CS483-04 Non-recursive and Recursive Algorithm **Analysis**

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http://www.cs.gmu.edu/~ lifei/teaching/cs483 fall07/

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Outline

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- Review and More
- ➤ Analysis of Non-recursive Algorithms
- Analysis of Recursive Algorithms
- Examples

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Review

• $O(g(n)) := \{f(n) \mid \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \le f(n) \le c \cdot g(n) \}$ for all $n \geq n_0$ }.

$$f(n) \in O(g(n))$$

f(n) grow no faster than g(n).

• $\Omega(g(n)) := \{f(n) \mid \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \le c \cdot g(n) \le f(n) \}$ for all $n \geq n_0$ }.

$$f(n) \in \Omega(g(n))$$

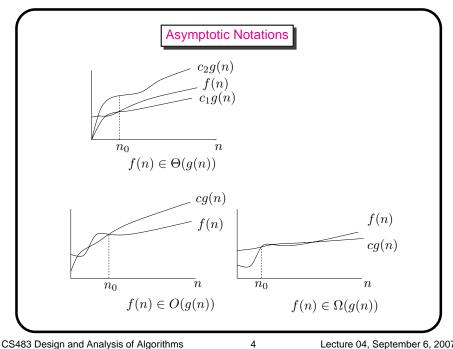
f(n) grows at least as fast as g(n).

• $\Theta(g(n)) := \{f(n) \mid \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that }$ $c_1 \cdot g(n) \le f(n) \le c_2 \cdot g(n)$ for all $n \ge n_0$ }.

$$f(n) \in \Theta(g(n))$$

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f(n) grows at the same rate as g(n).



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Review

- ➤ Tools and techniques to get asymptotic notation
 - L'Hopital's rule

If $\lim_{n\to\infty}f(n)=\lim_{n\to\infty}g(n)=\infty$ and the derivatives f' and g' exist, then

$$\lim_{n \to \infty} \frac{f(n)}{g(n)} = \lim_{n \to \infty} \frac{f'(n)}{g'(n)}$$

• Stirling's formula

$$n! \approx \sqrt{2\pi n} (\frac{n}{e})^n$$

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where e is the base of natural logarithm, $e \approx 2.718$. $\pi \approx 3.1415$.

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Exercises

- ightharpoonup All logarithmic functions $\log_a n$ belong to the same class $\Theta(\log n)$ no matter what the logarithmic base a>1 is.
- All polynomials of the same degree k belong to the same class $a_k n^k + a_{k-1} n^{k-1} + \cdots + a_1 n + a_0 \in \Theta(n^k)$.
- Exponential functions a^n have different orders of growth for different a's, i.e., $2^n \notin \Theta(3^n)$.

 $\ln n < (\ln n)^2 < \sqrt{n} < n < n \cdot \ln n < n^2 < n^3 < 2^n < n! < n^n$

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Some Properties of Asymptotic Order of Growth

- ➤ Transitivity
 - $f(n) \in O(g(n))$ and $g(n) \in O(h(n)) \Rightarrow f(n) \in O(h(n))$
 - $f(n) \in \Theta(g(n))$ and $g(n) \in \Theta(h(n)) \Rightarrow f(n) \in \Theta(h(n))$
 - $f(n) \in \Omega(g(n))$ and $g(n) \in \Omega(h(n)) \Rightarrow f(n) \in \Omega(h(n))$
- Reflexivity
 - $f(n) \in O(f(n))$
 - $f(n) \in \Theta(f(n))$
 - $f(n) \in \Omega(f(n))$
- Symmetry and Transpose Symmetry
 - $f(n) \in \Theta(g(n))$ if and only if $g(n) \in \Theta(f(n))$
 - $f(n) \in O(g(n))$ if and only if $g(n) \in \Omega(f(n))$

Outline

- Review
- Analysis of Non-recursive Algorithms
- Analysis of Recursive Algorithms
- Examples

Time Efficiency of Non-recursive Algorithms

- ullet Decide on parameter n indicating input size.
- Identify algorithm's basic operation.
- Determine worst, average, and best cases for input of size *n*.
- Sum the number of basic operations executed.
- ullet Simplify the sum using standard formula and rules (see Appendix A).

