

# Recursion



- Recursion is more than just a programming technique. It has two other uses in computer science and software engineering, namely:
  - as a way of describing, defining, or specifying things.
  - as a way of designing solutions to problems (divide and conquer).



# Recursive definitions

- A recursive definition is one in which something is defined in terms of itself
- Almost every algorithm that requires looping can be defined iteratively or recursively
- All recursive definitions require two parts:
  - *Base case*
  - *Recursive step*
- The recursive step is the one that is defined in terms of itself
- The recursive step must always move closer to the base case



- Factorial

- $n! = n * (n-1)!$

```
int fact(int n) {  
    if (n <= 1) return 1;  
    else return n * fact(n-1);  
} // fact
```

# Recursion Definition

- In general, we can define the factorial function in the following way:

$$\text{Factorial}(n) = \begin{cases} 1 & \text{if } n = 0 \\ n \times (n - 1) \times (n - 2) \times \dots \times 3 \times 2 \times 1 & \text{if } n > 0 \end{cases}$$

- 1! = 1**
    - 2! = 1 × 2 = 2**
    - 3! = 1 × 2 × 3 = 6**
    - 4! = 1 × 2 × 3 × 4 = 24**
    - 5! = 1 × 2 × 3 × 4 × 5 = 120**

# Iterative Definition



# Iteration vs. recursion

- Some things (e.g. reading from a file) are easier to implement iteratively
- Other things (e.g. mergesort) are easier to implement recursively
- Others are just as easy both ways
- It can be proved that two methods performing the same task, one implementing an iteration algorithm and one implementing a recursive version, are equivalent

# مضروب، Factorial

$$\text{Factorial}(5) = 5 * \text{Factorial}(4)$$
$$= 5 * 4 * 3 * 2 * 1$$

$$\text{Factorial}(4) = 4 * \text{Factorial}(3)$$
$$= 4 * 3 * 2 * 1$$

$$\text{Factorial}(3) = 3 * \text{Factorial}(2)$$
$$= 3 * 2 * 1$$

$$\text{Factorial}(2) = 2 * \text{Factorial}(1)$$
$$= 2 * 1$$

$$\text{Factorial}(1) = 1$$

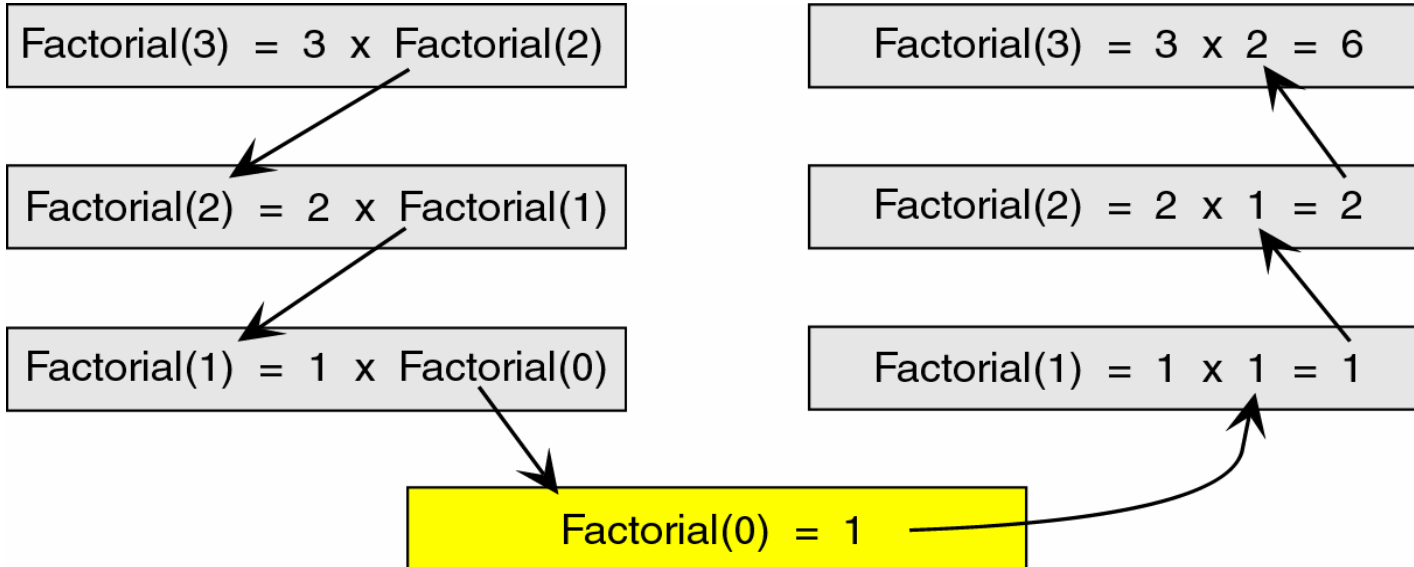
جزء متكررة (العدد ضرب  
factorial / العدد الذي قبله)

