Recursion

The Pattern

- : when a method calls itself
- Classic example--the factorial function:

-
$$n! = 1 \cdot 2 \cdot 3 \cdot \cdot \cdot (n-1) \cdot n$$

• Recursive definition:

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

• As a Python method:

```
1  def factorial(n):
2   if n == 0:
3    return 1
4   else:
5   return n * factorial(n-1)
```

Content of a Recursive Method

• Base case(s)

- Values of the input variables for which we perform no recursive calls are called base cases (there should be at least one base case).
- Every possible chain of recursive calls must eventually reach a base case.

• Recursive calls

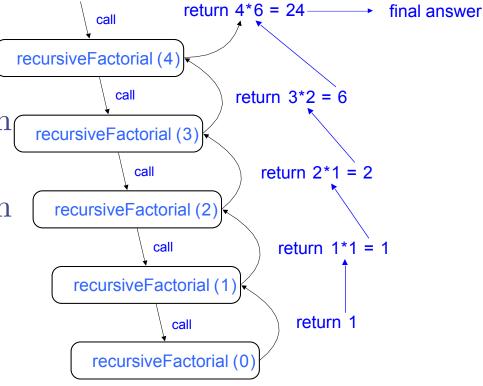
- Calls to the current method.
- Each recursive call should be defined so that it makes progress towards a base case.

Visualizing

trace

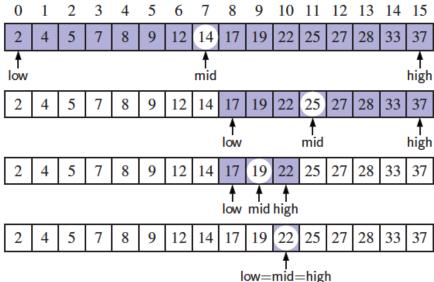
- A box for each recursive call
- An arrow from each caller to callee
- An arrow from each callee to caller showing return value

Example



Visualizing Binary Search

- We consider three cases:
 - If the target equals data[mid], then we have found the target.
 - If target < data[mid], then we recur on the first half of the sequence.
 - If target > data[mid], then we recur on the second half of the sequence.



Example of Linear

Algorithm LinearSum(A, n):

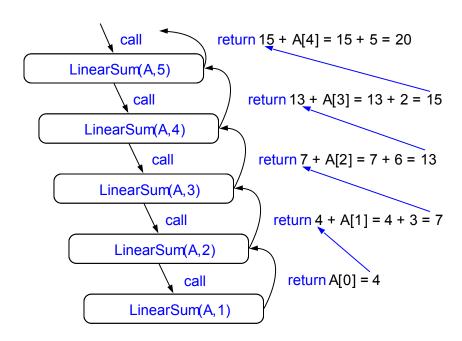
Input:

A integer array A and an integer n = 1, such that A has at least n elements

Output:

The sum of the first n integers in A

Example 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

Defining Arguments for

- In creating recursive methods, it is important to define the methods in ways that facilitate.
- For example, we defined the array reversal method as ReverseArray(A, i, j), not ReverseArray(A).

Python version:

```
def reverse(S, start, stop):
    """Reverse elements in implicit slice S[start:stop]."""
    if start < stop - 1:  # if at least 2 elements:
        S[start], S[stop-1] = S[stop-1], S[start] # swap first and last
        reverse(S, start+1, stop-1) # recur on rest</pre>
```

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 (for we make n recursive calls).
- We can do better than this, however.

