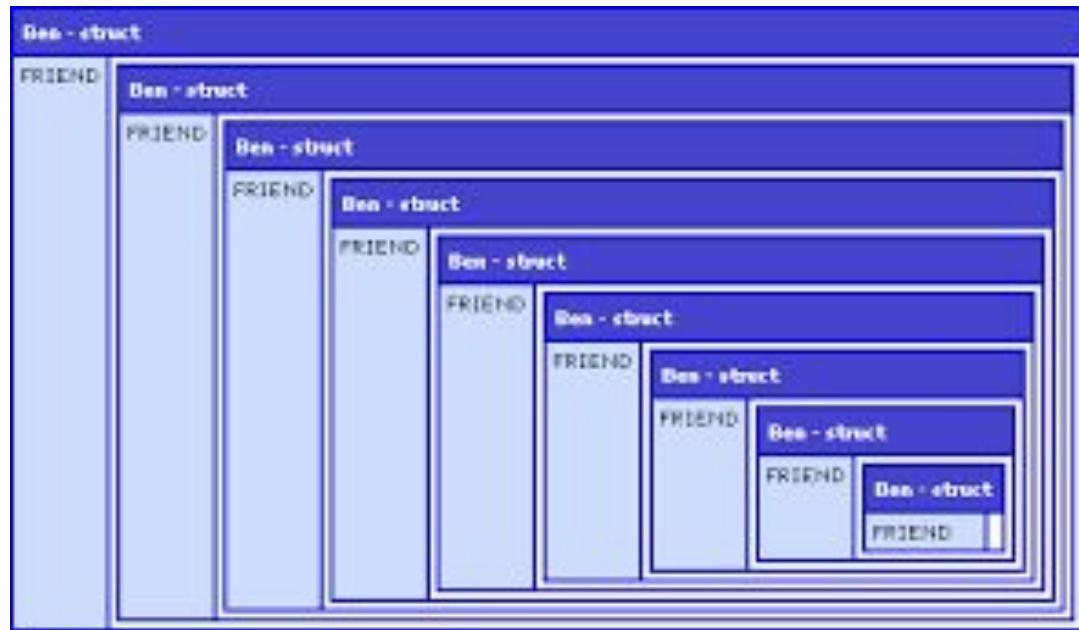
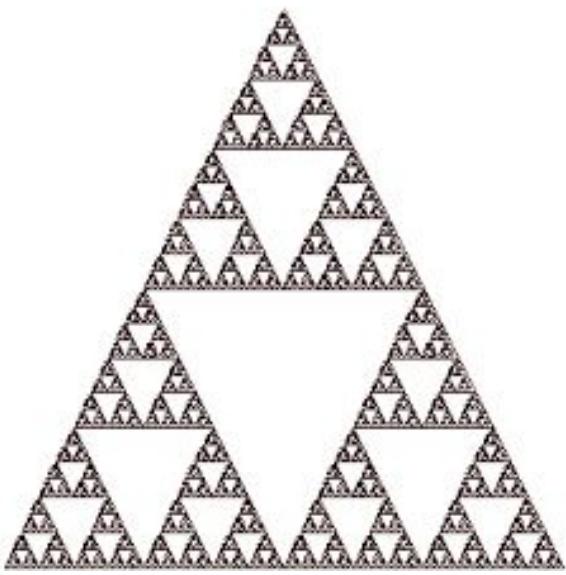


# Recursion

# Recursion: Basic idea

- We have a bigger problem whose solution is difficult to find
- We divide/decompose the problem into smaller (sub) problems
  - Keep on decomposing until we reach to the smallest sub-problem (base case) for which a solution is known or easy to find
  - Then go back in reverse order and build upon the solutions of the sub-problems
- Recursion is applied when the solution of a problem depends on the solutions to smaller instances of the same problem