$$\text{Factorial}(x) = x * \text{Factorial}(x-1)$$

جزء يوقف فيه التكرار base

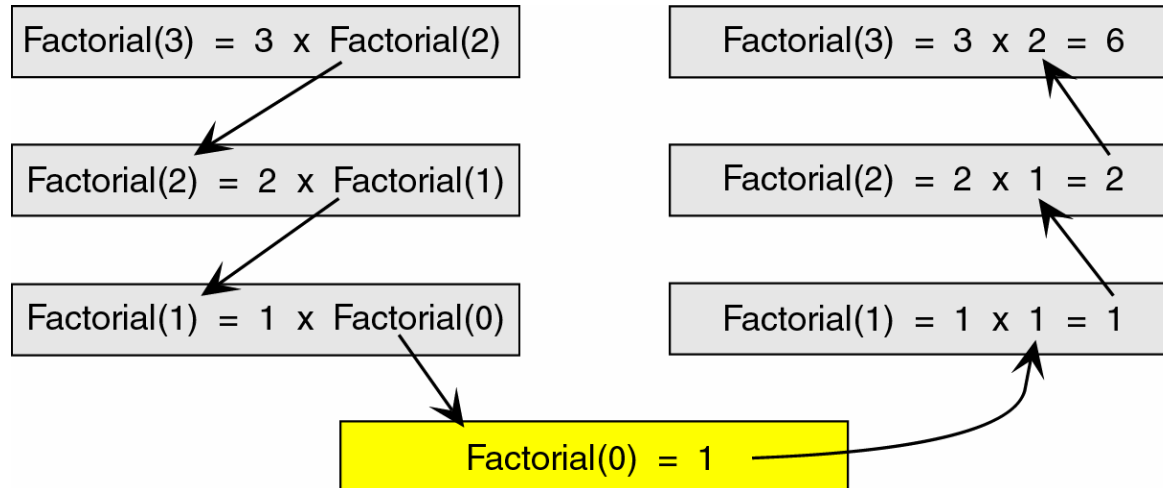


# Breakdown



- Here, we see that we start at the top level, factorial(3), and simplify the problem into  $3 \times \text{factorial}(2)$ .
- Now, we have a slightly less complicated problem in factorial(2), and we simplify this problem into  $2 \times \text{factorial}(1)$ .

# Breakdown



- We continue this process until we are able to reach a problem that has a known solution.
- In this case, that known solution is factorial(0) = 1.
- The functions then return in reverse order to complete the solution.



# Breakdown

- This known solution is called the ***base case***.
- Every recursive algorithm must have a base case to simplify to.
- Otherwise, the algorithm would run forever (or until the computer ran out of memory).



# Iterative Algorithm



```
factorial(n) {  
    i = 1  
    factN = 1  
    loop (i <= n)  
        factN = factN * i  
        i = i + 1  
    end loop  
    return factN  
}
```

The iterative solution is very straightforward. We simply loop through all the integers between 1 and n and multiply them together.



# Recursive Algorithm

```
factorial(n) {  
  if (n = 0)  
    return 1  
  else  
    return n*factorial(n-1)  
  end if  
}
```

Note how much simpler the code for the recursive version of the algorithm is as compared with the iterative version →

we have eliminated the loop and implemented the algorithm with 1 'if' statement.

# How Recursion Works



- When the function is finished, it needs to return to the function that called it.
- The calling function then 'wakes up' and continues processing.

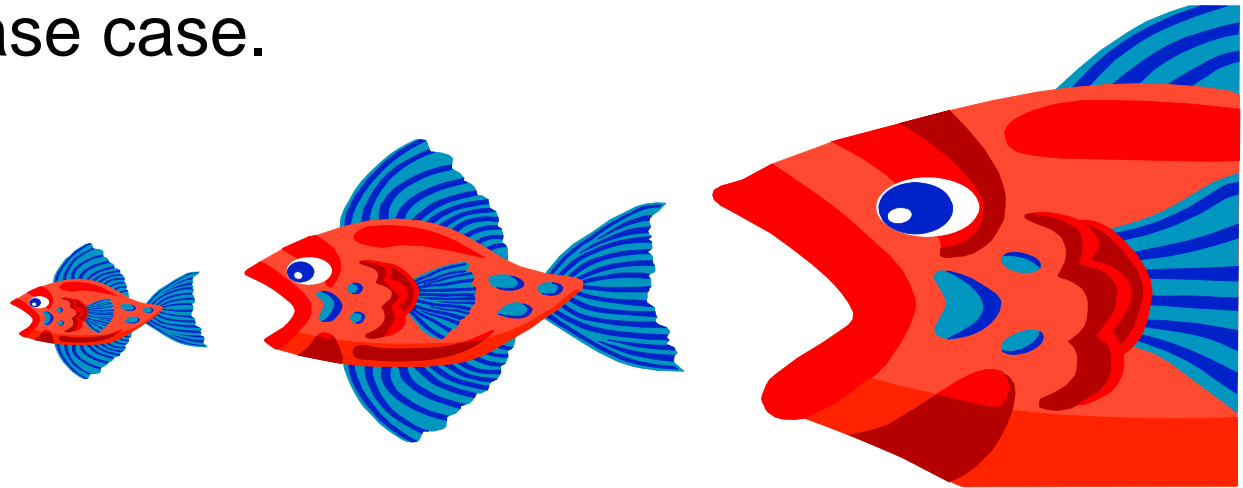
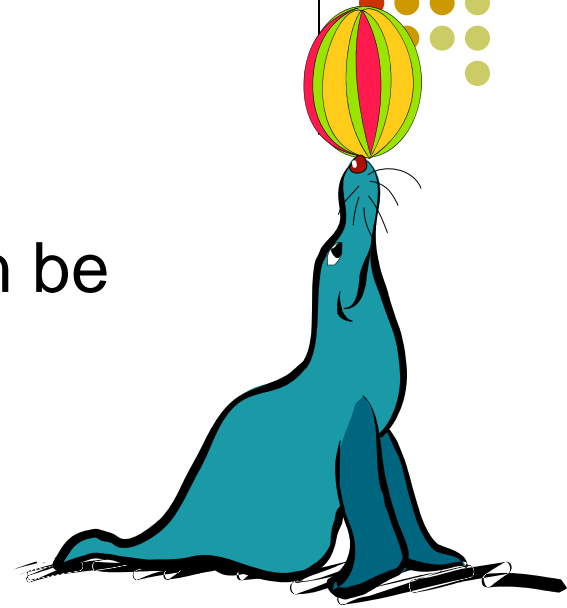
# How Recursion Works



- To do this we use a stack.
- Before a function is called, all relevant data is stored in a ***stackframe***.
- This stackframe is then pushed onto the system stack.
- After the called function is finished, it simply pops the system stack to return to the original state.

