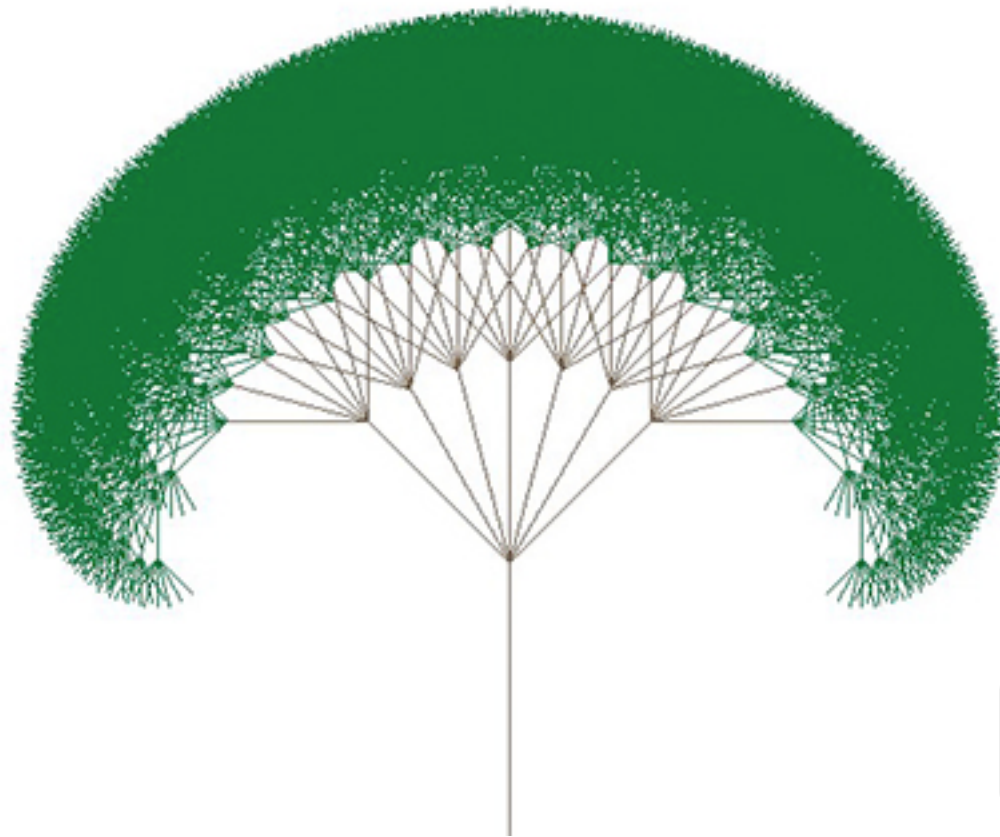


# Recursion: Introduction



# Announcements

- Deadlines
- Exam on Thursday: Units 1 – 5 (inclusive)
- PA 4 due tonight
- OLI Recursion over the weekend
- Monday: PA5 is due.

# Today

- Review of Big-O
- Recursion:
  - Introduction to recursion
  - What it is
  - Recursion and the stack
  - Recursion and iteration
  - Examples of simple recursive functions
  - Geometric recursion: fractals

# Big-O Review



# Asymptotic Analysis

- Beyond number of operations
- Goal: understanding behavior of program over the long run, with increasingly large inputs
- We are not concerned with constants factors:
  - How many iterations?
  - Not operations in each iteration
- Gives a useful approximation, suppresses details
- Worst-case

# Order of Complexity

- We express this as the (time) order of complexity
- Normally expressed using Big-O notation.
- Big-0 is ignores constants, focuses on **highest power of n**

## Number of iterations

- $n$
- $3n+3$
- $2n+8$

## Order of Complexity

- $O(n)$
- $O(n)$
- $O(n)$

# Linear Search: Worst Case

```
# let n = the length of list.
```

```
def search(list, key):
```

```
    index = 0 1
```

```
    while index < len(list): n+1
```

```
        if list[index] == key: n
```

```
            return index
```

```
        index = index + 1 n
```

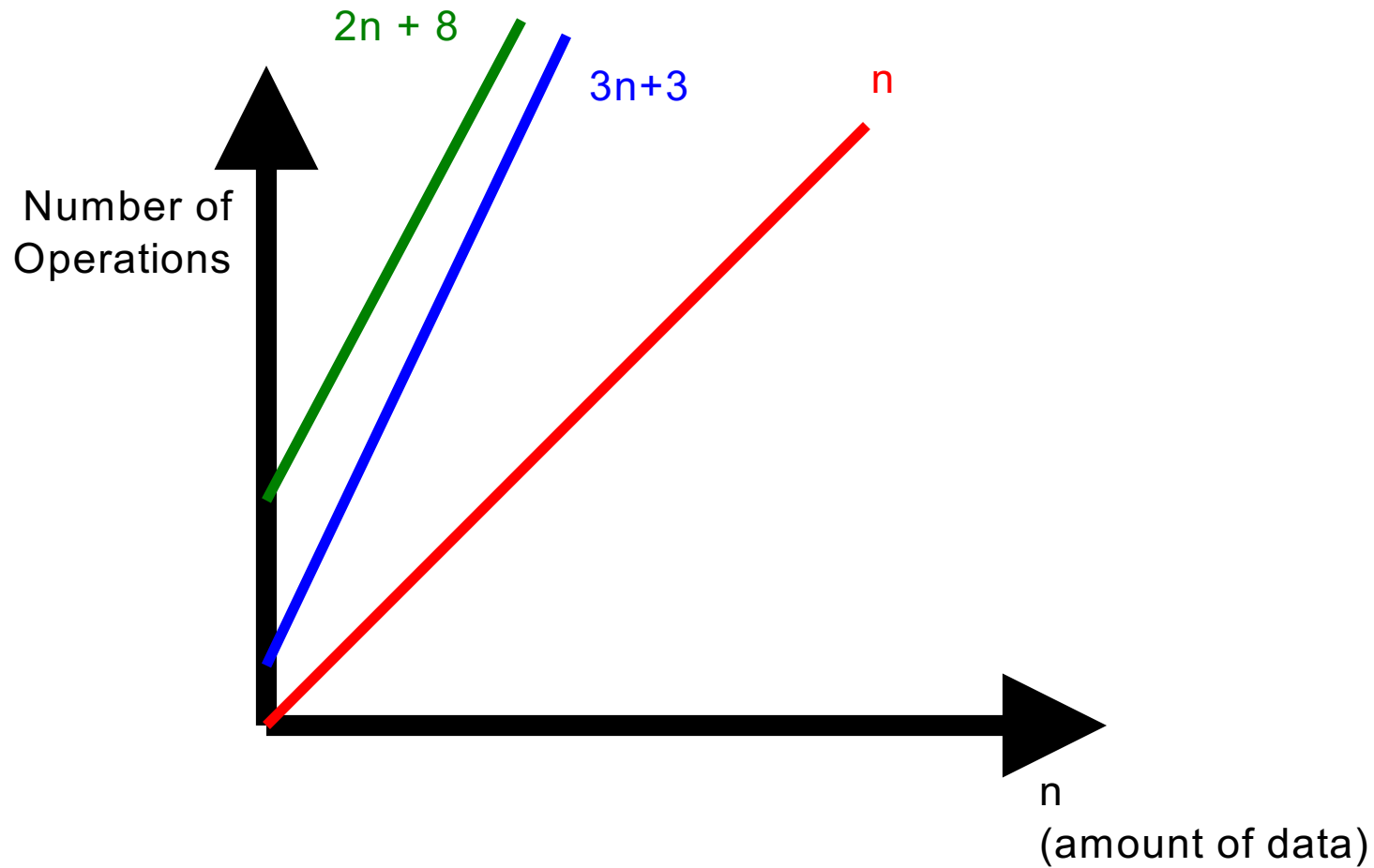
```
    return None 1
```

```
Total: 3n+3
```

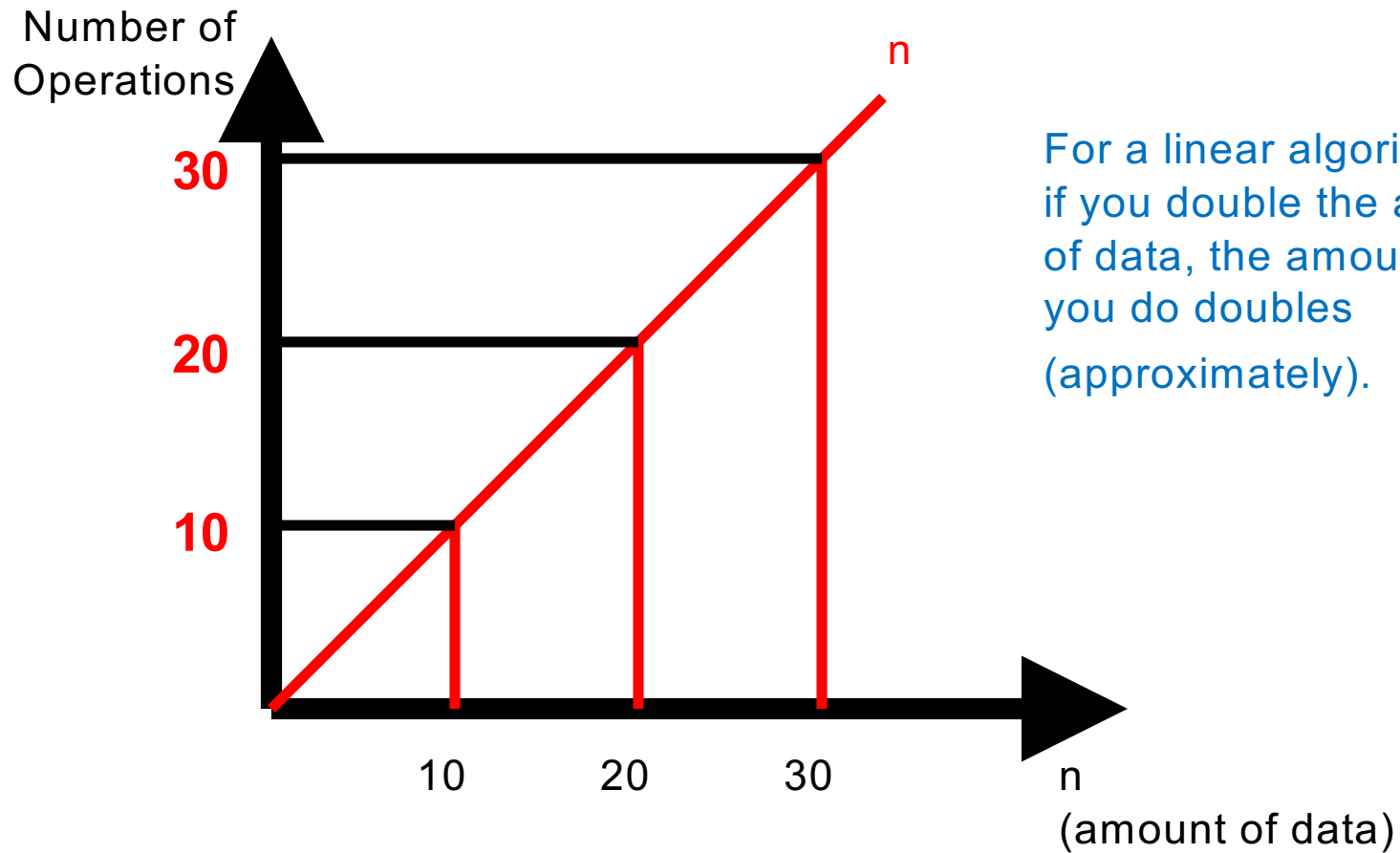
# Linear Search: Worst Case Simplified

```
# let n = the length of list.  
def search(list, key):  
    index = 0  
    while index < len(list):           n iterations  
        if list[index] == key:  
            return index  
        index = index + 1  
    return None
```

