



# Recursion

## Chapter 5

# Outline

- Induction
- Linear recursion
  - Example 1: Factorials
  - Example 2: Powers
  - Example 3: Reversing an array
- Binary recursion
  - Example 1: The Fibonacci sequence
  - Example 2: The Tower of Hanoi
- Drawbacks and pitfalls of recursion

# Outcomes

- By understanding this lecture you should be able to:
  - Use induction to prove the correctness of a recursive algorithm.
  - Identify the base case for an inductive solution
  - Design and analyze linear and binary recursion algorithms
  - Identify the overhead costs of recursion
  - Avoid errors commonly made in writing recursive algorithms

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- **Induction**
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# Divide and Conquer

- When faced with a difficult problem, a classic technique is to break it down into smaller parts that can be solved more easily.
- Recursion uses induction to do this.



# History of Induction

- Implicit use of induction goes back at least to Euclid's proof that the number of primes is infinite (c. 300 BC).
- The first explicit formulation of the principle is due to Pascal (1665).



Euclid of Alexandria,  
"The Father of Geometry"  
c. 300 BC



Blaise Pascal, 1623 - 1662

# Induction: Review

- Induction is a mathematical method for proving that a statement is true for a (possibly infinite) sequence of objects.
- There are two things that must be proved:
  1. **The Base Case:** The statement is true for the first object
  2. **The Inductive Step:** If the statement is true for a given object, it is also true for the next object.
- If these two statements hold, then the statement holds for all objects.

# Induction Example: An Arithmetic Sum

- Claim:  $\sum_{i=0}^n i = \frac{1}{2}n(n+1) \quad " \quad n \in \mathbb{N}$
- Proof:

**1. Base Case.** The statement holds for  $n = 0$ :

$$\sum_{i=0}^n i = \sum_{i=0}^0 i = 0$$

$$\frac{1}{2}n(n+1) = \frac{1}{2}0(0+1) = 0$$



**1. Inductive Step.** If the claim holds for  $n = k$ , then it also holds for  $n = k+1$ .

$$\sum_{i=0}^{k+1} i = k+1 + \sum_{i=0}^k i = k+1 + \frac{1}{2}k(k+1) = \frac{1}{2}(k+1)(k+2) \quad \checkmark$$



# Recursive Divide and Conquer

- You are given a problem input that is too big to solve directly.
- You imagine,
  - “Suppose I had a friend who could give me the answer to the same problem with slightly smaller input.”
  - “Then I could easily solve the larger problem.”
- In recursion this “friend” will actually be another instance (clone) of yourself.

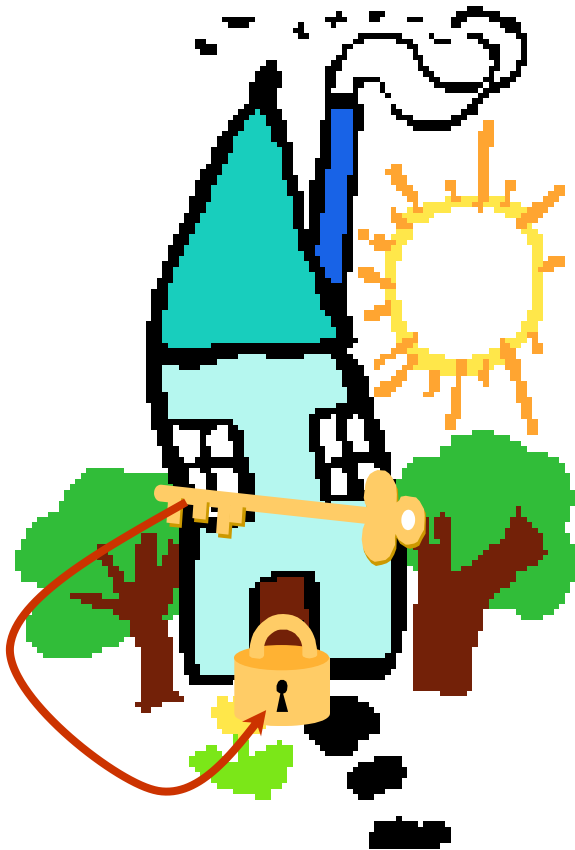


Tai (left) and Snuppy (right): the first puppy clone.

# Friends & Induction

Recursive Algorithm:

- Assume you have an algorithm that works.
- Use it to write an algorithm that works.