# Basic Recursion

- 1. Base cases:
  - Always have at least one case that can be solved without using recursion.
- 2. Make progress:
  - Any recursive call must make progress toward a base case.



# Advantage and Limitations of Recursion



- Recursive solutions can be easier to understand and to describe than iterative solutions.
- Recursion works the best when the algorithm and/or data structure that is used naturally supports recursion.
- One such data structure is the tree (more to come).
- One such algorithm is the binary search algorithm that we discussed earlier in the course.

# Fibonacci Numbers

0, 1, 1, 2, 3, 5, 8, 13, 21 .....

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	...
0	1	1	2	3	5	8	13	21	34	55	89	144	233	377	610	...

Fibonacci (0) = 0

Fibonacci (1) = 1

Fibonacci (x) = Fibonacci (x-1) + Fibonacci (x-2)

Fib(7) = ?

5 4 3 2 1

return 0

return 1

return 1

return Fib(x-1) + Fib(x-2)

$$\text{Fib}(7) = \text{Fib}(6) + \text{Fib}(5)$$

$$\text{Fib}(6) = \text{Fib}(5) + \text{Fib}(4)$$

$$\text{Fib}(5) = \text{Fib}(4) + \text{Fib}(3)$$

$$\text{Fib}(4) = \text{Fib}(3) + \text{Fib}(2)$$

$$\text{Fib}(3) = \text{Fib}(2) + \text{Fib}(1)$$

$$\text{Fib}(2) = \text{Fib}(1) + \text{Fib}(0)$$

$$\text{Fib}(1) = 1$$

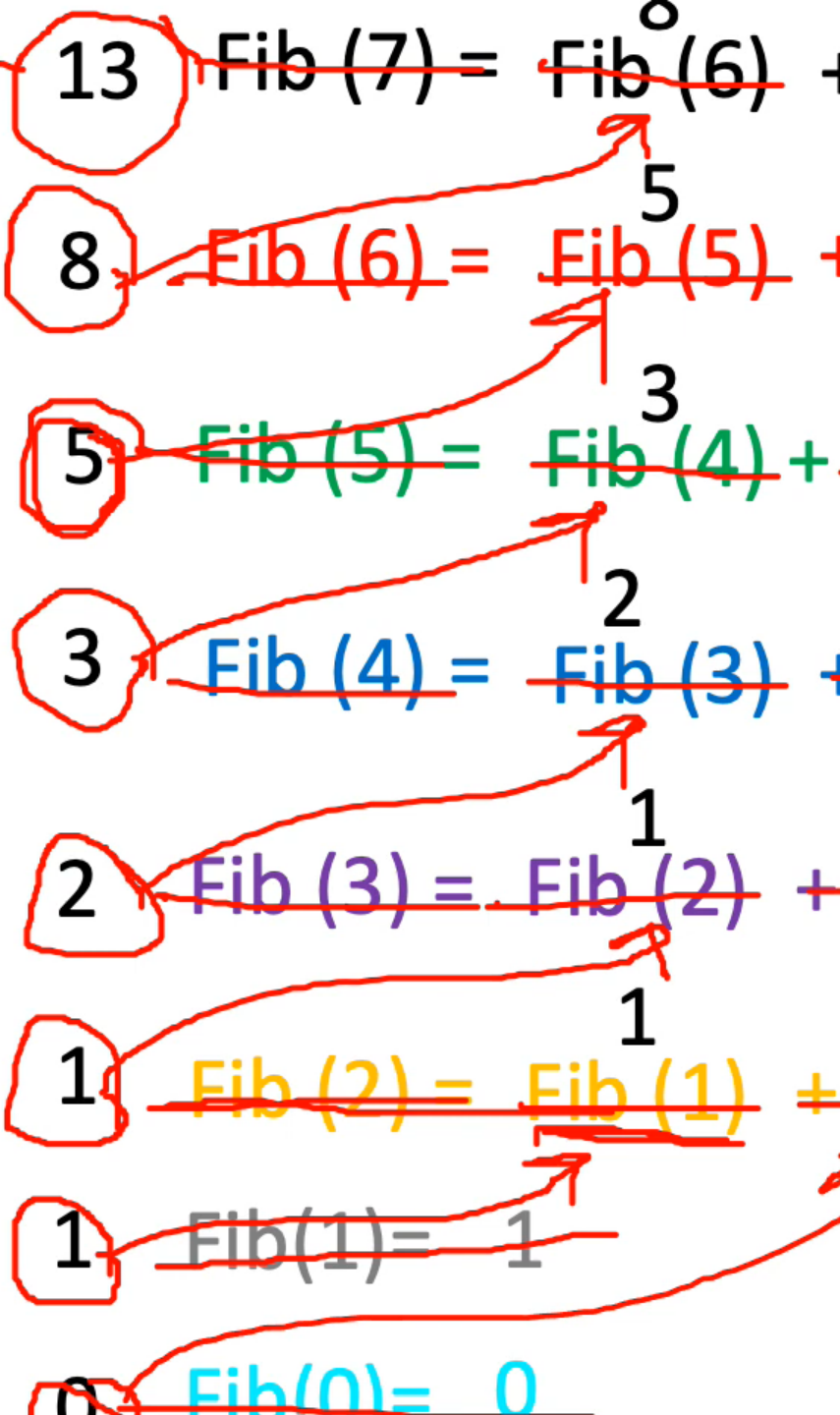


Fib(7) = ?  
5 4 3 2 1 0

0	1	2	3	4	5	6	7	8
0	1	1	2	3	5	8	13	21

return 0  
return 1

return Fib(x-1) + Fib(x-2)