# $O(n)$ (“Linear”)



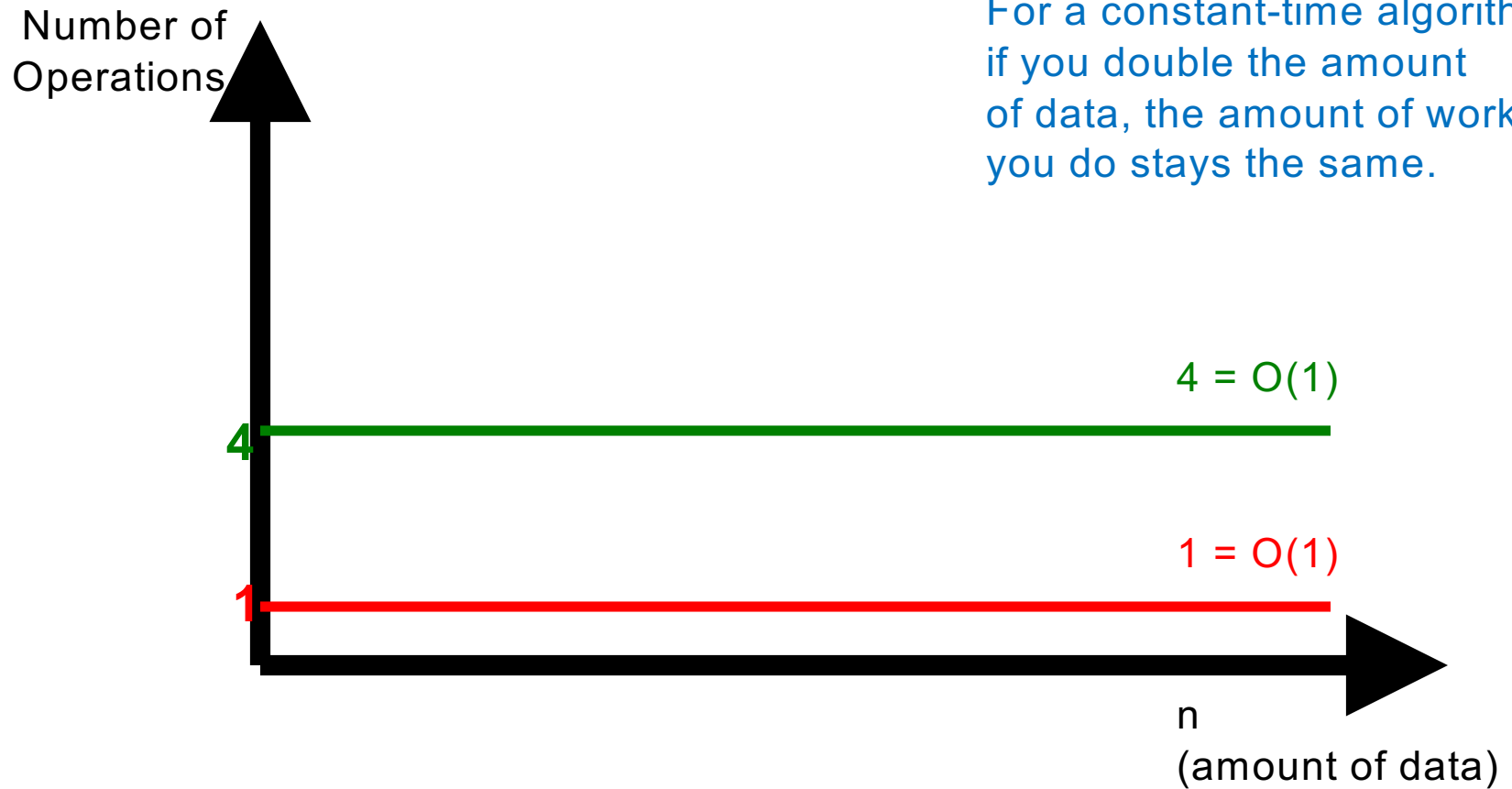
# $O(n)$



For a linear algorithm, if you double the amount of data, the amount of work you do doubles (approximately).

# $O(1)$ (“Constant-Time”)

For a constant-time algorithm, if you double the amount of data, the amount of work you do stays the same.



# Insertion Sort: worst case

```
# let n = the length of list.  
def isort(list):  
    i = 1  
    while i != len(list):      #n-1 iterations  
        move_left(list,i)  
        i = i + 1  
    return list
```

What is the cost of move\_left?



# Insertion Sort: cost of move left

```
# let n = the length of list.  
def move_left(a, i):  
    x = a.pop(i)           n iterations  
    j = i - 1  
    while j >= 0 and a[j] > x: i iterations  
        j = j - 1  
    a.insert(j + 1, x)     n iterations
```

Total cost (at most):  $n + i + n$

But what is  $i$ ? To find out, look at `isort`, which calls `move_left`, supplying a value for  $i$

# Insertion Sort: worst case

```
# let n = the length of list.  
def isort(list):  
    i = 1  
    while i != len(list):      #n-1 iterations  
        move_left(list,i)    #i goes from 1 to n-1  
        i = i + 1  
    return list
```

Total cost: *cost of move\_left as i goes from 1 to n-1*

# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list,i)  
        i = i + 1  
    return list
```


# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):      n-1  
        move_left(list,i)  
        i = i + 1  
    return list
```

# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list,i)  
        pop  
        while loop  
        insert  
        i = i + 1  
    return list
```

n-1



# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list, i)           n-1  
        pop..... } n  
        while loop  
        insert  
        i = i + 1  
    return list
```

# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list,i)  
        pop..... } n  
        while loop }  
        insert ..... } n  
        i = i + 1  
    return list
```

The diagram shows a code snippet for an insertion sort function. The code is annotated with complexity markers. A green bracket on the right side of the code indicates the cost of the operations within the while loop. The cost of the while loop is  $n-1$ . The cost of the pop operation is  $n$ , and the cost of the insert operation is  $n$ .

# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list, i)           n-1  
        pop & insert .....  
        while loop                   } n + n  
        i = i + 1  
    return list
```



# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list, i)           n-1  
        pop & insert .....  
        while loop                   } 2n  
    i = i + 1  
    return list
```

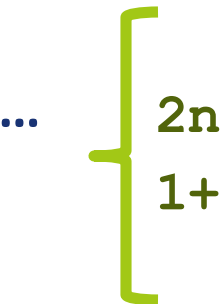
# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list, i)  
        pop & insert .....  
        while loop  
        i = i + 1  
    return list
```

$n-1$

$2n$

$1+$



# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list,i)  
        pop & insert .....  
        while loop  
        i = i + 1  
    return list
```

$n-1$

$2n$

$1+2+$

# Examining the cost

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list, i)  
        pop & insert .....  
        while loop  
        i = i + 1  
    return list
```

$n-1$

$2n$   
 $1+2+3\dots$

# How can we express this?

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

```
        move_left(list, i)
```

```
            pop & insert .....
```

```
            while loop
```

```
        i = i + 1
```

```
    return list
```

$n-1$

$2n$

$1+2+3\dots n-1$

# How can we express this?

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

```
        move_left(list, i)
```

```
        pop & insert .....
```

```
        while loop
```

```
        i = i + 1
```

```
    return list
```

n-1

2n

1+2+3...n-1



**1 + 2 + 3 ... n - 1**

Test for  $n = 7$

$$1 + 2 + 3 \dots n - 1$$

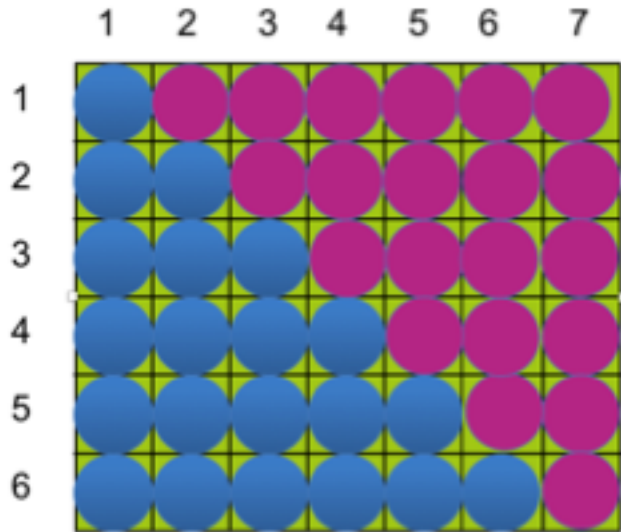
Test for  $n = 7$ .

$$1 + 2 + 3 + 4 + 5 + 6$$

$$1 + 2 + 3 \dots n - 1$$



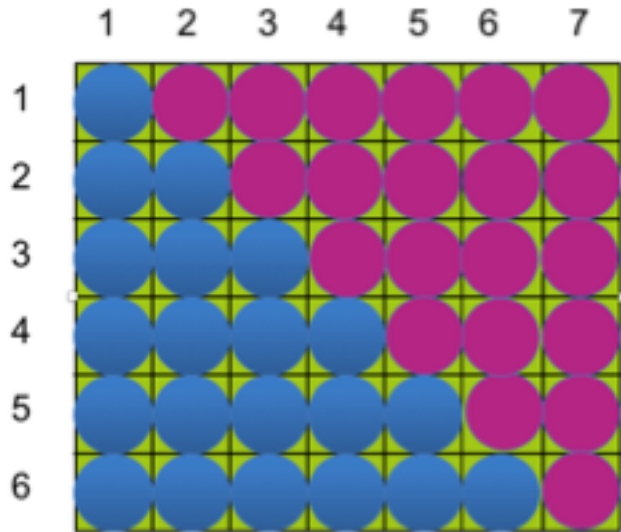
Test for  $n = 7$ .



$$1 + 2 + 3 + 4 + 5 + 6$$

$$1 + 2 + 3 \dots n - 1$$

Test for  $n = 7$ .

