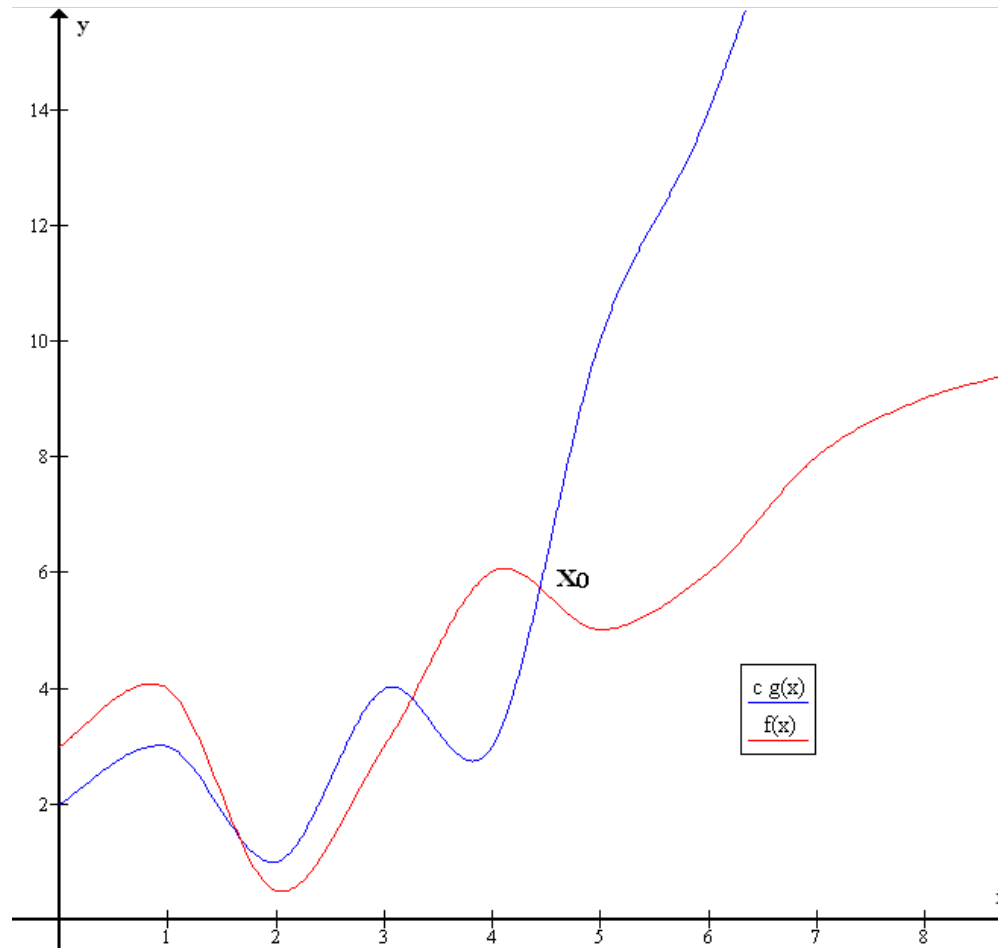
# Asymptotic Bounds (Big-O Analysis)

- A function  $f(n)$  is  $O(g(n))$  if and only if there exists positive constants  $c$  and  $n_0$  such that

$$|f(n)| \leq c * g(n) \text{ for all } n \geq n_0$$

- “g” is bigger than “f” **for large n**
- Example:
  - $f(n) = n^2/2$  is  $O(n^2)$
  - $f(n) = 1000n^3$  is  $O(n^3)$
  - $f(n) = n/2$  is  $O(n)$

$$|f(n)| \leq c * g(n) \text{ for all } n \geq n_0$$



# Determining Upper Bound

- We usually want the smallest upper bound to estimate running time
- Example:
  - $f(n) = 3n^2$
  - $f(n)$  is  $O(n^2)$
  - $f(n)$  is  $O(n^3)$
  - $f(n)$  is  $O(2^n)$
- Best estimate of running time is  $O(n^2)$
- We might care about  $c$  and  $n_0$  **in practice**, but focus on size of  $g$  when designing structures

# Vector Operations

- For Object `o`, int `i`, and `n` elements:
  - `set(i, o)`
  - `add(o)`
  - `add(i, o)`
  - `remove(i)`
  - `add(o)` executed `n` times
  - `add(i, o)` executed `n` times

# Vector Operations

- For Object  $o$ , int  $i$ , and  $n$  elements:
  - $\text{set}(i, o) - O(1)$
  - $\text{add}(o) - O(1)$
  - $\text{add}(i, o) - O(n)$
  - $\text{remove}(i) - O(n)$
  - $\text{add}(o)$  executed  $n$  times  $- O(n)$
  - $\text{add}(i, o)$  executed  $n$  times  $- O(n^2)$



# Common Functions

For  $n$  = number of elements:

- $O(1)$ : constant time and space
- $O(\log n)$ : divide and conquer algorithms, binary search
- $O(n)$ : linear dependence, simple list lookup
- $O(n \log n)$ : divide and conquer sorting algorithms
- $O(n^2)$ : matrix addition, selection sort
- $O(n^3)$ : matrix multiplication
- $O(n^k)$ : cell phone switching algorithms
- $O(2^n)$ : color graph with 3 colors, satisfiability
- $O(n!)$ : traveling salesman problem

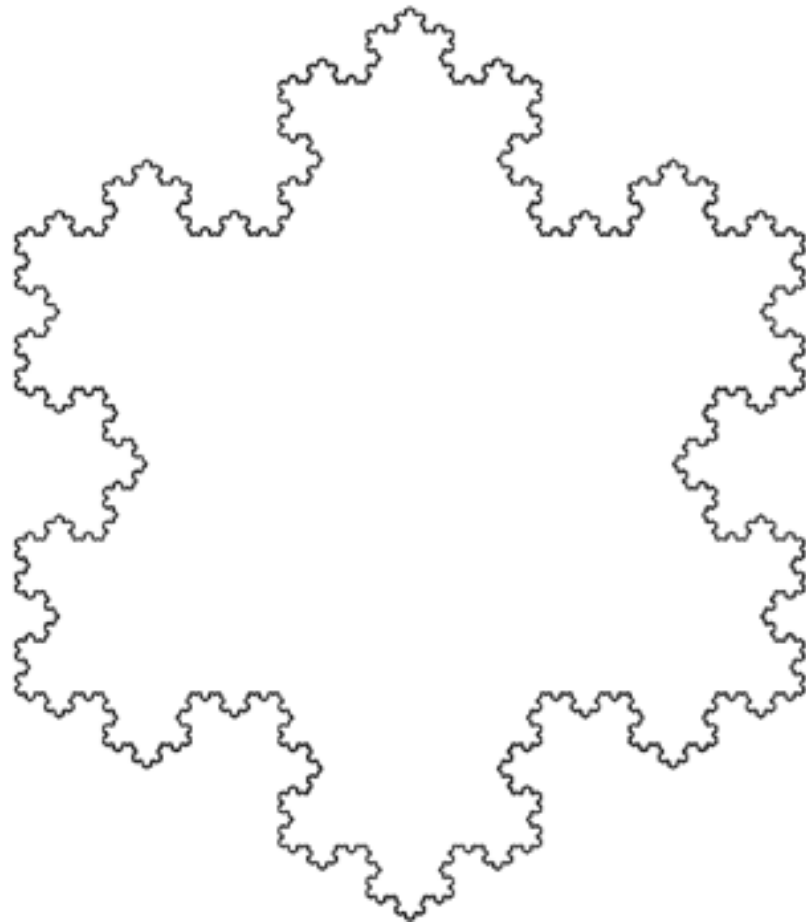
# Input-dependent Running Times

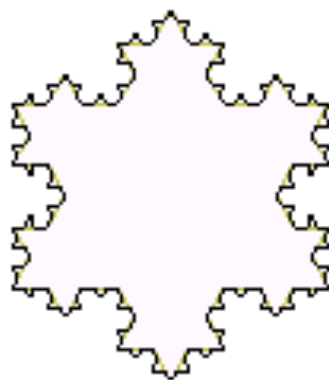
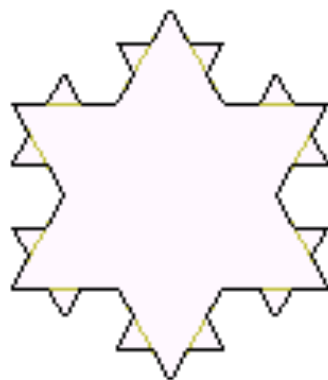
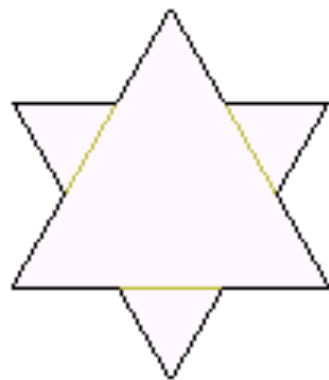
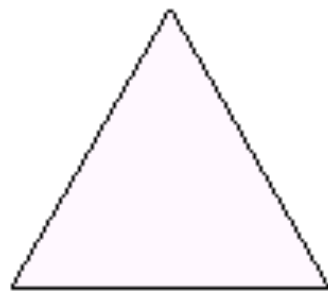
- Algorithms may have different running times for different input values
- Best case
  - Sort already sorted array in  $O(n)$
  - Find item in first place that we look  $O(1)$
- Worst case
  - Don't find item in list  $O(n)$
  - Reverse order sort  $O(n^2)$
- Average case
  - Linear search  $O(n)$
  - Sort random array  $O(n \log n)$

Moving on...