# Fibonacci function:

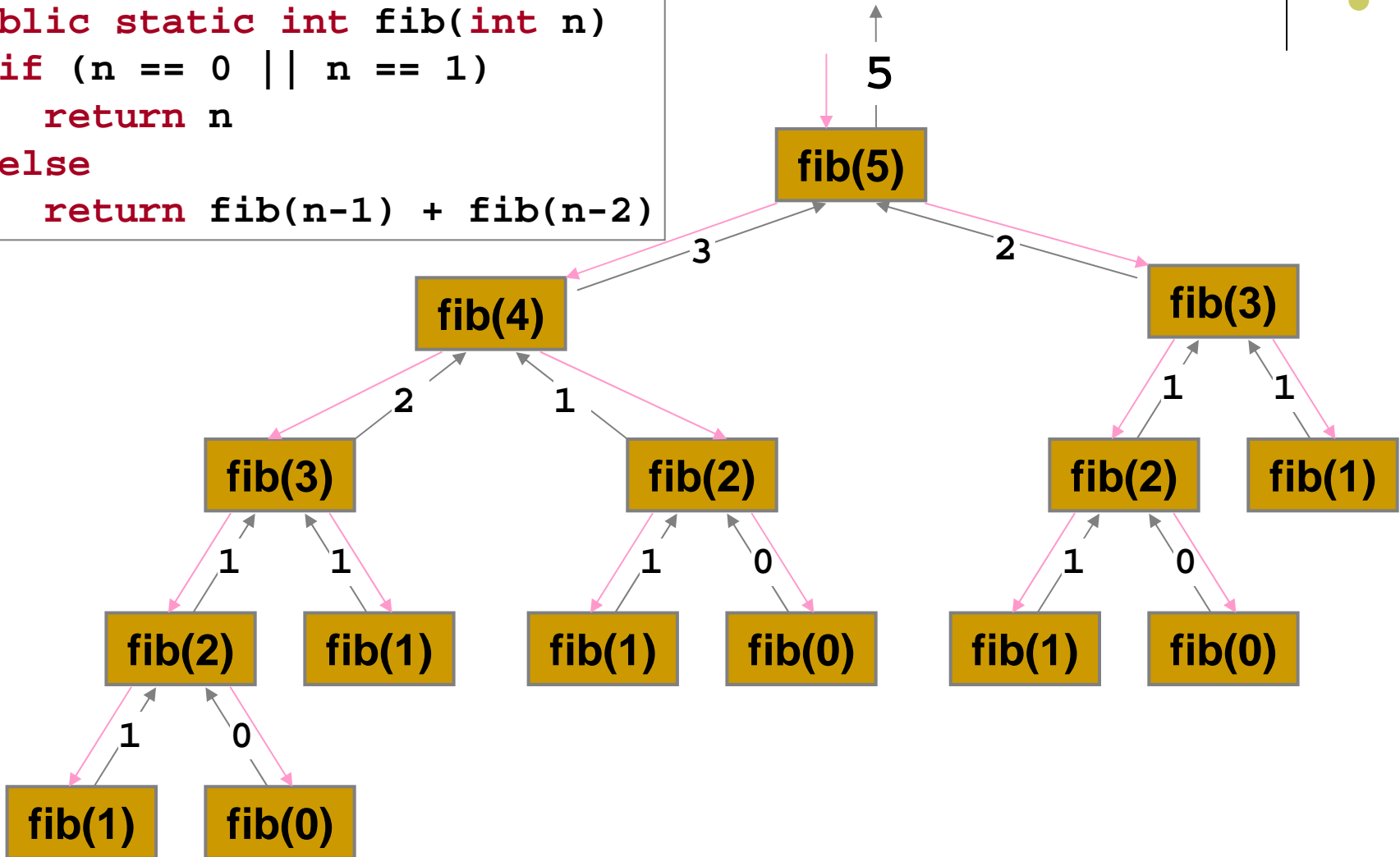


- $\text{fibonacci}(0) = 1$
- $\text{fibonacci}(1) = 1$
- $\text{fibonacci}(n) = \text{fibonacci}(n-1) + \text{fibonacci}(n-2)$
- [for  $n > 1$ ]
- **This definition is a little different than the previous ones because it has two base cases, not just one; in fact, you can have as many as you like.**
- **In the recursive case, there are two recursive calls, not just one. There can be as many as you like.**

# Function Analysis for call `fib(5)`



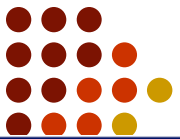
```
public static int fib(int n)
  if (n == 0 || n == 1)
    return n
  else
    return fib(n-1) + fib(n-2)
```