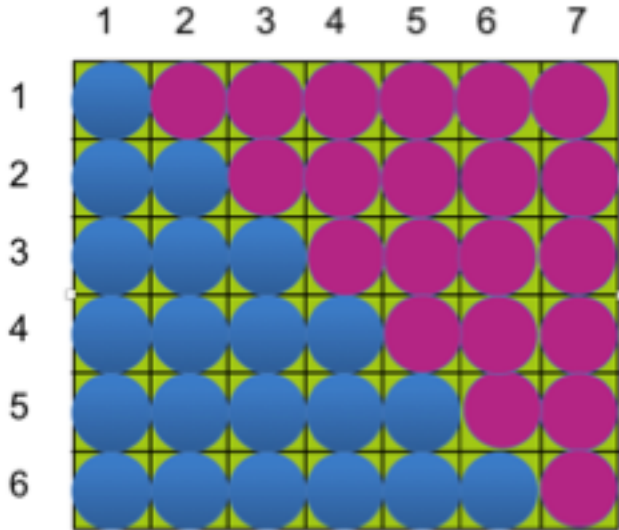


$(6) * (7) / 2$  blue circles

$$1 + 2 + 3 + 4 + 5 + 6$$

$$1 + 2 + 3 \dots n - 1$$

Test for  $n = 7$ .



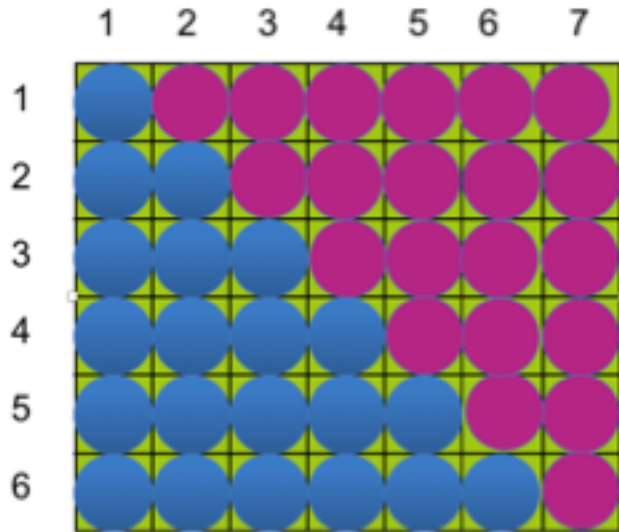
$(6) * (7) / 2$  blue circles

$(n-1) * (n) / 2$  blue circles

$$1 + 2 + 3 + 4 + 5 + 6$$

$$1 + 2 + 3 \dots n - 1$$

# Our equation ...



$(6) * (7) / 2$  blue circles

$(n-1) * (n) / 2$  blue circles

$$(n-1) * n / 2$$

$$1 + 2 + 3 \dots n - 1$$

Our equation ...

$$(n-1) * n / 2$$

$$1+2+3...n-1$$

# How can we express this?

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

```
        move_left(list, i)
```

```
        pop & insert .....
```

```
        while loop
```

n-1

2n

1+2+3...n-1

`i = i + 1`  
`return list`

$$(n-1) * n / 2$$
$$1+2+3...n-1$$

# How can we express this?

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

```
        move_left(list,i)
```

```
        pop & insert .....
```

```
        while loop
```


```
        i = i + 1
```

```
    return list
```

n-1

2n

1+2+3...n-1


$$(n-1) * n / 2$$

# Combine to calculate

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

```
        move_left(list, i)
```

```
            pop & insert .....
```

```
            while loop
```

```
        i = i + 1
```

```
    return list
```

$n-1$

$2n +$

$(n-1)*n/2$



# How can we express this?

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

```
        move_left(list, i)
```

```
        pop & insert & while
```

```
        i = i + 1
```

```
    return list
```

n-1

$2n + (n-1)*n/2$

# How can we express this?

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

```
        move_left(list, i)
```

n-1

{  $2n + (n-1)*n/2$

```
        i = i + 1
```

```
    return list
```

# How can we express this?

```
def isort(list):  
    i = 1  
    while i != len(list):  
        move_left(list, i)  
        i = i + 1  
    return list
```

$n-1$   
 $\left\{ (2n + (n-1)*n/2) \right.$

# Total number of operations

```
def isort(list):
```

```
    i = 1
```

```
    while i != len(list):
```

$(n-1) * (2n + (n-1)*n/2)$

```
        move_left(list,i)
```

```
        i = i + 1
```

```
    return list
```

# Generalizing...

$$(n-1) * (2n + (n-1)*n/2)$$

$$\square = 2n^2 - 2n + (n^2 - n) / 2$$

$$\square = (5n^2 - 5n) / 2$$

$$\square = (5/2)n^2 - (5/2)n$$

Highest order term? ...

$$(5/2)n^2 - (5/2)n$$

$$n^2$$

# Order of Complexity

Number of operations

$$n^2$$

$$(5/2)n^2 - (1/2)n$$

$$2n^2 + 7$$

Order of Complexity

$$O(n^2)$$

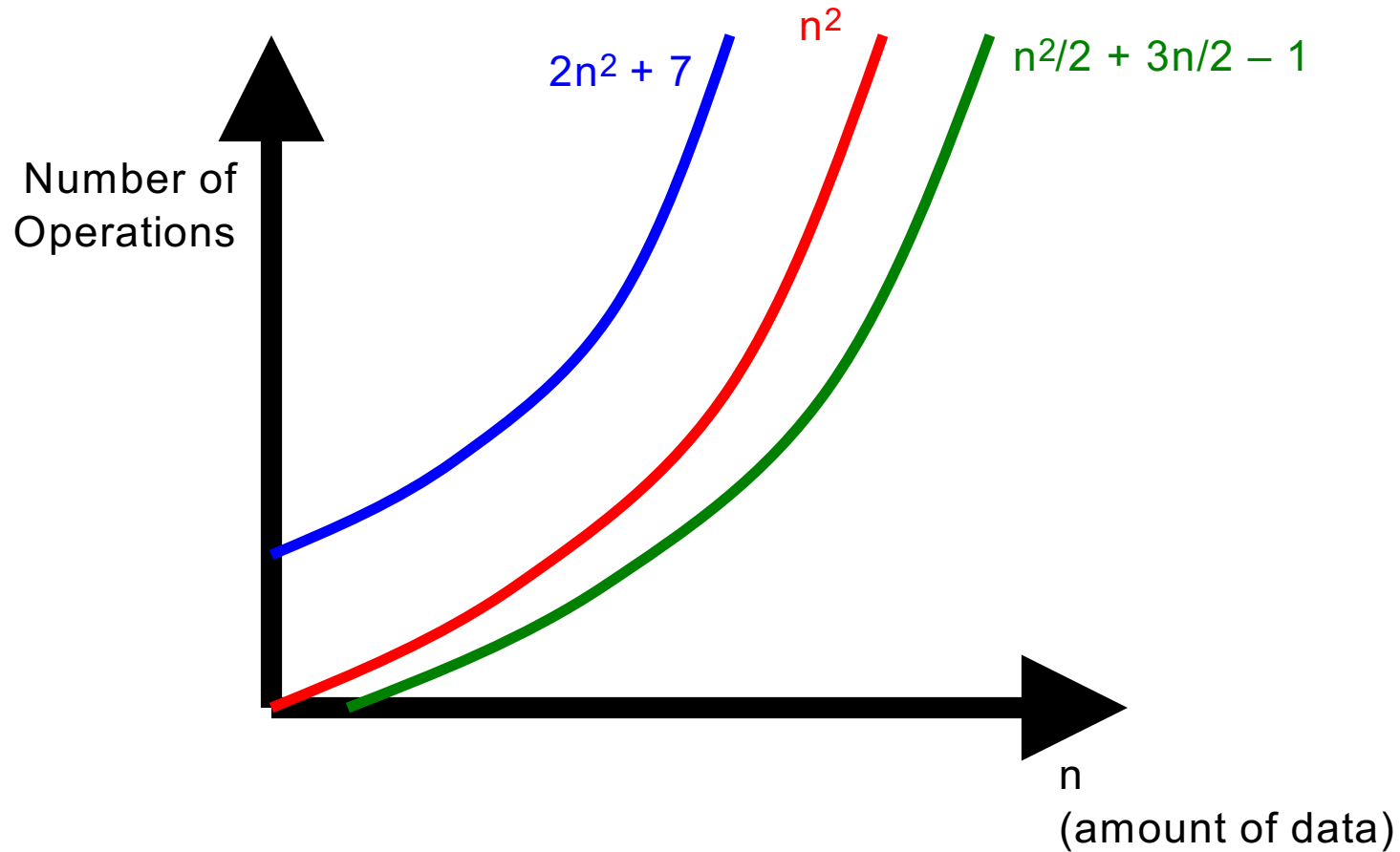
$$O(n^2)$$

$$O(n^2)$$

**Usually doesn't matter what the constants are... we are only concerned about the highest power of  $n$ .**

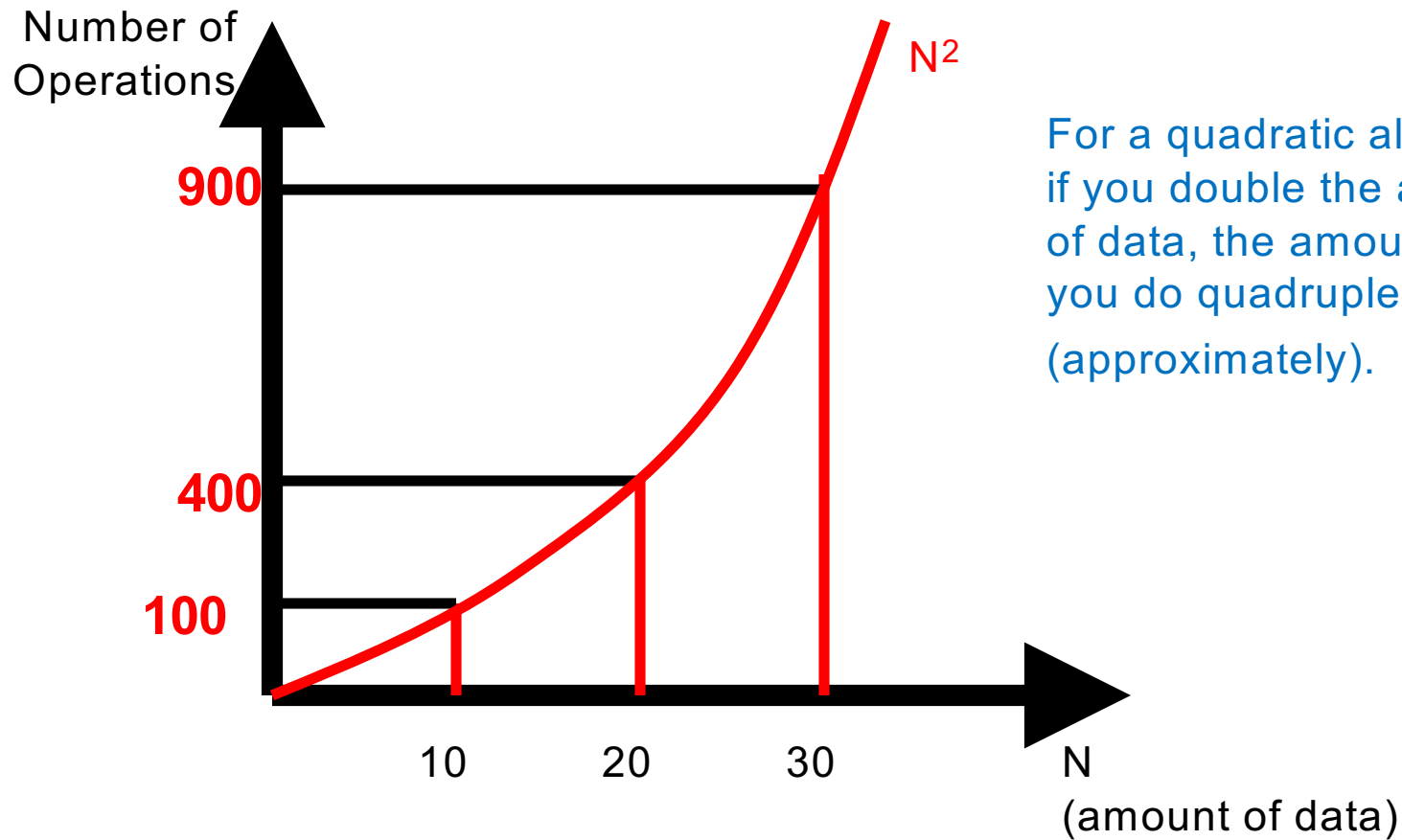
**$f(n)$  is  $O(g(n))$  means  $f(n) < g(n) \cdot k$  for some positive  $k$**