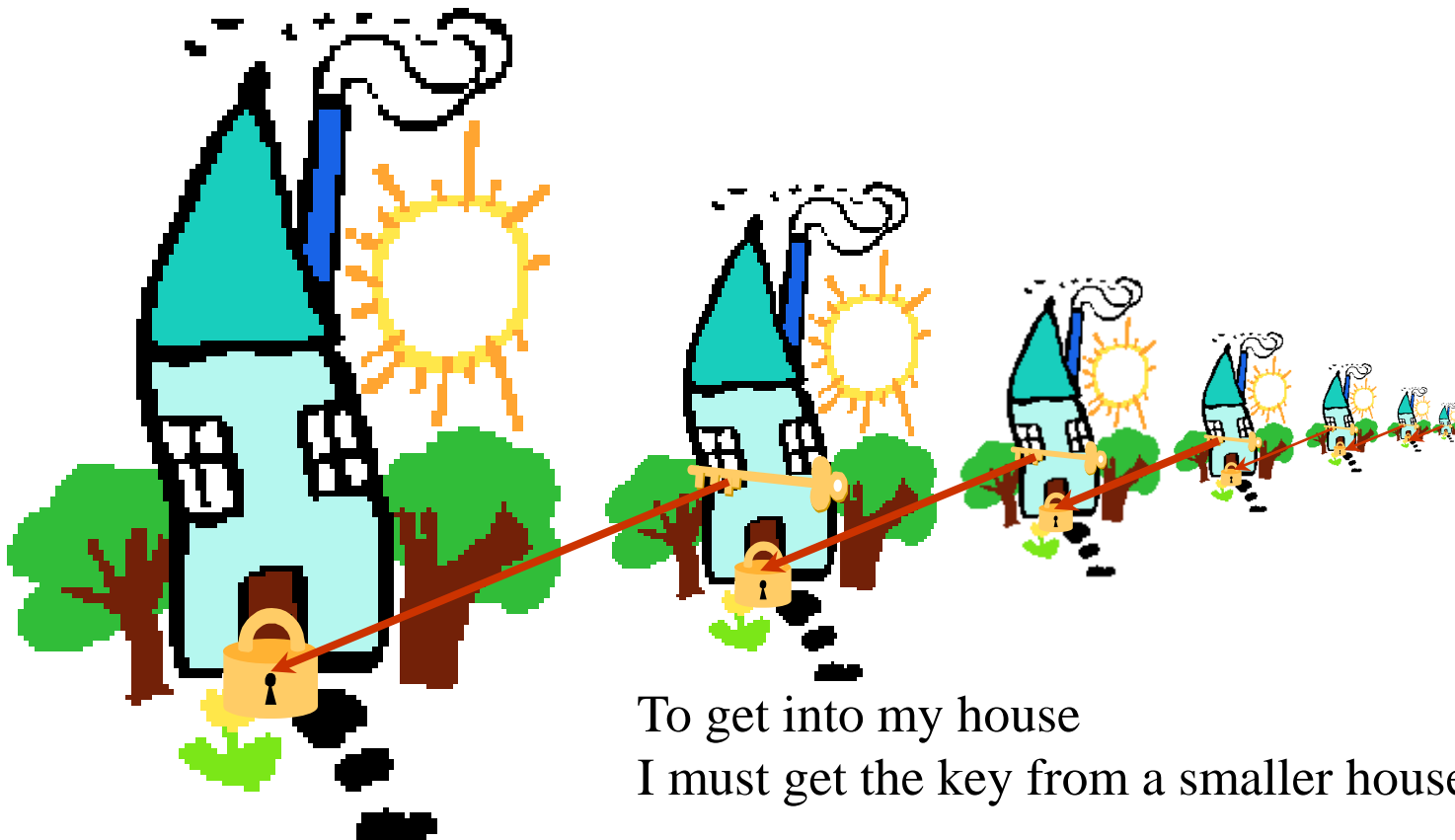
If I could get in,  
I could get the key.  
Then I could unlock the door  
so that I can get in.

**Circular Argument!**

# Friends & Induction

Recursive Algorithm:

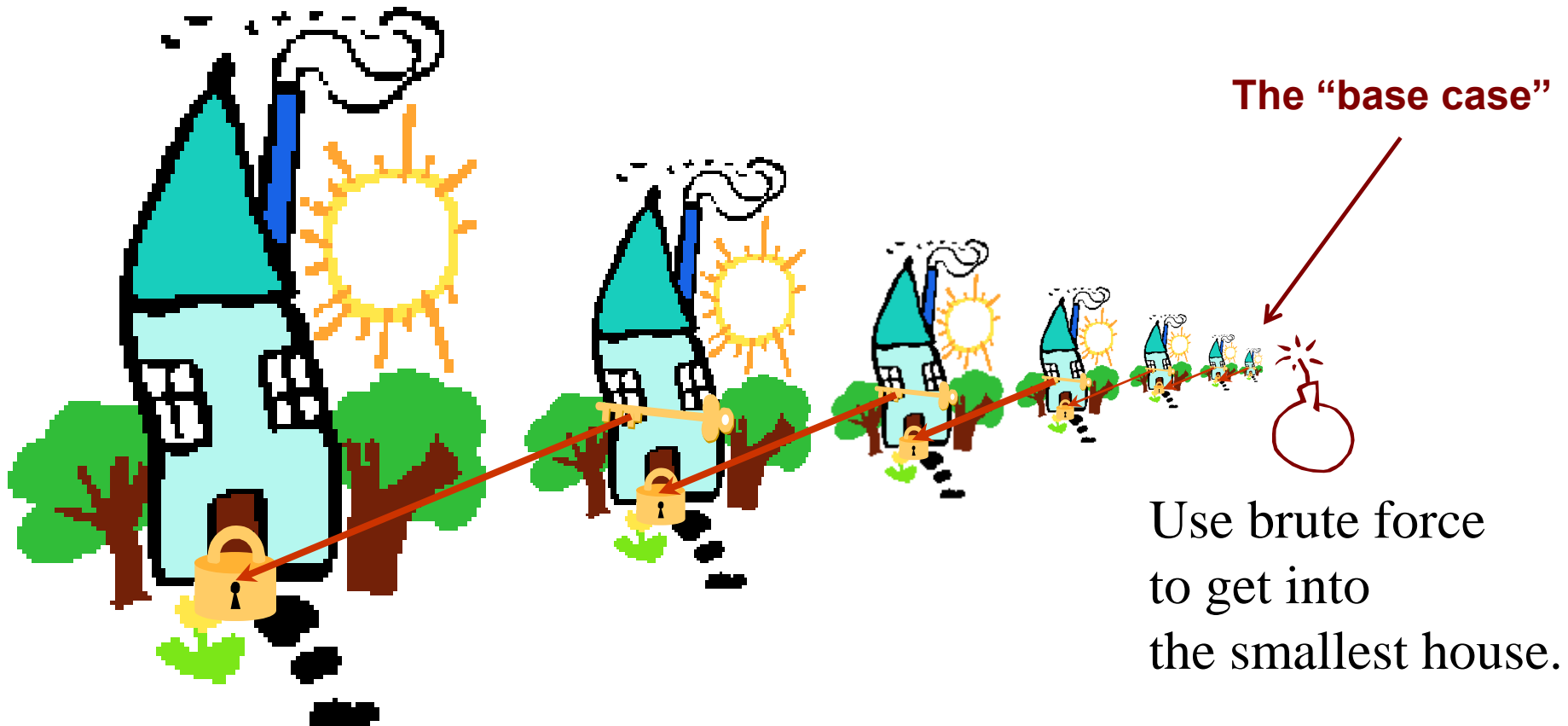
- Assume you have an algorithm that works.
- Use it to write an algorithm that works.



# Friends & Induction

Recursive Algorithm:

- Assume you have an algorithm that works.
- Use it to write an algorithm that works.



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# Recall: Design Pattern

- A template for a software solution that can be applied to a variety of situations.
- Main elements of solution are described in the abstract.
- Can be specialized to meet specific circumstances.