# Example 1: Factorial

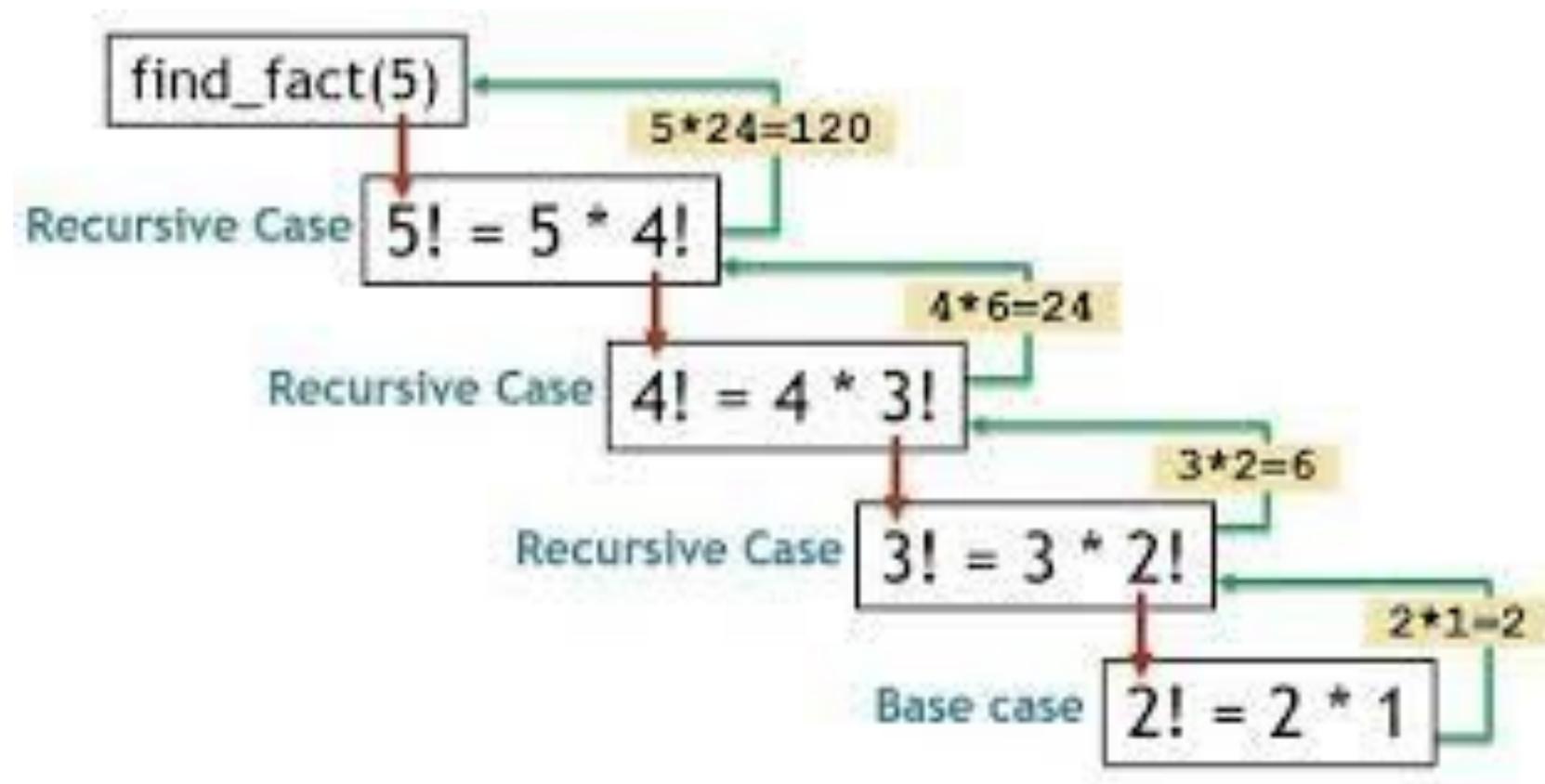
## □ A function which calls itself

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

# Finding a recursive solution

- Each successive recursive call should bring you **closer** to a situation in which the answer is **known** (cf.  $n-1$  in the previous slide)
- A case for which the answer is known (and can be expressed without recursion) is called a **base case**
- Each recursive algorithm must have **at least one base case**, as well as the **general recursive case**

- The factorial of a positive integer  $n$ , denoted  $n!$ , is defined as the product of the integers from 1 to  $n$ . For example,  $4! = 4 \cdot 3 \cdot 2 \cdot 1 = 24$ .



# Recursion in Action: *factorial(n)*

```
factorial (5) = 5 × factorial (4)
                = 5 × (4 × factorial (3))
                = 5 × (4 × (3 × factorial (2)))
                = 5 × (4 × (3 × (2 × factorial (1))))
                = 5 × (4 × (3 × (2 × (1 × factorial (0)))))
                = 5 × (4 × (3 × (2 × (1 × 1))))
                = 5 × (4 × (3 × (2 × 1)))
                = 5 × (4 × (3 × 2))
                = 5 × (4 × 6)
                = 5 × 24
                = 120
```

**Base case arrived**  
Some concept  
from elementary  
maths: Solve the  
inner-most  
bracket, first, and  
then go outward

# Recursion vs. Iteration: Computing N!

- The factorial of a positive integer  $n$ , denoted  $n!$ , is defined as the product of the integers from 1 to  $n$ . For example,  $4! = 4 \cdot 3 \cdot 2 \cdot 1 = 24$ .