# $O(n^2)$ ("Quadratic")





$$O(n^2)$$



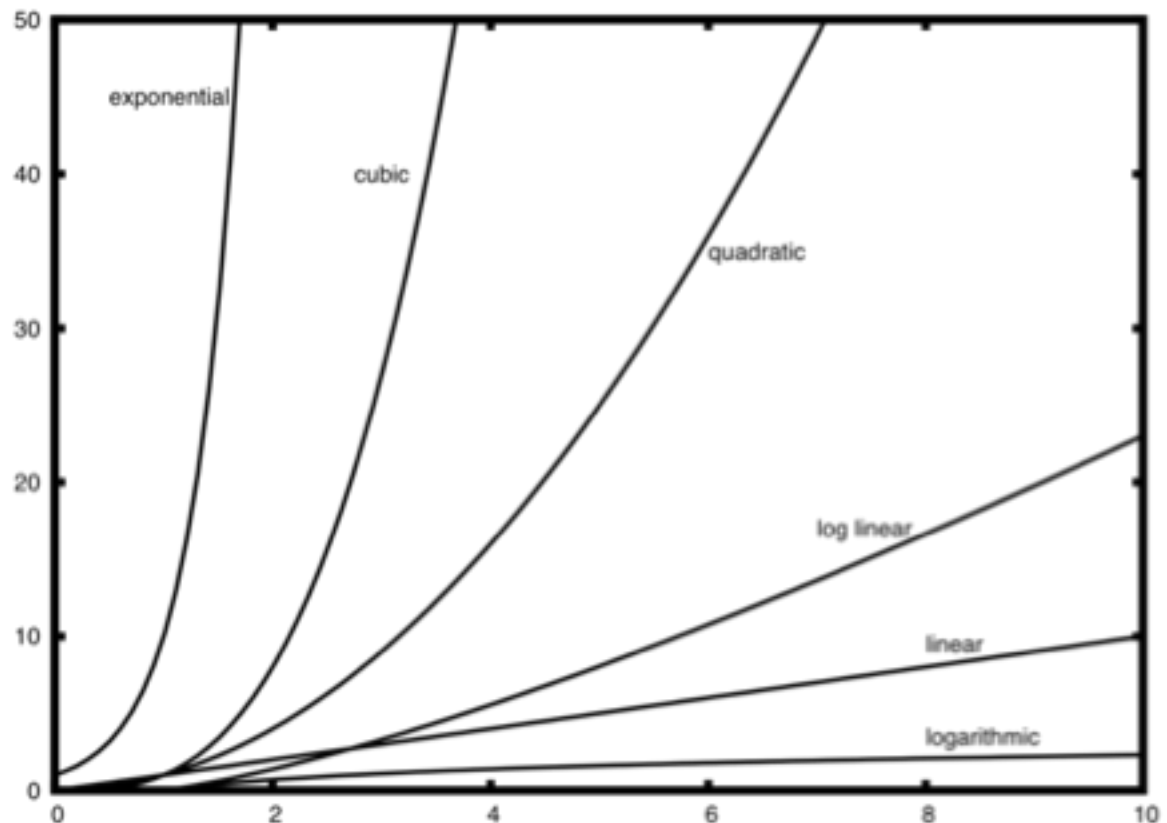
For a quadratic algorithm, if you double the amount of data, the amount of work you do quadruples (approximately).

# Two Examples

- Linear Sort       $O(n)$       linear
- Insertion Sort       $O(n^2)$       quadratic

# Big O

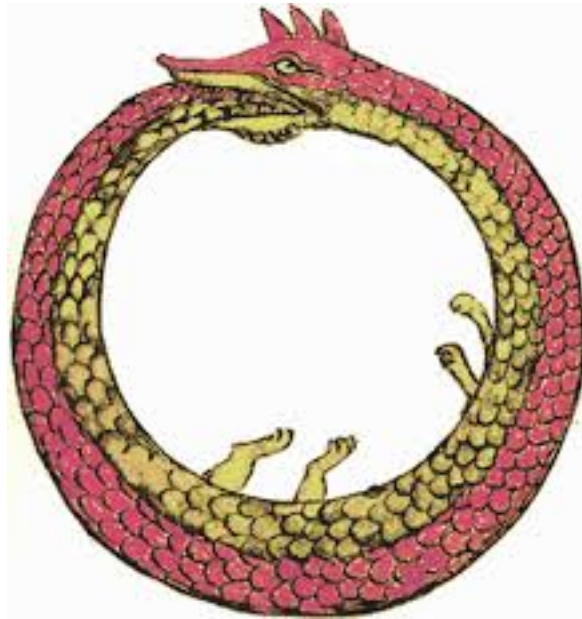
- $O(1)$  constant
- $O(\log n)$  logarithmic
- $O(n)$  linear
- $O(n \log n)$  log linear
- $O(n^2)$  quadratic
- $O(n^3)$  cubic
- $O(2^n)$  exponential



# How work increases

Input Size	$O(n)$	$O(n^2)$	$O(n^3)$	$O(2^n)$
2	2	4	8	4
4	4	16	64	16
8	8	64	512	256
16	16	256	4096	65536
32	32	1024	32768	4294967296

# Recursion



**THE LOOPLESS LOOP**

# Recursion

- A recursive function is one that **calls itself**.
- ```
def i_am_recursive(x):  
    maybe do some work  
    if there is more work to do:  
        i_am_recursive(next(x))  
    return the desired result
```
- Infinite loop? Not necessarily, not if `next(x)` needs less work than `x`.

# Recursive Definitions

- Every recursive function definition includes two parts:
  - **Base case(s) (non-recursive)**  
One or more simple cases that can be done directly or immediately
  - **Recursive case(s)**  
One or more cases that require solving “simpler” version(s) of the original problem.
    - By “simpler”, we mean “smaller” or “shorter” or “closer to the base case”.

# Example: Factorial

- $n! = n \times (n-1) \times (n-2) \times \dots \times 1$   
 $2! = 2 \times 1$   
 $3! = 3 \times 2 \times 1$   
 $4! = 4 \times 3 \times 2 \times 1$   
 $9! = 362,880$   
 $10! = 3,628,800$   
 $10! = 10 \times 9!$

▣ *alternatively:* (Recursive case)

$$0! = 1 \quad (\text{Base case})$$

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

$$\text{So } 4! = 4 \times 3! \rightarrow 3! = 3 \times 2! \rightarrow 2! = 2 \times 1! \rightarrow$$

$$1! = 1 \times 0! \rightarrow 0! = 1$$



# Recursion conceptually

$$4! = 4(3!)$$

$$3! = 3(2!)$$

$$2! = 2(1!)$$

$$1! = 1(0!)$$




Base case



make smaller instances  
of the same problem


# Recursion conceptually


$$4! = 4(3!)$$

$$3! = 3(2!)$$

$$2! = 2(1!)$$

$$1! = 1(0!) = 1(1) = 1$$



Compute the base case

make smaller instances  
of the same problem

# Recursion conceptually

$$4! = 4(3!)$$


$$3! = 3(2!)$$

$$2! = 2(1!) = 2$$

$$1! = 1(0!) = 1(1) = 1$$

  
Compute the base case

  
make smaller instances  
of the same problem

  
build up  
the result

# Recursion conceptually

$$4! = 4(3!)$$

$$3! = 3(2!)$$

$$2! = 2(1!)$$

$$1! = 1(0!) = 1(1) = 1$$

$$= 6$$

$$= 2$$

  
Compute the base case

make smaller instances  
of the same problem

build up  
the result

# Recursion conceptually

$$4! = 4(3!)$$

$$3! = 3(2!)$$

$$2! = 2(1!)$$

$$1! = 1(0!) = 1(1) = 1$$

$$= 2$$

$$= 6$$


$$= 24$$



make smaller instances  
of the same problem



Compute the base case



build up  
the result

# Recipe for Writing Recursive Functions

(by Dave Feinberg)

## 1. Write `if`. (Why?)

There must be at least 2 cases: base and recursive

## 2. Handle simplest case(s).

No recursive call needed (base case).

## 3. Write recursive calls(s).

Input is slightly simpler to get closer to base case.

## 4. Assume the recursive call works!

Ask yourself: What does it do?

Ask yourself: How does it help?