# Linear Recursion Design Pattern

- **Test for base cases**

- Begin by testing for a set of base cases (there should be at least one).
- Every possible chain of recursive calls **must** eventually reach a base case, and the handling of each base case should not use recursion.

- ***Recurse once***

- Perform a single recursive call. (This recursive step may involve a test that decides which of several possible recursive calls to make, but it should ultimately choose to make just one of these calls each time we perform this step.)
- Define each possible recursive call so that it makes **progress** towards a base case.

# Example 1

- The factorial function:
  - $n! = 1 \cdot 2 \cdot 3 \cdot \dots \cdot (n-1) \cdot n$

- Recursive definition:

$$f(n) = \begin{cases} 1 & \text{if } n = 0 \\ n \cdot f(n-1) & \text{else} \end{cases}$$

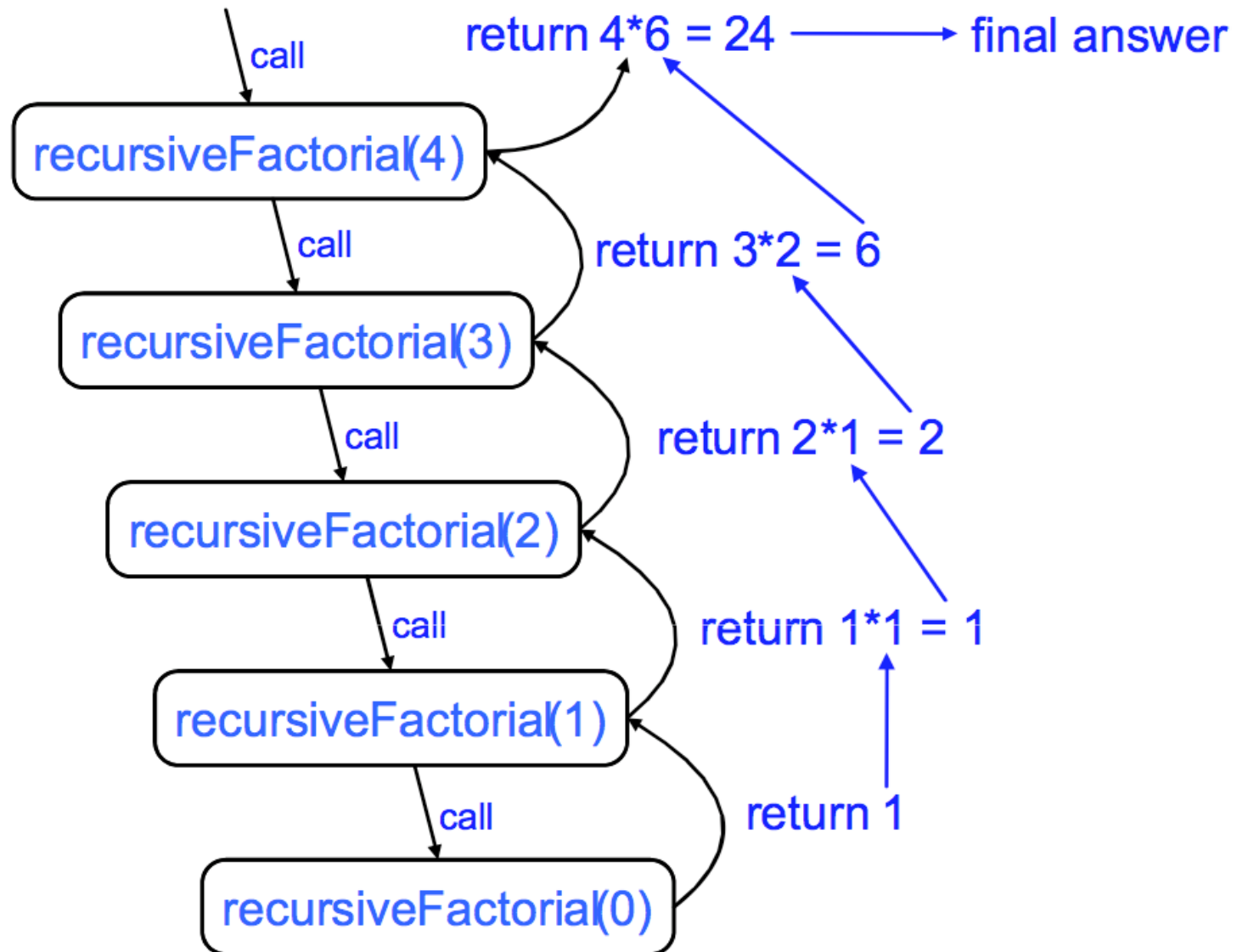
- As a Java method:

// recursive factorial function

```
public static int recursiveFactorial(int n) {  
    if (n == 0) return 1;    // base case  
    else return n * recursiveFactorial(n-1); // recursive case  
}
```



# Tracing Recursion



# Linear Recursion

- recursiveFactorial is an example of **linear** recursion: only one recursive call is made per stack frame.
- Since there are  $n$  recursive calls, this algorithm has  $O(n)$  run time.

// recursive factorial function

```
public static int recursiveFactorial(int n) {  
    if (n == 0) return 1;    // base case  
    else return n * recursiveFactorial(n-1); // recursive case  
}
```

## Example 2: Computing Powers

- The power function,  **$p(x,n) = x^n$** , can be defined recursively:

$$p(x,n) = \begin{cases} 1 & \text{if } n = 0 \\ x \cdot p(x,n-1) & \text{otherwise} \end{cases}$$

- Assume multiplication takes constant time (independent of value of arguments).
- This leads to a power function that runs in  $O(n)$  time (for we make  $n$  recursive calls).
- Can we do better than this?**

# Recursive Squaring

- We can derive a more efficient linearly recursive algorithm by using repeated squaring:

$$p(x,n) = \begin{cases} 1 & \text{if } n = 0 \\ x \cdot p(x, (n-1)/2)^2 & \text{if } n > 0 \text{ is odd} \\ p(x, n/2)^2 & \text{if } n > 0 \text{ is even} \end{cases}$$