- Iterative Solution

$$n! = \begin{cases} 1 & \text{if } n = 0 \\ n \cdot (n - 1) \cdot (n - 2) \cdots 3 \cdot 2 \cdot 1 & \text{if } n \geq 1 \end{cases}$$

- Recursive Solution

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

# Recursion: Do we really need it?

- In some programming languages recursion is imperative
  - For example, in declarative/logic languages (LISP, Prolog etc.)
  - Variables can't be updated more than once, so no looping
  - Heavy backtracking

# Recursion in Action: *factorial(n)*

```
factorial (5) = 5 × factorial (4)
                = 5 × (4 × factorial (3))
                = 5 × (4 × (3 × factorial (2)))
                = 5 × (4 × (3 × (2 × factorial (1))))
                = 5 × (4 × (3 × (2 × (1 × factorial (0)))))
                = 5 × (4 × (3 × (2 × (1 × 1))))
                = 5 × (4 × (3 × (2 × 1)))
                = 5 × (4 × (3 × 2))
                = 5 × (4 × 6)
                = 5 × 24
                = 120
```

Base case

arrived

Some concept from elementary maths: Solve the inner-most bracket, first, and then go outward

# How to write a recursive function?

- Determine the size factor (e.g.  $n$  in  $\text{factorial}(n)$ )
- Determine the base case(s)
  - the one for which you know the answer (e.g.  $0! = 1$ )
- Determine the general case(s)
  - the one where the problem is expressed as a smaller version of itself (must converge to base case)
- Verify the algorithm
  - use the "Three-Question-Method" – next slide

# Linear Recursion

- The simplest form of recursion is *linear recursion*, where a method is defined so that it makes at most one recursive call each time it is invoked
- This type of recursion is useful when we view an algorithmic problem in terms of a **first or last element plus a remaining set** that has the same structure as the original set

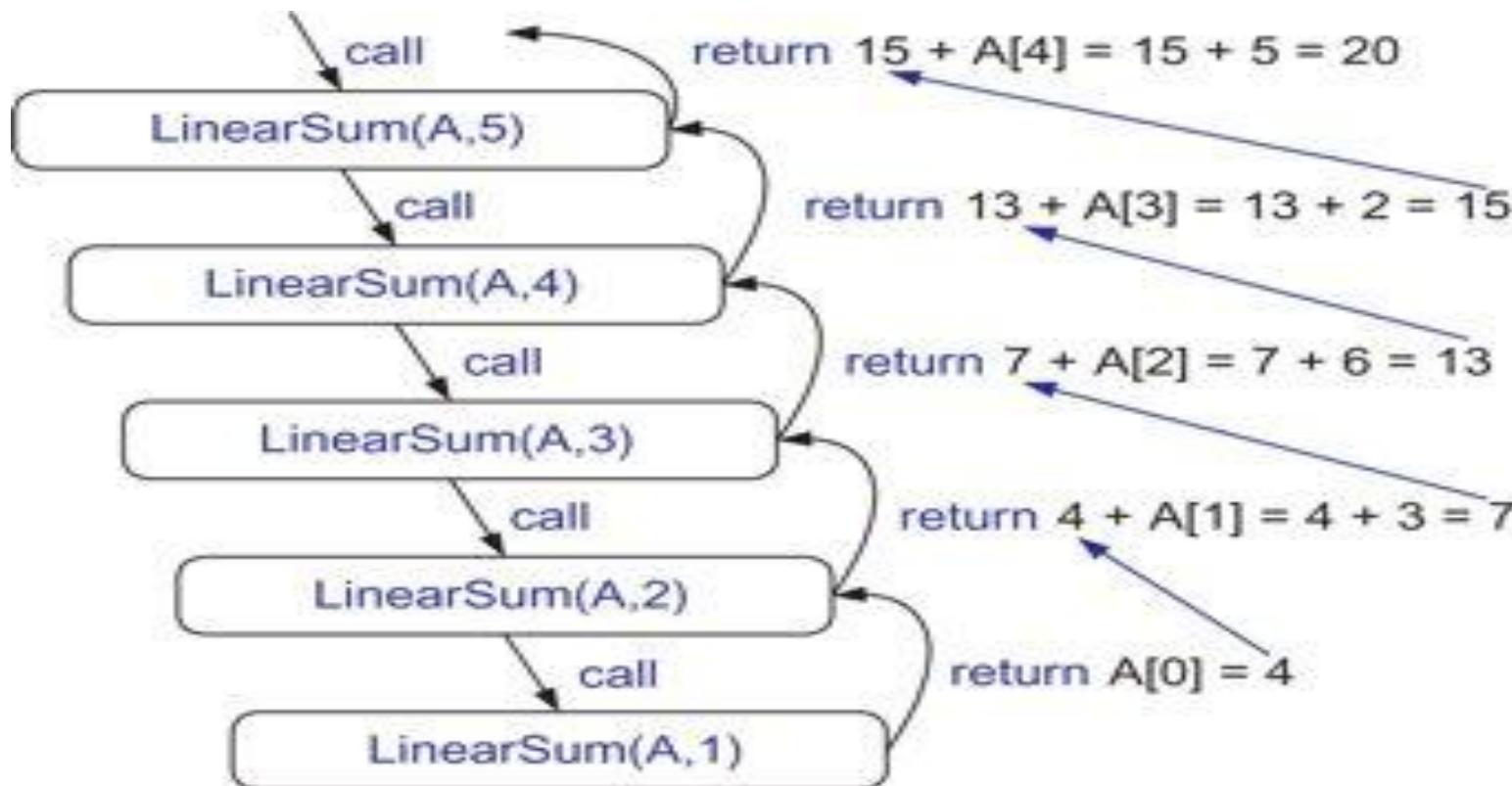
## Example 2: Summing the Elements of an Array