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 = \operatorname{Power}(x, (n-1)/2)
         return x \cdot y \cdot y
    else
         y = Power(x, n/2)
         return y \cdot y
```

Tail Recursion

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

```
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 \operatorname{do}

Swap A[i] and A[j]

i = i + 1

j = j - 1
```

return

Another Binary Recusive Method

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

```
Algorithm BinarySum(A, i, n):

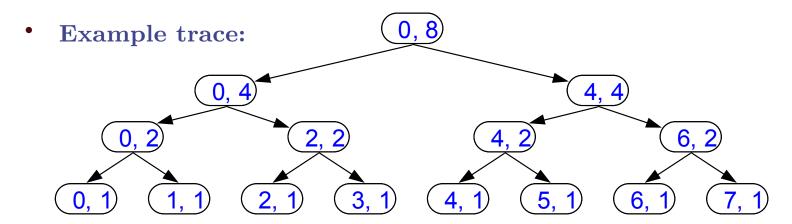
Input: An array A and integers i and n

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

if n = 1 then

return A[i]

return BinarySum(A, i, n/2) + BinarySum(A, i + n/2, n/2)
```



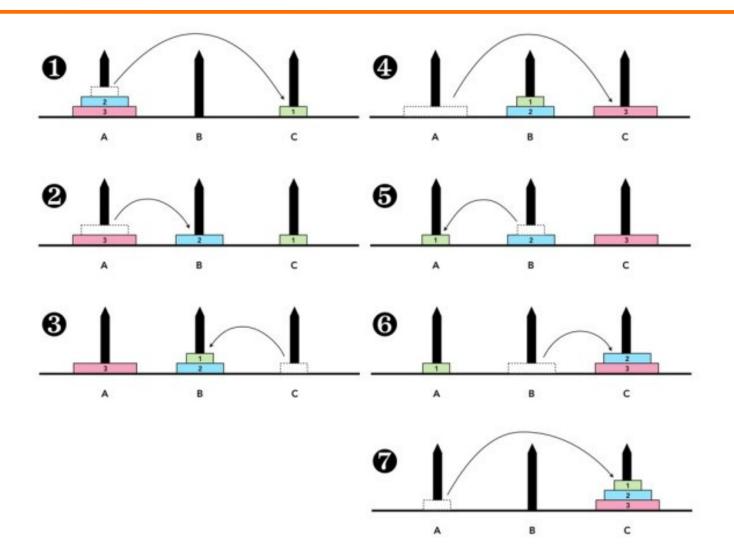
Tower of Hanoi



Procedure for moving a tower of n disks from a peg A onto a peg C, with B serving as an auxiliary peg:

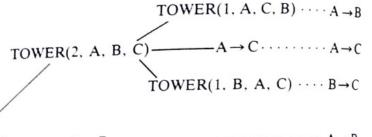
- 1)If n > 1, then first use this procedure to move the n 1 smaller disks from peg A to peg B.
- 2) Now the largest disk, i.e. disk n can be moved from peg A to peg C.
- 3)If n > 1, then again use this procedure to move the n 1 smaller disks from peg B to peg C.

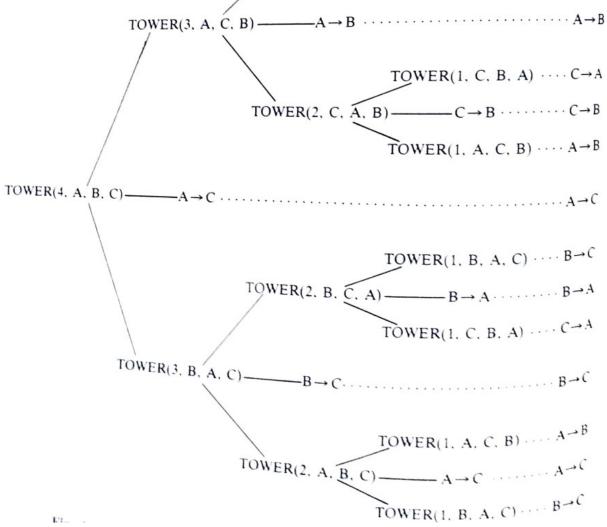
Tower of Hanoi for 3 disks



Tower of Hanoi

Recursion trace for 4 disks





Divide & Conquer methods

- **Divide** the problem into one or more subproblems that are smaller instances of the same problem.
- Conquer the subproblems by solving them recursively.
- Combine the subproblem solutions to form a solution to the original problem.

Merge Sort Analysis

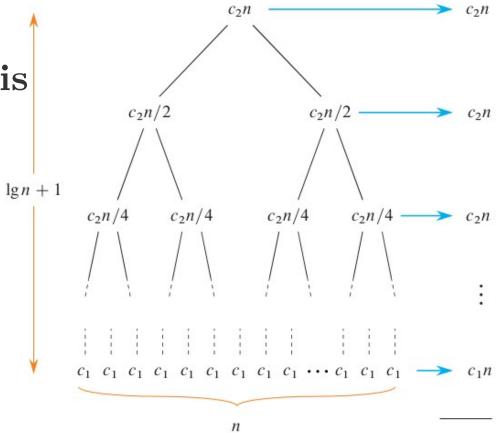
$$T(n) = \begin{cases} \Theta(1) & \text{if } n < n_0, \\ D(n) + aT(n/b) + C(n) & \text{otherwise}. \end{cases}$$

$$T(n) = \begin{cases} c_1 & \text{if } n = 1, \\ 2T(n/2) + c_2 n & \text{if } n > 1, \end{cases}$$