- For example,

$$2^4 = 2^{(4/2)^2} = (2^{4/2})^2 = (2^2)^2 = 4^2 = 16$$

$$2^5 = 2^{1+(4/2)^2} = 2(2^{4/2})^2 = 2(2^2)^2 = 2(4^2) = 32$$

$$2^6 = 2^{(6/2)^2} = (2^{6/2})^2 = (2^3)^2 = 8^2 = 64$$

$$2^7 = 2^{1+(6/2)^2} = 2(2^{6/2})^2 = 2(2^3)^2 = 2(8^2) = 128.$$

# A Recursive Squaring Method

**Algorithm** Power( $x, n$ ):

**Input:** A number  $x$  and integer  $n$

**Output:** The value  $x^n$

**if**  $n = 0$  **then**

**return** 1

**if**  $n$  is odd **then**

$y = \text{Power}(x, (n - 1)/2)$

**return**  $x \cdot y \cdot y$

**else**

$y = \text{Power}(x, n/2)$

**return**  $y \cdot y$

# Analyzing the Recursive Squaring Method

**Algorithm** Power( $x, n$ ):

**Input:** A number  $x$  and integer  $n = 0$

**Output:** The value  $x^n$

```
if  $n = 0$  then
    return 1
if  $n$  is odd then
     $y = \text{Power}(x, (n - 1)/2)$ 
    return  $x \cdot y \cdot y$ 
else
     $y = \text{Power}(x, n/2)$ 
    return  $y \cdot y$ 
```

Although there are 2 statements that recursively call Power, only one is executed per stack frame.

Each time we make a recursive call we halve the value of  $n$  (roughly).

Thus we make a total of  $\log n$  recursive calls. That is, this method runs in  $O(\log n)$  time.

# Tail Recursion

- Tail recursion occurs when a linearly recursive method makes its recursive call as its **last** step.
- Such a method can easily be converted to an iterative methods (which saves on some resources).

# Example: Recursively Reversing an Array

**Algorithm** ReverseArray( $A, i, j$ ):

***Input:*** An array  $A$  and nonnegative integer indices  $i$  and  $j$

***Output:*** The reversal of the elements in  $A$  starting at index  $i$  and ending at  $j$

**if**  $i < j$  **then**

    Swap  $A[i]$  and  $A[j]$

    ReverseArray( $A, i + 1, j - 1$ )

**return**



# Example: Iteratively Reversing an Array

**Algorithm** IterativeReverseArray( $A, i, j$ ):

***Input:*** An array  $A$  and nonnegative integer indices  $i$  and  $j$

***Output:*** The reversal of the elements in  $A$  starting at index  $i$  and ending at  $j$

**while**  $i < j$  **do**

    Swap  $A[i]$  and  $A[j]$

$i = i + 1$

$j = j - 1$

**return**

# Defining Arguments for Recursion

- Solving a problem recursively sometimes requires passing additional parameters.
- **ReverseArray** is a good example: although we might initially think of passing only the array **A** as a parameter at the top level, lower levels need to know where in the array they are operating.
- Thus the recursive interface is **ReverseArray(A, i, j)**.
- We then invoke the method at the highest level with the message **ReverseArray(A, 1, n)**.

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# Binary Recursion

- Binary recursion occurs whenever there are **two** recursive calls for each non-base case.
- Example 1: **The Fibonacci Sequence**

# The Fibonacci Sequence

- Fibonacci numbers are defined recursively:

$$F_0 = 0$$

$$F_1 = 1$$

$$F_i = F_{i-1} + F_{i-2} \quad \text{for } i > 1.$$

The ratio  $F_i / F_{i-1}$  converges to  $\phi = \frac{1 + \sqrt{5}}{2} = 1.61803398874989\dots$

(The “**Golden Ratio**”)

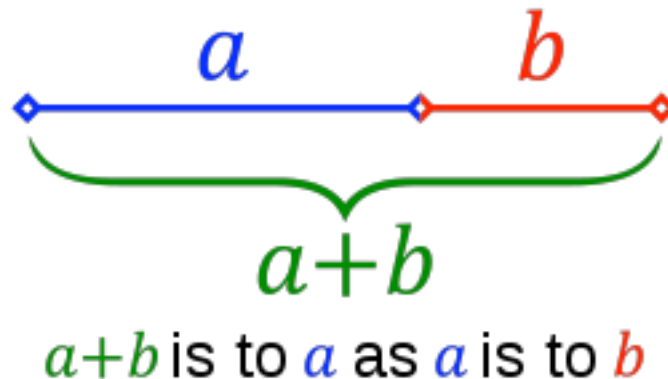


Fibonacci (c. 1170 - c. 1250)  
(aka Leonardo of Pisa)

# The Golden Ratio

- Two quantities are in the **golden ratio** if the ratio of the sum of the quantities to the larger quantity is equal to the ratio of the larger quantity to the smaller one.

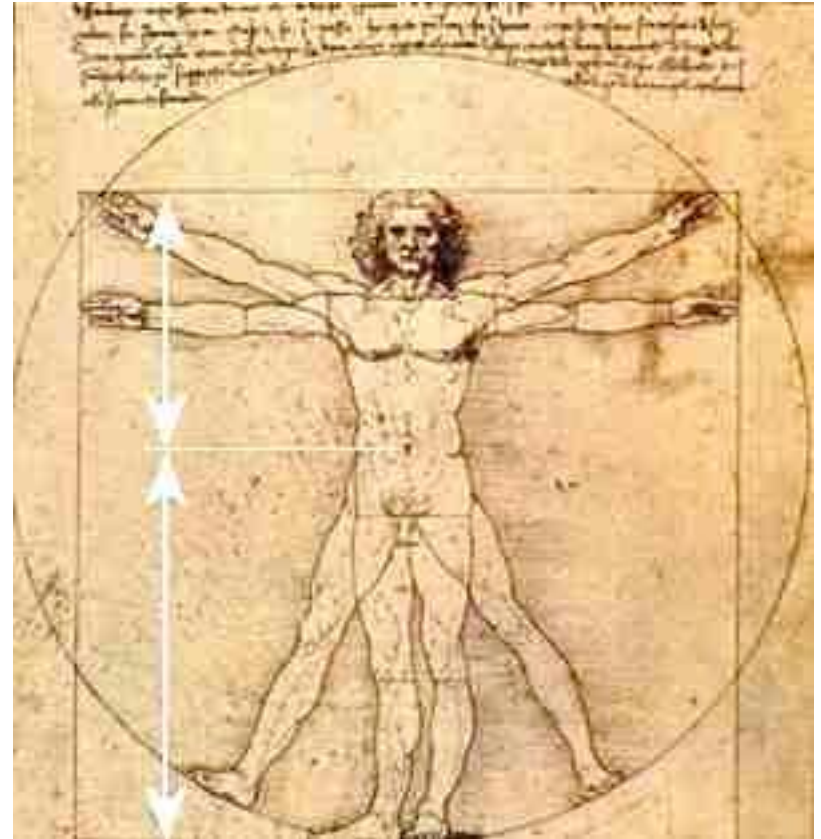
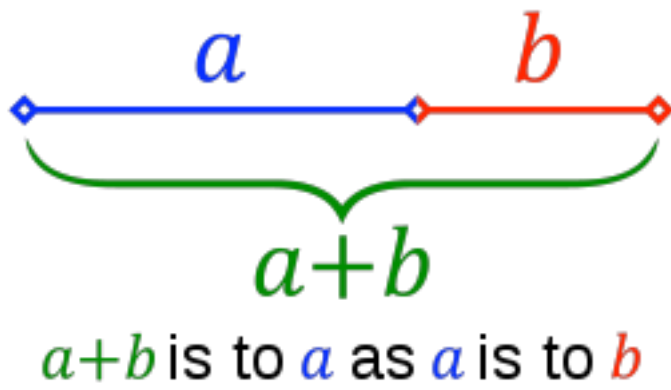
$j$  is the unique positive solution to  $j = \frac{a+b}{a} = \frac{a}{b}$ .