# Recursive Factorial in Python

```
# Assumes n >= 0
```

```
def factorial(n):
```

```
    if n == 0:        # base case
```

```
        return 1
```

```
    else:             # recursive case
```

```
        result = factorial(n-1)
```

```
        return n * result
```

$0! = 1$

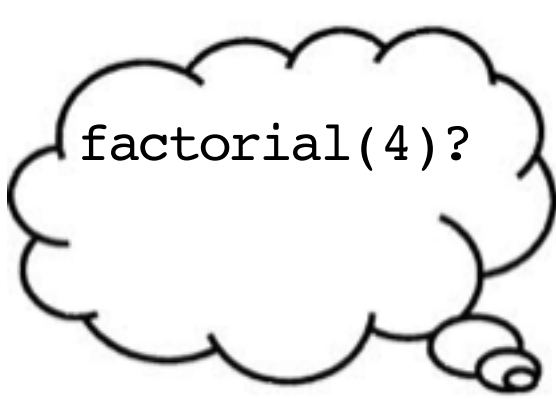
(Base case)

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

(Recursive case)

S

n=4



T

A

C

K





S

n=4

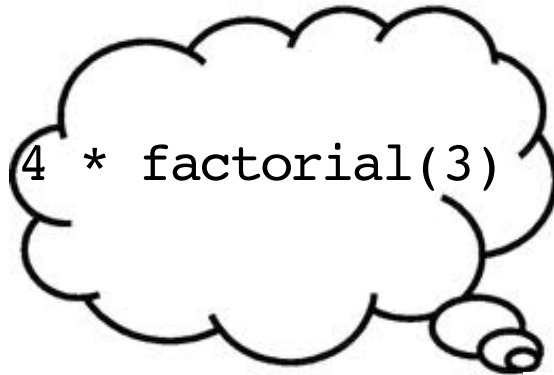
factorial(4)? = 4 \* factorial(3)

T

A

C

K



S

n=4     `factorial(4)? = 4 * factorial(3)`

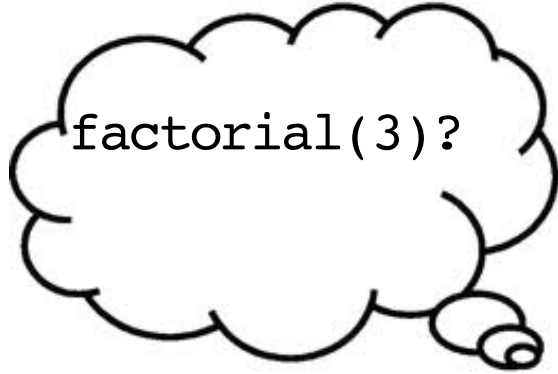
T

n=3     `factorial(3)?`

A

C

K



S

$$n=4 \quad \text{factorial}(4)? = 4 * \text{factorial}(3)$$

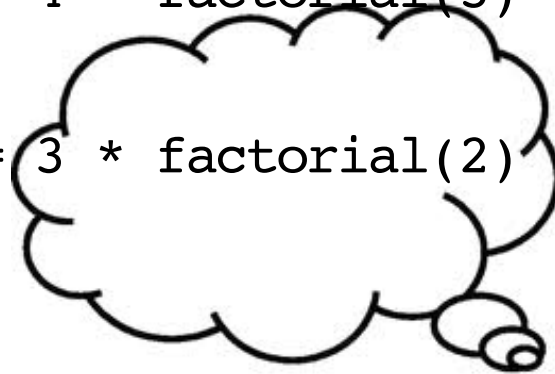
T

$$n=3 \quad \text{factorial}(3)? = 3 * \text{factorial}(2)$$

A

C

K



S

$$n=4 \quad \text{factorial}(4)? = 4 * \text{factorial}(3)$$

T

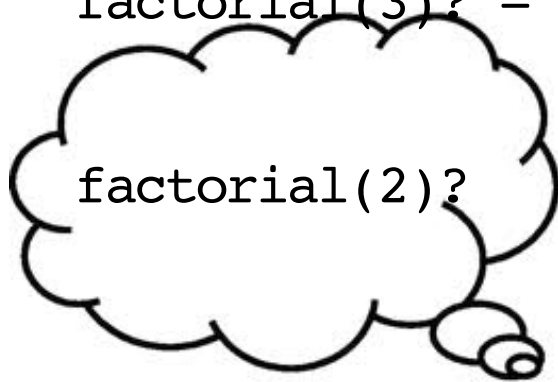
$$n=3 \quad \text{factorial}(3)? = 3 * \text{factorial}(2)$$

A

$$n=2 \quad \text{factorial}(2)?$$

C

K



S

$$n=4 \quad \text{factorial}(4)? = 4 * \text{factorial}(3)$$

T

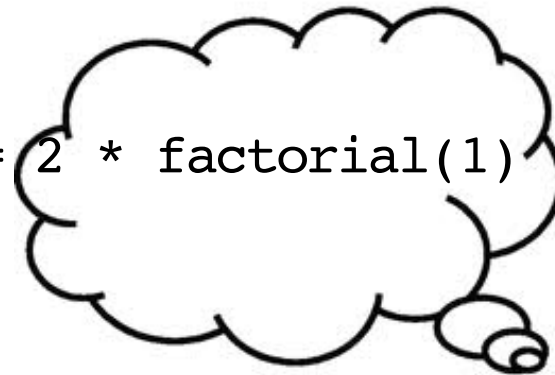
$$n=3 \quad \text{factorial}(3)? = 3 * \text{factorial}(2)$$

A

$$n=2 \quad \text{factorial}(2)? = 2 * \text{factorial}(1)$$

C

K



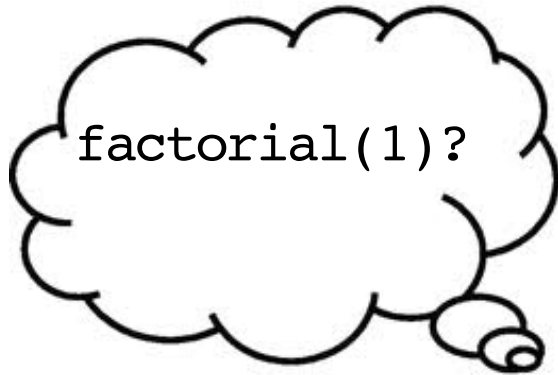
S  $n=4$  factorial(4)? = 4 \* factorial(3)

T  $n=3$  factorial(3)? = 3 \* factorial(2)

A  $n=2$  factorial(2)? = 2 \* factorial(1)

C  $n=1$  factorial(1)?

K



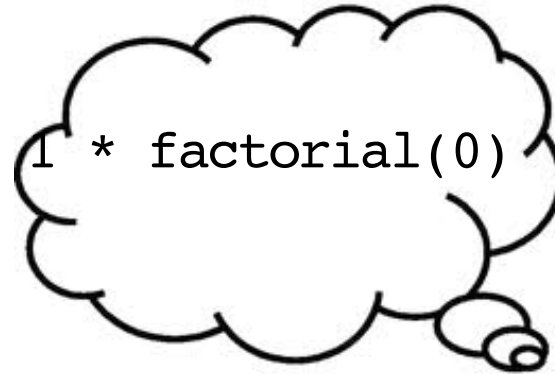
S  $n=4$  factorial(4)? = 4 \* factorial(3)

T  $n=3$  factorial(3)? = 3 \* factorial(2)

A  $n=2$  factorial(2)? = 2 \* factorial(1)

C  $n=1$  factorial(1)? = 1 \* factorial(0)

K



S  $n=4$  factorial(4)? = 4 \* factorial(3)

T  $n=3$  factorial(3)? = 3 \* factorial(2)

A  $n=2$  factorial(2)? = 2 \* factorial(1)

C  $n=1$  factorial(1)? = 1 \* factorial(0)

K  $n=0$  factorial(0) = 1





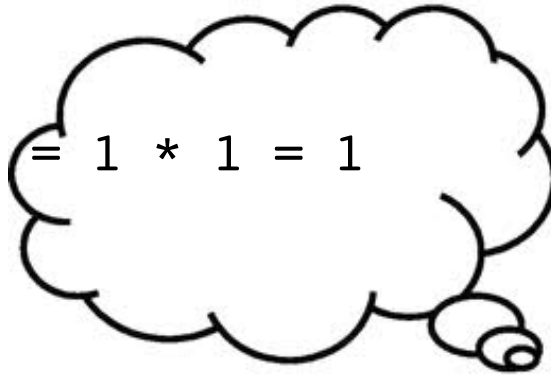


S  $n=4$  factorial(4)? = 4 \* factorial(3)

T  $n=3$  factorial(3)? = 3 \* factorial(2)

A  $n=2$  factorial(2)? = 2 \* factorial(1)

C  $n=1$  factorial(1) = 1 \* 1 = 1



S

$$n=4 \quad \text{factorial}(4)? = 4 * \text{factorial}(3)$$

T

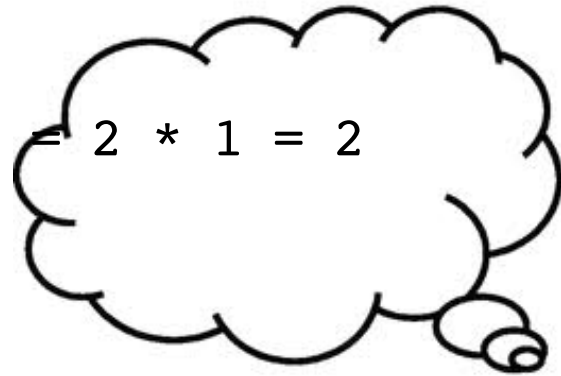
$$n=3 \quad \text{factorial}(3)? = 3 * \text{factorial}(2)$$

A

$$n=2 \quad \text{factorial}(2) = 2 * 1 = 2$$

C

K



S

$$n=4 \quad \text{factorial}(4)? = 4 * \text{factorial}(3)$$

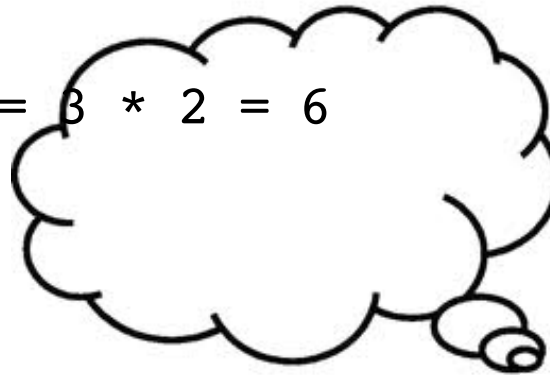
T

$$n=3 \quad \text{factorial}(3) = 3 * 2 = 6$$

A

C

K



S

n=4

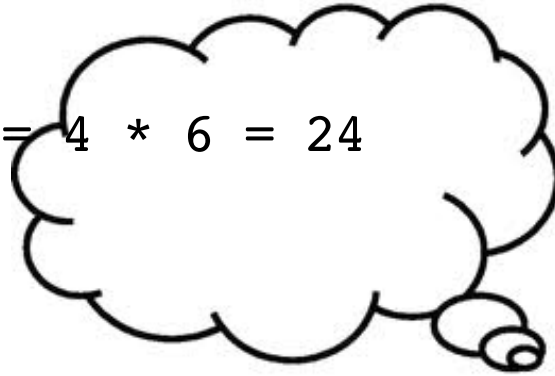
$$\text{factorial}(4) = 4 * 6 = 24$$

T



A

C

K



# Recursive vs. Iterative Solutions

- For **every recursive function**,  calls itself  
there is an equivalent iterative solution.
- For **every iterative function**,  for loop,  
while loop  
there is an equivalent recursive solution.
- But **some problems** are easier to solve one way than the other way.
- And be aware that **most recursive** programs need space for the stack, behind the scenes

# Factorial Function (Iterative)

```
def factorial(n):  
    result = 1    # initialize accumulator var  
    for i in range(1, n+1):  
        result = result * i  
    return result
```

## Versus (Recursive):

```
def factorial(n):  
    if n == 0:    # base case  
        return 1  
    else:        # recursive case  
        return n * factorial(n-1)
```

# A Strategy for Recursive Problem Solving (hat tip to Dave Evans)

- Think of the smallest size of the problem and write down the solution (base case)
- **Be optimistic.** Assume you magically have a working function to solve any size. How could you use it on a smaller size and **use the answer** to solve a bigger size? (recursive case)
- Combine the base case and the recursive case

# Recursion on Lists

Do we know how to use iteration to sum the elements in a list?



