

# CSCI 373 Textbook Notes

---

## Recursion

---

- *Repetition* is a key feature of high level languages, and we have seen that repetition can be achieved through for and while loops
- Another way to achieve repetition is through **recursion**, which occurs whenever a function calls itself within its own definition
- **The Factorial Function**
  - Let us define the factorial function
    - $n! = 1$ , if  $n = 0$
    - $n! = n \cdot (n - 1) \cdot (n - 2) \cdot \dots \cdot 3 \cdot 2 \cdot 1$  if  $n \geq 1$
  - For example,
    - $5! = 5 * 4 * 3 * 2 * 1 = 120$
  - From this, we can see,
    - $5! = 5 * (4!)$
    - and then, since  $4! = 4 * 3 * 2 * 1$ 
      - $5! = 5 * 4 * (3!)$
  - So this leads to the following recursive definition of the recursive factorial function
    - $\text{factorial}(n) = 1$ , if  $n = 0$
    - $\text{factorial}(n) = n \cdot \text{factorial}(n - 1)$ , if  $n \geq 1$
  - As we can see, this function contains one or more **base cases**
    - In this case, the base case is 1 when  $n = 0$

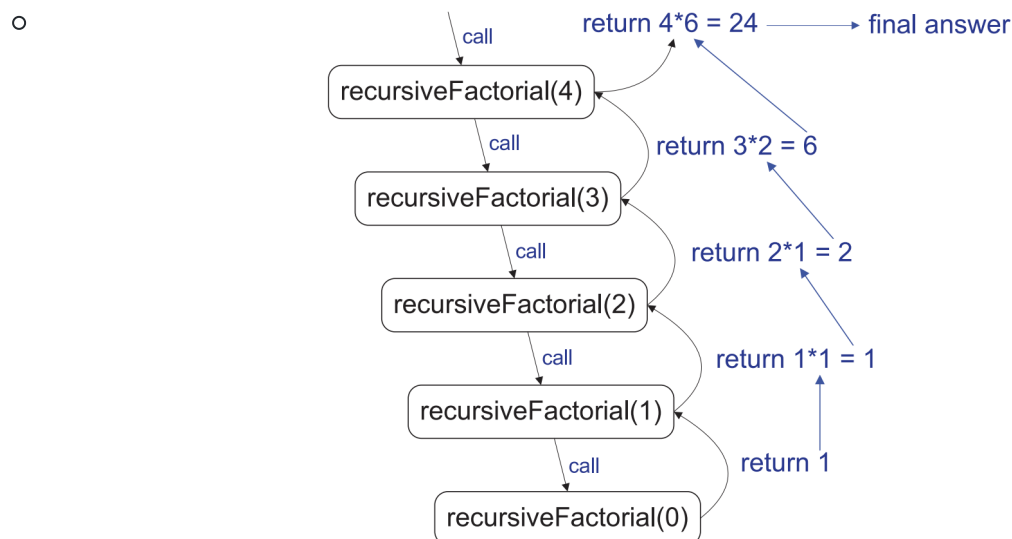
- There is no circularity in this definition because each time the function is invoked, its argument is smaller by one

- C++ Implementation of Recursion in the Factorial Function**

- Note that in the following definition, no loops are necessary since recursion is used

```
int recursiveFactorial(int n)
{
    if(n==0)
        return 1; //basis case which will always be called at the end
    else
        return n * recursiveFactorial(n-1); //recursive case
}
```

- This function can be illustrated using the following recursion trace



**Figure 3.16:** A recursion trace for the call recursiveFactorial(4).

- Recursive Example using an English Ruler**

- An English ruler is broken into intervals and each interval contains a set of *ticks*
- These ticks are placed at intervals of  $\frac{1}{2}$  inch,  $\frac{1}{4}$  inch, and so on
- As the size interval decreases by half, the tick length decreases by one
- Below are some representations of English Rulers
-

(a) (b) (c)

**Figure 3.17:** Three sample outputs of an English ruler drawing: (a) a 2-inch ruler with major tick length 4; (b) a 1-inch ruler with major tick length 5; (c) a 3-inch ruler with major tick length 3.