# The Golden Ratio



The Parthenon



Leonardo

# Computing Fibonacci Numbers

$$F_0 = 0$$

$$F_1 = 1$$

$$F_i = F_{i-1} + F_{i-2} \quad \text{for } i > 1.$$

- A recursive algorithm (first attempt):

**Algorithm** BinaryFib( $k$ ):

**Input:** Positive integer  $k$

**Output:** The  $k$ th Fibonacci number  $F_k$

**if**  $k < 2$  **then**

**return**  $k$

**else**

**return** BinaryFib( $k - 1$ ) + BinaryFib( $k - 2$ )



# Analyzing the Binary Recursion Fibonacci Algorithm

- Let  $n_k$  denote number of recursive calls made by BinaryFib(k).  
Then
  - $n_0 = 1$
  - $n_1 = 1$
  - $n_2 = n_1 + n_0 + 1 = 1 + 1 + 1 = 3$
  - $n_3 = n_2 + n_1 + 1 = 3 + 1 + 1 = 5$
  - $n_4 = n_3 + n_2 + 1 = 5 + 3 + 1 = 9$
  - $n_5 = n_4 + n_3 + 1 = 9 + 5 + 1 = 15$
  - $n_6 = n_5 + n_4 + 1 = 15 + 9 + 1 = 25$
  - $n_7 = n_6 + n_5 + 1 = 25 + 15 + 1 = 41$
  - $n_8 = n_7 + n_6 + 1 = 41 + 25 + 1 = 67.$
- Note that  $n_k$  more than doubles for every other value of  $n_k$ . That is,  $n_k > 2^{k/2}$ . It increases exponentially!

# A Better Fibonacci Algorithm

- Use **linear** recursion instead:

**Algorithm** LinearFibonacci( $k$ ):

**Input:** A positive integer  $k$

**Output:** Pair of Fibonacci numbers ( $F_k, F_{k-1}$ )

**if**  $k = 1$  **then**

**return** ( $k, 0$ )

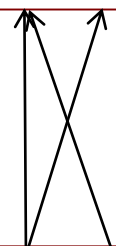
**else**

$(i, j) = \text{LinearFibonacci}(k - 1)$

**return** ( $i + j, i$ )

LinearFibonacci( $k$ ):  $F_k, F_{k-1}$

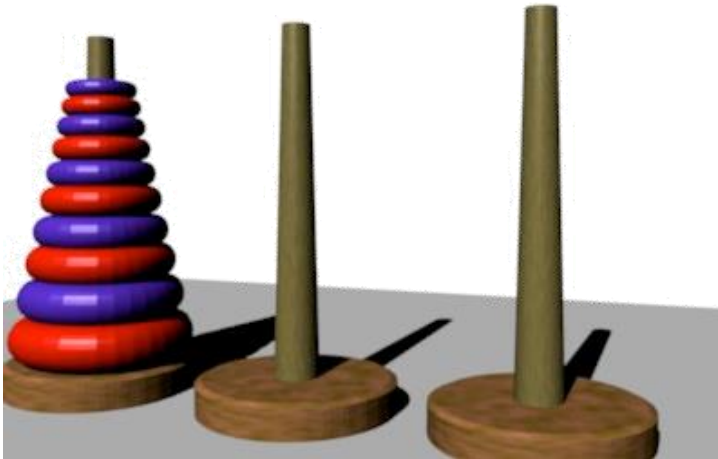
LinearFibonacci( $k-1$ ):  $F_{k-1}, F_{k-2}$



- Runs in  **$O(k)$**  time.