# Recursion on Lists

- First we need a way of getting a smaller input from a larger one:

- Forming a sub-list of a list:

```
>>> a = [1, 11, 111, 1111, 11111, 111111]
```

```
>>> a[1:]  the "tail" of list a  
[11, 111, 1111, 11111, 111111]
```

```
>>> a[2:]  
[111, 1111, 11111, 111111]
```

```
>>> a[3:]  
[1111, 11111, 111111]
```

```
>>> a[3:5]  
[1111, 11111]
```

# Recursive sum of a list

```
def sumlist(items):  
    if      :
```

What is the smallest size  
list?

# Recursive sum of a list

```
def sumlist(items):  
    if items == []:
```

The smallest size list is the  
empty list.

What is the sum of an empty list?

# Recursive sum of a list

```
def sumlist(items):  
    if items == []:  
        return 0
```

Base case:  
The sum of an **empty list** is 0.



# Recursive sum of a list

```
def sumlist(items):
```

```
    if items == []:
```

```
        return 0
```

```
    else:
```



Recursive case:  
the list is not empty

# Recursive sum of a list


```
def sumlist(items):  
    if items == []:  
        return 0  
    else:  
        ... sumlist( ) ...
```



What is a simpler/smaller case?

# Recursive sum of a list

```
def sumlist(items):  
    if items == []:  
        return 0  
    else:  
        ... sumlist(items[1:]) ...
```

  
"tail" of list

What if **we already know**  
the sum of the list's tail?



# Recursive sum of a list

```
def sumlist(items):  
    if items == []:  
        return 0  
    else:  
        return items[0] + sumlist(items[1:])
```



What if **we already know** the sum of the list's tail?

We can just add in the list's first element!



# Tracing sumlist

```
def sumlist(items):  
    if items== []:  
        return 0  
    else:  
        return items[0] + sumlist(items[1:])
```

```
>>> sumlist([2,5,7])
```

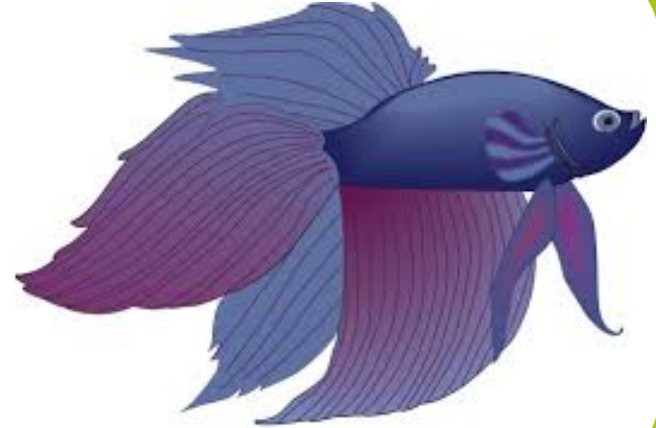
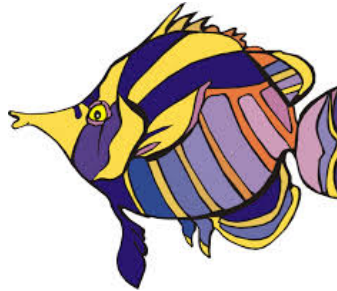
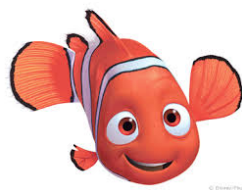
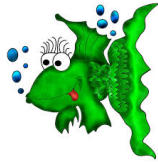
```
sumlist([2,5,7]) = 2 + sumlist([5,7])  
                   5 + sumlist([7])  
                       7 + sumlist([])  
                           0
```

After reaching the base case, the final result is built up by the computer by adding 0+7+5+2.

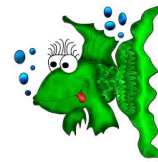
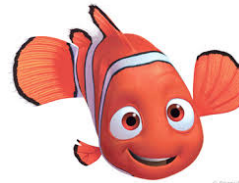
# List Recursion: exercise

- Let's create a recursive function `rev ( items )`
- **Input:** a list of items
- **Output:** another list, with all the same items, but in reverse order
- **Remember:** it's usually sensible to break the list down into its *head* (first element) and its *tail* (all the rest). The tail is a smaller list, and so "closer" to the base case.
- Soooo... (picture on next slide)

# Reversing a list: recursive case



see [file rev\\_list.py](#)

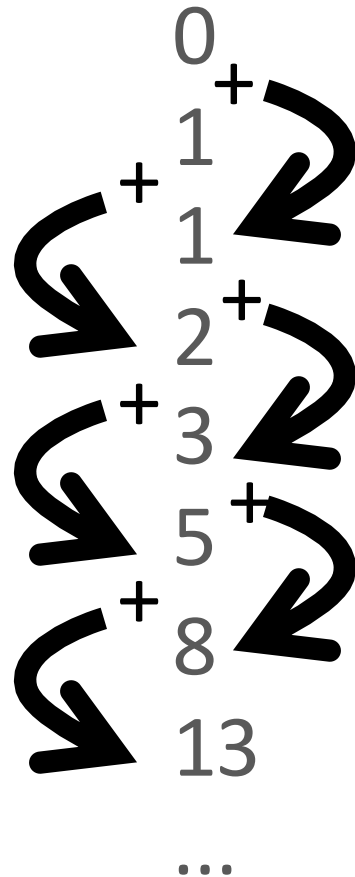


# Multiple Recursive Calls

- So far we've used just one recursive call to build up our answer
- The real **conceptual** power of recursion happens when we need more than one!
- Example: Fibonacci numbers

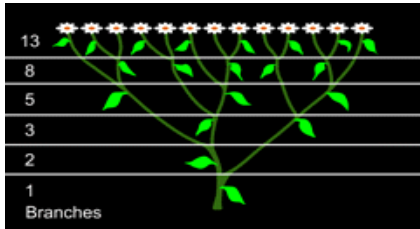
# Fibonacci Numbers

□ A sequence of numbers:



# Fibonacci Numbers in Nature

- 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, 144, 233, etc.
- Number of branches on a tree, petals on a flower, spirals on a pineapple.
- [Vi Hart's video on Fibonacci numbers](http://www.youtube.com/watch?v=ahXIMUkSXX0)  
(<http://www.youtube.com/watch?v=ahXIMUkSXX0>)




# Recursive Fibonacci

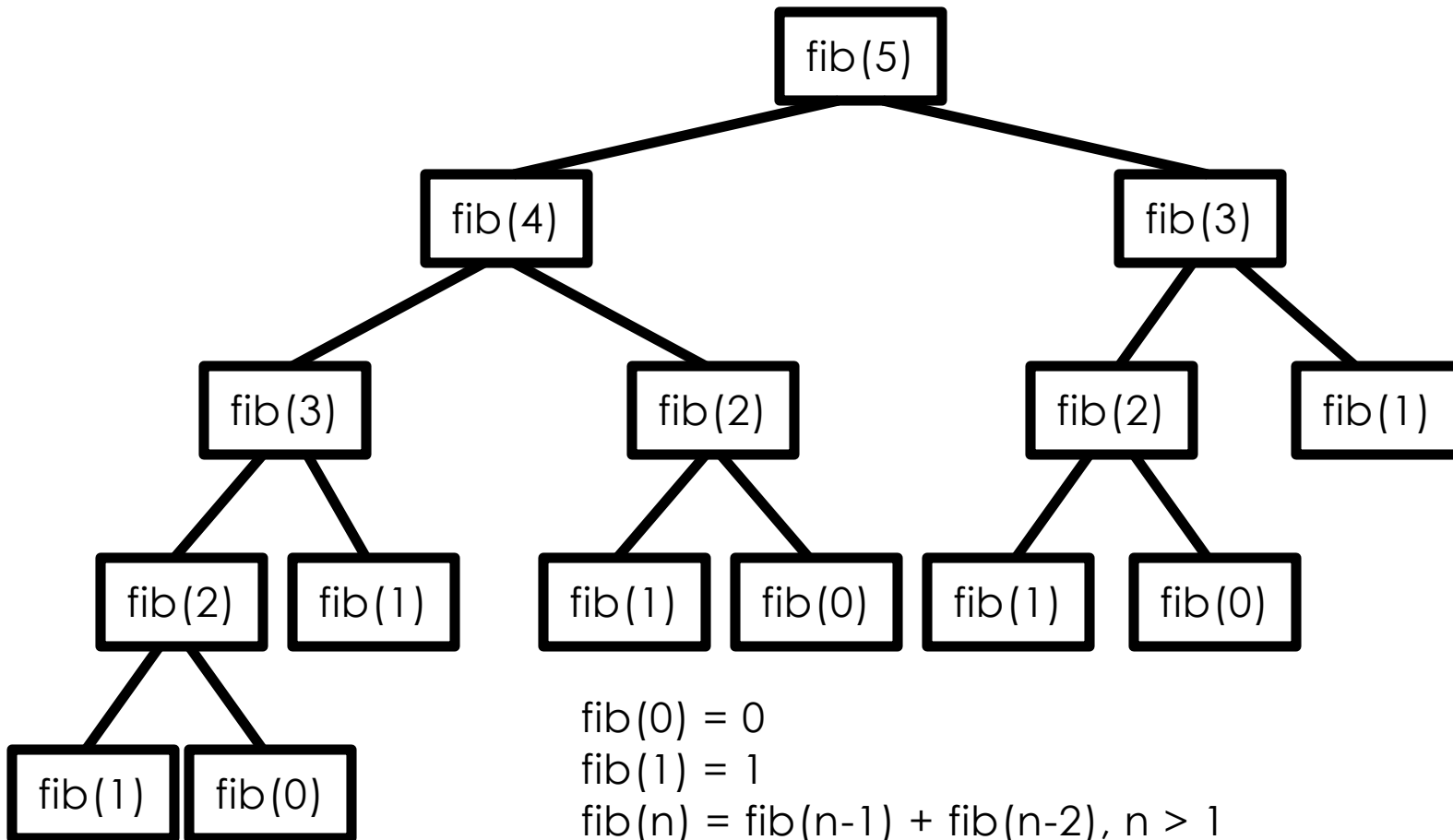
- ▣ Let  $\text{fib}(n)$  = the  $n$ th Fibonacci number,  $n \geq 0$ 
  - $\text{fib}(0) = 0$  (base case)
  - $\text{fib}(1) = 1$  (base case)
  - $\text{fib}(n) = \text{fib}(n-1) + \text{fib}(n-2), n > 1$

```
def fib(n):  
    if n == 0 or n == 1:  
        return n  
    else:  
        return fib(n-1) + fib(n-2)
```