# Recursion

- **General problem solving strategy**
  - Break problem into smaller pieces
  - Sub-problems may look a lot like original - may in fact be smaller versions of same problem
- **Examples**





# Recursion

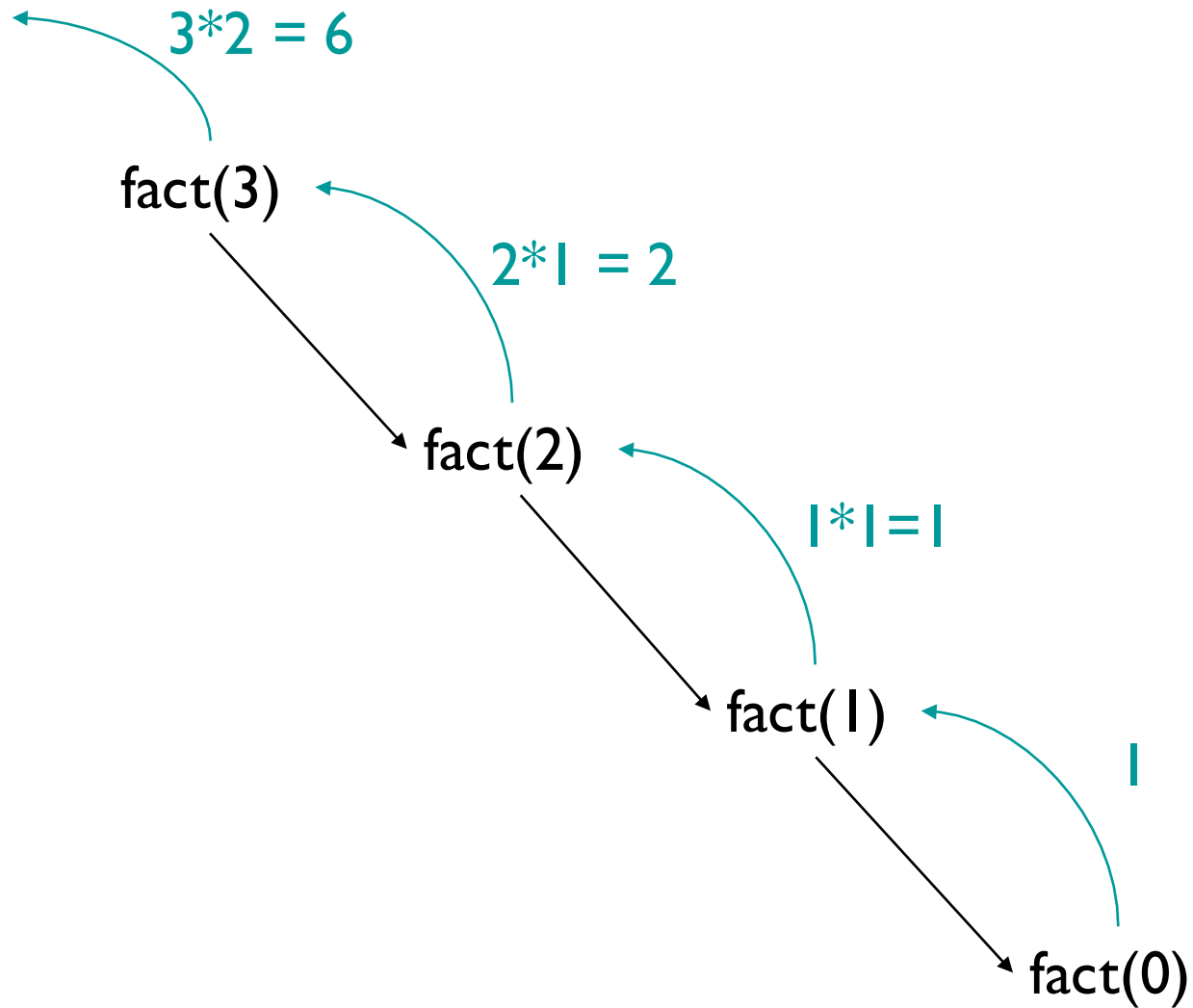
- Many algorithms are recursive
  - Can be easier to understand (and prove correctness/state efficiency of) than iterative versions
- Today we will review recursion and Wednesday we will talk about techniques for reasoning about recursive algorithms

# Factorial

- $n! = n \cdot (n-1) \cdot (n-2) \cdot \dots \cdot 1$
- How can we implement this?
  - We could use a while loop...
  
- But we could also write it recursively
  - $n! = n \cdot (n-1)!$



# Factorial



# Factorial

- In recursion, we always use the same basic approach
- What's our base case?
  - $n=0$ ;  $\text{fact}(0) = 1$
- What's our recursive case?
  - $n>0$ ;  $\text{fact}(n) = n \bullet \text{fact}(n-1)$

# fact.java

```
public class fact{

    public static int fact(int n) {
        if (n==0) {
            return 1;
        }
        else {
            return n*fact(n-1);
        }
    }

    public static void main(String args[]) {
        System.out.println(fact(Integer.valueOf(args[0]).intValue()));
    }

}
```