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Time Efficiency of Non-recursive Algorithms

- $\sum_{i=l}^{u} 1 = 1 + 1 + \dots + 1 = (u-l) + 1$.
- $\sum_{i=1}^{n} i = 1 + 2 + \dots + n = \frac{n(n+1)}{2} \approx \frac{n^2}{2} \in \Omega(n^2).$
- $\sum_{i=1}^{n} i^2 = 1^2 + 2^2 + \dots + n^2 = \frac{n(n+1)(2n+1)}{6} \approx \frac{n^3}{3} \in \Omega(n^3).$
- $\sum_{i=1}^{n} a^i = 1 + a + a^2 + \dots + a^n = \frac{a^{n+1}-1}{a-1}, \forall a \neq 1.$

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$\ \, \hbox{Example $1:$ Maximum Element} \\$

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- Determine the value of the largest element in a given array.
- Input: An array $A[0, \cdots, n-1]$ of real numbers.
- ullet Output: The value of the largest element in A.

Algorithm 0.1: $\text{MAXELEMENT}(A[0,\cdots n-1])$

$$\begin{aligned} max &= A[0] \\ &\text{for } i = 1 \text{ to } n-1 \\ &\text{do } \begin{cases} &\text{if } A[i] > max \\ &\text{then } max = A[i] \end{cases} \end{aligned}$$

Example 2: Element Uniqueness Problem

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- Determine whether all the elements in a given array are distinct.
- $\bullet \ \ \text{Input: An array} \ A[0,\dots,n-1].$
- Output: Returns "true" if all the elements in A are distinct and "false" otherwise.

Algorithm 0.2: UniqueElements($A[0,\cdots n-1]$)

$$\begin{aligned} & \text{for } i = 0 \text{ to } n-2 \\ & \text{do } \begin{cases} & \text{for } j = i+1 \text{ to } n-1 \\ & \text{do } \end{cases} & \text{if } A[i] = A[j] \\ & \text{then return } (false) \end{aligned}$$
 return $(true)$

Example 3: Matrix Multiplication

- Multiply 2 n-by-n matrices by the definition-based algorithm.
- ullet Input: 2 n-by-n matrices A and B.
- Output: Matrix $C = A \cdot B$.

Algorithm 0.3: MATRIXMULTI(A, B)

$$\begin{aligned} & \text{for } i=0 \text{ to } n-1 \\ & \text{do} \left\{ \begin{aligned} & \text{for } j=0 \text{ to } n-1 \\ & \text{do} \left\{ \begin{aligned} & C[i,j]=0 \\ & \text{for } k=0 \text{ to } n-1 \\ & \text{do } C[i,j]=C[i,j]+A[i,k] \cdot B[k,j] \end{aligned} \right. \end{aligned} \end{aligned}$$
 return (C)

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Example 4: Counting Binary Bits

- ullet Input: A positive decimal integer n.
- Output: The number of binary digits in *n*'s binary representation.

Algorithm 0.4: COUNTBINARYBITS(n)

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- ➤ Review
- ➤ Analysis of Non-recursive Algorithms
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Recurrences

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➤ A recurrence is an equation or inequality that describes a function in terms of its value over a smaller value.

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ightharpoonup Example: Find n!

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Recurrences

- ➤ A recurrence is an equation or inequality that describes a function in terms of its value over a smaller value.
- **Example**: Find n!
 - Non-recursive: $1 \cdot 2 \cdot 3 \cdot \cdot \cdot n$

Algorithm 0.5: FINDFACTORIAL- $\alpha(n)$

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\begin{aligned} &factorial = 1\\ &\text{for } i = 1 \text{ to } n\\ &\text{do } factorial = factorial \cdot i\\ &\text{return } (factorial) \end{aligned}
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• Recurrence: $f(n) = n \cdot f(n-1)$

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Algorithm 0.6: FINDFACTORIAL- $\beta(n)$

$$\quad \text{if } n=0$$

then return (1)

else return $(n \cdot \text{FindFactorial} - \beta(n-1))$

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Example: Counting Number of Bits

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- Input: A positive decimal integer n.
- ullet Output: The number of binary digits in n's binary representation.