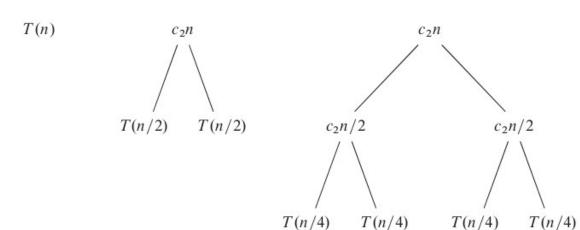
$$\begin{array}{c|ccccc}
p & q & & r \\
5 & 6 & 7 & 8 \\
\hline
14 & 6 & 11 & 2
\end{array}$$

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Merge Sort Analysis



Total: $c_2 n \lg n + c_1 n$



Iterative Substitution

In the iterative substitution, or "plug-and-chug," technique, we iteratively apply the recurrence equation to itself and see if we can find a pattern: T(x) = 2T(x/2) + 2T(x/2)

$$T(n)=2T(n/2)+c_{2}n$$

$$2(2T(n/2^{2}))+c_{2}(n/2))+c_{2}n$$

$$T(n)=2^{2}T(n/2^{2})+2c_{2}n$$

$$T(n)=2^{3}T(n/2^{3})+3c_{2}n$$

$$T(n)=2^{4}T(n/2^{4})+4c_{2}n$$
...
$$T(n)=2^{i}T(n/2^{i})+ic_{2}n$$

Note that base, $T(1)=c_1$, case occurs when $2^i=n$. That is, $i=\log n$. So, $T(n)=c_1n+c_2n\log n$

Thus, T(n) is $O(n \log n)$.

Master Method

Many divide-and-conquer recurrence equations have the form:

$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

The Master Theorem:

- 1. if f(n) is $O(n^{\log_b a \varepsilon})$, then T(n) is $\Theta(n^{\log_b a})$
- 2. if f(n) is $\Theta(n^{\log_b a} \log^k n)$, then T(n) is $\Theta(n^{\log_b a} \log^{k+1} n)$
- 3. if f(n) is $\Omega(n^{\log_b a + \varepsilon})$, then T(n) is $\Theta(f(n))$, provided $af(n/b) \le \delta f(n)$ for some $\delta < 1$.

Master Method, Example 1

The form:
$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

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- 3. if f(n) is $\Omega(n^{\log_b a + \varepsilon})$, then T(n) is $\Theta(f(n))$, provided $af(n/b) \le \delta f(n)$ for some $\delta < 1$.

Example:

$$T(n) = 4T(n/2) + n$$

Solution: $\log_b a = 2$, so case 1 says T(n) is O(n²).

References

Data Structures and Algorithms in Python Michael T. Goodrich, Roberto Tamassia, Michael H. Goldwasser

Introduction to Algorithms Leiserson, Stein, Rivest, Cormen

Algorithms, 4th Edition Robert Sedgewick and Kevin Wayne

Some images from the internet