# Binary Recursion

- Second Example: **The Tower of Hanoi**



Example

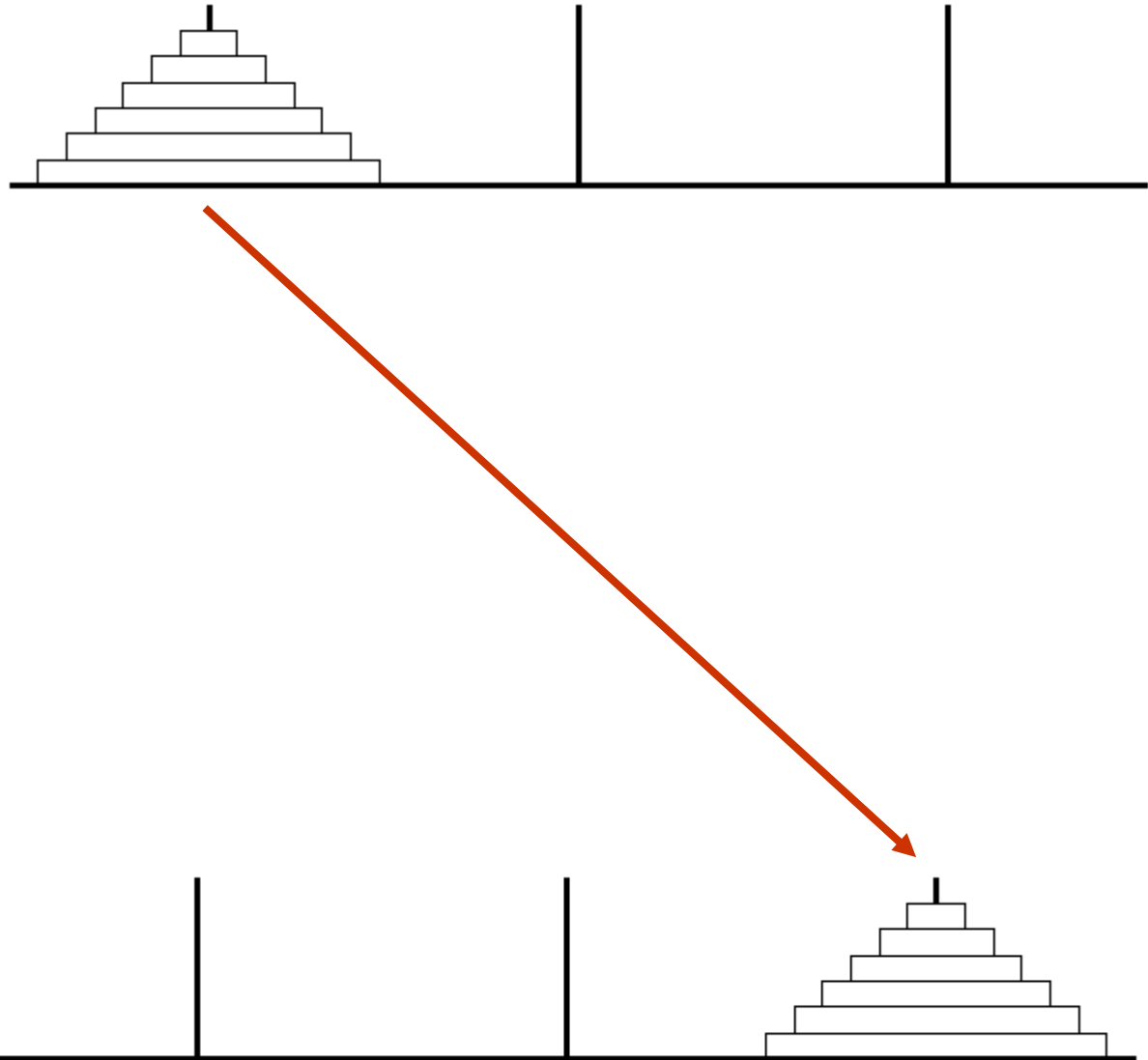
# Tower of Hanoi



This job of mine  
is a bit daunting.  
Where do I start?

And I am lazy.

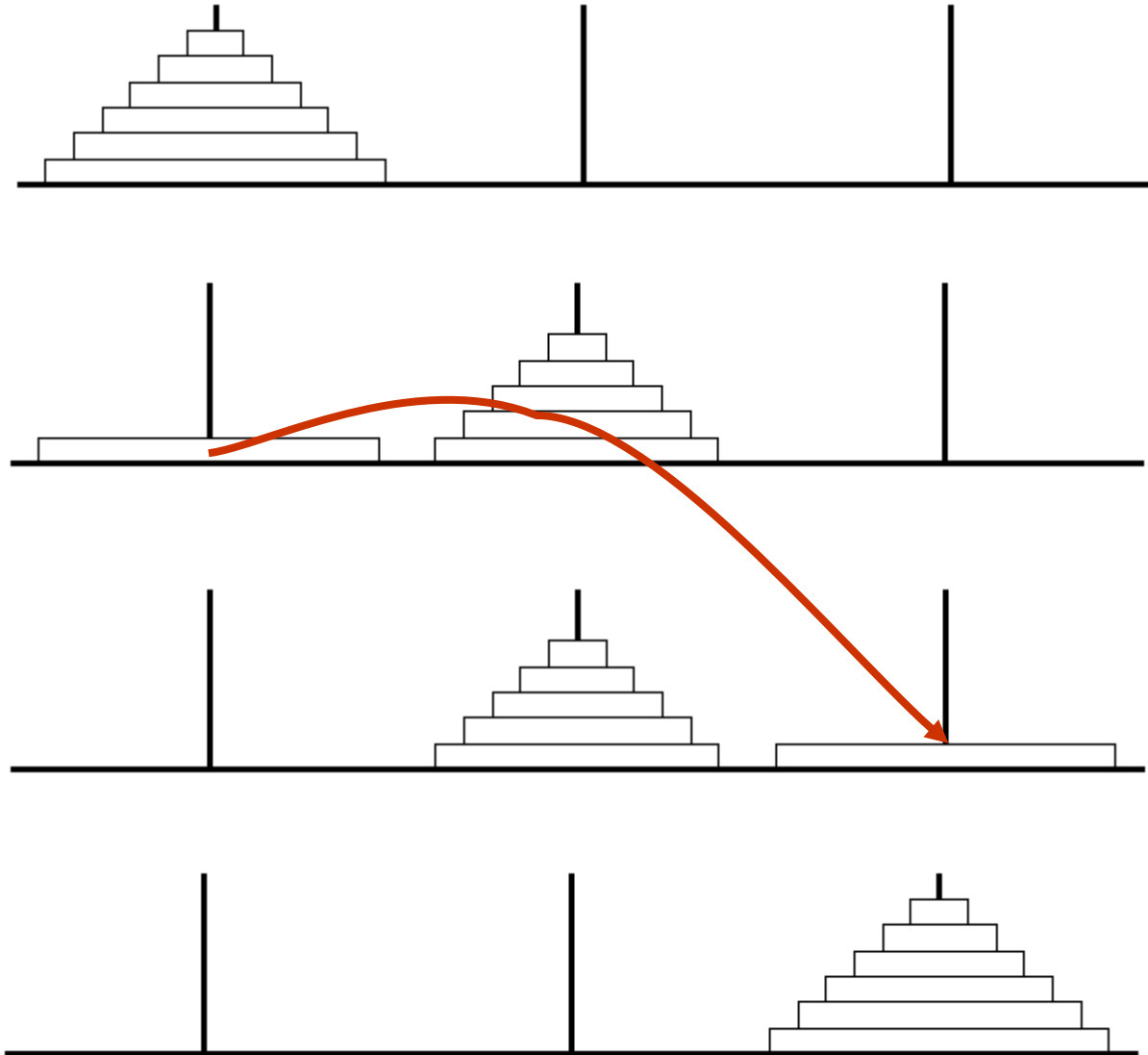
Example from J. Edmonds – Thanks Jeff!