Algorithm 0.7: Non-RecursiveCount(n)

$$\begin{aligned} & count = 1 \\ & \text{while } n > 1 \\ & \text{do} \begin{cases} & count = count + 1 \\ & n = \lfloor n/2 \rfloor \end{cases} \end{aligned}$$

$$\text{return } (count)$$

Algorithm 0.8: RECURSIVECOUNT(n)

if
$$i=1$$

$$\label{eq:continuous} \mbox{do return } (1)$$

$$\mbox{else return } (RecursiveCount(\lfloor n/2 \rfloor) + 1)$$

Example: Fibonacci Numbers

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• Output: A sequence of numbers $F_0, F_1, F_2, \ldots, F_n, \ldots$ such that

$$F_n = \begin{cases} F_{n-1} + F_{n-2}, & \text{if } n > 1 \\ 1, & \text{if } n = 1 \\ 0, & \text{if } n = 0. \end{cases}$$

Example: Fibonacci Numbers

• Output: A sequence of numbers $F_0, F_1, F_2, \ldots, F_n, \ldots$ such that

$$F_n = \begin{cases} F_{n-1} + F_{n-2}, & \text{if } n > 1 \\ 1, & \text{if } n = 1 \\ 0, & \text{if } n = 0. \end{cases}$$

Algorithm 0.9: FIBNUMBER(n)

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if n=0 return (0) if n=1 return (1) else return (FibNumber(n-1)+FibNumber(n-2))
```

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Example: Hanoi Tower Problem

- Move all the disks from peg a to peg c. Large disk cannot be on top of a smaller one.
- Input: n disks in order of sizes on peg a. 3 pegs a, b, and c

http://en.wikipedia.org/wiki/Tower of Hanoi

Algorithm 0.10: HanoiTower(n, a, c, b)

if n=1

Move the disk from a to c

HanoiTower(n-1,a,b,c)

else \langle Move the largest disk from a to c

HanoiTower(n-1,b,c,a)

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Analysis of Recursive Algorithms

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- ➤ The iteration method
 - Expand (iterate) the recurrence and express it as a summation of terms depending only on n and the initial conditions.
- > The substitution method
- Master Theorem

(To be introduced in Chapter 4.)

Iteration Method: Examples

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 \bullet n!

$$T(n) = T(n-1) + 1$$

Tower of Hanoi

$$T(n) = 2T(n-1) + 1$$

Iteration: Example

•
$$n! (T(n) = T(n-1) + 1)$$

$$T(n) = T(n-1)+1$$

= $(T(n-2)+1)+1$
= $T(n-2)+2$
... = $T(n-i)+i$
... = $T(0)+n=n$

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• Tower of Hanoi (T(n) = 2T(n-1) + 1) ???

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Assignment 1

- ➤ Problems
 - 1. Prove or find a counter-example:

$$\left(\frac{n}{3}\right)^n < n! < \left(\frac{n}{2}\right)^n, \quad \text{if } n \ge 6.$$

- 2. p. 8, Exercises (1.1) 5, 6.
- 3. p. 60, Exercises (2.2) 5, 6
- 4. p. 67, Exercises (2.3) 2, 4
- 5. p. 76, Exercises (2.4) 1, 3, 5
- Due date: September 20, 2007. In class

Iteration: Example

- n! (T(n) = T(n-1) + 1)
- Tower of Hanoi (T(n) = 2T(n-1) + 1)

$$T(n) = 2T(n-1)+1$$

$$= 2(2T(n-2)+1)+1$$

$$= 2^{2}T(n-2)+2+1$$
...
$$= 2^{i}T(n-i)+2^{i-1}+\dots+1$$
...
$$= 2^{n-1}T(1)+2^{n-1}+2^{n-1}+\dots+1$$

$$= 2^{n-1}T(1)+\sum_{i=0}^{n-2} 2^{i}$$

$$= 2^{n-1}+2^{n-1}-1$$

$$= 2^{n}-1$$

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