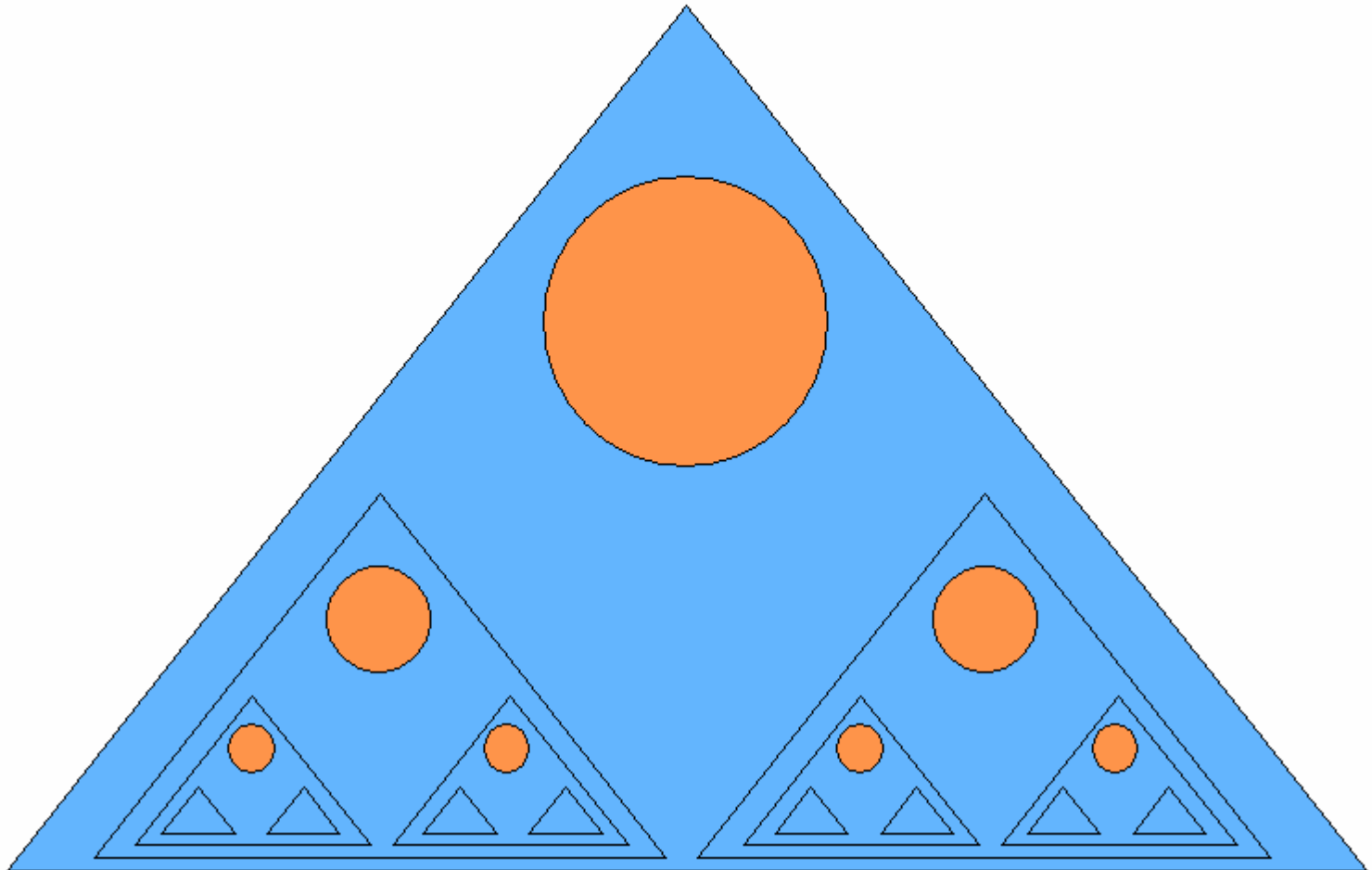
# Conclusion



- A recursive solution solves a problem by solving a smaller instance of the same problem.
- It solves this new problem by solving an even smaller instance of the same problem.
- Eventually, the new problem will be so small that its solution will be either obvious or known.
- This solution will lead to the solution of the original problem.



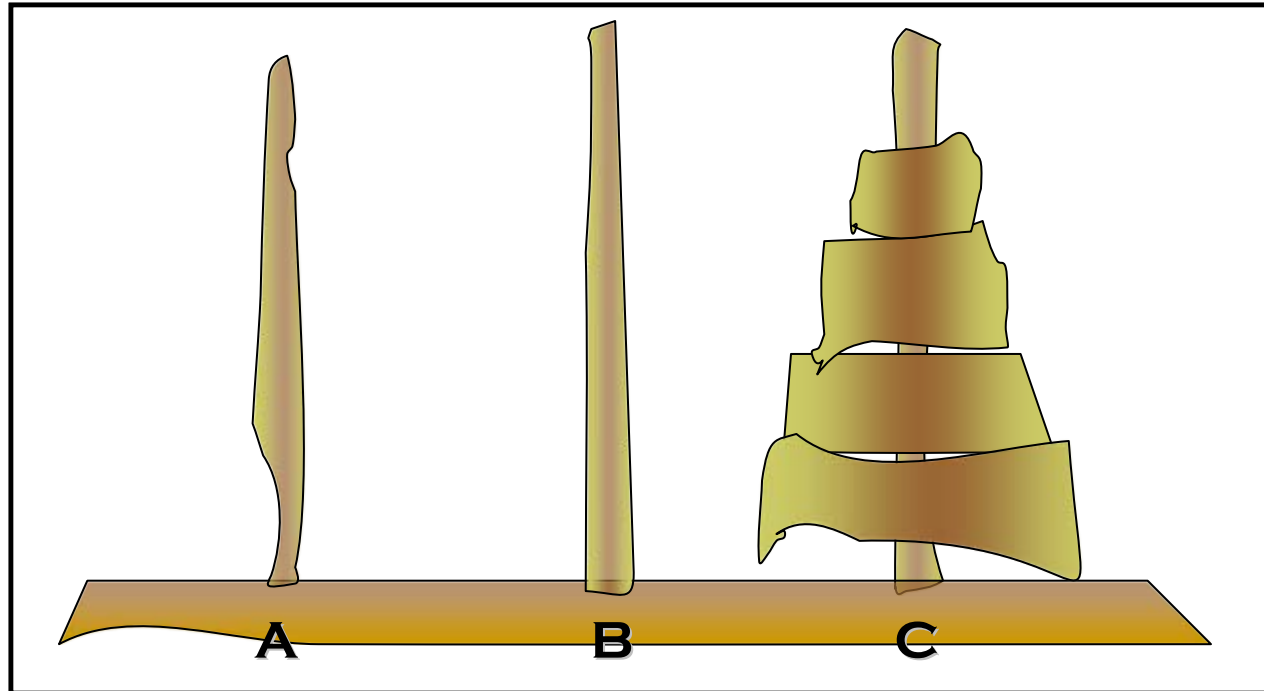
A *binary tree* is either empty, or it consists of a node part (an element) and two subtree parts. Each of the subtree parts are themselves binary trees.