- The longest tick length of an English Ruler will be referred to as the *major tick length*
- One approach to drawing this consists of three functions
  - `drawRuler()` draws the entire ruler and takes the number of inches, `nInches`, and the major tick length, `majorLength` as arguments
  - The utility function, `drawOneTick()`, draws a single tick of the given length
  - `drawTicks`, which is the recursive function which draws the sequence of ticks within some interval
- Here is a C++ implementation of what is described above

```
void drawOneTick(int tickLength, int tickLabel = -1)
{
    for(int i=0; i<tickLength; i++)
        cout<<"-";
    if(tickLabel>=0)
        cout<<" "<<tickLabel;
    cout<<endl;
}
```

- This function draws one tick with an optional label

```

○ void drawTicks(int tickLength)
{
    if(tickLength>0)
    {
        drawTicks(tickLength-1);
        drawOneTick(tickLength);
        drawTicks(tickLength-1);
    }
}

```

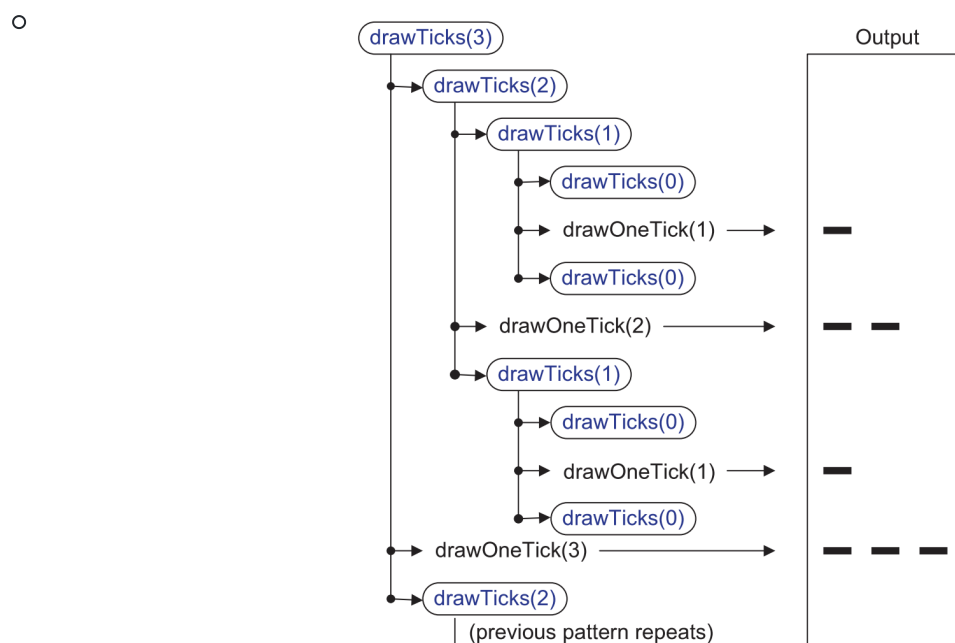
- This function recursively draws ticks between two major ticks

```

○ void drawRuler(int nInches, int majorLength)
{
    drawOneTick(majorLength, 0);
    for(int i=1; i<nInches; i++)
    {
        drawTicks(majorLength-1);
        drawOneTick(majorLength, i)
    }
}

```

- This function can be used to draw the ruler as a whole



**Figure 3.18:** A partial recursion trace for the call `drawTicks(3)`. The second pattern of calls for `drawTicks(2)` is not shown, but it is identical to the first.

- The above recursion trace provides an illustration of what will occur when `drawRuler` is run with a major tick length of 3

- **More Examples of Recursion**

- Recursion can be beneficial by allowing us to exploit a more *natural* form of repetition that does not involve complex nested loops or case analyses
- *Example 3.1:* Modern OSes operate file-system directories in a recursive manner, meaning folders can be nested inside of folders in an arbitrarily deep fashion so long as there is sufficient space in memory
- *Example 3.2:* The syntax in modern programming languages is most often defined in a recursive manner

- **3.5.1: Linear Recursion**

- Linear recursion is the simplest form of recursion
- Linear recursion refers to a recursive function that makes, at most, one recursive call each time that it is invoked
- *Summing the elements of an array recursively*
  - Suppose we have an array,  $A$ , of  $n$  integers which we want to sum together
  - Since we know that the sum of all integers in  $A$  is equal to  $A[0]$  when  $n = 1$ , we can solve this problem recursively with the following algorithm

