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It's programming by faith!

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- solves a problem with smaller recursive cases
- has one or more non-recursive base cases

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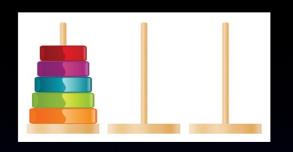
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 base case

The "Tower of Hanoi" puzzle was invented by Édouard Lucas in 1883:



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Recursive case

We can move the top N-1 disks to the other position, then the single largest disk to the target position, then move the N-1 disks to the target.

Let's write some python:

```
def hanoi(n, src, dst, tmp):
    if n == 1:
        print('move %s to %s' % (src, dst))
    else:
        hanoi(n-1, src, tmp, dst)
        hanoi(1, src, dst, tmp)
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This solution is easy to split into base and recursive cases. It fails if N is less than zero.

A more robust and more compact solution:

```
def hanoi(n, src, dst, tmp):
    if n > 0:
        hanoi(n-1, src, tmp, dst)
        print('move %s to %s' % (src, dst))
        hanoi(n-1, tmp, dst, src)
```

More efficient than the first solution?

```
import sys

def hanoi(n, src, dst, tmp):
    if n > 0:
        hanoi(n-1, src, tmp, dst)
        print('move %s to %s' % (src, dst))
        hanoi(n-1, tmp, dst, src)

number = int(sys.argv[1])
hanoi(number, 'A', 'B', 'C')
```

```
% python ./hanoi.py 3
move A to B
move A to C
move B to C
move A to B
move C to A
move C to B
move A to B
```

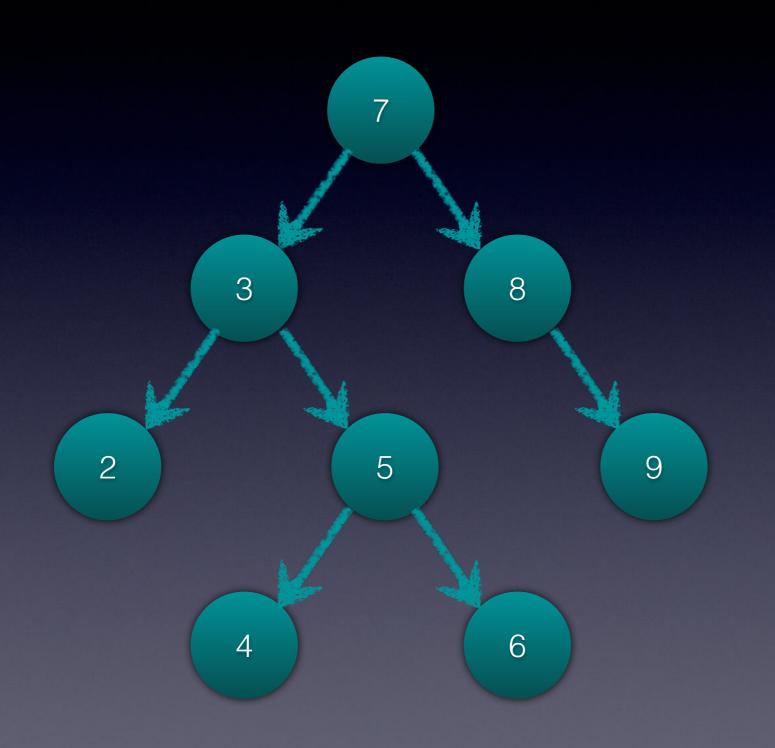
Why are recursive algorithms useful?

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Some data structures are recursively defined, so recursive algorithms are a natural way to process them.

For example, binary trees.

A binary tree looks like this:

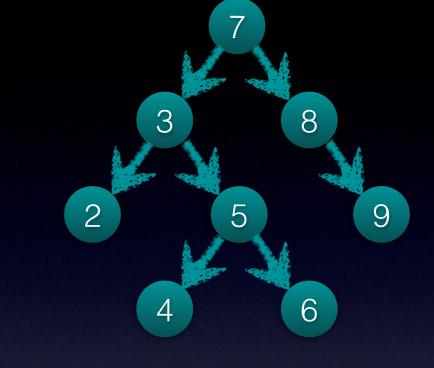


A binary tree consists of a *node* which has a *value* and *left* and *right* pointers, which may be *None* or refer to another binary tree.

Suppose we want to print out the values in the tree nodes.

One method is, for each node:

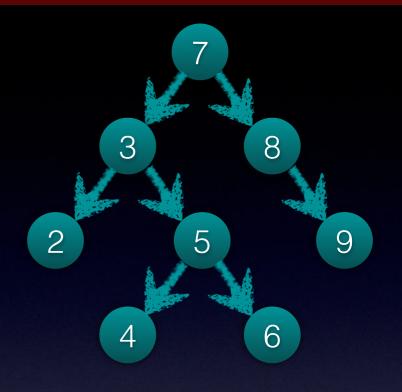
- print the left sub-tree
- print the node value
- print the right sub-tree



This is called an *in-order* walk of the tree. We print the node value in between the left and right sub-tree values.

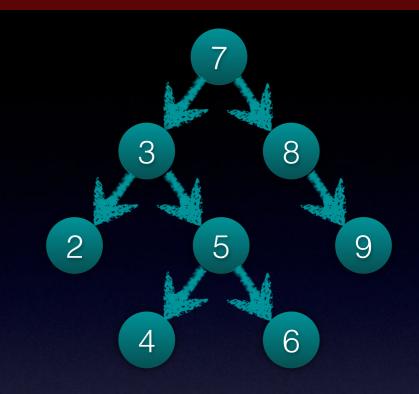
Let's define our tree:

```
class Node(object):
   def init (self, value,
                left=None, right=None):
      self.value = value
      self.left = left
      self.right = right
tree = Node(value=7)
tree.left = Node(value=3)
tree.left.left = Node(value=2)
tree.left.right = Node(value=5)
tree.left.right.left = Node(value=4)
tree.left.right.right = Node(value=6)
tree.right = Node(value=8)
tree.right.right = Node(value=9)
```



The code to walk the tree is:

```
def walk_inorder(node):
    if node is not None:
       walk_inorder(node.left)
       print(node.value)
       walk_inorder(node.right)
```

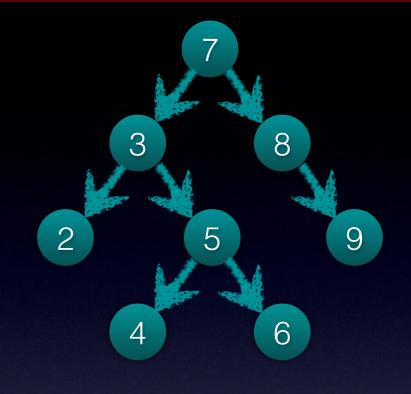


Executing this function on the example tree:

```
2 3 4 5 6 7 8 9
```

There are two other ways to walk through a binary tree:

```
def walk preorder(node):
   if node is not None:
      print(node.value)
      walk inorder(node.left)
      walk inorder(node.right)
def walk postorder(node):
   if node is not None:
      walk inorder(node.left)
      walk inorder(node.right)
      print(node.value)
```



Executing these functions on the example tree:

```
7 3 2 5 4 6 8 9 # preorder
2 4 6 5 3 9 8 7 # postorder
```

What are the costs of a recursive solution?

- every nested function call costs memory
- every function call costs time

When the costs of using recursion are too high we might use an iterative algorithm. There are algorithms to iteratively walk through a binary tree (eg, Shorr Waite), but they are not as simple as the recursive approach.

It is possible to combine recursion with various techniques to achieve an efficient solution.

The Fibonacci function:

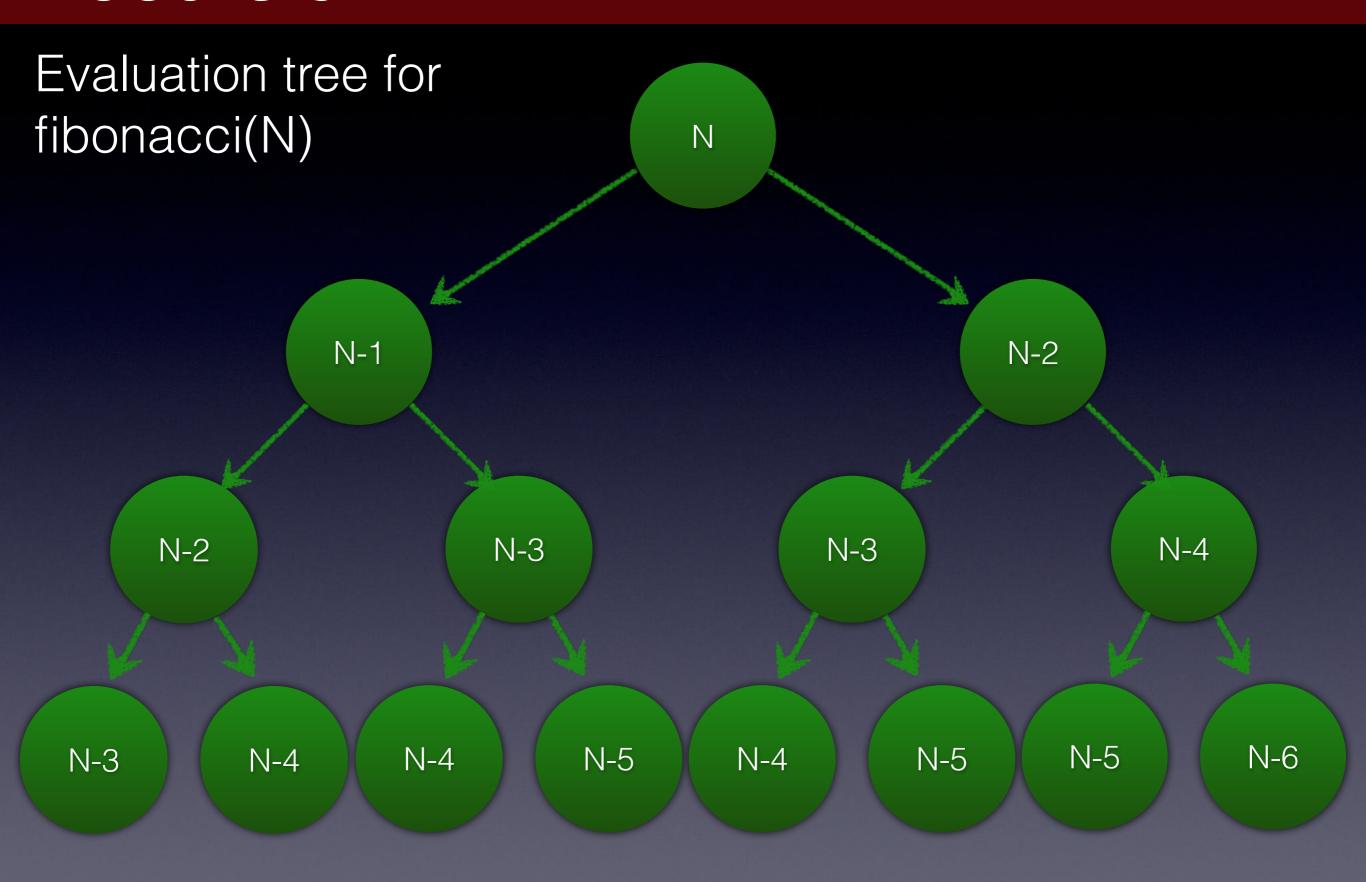
```
def fibonacci(n):
    if n == 0:
        return 0
    if n == 1:
        return 1
        return fibonacci(n-1) + fibonacci(n-2)
```

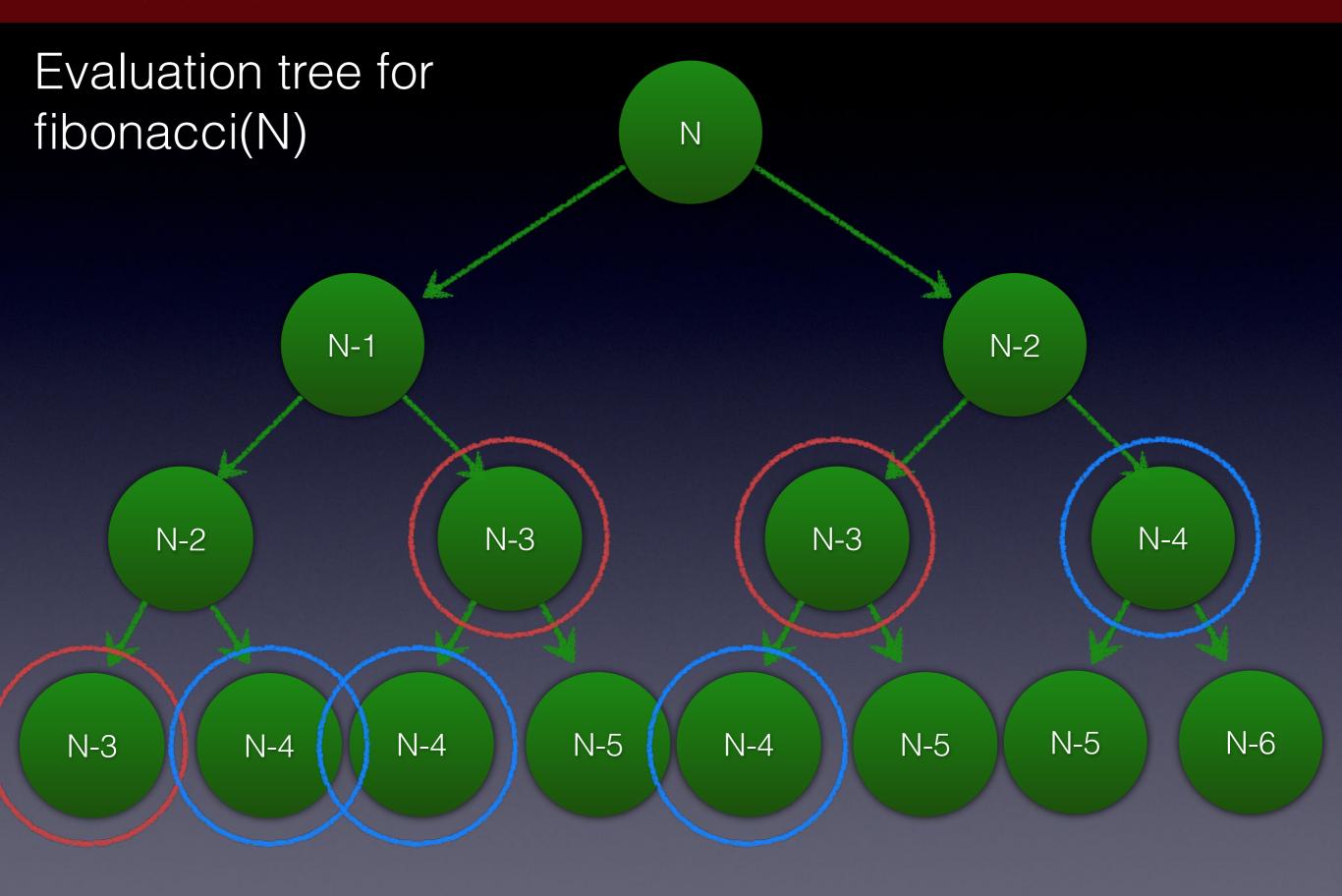
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```

This naive Fibonacci function is very inefficient. Try evaluating fibonacci(40). It's really slow!

```
fibonacci(40)=102334155 took 57.7889390s
```





If we could evaluate fibonacci(m) once and use the remembered result, we would save time.

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```
A useful speedup - more than 2,000,000 times faster!
```

fibonacci memo(40)=102334155 took 0.000025s

This approach is called memoisation. It's part of what is called dynamic programming.

There are other ways to memoise using decorators, but that's another talk.



Code and PDF at:

code.google.com/p/rzzzwilson/source/browse/talks/recursion_1