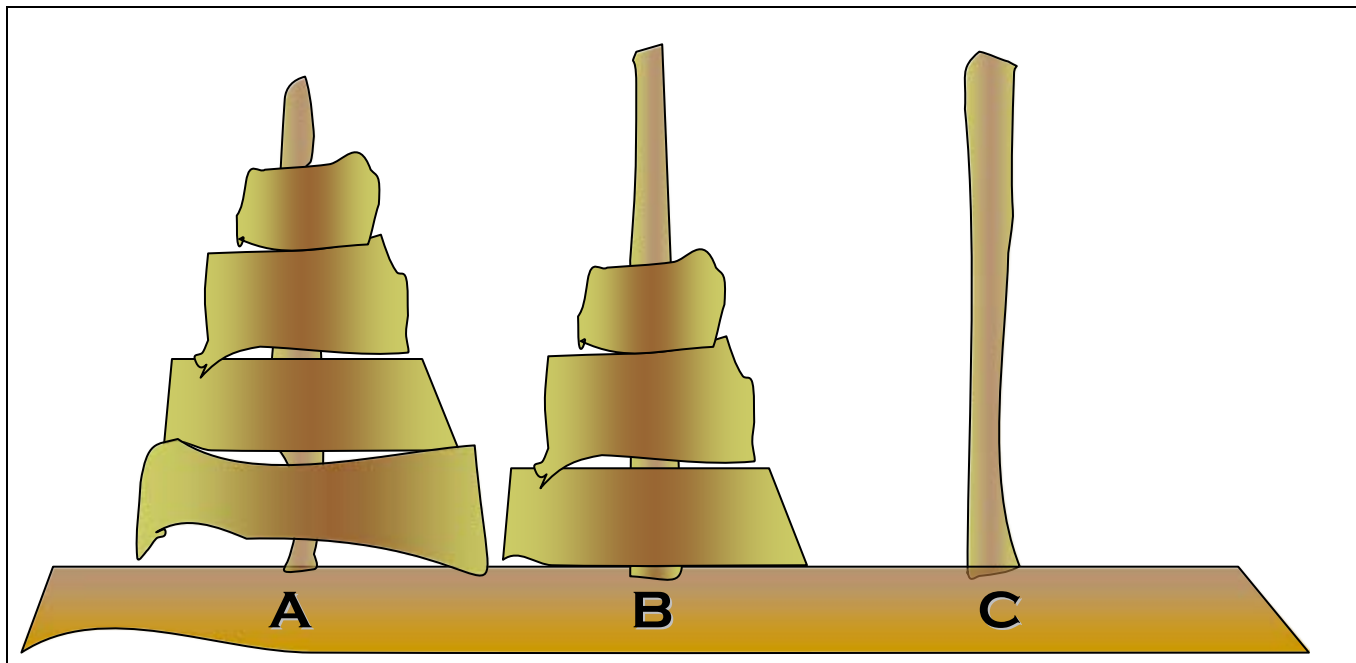


# Towers of Hanoi

- Move  $n$  (4) disks from pole A to pole C
- such that a disk is never put on a smaller disk

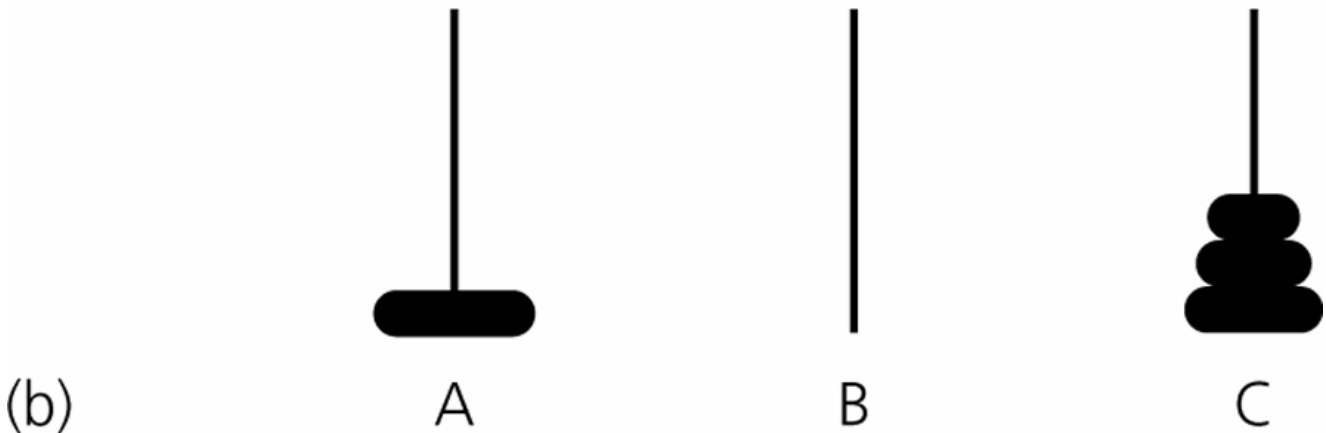
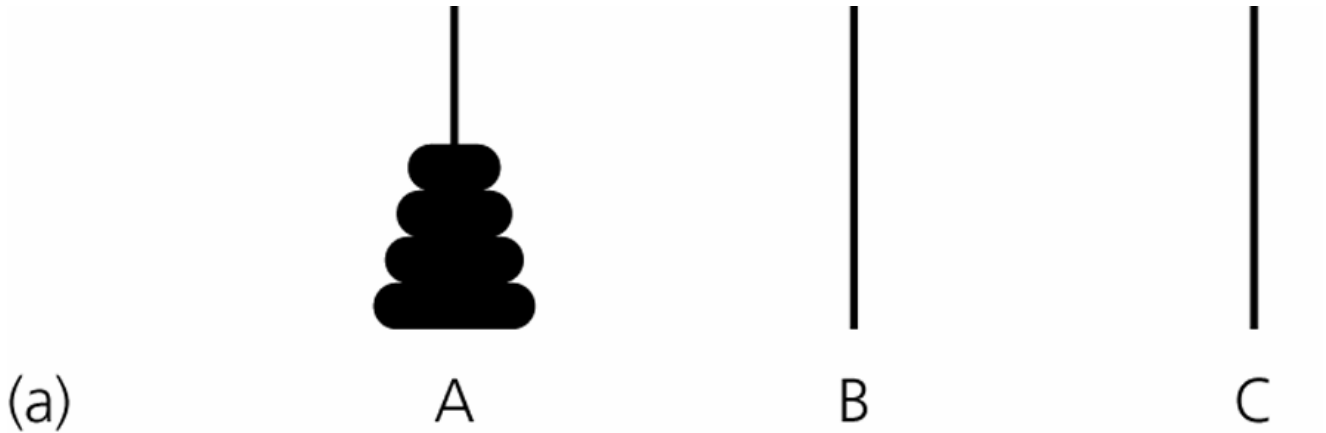