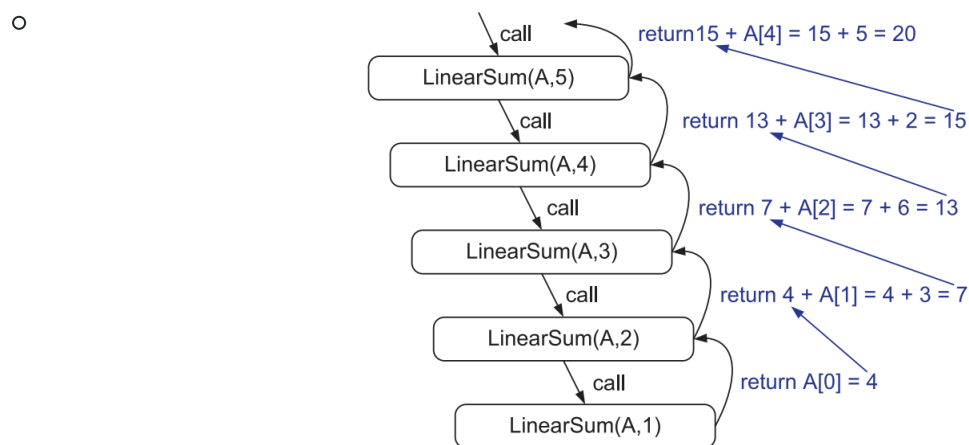
- **Algorithm for LinearSum( $A$ ,  $n$ )**

```
Input: an integer Array, A, and int n >= 1
Output: The sum of the first n integers in A

if n=1 then
    return A[0]
else
    return LinearSum(A, n-1)+A[n-1]
```

- This illustrates one very important aspect of *all* recursive functions - the fact that it terminates
    - This can be fairly easily achieved by writing a non recursive statement for the base case, in this case the `if n=1` statement achieves this
- In fact, an algorithm that employs linear recursion generally adheres to the following form

- *Test for base cases*, where the function reaches a pre-defined base case for which a recursive call is not needed
  - Base cases should be defined such that every possible chain of recursive calls eventually reaches a base case
- *Recursion*, where after testing for base cases, the function will recursively call itself
  - It might have to decide between different recursive steps, but a linear recursive algorithm should call itself recursively only once each time it is invoked
- Now, let us consider the recursion trace, or visual diagram representing a system's logic during a recursive linear summation



**Figure 3.19:** Recursion trace for an execution of  $\text{LinearSum}(A, n)$  with input parameters  $A = \{4, 3, 6, 2, 5\}$  and  $n = 5$ .

- Recursive algorithms can, however, take up more space in the memory due to their need to store each prior recursive call until the function terminates
- Therefore, it can sometimes be useful to be able to derive non-recursive algorithms from recursive ones
- *Tail recursion* occurs when a recursive algorithm initiates the recursive call as the last thing it does other than base case evaluation
- Algorithms that utilize tail recursion are simple to convert from recursive to non-recursive
  - This can be achieved by iterating through the recursive calls rather than calling them explicitly
- Here is the algorithm for `IterativeReverseArray()`

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

Input: An array A, and non-negative integer indices i and j

Output: Reversal of A from i to j

```

while i < j
    Swap A[i] and A[j]
    i <- i+1
    j <- j-1
return
      
```

### • 3.5.2: Binary Recursion

- When a function makes two recursive calls, we can refer to this as *binary recursion*
- Let us look at the algorithm for a Binary sum

- Algorithm **BinarySum**(A, i, n)
 

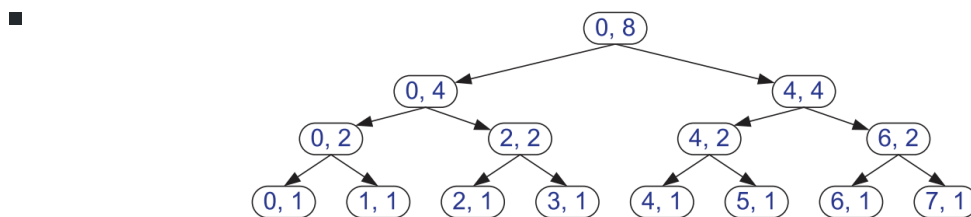
Input: An array A and integers i and n

Output: The sum of the n integers in A starting at i

```

if n=1
    return A[i]
return
    BinarySum(A, i, [n/2]) + BinarySum(A, i+[n/2], [n/2])
      
```