- We can solve this summation problem using linear recursion by observing that the sum of all  $n$  integers in an array  $A$  is:
  - Equal to  $A[0]$ , if  $n = 1$  (The array has one element), or
  - The sum of the first  $n - 1$  integers in  $A$  plus the last element

```
int LinearSum(int A[], n) {  
    if n = 1 then  
        return A[0]; // base case  
    else  
        return A[n-1] + LinearSum(A, n-1)  
}
```

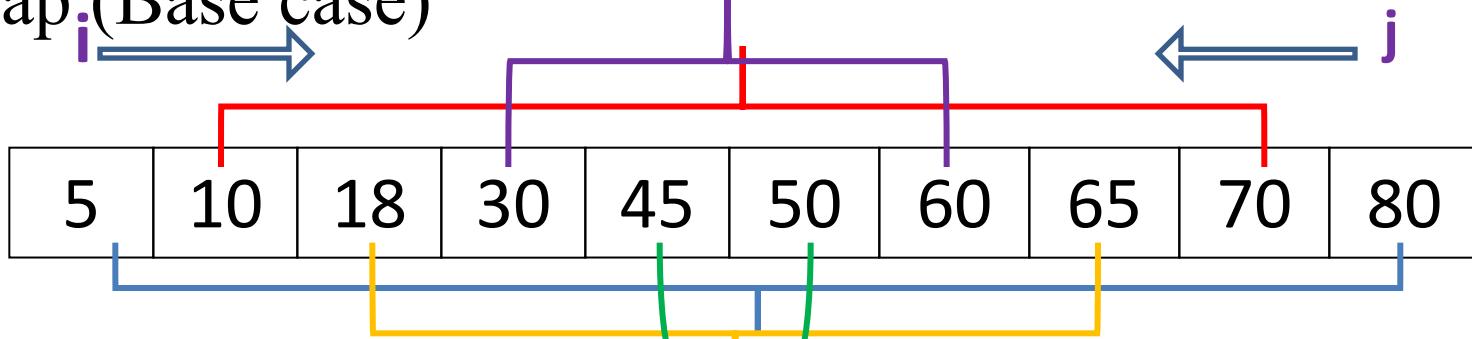
# Analyzing Recursive Algorithms using Recursion Traces

- Recursion trace for an execution of  $\text{LinearSum}(A, n)$  with input parameters  $A = [4,3,6,2,5]$  and  $n = 5$



# Linear recursion: Reversing an Array

- Swap 1<sup>st</sup> and last elements, 2<sup>nd</sup> and second to last, 3<sup>rd</sup> and third to last, and so on
- If an array contains only one element no need to swap **i**(Base case)



- Update *i* and *j* in such a way that they converge to the base case (*i* = *j*)

# Example 3: Reversing an Array

```
void reverseArray(int A[], i, j){  
    if (i < j){  
        int temp = A[i];  
        A[i] = A[j];  
        A[j] = temp;  
        reverseArray(A, i+1, j-1)  
    }  
    // in base case, do nothing  
}
```

# Linear recursion: run-time analysis

- Time complexity of linear recursion is proportional to the problem size
  - Normally, it is equal to the number of times the function calls itself
- In terms of Big-O notation time complexity of a linear recursive function/algorithm is  $O(n)$