**Two** recursive calls!

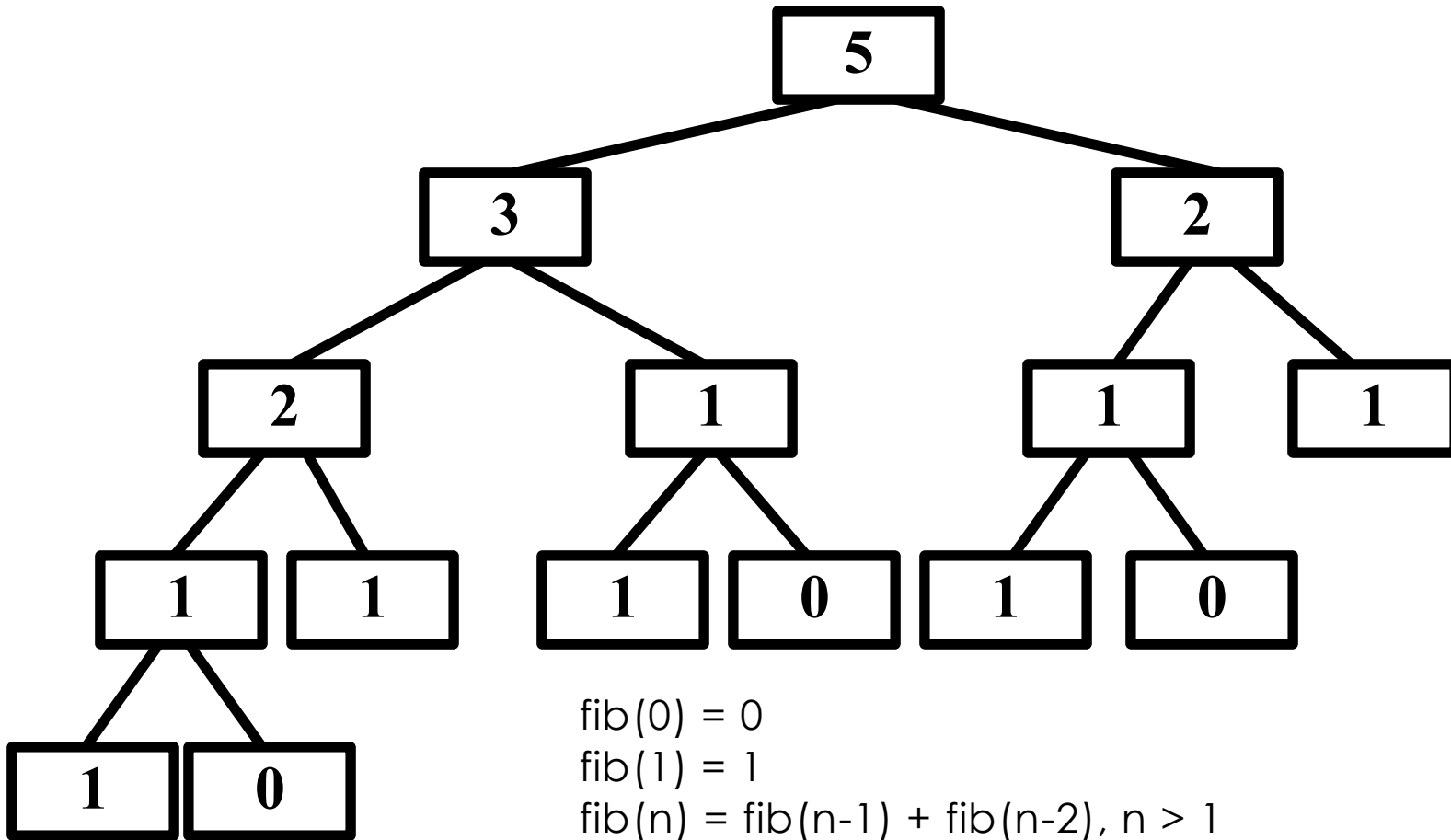


# Recursive Call Tree





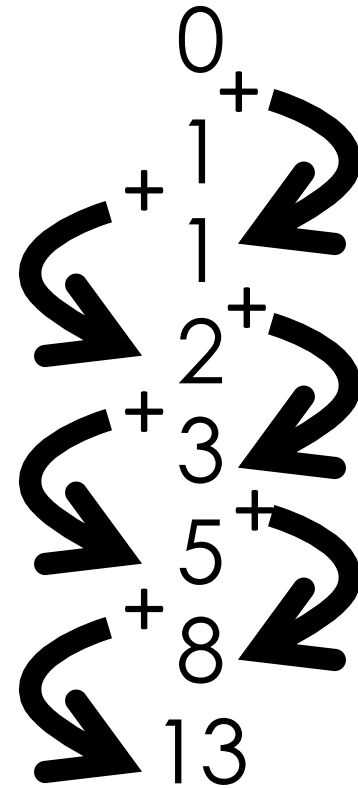
# Recursive Call Tree



# Iterative Fibonacci

```
def fib(n):  
    x = 0  
    next_x = 1  
    for i in range(1, n+1):  
        old_x = x  
        x = next_x  
        next_x = old_x + x  
    return x
```

sequence:



...

# Simultaneous Assignment

Assign values to multiple variables in a single statement:

```
sum, diff = x + y, x - y
```

```
x, y = y, x
```

# Iterative Fibonacci

```
def fib(n):  
    x = 0  
    next_x = 1  
    for i in range(1, n+1):  
        x, next_x = next_x, x + next_x  
    return x
```

**SIMULTANEOUS  
ASSIGNMENT**



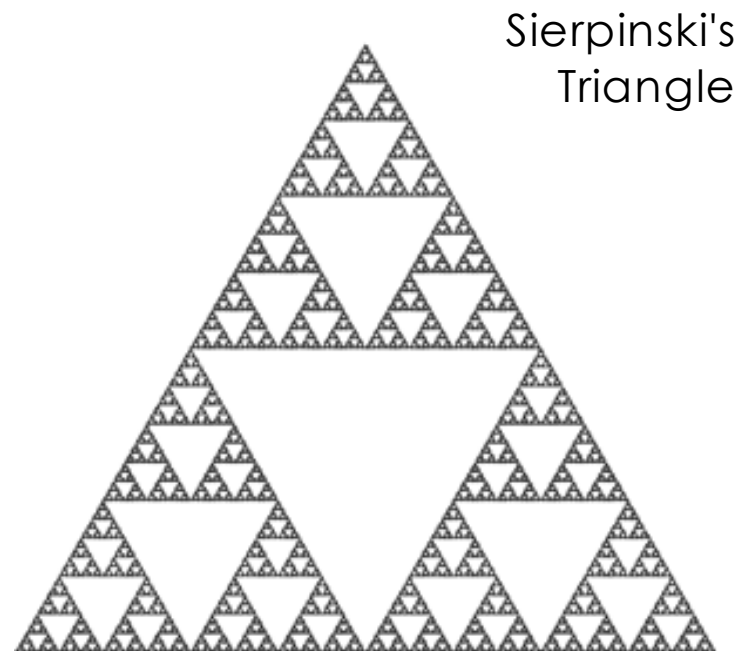
**Faster than the  
recursive  
version. Why?**

# Geometric Recursion (Fractals)

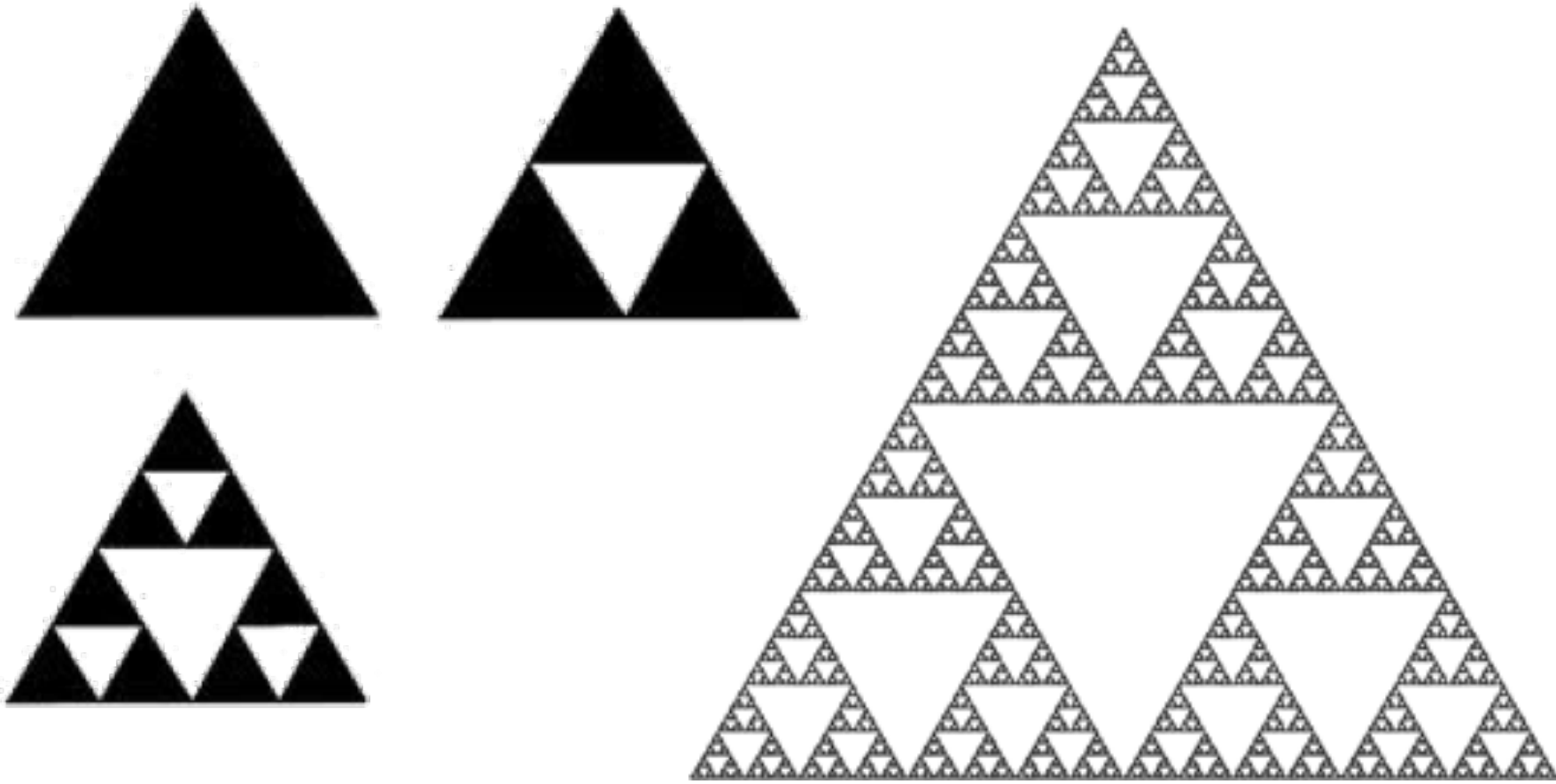
- A recursive operation performed on successively smaller regions.