# Tower of Hanoi



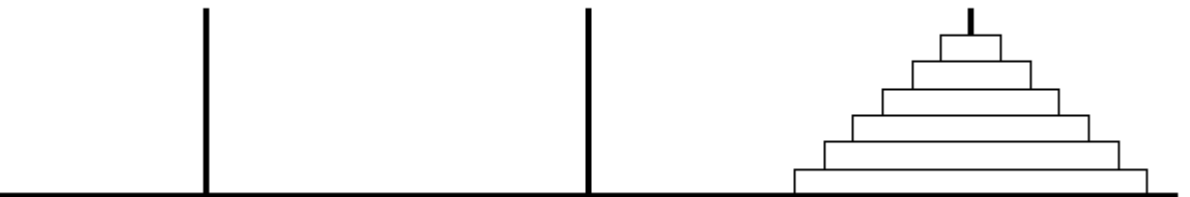
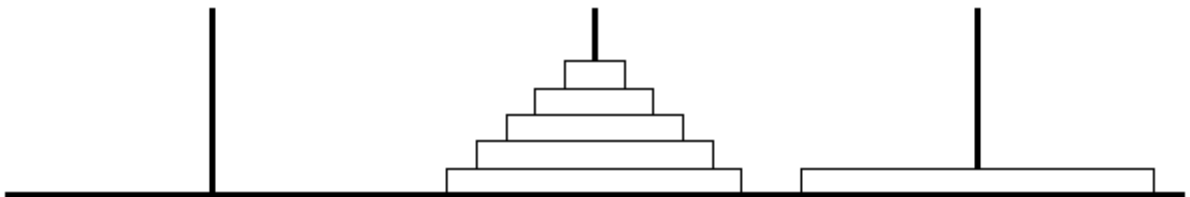
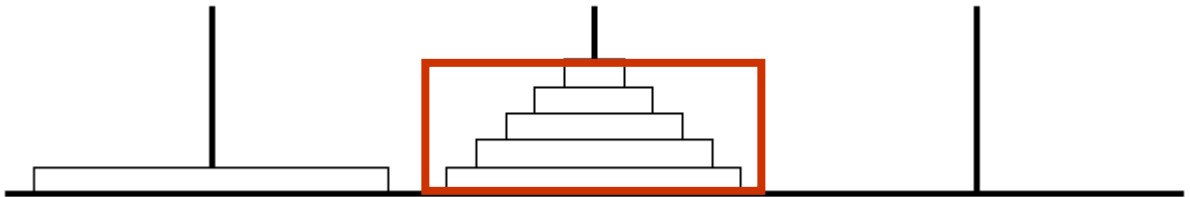
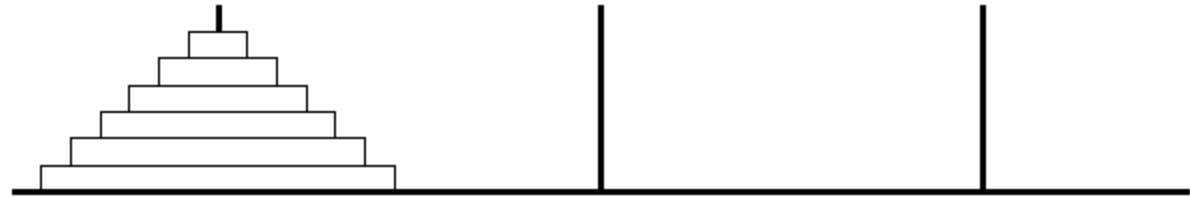
At some point,  
the biggest disk  
moves.  
I will do that job.



# Tower of Hanoi



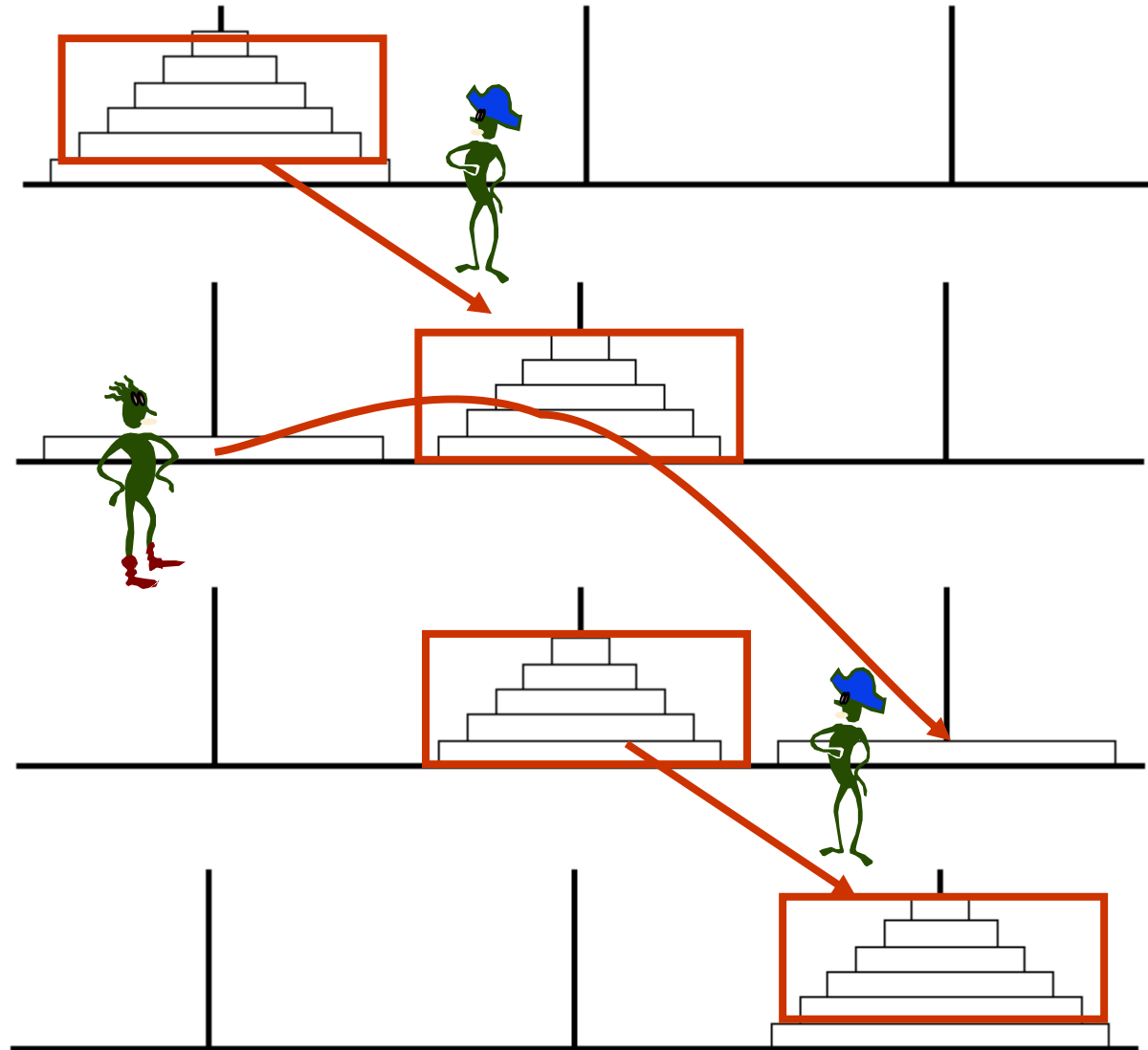
To do this,  
the other disks  
must be in the  
middle.



