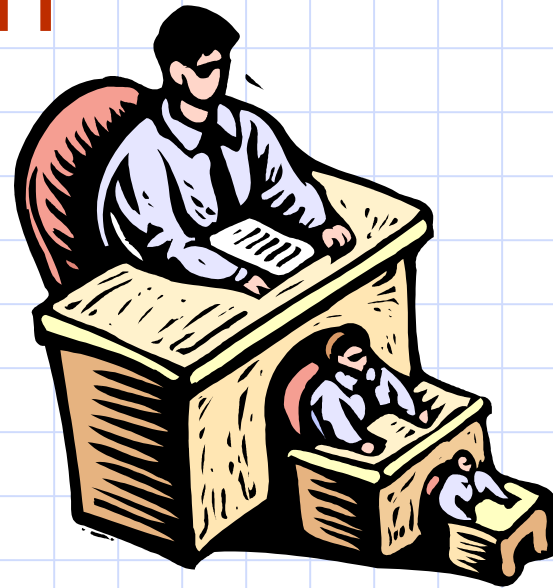


Using Recursion



The Recursion Pattern

- ❑ **Recursion:** when a method calls itself
- ❑ Classic example: 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; // basis case
    return n * recursiveFactorial(n-1); //
}
```

Linear Recursion

- ❑ 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** reach a base case
- ❑ Recur once
 - Perform a **single recursive call**
 - This step may have a test that decides which of several possible recursive calls to make, but it should ultimately make **just one** of these calls
 - Define each possible recursive call so that **it makes progress towards** a base case.

Example of Linear Recursion

LinearSum(A, n):

Input:

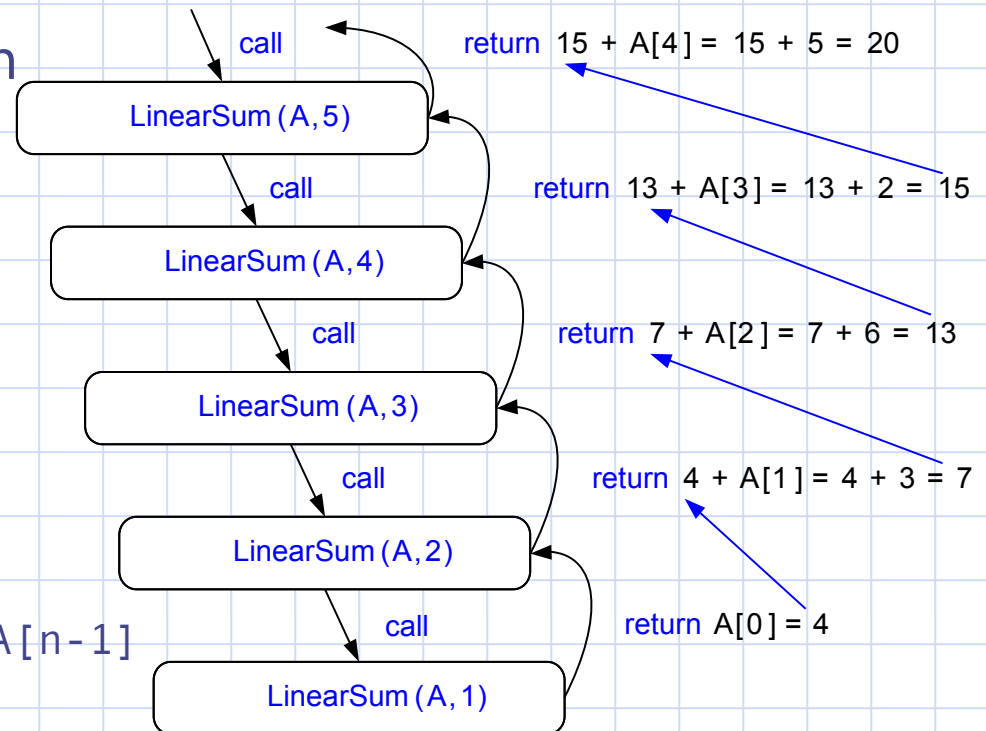
An integer array A and an integer n such that A has at least n elements

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]
```

Example recursion trace:



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
    temp = A[i]    // Swap A[i] and A[j]
    A[i] = A[j]
    A[j] = temp
    ReverseArray(A, i+1, j-1)
return
```

Computing Powers

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

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

- This leads to an power function that runs in **$O(n)$** time (and make **n** recursive calls).
- We can do better than this

Recursive Squaring

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

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

- For example,

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

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

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

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

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 $y \cdot y$

Analysis

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 $y \cdot y$

The recurrence that describes the complexity of this algorithm is

$$T(n) = T(n/2) + c$$

Each time we make a recursive call we halve the value of n ; hence, we make $\log n$ recursive calls. That is, this method runs in $O(\log n)$ time.

It is important that we use a variable twice here rather than calling the method twice.

Tail Recursion

- Tail recursion occurs when a linearly recursive method makes its recursive call as its last step (an example: the array reversal method).
- Such methods can be easily converted to non-recursive methods (which saves on some resources).

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

Binary Recursion

- Binary recursion occurs whenever there are **two** recursive calls for each non-base case.

A Binary Recursive Method

- Problem: add all the numbers in an integer array A:

Algorithm BinarySum(A, i, j):

Input: An array A, two indexes *i* and *j*

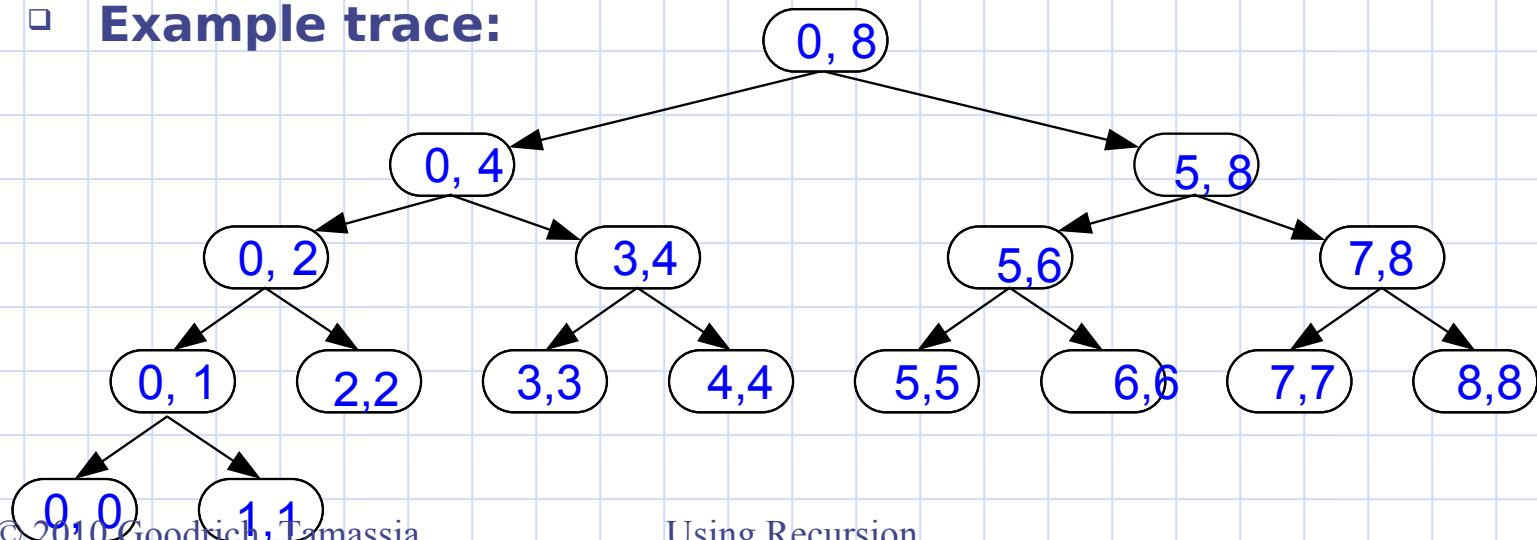
Output: The sum of the integers in A between *i* and *j*

if (*i=j*) **then**

return A[*i*]

return BinarySum(A, *i*, (*i+j*)/2) + BinarySum(A, (*i+j*)/2 +1, *j*)

- **Example trace:**



Computing Fibonacci Numbers

- Fibonacci numbers are defined recursively:

$$F_0 = 0$$

$$F_1 = 1$$

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

- Recursive algorithm (first attempt):

Algorithm **BinaryFib**(k):

Input: Non-negative integer k

Output: The k th Fibonacci number F_k

if $k \leq 1$ **return** k

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

Analysis

- Let n_k be the number of recursive calls by **BinaryFib**(k)
 - $n_0 = 1, n_1 = 1$
 - $n_2 = n_1 + n_0 + 1 = 1 + 1 + 1 = 3$
 - $n_3 = n_2 + n_1 + 1 = 3 + 1 + 1 = 5$
 - $n_4 = n_3 + n_2 + 1 = 5 + 3 + 1 = 9$
 - $n_5 = n_4 + n_3 + 1 = 9 + 5 + 1 = 15$
 - $n_6 = n_5 + n_4 + 1 = 15 + 9 + 1 = 25$
 - $n_7 = n_6 + n_5 + 1 = 25 + 15 + 1 = 41$
 - $n_8 = n_7 + n_6 + 1 = 41 + 25 + 1 = 67$
- That is, $n_k > 2^{k/2}$. It is exponential!

A Better Fibonacci Algorithm

- Use linear recursion instead

Algorithm `LinearFibonacci(k)`:

Input: A non-negative integer k

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

if $k = 1$ **then return** $(1, 0)$

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

return $(i + j, i)$

- `LinearFibonacci` makes $k-1$ recursive calls

An iterative Fibonacci Algorithm

- **Algorithm** `IterativeFibonacci(k)`:
 Input: A non-negative integer k
 Output: The k -th Fibonacci number

```
int A[] 0 = new int[k+1];  
A[0]=0; A[1] = 1;  
for (int i=2; i<=k; i++)  
    A[i] = A[i-1] + A[i-2]  
return A[k]
```