- Move  $n$  (4) disks from A to C
  - Move  $n-1$  (3) disks from A to B
  - Move 1 disk from A to C
  - Move  $n-1$  (3) disks from B to C



# Figure 2.19a and b

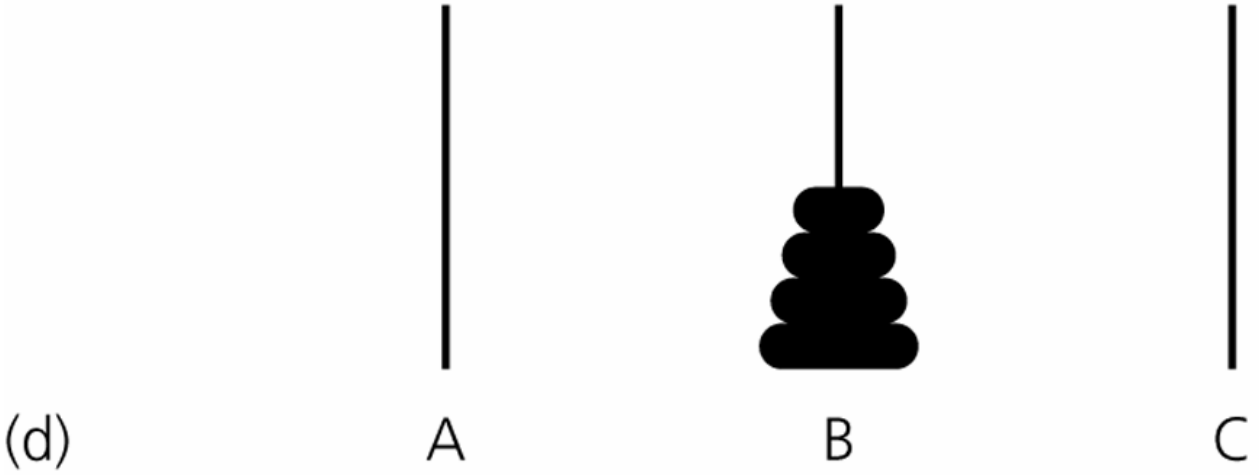
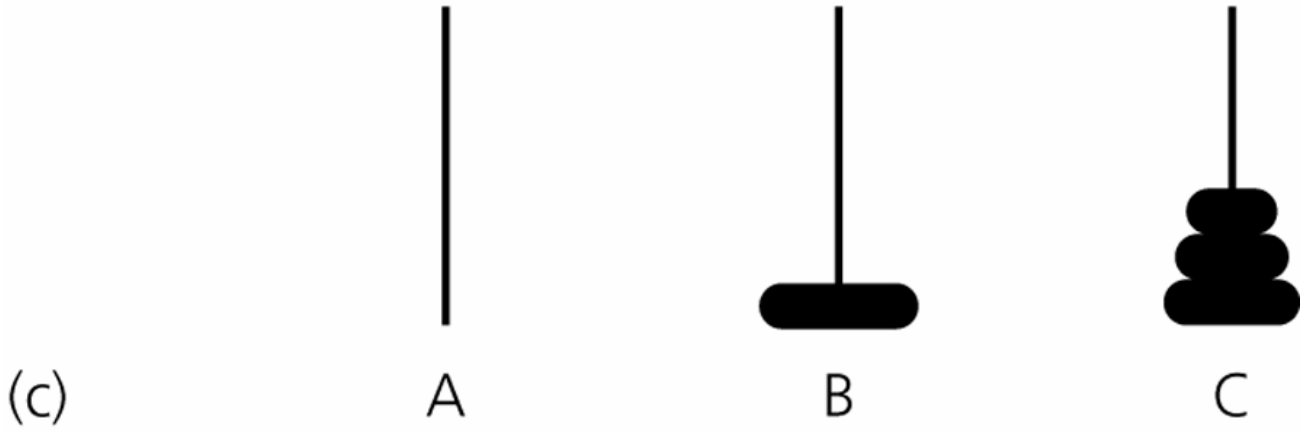
a) The initial state; b) move  $n - 1$  disks from  $A$  to  $C$