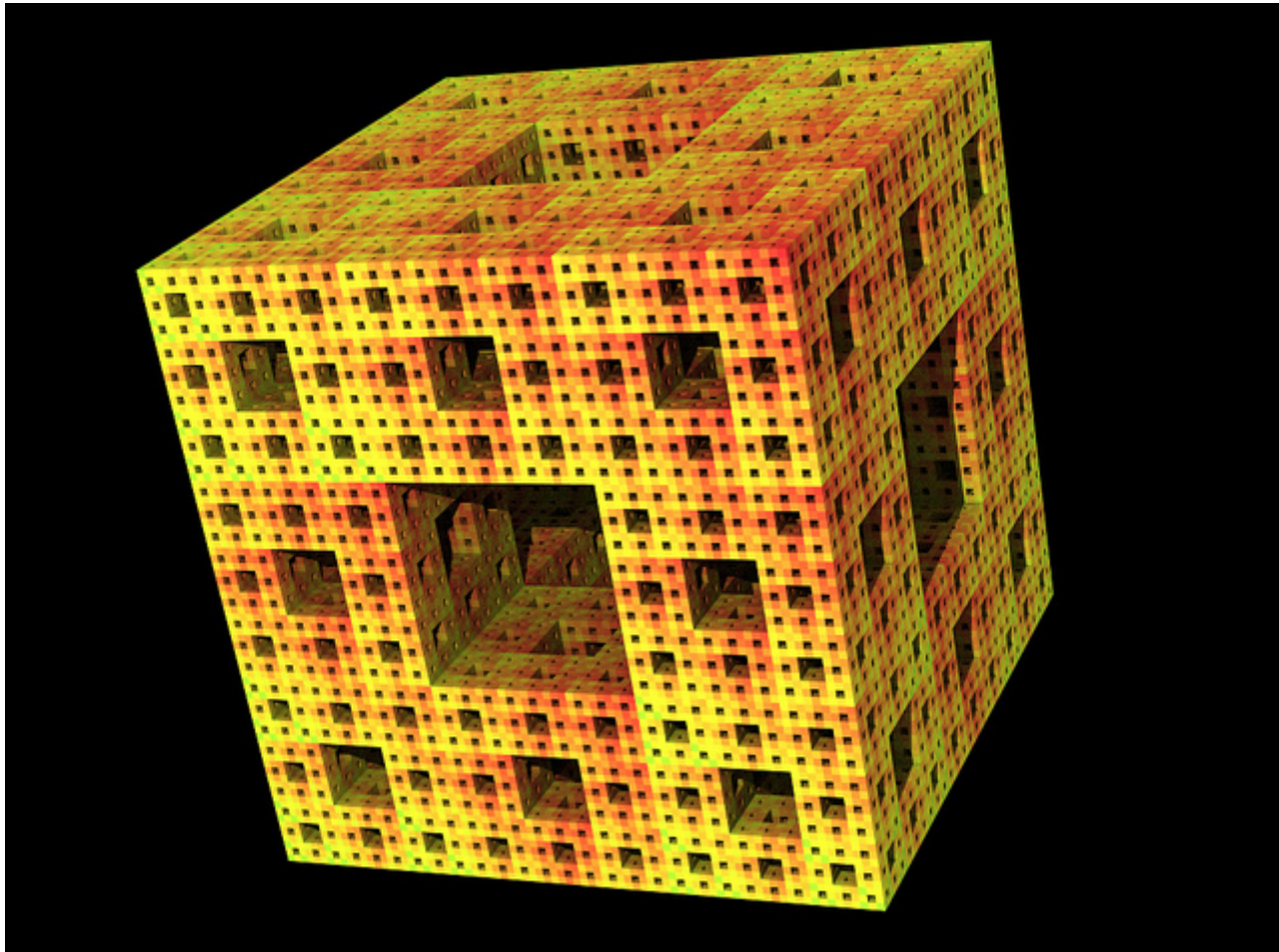
<http://fusionanomaly.net/recursion.jpg>



# Sierpinski's Triangle



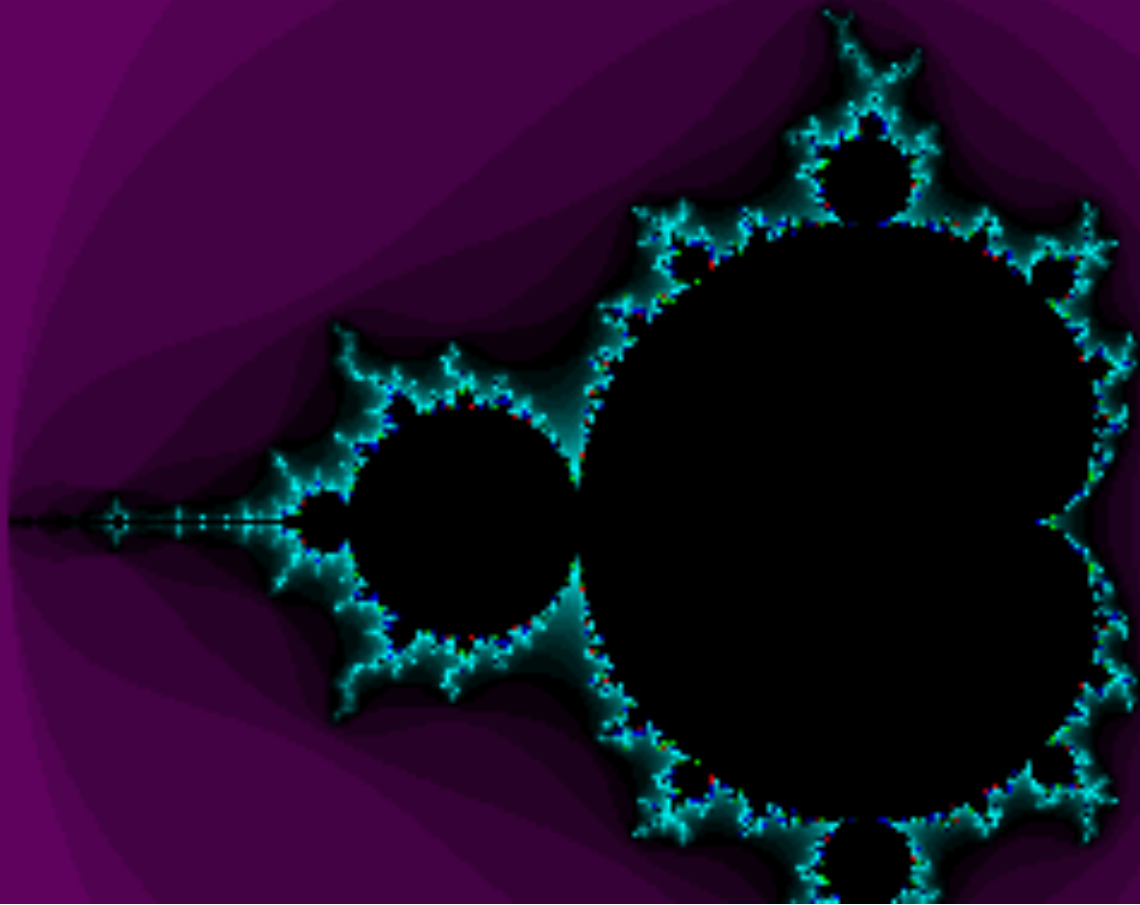
# Sierpinski's Carpet



(the next slide shows an animation that could give some people headaches)

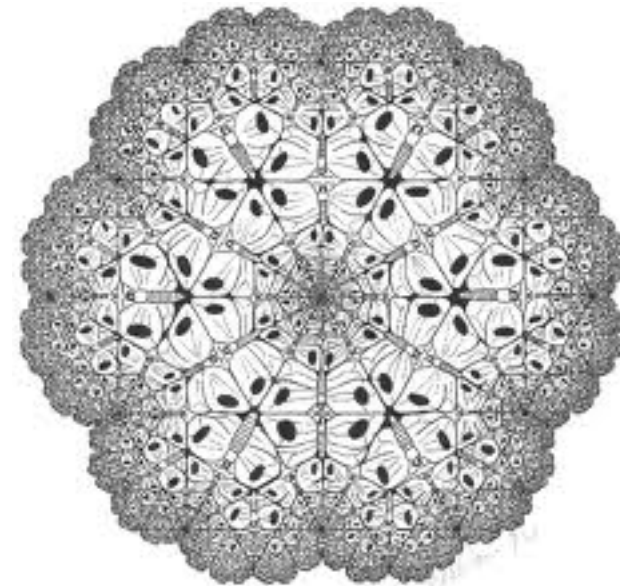
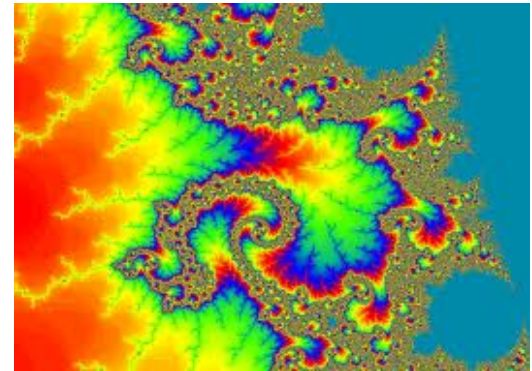
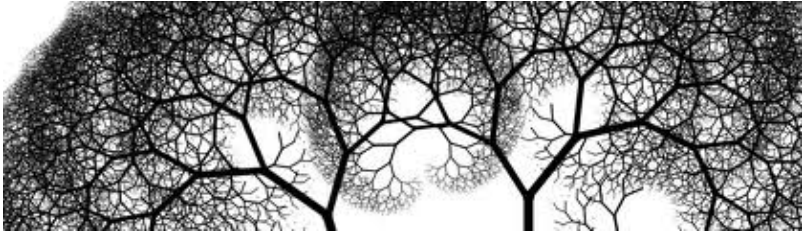


# Mandelbrot set



Source: Clint Sprott, <http://sprott.physics.wisc.edu/fractals/animated/>

# Fancier fractals



# Next Lecture

recursion for  
search