- Below is the trace for Binary Sum



**Figure 3.20:** Recursion trace for the execution of **BinarySum**(0,8).

### • Computing Fibonacci Numbers via Binary Recursion

- Algorithm BinaryFib(k):

Input: Nonnegative integer k  
Output: The kth Fibonacci number  $F_k$

```
if k ≤ 1 then
    return k
else
    return BinaryFib(k-1) + BinaryFib(k-2)
```

- However, this is of time complexity  $O(n^2)$ , because the number of recursive calls more than doubles with each consecutive index
- Therefore, it is actually more efficient to compute the  $k_{th}$  Fibonacci number using *linear recursion*

- Computing Fibonacci Numbers via Linear Recursion

- Algorithm LinearFibonacci(k):

Input: A nonnegative integer k  
Output: Pair of Fibonacci numbers  $(F_k, F_{k-1})$

```
if k ≤ 1 then
    return (k, 0)
else
    (i, j) ← LinearFibonacci(k-1)
    return (i+j, i)
```

- For this algorithm, the time complexity is  $O(n)$  so it is far more efficient than binary recursion for Fibonacci calculations

- 3.5.3: Multiple Recursion

- If we generalize the jump from linear to binary recursion, we can arrive at *multiple recursion*
  - Multiple recursion algorithms may make multiple recursive calls, with that number being possibly more than two
- Below is the algorithm and recursion trace for an algorithm written to solve *summation puzzles* where different letters represent integers in an equation



■ Algorithm **PuzzleSolve**( $k, S, U$ ):

Input: An integer  $k$ , sequence  $S$ , and set  $U$   
Output: An enumeration of all  $k$ -length extensions to  $S$  using elements in  $U$  without repetitions

```

for each  $e$  in  $U$  do
  Remove  $e$  from  $U$  { $e$  is now being used} Add  $e$  to the end of  $S$ 
  if  $k = 1$  then
    Test whether  $S$  is a configuration that solves the puzzle
    if  $S$  solves the puzzle then
      return "Solution found: "  $S$ 
  else
    PuzzleSolve( $k-1, S, U$ )
  Add  $e$  back to  $U$  { $e$  is now unused}
  Remove  $e$  from the end of  $S$ 

```

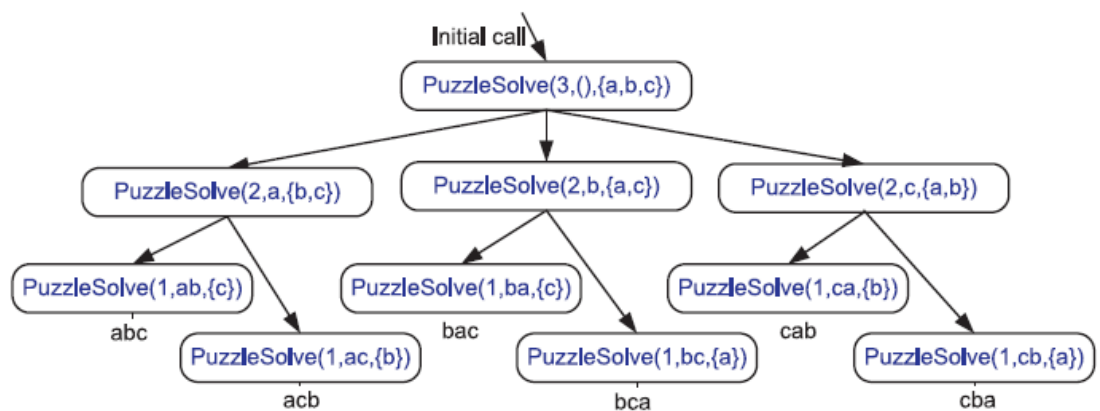


Figure 3.21: Recursion trace for an execution of **PuzzleSolve**( $3, S, U$ ), where  $S$  is empty and  $U = \{a, b, c\}$ . This execution generates and tests all permutations of  $a, b$ , and  $c$ . We show the permutations generated directly below their respective boxes.

■