# Figure 2.19c and d

c) move one disk from *A* to *B*; d) move  $n - 1$  disks from *C* to *B*



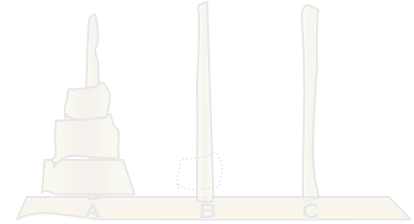
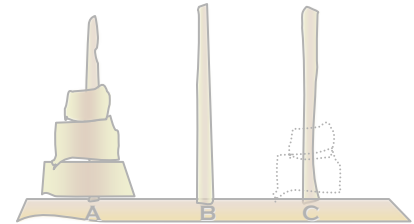
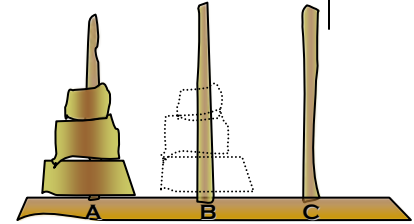
# Hanoi towers



```
public static void solveTowers(int count, char source,
                                char destination, char spare) {
    if (count == 1) {
        System.out.println("Move top disk from pole " + source +
                            " to pole " + destination);
    }
    else {
        solveTowers(count-1, source, spare, destination); // X
        solveTowers(1, source, destination, spare);       // Y
        solveTowers(count-1, spare, destination, source); // Z
    } // end if
} // end solveTowers
```

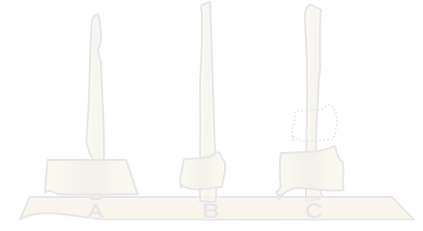
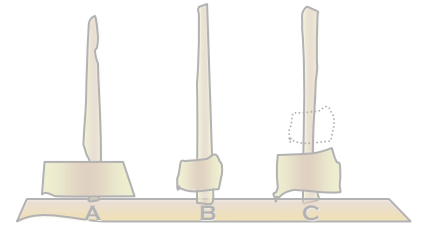
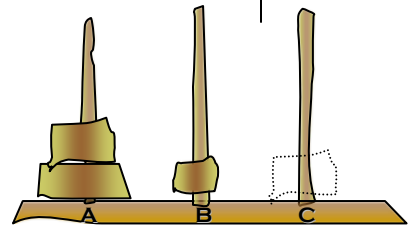
# Figure 2.21a

Box trace of *solveTowers(3, 'A', 'B', 'C')*



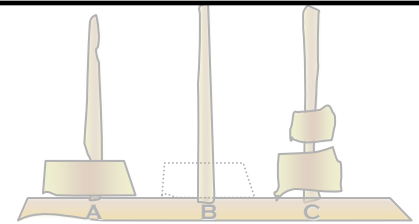
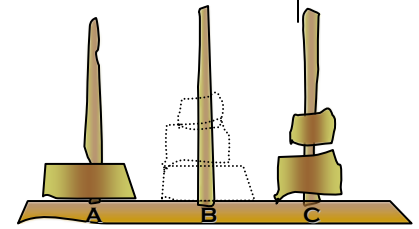
# Figure 2.21b

Box trace of `solveTowers(3, 'A', 'B', 'C')`



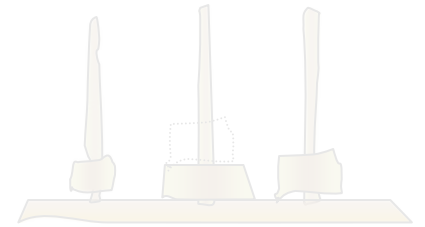
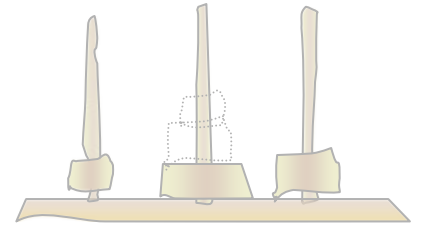
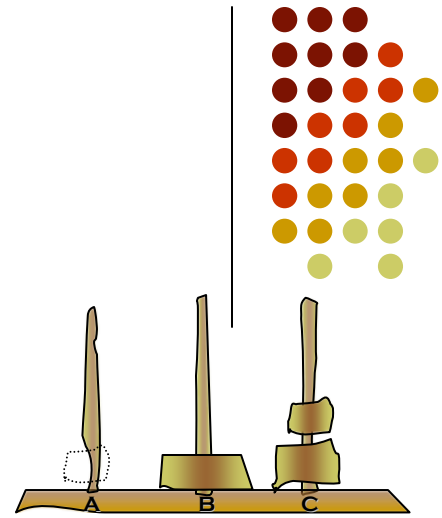
# Figure 2.21c

Box trace of `solveTowers(3, 'A', 'B', 'C')`



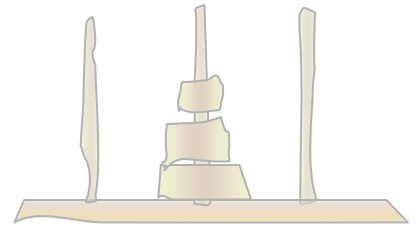
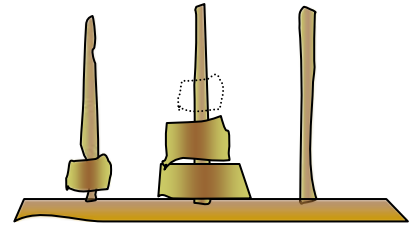
# Figure 2.21d

Box trace of `solveTowers(3, 'A', 'B', 'C')`



# Figure 2.21e

Box trace of *solveTowers(3, 'A', 'B', 'C')*





# Cost of Hanoi Towers

- How many moves is necessary to solve Hanoi Towers problem for N disks?
- $\text{moves}(1) = 1$
- $\text{moves}(N) = \text{moves}(N-1) + \text{moves}(1) + \text{moves}(N-1)$
- i.e.  
 $\text{moves}(N) = 2 * \text{moves}(N-1) + 1$
- Guess solution and show it's correct with Mathematical Induction!