

Laboratório de Programação Avançada 2018/19

Week 2 – Recursion



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Recursive methods are either:

- Tail recursive: recursive call is the last statement
- Nontail recursive

Advantages of tail recursive methods:

- Tail recursive can be transformed into iterative: gain in efficiency
- Some compilers can detect tail recursion to optimize code
- Used to implement loops in languages that do not support loop structures explicitly (e.g. prolog)

Recursion and iteration

A nontail recursive method for adding up from 1 to n :

Function $sum(n)$

if $n = 0$ then {base case}

return 0

else {recursive step}

return $sum(n - 1) + n$

A tail recursive method ($sum(n, 0)$ in the first call):

Function $sum(n, res)$

if $n = 0$ then {base case}

return res

else {recursive step}

return $sum(n - 1, n + res)$

Recursion and iteration

A nontail recursive method for computing the factorial of n :

Function $fact(n)$

if $n \leq 1$ then {base case}

return 1

else {recursive step}

return $n \cdot fact(n - 1)$

A tail recursive method ($fact(n, 1)$ in the first call):

Function $fact(n, res)$

if $n \leq 1$ then {base case}

return res

else {recursive step}

return $fact(n - 1, n \cdot res)$

Time complexity of recursion

- The time complexity of recursive programs may not be easy to derive
- Guess the recurrence (e.g., by *unrolling* it) and show that is correct by induction or use the Master Theorem, if applicable
- Reading material: Cormen et al, Introduction to Algorithms, chapter 4.

Time complexity of recursion

A recursive algorithm to print all binary strings of size n

Function *binaryString*(n , Q)

if $n = 0$ **then** {base case}

print Q

else

push('0', Q)

binaryString($n - 1$, Q) {1st recursive step}

pop(Q)

push('1', Q)

binaryString($n - 1$, Q) {2nd recursive step}

pop(Q)

return

What is the time complexity of this algorithm?

Time complexity of recursion

A first guess: Unrolling the recurrence

Let T_q denote the cost of queue operations at each recursive step.

$$\begin{aligned}T(n) &= 2T(n-1) + T_q \\&= 2(2T(n-2) + T_q) + T_q \\&= 4T(n-2) + 3T_q \\&= 4(2T(n-3) + T_q) + 3T_q \\&= 8T(n-3) + 7T_q \\&\dots \\&= 2^k T(n-k) + (2^k - 1)T_q \\&\dots \\&= 2^n T(0) + (2^n - 1)T_q \in O(2^{n+1})\end{aligned}$$

Time complexity of recursion

Master Theorem

The time complexity of some recursive programs can be defined as:

$$T(n) = \begin{cases} aT(n/b) + n^c & \text{if } n > 1 \\ d & \text{if } n = 1 \end{cases}$$

- n is the problem size
- $a \geq 1$ is the number of subproblems at each recursive step
- $b > 1$ is the size of each subproblem
- $d \geq 0$ is cost of the base case
- n^c is the cost of each recursive step

Master Theorem

$$T(n) = \begin{cases} aT(n/b) + n^c & \text{if } n > 1 \\ d & \text{if } n = 1 \end{cases}$$

- If $\log_b a < c$, $T(n) = \Theta(n^c)$
- If $\log_b a = c$, $T(n) = \Theta(n^c \log n)$
- If $\log_b a > c$, $T(n) = \Theta(n^{\log_b a})$

Time complexity of recursion

Merge Sort

Function *MergeSort*(*A*, *low*, *high*)

if *low* = *high* **then** {base case}

return

else

$mid = (low + high) / 2$

MergeSort(*A*, *low*, *mid*) {1st recursive step}

MergeSort(*A*, *mid* + 1, *high*) {2nd recursive step}

 Merge the sorted lists from the two previous steps

return

What is the time complexity for an array of size n ?

Master Theorem

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

- If $\log_b a < c$, $T(n) = \Theta(n^c)$
- If $\log_b a = c$, $T(n) = \Theta(n^c \log n)$
- If $\log_b a > c$, $T(n) = \Theta(n^{\log_b a})$

The time complexity of merge sort is $\Theta(n \log n)$.

Time complexity of recursion

Binary Search

Function *binarySearch*(*A*, *low*, *high*, *v*)

$mid = (low + high) / 2$

if $A[mid] = v$ **then** {base case}

return *mid*

if $low = high$ **then** {base case}

return -1

if $A[mid] \leq v$ **then**

return *binarySearch*(*A*, *low*, *mid*, *v*) {1st recursive step}

else

return *binarySearch*(*A*, *mid* + 1, *high*, *v*) {2nd recursive step}

What is the time complexity for an array of size n ?

Master Theorem

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

- If $\log_b a < c$, $T(n) = \Theta(n^c)$
- If $\log_b a = c$, $T(n) = \Theta(n^c \log n)$
- If $\log_b a > c$, $T(n) = \Theta(n^{\log_b a})$

The time complexity of binary search is $\Theta(\log n)$.

Balanced Binary Tree Traversal

Function *binaryTreeSearch*(*node*)

print(*node*)

if *left*(*node*) \neq NULL **then**

binaryTreeSearch(*left*(*node*))

{1st recursive step}

if *right*(*node*) \neq NULL **then**

binaryTreeSearch(*right*(*node*))

{2nd recursive step}

What is the time complexity for a tree with n nodes?

Master Theorem

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

- If $\log_b a < c$, $T(n) = \Theta(n^c)$
- If $\log_b a = c$, $T(n) = \Theta(n^c \log n)$
- If $\log_b a > c$, $T(n) = \Theta(n^{\log_b a})$

The time complexity of balanced binary tree traversal is $\Theta(n)$.

Strassen's Algorithm for Matrix Multiplication

Function *Strassen* $\left(A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}, B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}, n \right)$

if $n = 1$ **then**

$$C = A \cdot B$$

else

$$M_1 = \text{Strassen}(A_{11} + A_{22}, B_{11} + B_{22}, n/2)$$

$$M_2 = \text{Strassen}(A_{21} + A_{22}, B_{11}, n/2)$$

$$M_3 = \text{Strassen}(A_{11}, B_{12} - B_{22}, n/2)$$

$$M_4 = \text{Strassen}(A_{22}, B_{21} - B_{11}, n/2)$$

$$M_5 = \text{Strassen}(A_{11} + A_{12}, B_{22}, n/2)$$

$$M_6 = \text{Strassen}(A_{21} - A_{11}, B_{11} + B_{12}, n/2)$$

$$M_7 = \text{Strassen}(A_{12} - A_{22}, B_{21} + B_{22}, n/2)$$

$$C = \begin{bmatrix} M_1 + M_4 - M_5 + M_7 & M_3 + M_5 \\ M_2 + M_4 & M_1 - M_2 + M_3 + M_6 \end{bmatrix}$$

return C

Master Theorem

$$T(n) = \begin{cases} 7T(n/2) + O(n^2) & \text{if } n > 1 \\ O(1) & \text{if } n = 1 \end{cases}$$

- If $\log_b a < c$, $T(n) = \Theta(n^c)$
- If $\log_b a = c$, $T(n) = \Theta(n^c \log n)$
- If $\log_b a > c$, $T(n) = \Theta(n^{\log_b a})$

The time complexity is $\Theta(n^{\log_2 7}) \approx O(n^{2.8074})$.