# Tower of Hanoi

How will these  
move?

I will get a  
friend to do it.  
And another to  
move these.  
I only move the  
big disk.





# Tower of Hanoi

Code:

algorithm *TowersOfHanoi*( $n$ ,  $source$ ,  $destination$ ,  $spare$ )

$\langle pre-cond \rangle$ : The  $n$  smallest disks are on  $pole_{source}$ .

$\langle post-cond \rangle$ : They are moved to  $pole_{destination}$ .

begin

if( $n = 1$ )

Move the single disk from  $pole_{source}$  to  $pole_{destination}$ .

else

*TowersOfHanoi*( $n - 1$ ,  $source$ ,  $spare$ ,  $destination$ )

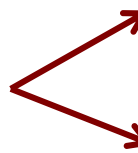
Move the  $n^{th}$  disk from  $pole_{source}$  to  $pole_{destination}$ .

*TowersOfHanoi*( $n - 1$ ,  $spare$ ,  $destination$ ,  $source$ )

end if

end algorithm

**2 recursive  
calls!**



# Tower of Hanoi

Code:

**algorithm** *TowersOfHanoi*(*n*, *source*, *destination*, *spare*)

$\langle pre-cond \rangle$ : The *n* smallest disks are on *pole<sub>source</sub>*.

$\langle post-cond \rangle$ : They are moved to *pole<sub>destination</sub>*.

  begin

    if(*n* = 1)

      Move the single disk from *pole<sub>source</sub>* to *pole<sub>destination</sub>*.

    else

*TowersOfHanoi*(*n* - 1, *source*, *spare*, *destination*)

      Move the *n<sup>th</sup>* disk from *pole<sub>source</sub>* to *pole<sub>destination</sub>*.

*TowersOfHanoi*(*n* - 1, *spare*, *destination*, *source*)

    end if

  end algorithm

Time:

$$T(1) = 1,$$

$$T(n) = 1 + 2T(n-1) \approx 2T(n-1)$$

$$\approx 2(2T(n-2)) \approx 4T(n-2)$$

$$\approx 4(2T(n-3)) \approx 8T(n-3)$$

$$\approx 2^i T(n-i)$$

$$\approx 2^n$$

# Binary Recursion: Summary

- Binary recursion spawns an exponential number of recursive calls.
- If the inputs are only declining **arithmetically** (e.g.,  $n-1$ ,  $n-2$ , ...) the result will be an exponential running time.
- In order to use binary recursion, the input must be declining **geometrically** (e.g.,  $n/2$ ,  $n/4$ , ...).
- We will see efficient examples of binary recursion with geometrically declining inputs when we discuss **heaps** and **sorting**.

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# The Overhead Costs of Recursion

- Many problems are naturally defined recursively.
- This can lead to simple, elegant code.
- However, recursive solutions entail a **cost in time and memory**: each recursive call requires that the current process state (variables, program counter) be **pushed** onto the system stack, and **popped** once the recursion unwinds.
- This typically affects the running time **constants**, but **not** the **asymptotic time complexity** (e.g.,  $O(n)$ ,  $O(n^2)$  etc.)
- Thus **recursive solutions may still be preferred** unless there are very strict time/memory constraints.

# The “Curse” in Recursion: Errors to Avoid

```
// recursive factorial function
```

```
public static int recursiveFactorial(int n) {  
    return n * recursiveFactorial(n-1);  
}
```

- **There must be a base condition: the recursion must ground out!**

# The “Curse” in Recursion: Errors to Avoid

// recursive factorial function

```
public static int recursiveFactorial(int n) {  
    if (n == 0) return recursiveFactorial(n); // base case  
    else return n * recursiveFactorial(n- 1); // recursive case  
}
```

- **The base condition must not involve more recursion!**

# The “Curse” in Recursion: Errors to Avoid

// recursive factorial function

```
public static int recursiveFactorial(int n) {  
    if (n == 0) return 1;    // base case  
    else return (n - 1) * recursiveFactorial(n);    // recursive  
    case  
}
```

- The input **must be converging** toward the base condition!



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# Outcomes

- By understanding this lecture you should be able to:
  - Use induction to prove the correctness of a recursive algorithm.
  - Identify the base case for an inductive solution
  - Design and analyze linear and binary recursion algorithms
  - Identify the overhead costs of recursion
  - Avoid errors commonly made in writing recursive algorithms