# CS2223 Algorithms B Term 2013

## Exam 1. November 15, 2013 - SOLUTIONS

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### Instructions:

- Show your work and justify your answers
- Use the space provided to write your answers
- Ask in case of doubt

## Problem I. [15 Points] Asymptotic Growth of Functions.

Prove in detail that for any constant  $a \ge 0$ :  $n^a log(n) = O(n^{a+1})$ . Show your work.

**Solution:** For illustration purposes, we include two alternative solutions.

**Solution 1:** using the definition of Big-Oh:

We need to find constants c > 0 and  $n_0 > 0$  such that for all  $n \ge n_0$ ,  $n^a \log(n) \le c n^{a+1}$ .

Note that since  $a \geq 0$  and n > 0 then  $n^a > 0$ . Also, we know that for all n > 0,  $\log(n) \leq n$ . Hence,  $n^a \log(n) \leq n^a n = n^{a+1}$ . Therefore for constants c = 1 and  $n_0 = 1$ , we have that for all  $n \geq n_0$ ,  $n^a \log(n) \leq c n^{a+1}$ . Hence,  $n^a \log(n) = O(n^{a+1})$ .

**Solution 2:** using the limit rule:

$$\lim_{n\to\infty} \frac{n^{a}\log(n)}{n^{a+1}} = \lim_{n\to\infty} \frac{\log(n)}{n} = \lim_{n\to\infty} \frac{1/n}{1} \text{ (using de L'Hôpital's rule)} = 0$$

Since this limit exists and is equal to 0, then  $n^{a}log(n) = O(n^{a+1})$ .

## Problem II. [30 points] Runtime Analysis

The bubleSort algorithm receives a list as its input and returns this list sorted in increasing order. (Algorithm below adapted from http://interactivepython.org/courselib/static/pythonds/SortSearch/sorting.html)

def bubbleSort(alist): Cost per instruction Number of repetitions  $\ldots \ldots c_1 \ldots \ldots \ldots \ldots$ 1. n = length(alist) 2. for j in [n-1, n-2, ..., 1]:  $\dots \dots n-1 \dots \dots$  $\ldots c_2 \ldots c_2$ for i in [0, 1, 2, ..., i-1]:  $\dots n(n-1)/2 \dots$ 3.  $\dots \dots C_3 \dots \dots$ if alist[i] > alist[i+1]:  $\ldots \ldots C_4 \ldots \ldots$  $\dots n(n-1)/2 \dots$ 4.  $\dots n(n-1)/2 \dots$ 5. temp = alist[i]  $\dots n(n-1)/2 \dots$  $\dots$   $C_6$   $\dots$   $C_6$ 6. alist[i] = alist[i+1]  $\dots n(n-1)/2 \dots$ 7. alist[i+1] = temp  $\ldots \ldots C_7 \ldots \ldots$ 8. return(alist)  $\ldots \ldots c_8 \ldots \ldots \ldots \ldots 1 \ldots \ldots 1$ 

- 1. [20 Points] Use worst case analysis to construct a function T(n) that gives the runtime of this algorithm as a function of n, the length of the input list. Notes:
  - Instructions 1 and 8: Assume that they are executed in constant time (as shown above).
  - Java's equivalent of instruction 2 is: for (int j = n-1; j >= 1; j--)
  - Java's equivalent of instruction 3 is: for (int i = 0; i < j; i++)</li>

Show your work step by step, and justify your answer.

**Solution:** We describe below our runtime analysis instruction by instruction:

- **1.** Provided in the problem statement: constant time.
- 2. j varies from n-1 to 1, so the condition of this loop is executed n-1 times. If we follow the textbook convention we'd add 1, for a total of n, to include the final check of the loop condition.
- 3-7 i varies from 0 to j-1. So the number of times that this loop is executed is:

$$j = n-1$$
:  $i = 0, 1, ..., n-4, n-3, n-2$ :  $n-1 \text{ times}$  (= j times)  $j = n-2$ :  $i = 0, 1, ..., n-4, n-3$ :  $n-2 \text{ times}$  (= j times)  $j = n-3$ :  $i = 0, 1, ..., n-4$ :  $n-3 \text{ times}$  (= j times) ... ... ... ... ...  $j = 1$ :  $i = 0$ :  $1 \text{ time}$  (= j times)

Hence, the total number of times that each of these instructions is executed is  $\sum_{j=1}^{n-1} j = \frac{(n-1)n}{2}$ .

**Note:** If we follow the convention in the textbook that the condition of the loop is executed 1 more time that the body of the loop (and hence each row count in the tabulation above will be incremented by 1) the number of times that instruction 3 would be executed is:

$$\sum_{j=2}^{n} j = \frac{(n+1)n}{2} - 1 = \frac{(n+2)(n-1)}{2}$$

8. Provided in the problem statement: constant time.

Hence, 
$$T(n)=c_1+c_2(n-1)+(c_3+c_4+c_5+c_6+c_7)\frac{(n-1)n}{2}+c_8=k_2n^2+k_1n+k_0$$
 for constants  $k_0=c_1-c_2+c_8$ ;  $k_1=c_2-\frac{c_3+c_4+c_5+c_6+c_7}{2}$ ;  $k_2=\frac{c_3+c_4+c_5+c_6+c_7}{2}$ .

2. [10 points] Provide an asymptotic upper bound g(n) as tight as possible for your T(n) function above and prove in detail that T(n) = O(g(n)).

**Solution:** Let  $g(n) = n^2$ .

Claim: 
$$T(n) = k_2 n^2 + k_1 n + k_0 = O(n^2)$$

<u>Proof:</u> We need to find constants c>0 and  $n_0>0$  such that for all  $n\geq n_0$ ,  $T(n)\leq c\ g(n)$ . Note that  $k_1n\leq k_1n^2$  and  $k_0\leq k_0n^2$  for all  $n\geq 1$ . Take  $c=k_0+k_1+k_2$  and  $n_0=1$ . Then,

for all 
$$n \ge n_0$$
,  $T(n) \le c g(n)$ , and so  $T(n) = O(g(n)) = O(n^2)$ .

## Problem III. [25 points] Transpose Symmetry of Big-O and Big-Omega

Let f(n) and g(n) be asymptotically positive functions. Use the definition of Big-O and Big-Omega to prove in detail that

$$f(n) = O(g(n))$$
 if and only if  $g(n) = \Omega(f(n))$ 

1. [10 points] Prove that if f(n) = O(g(n)) then  $g(n) = \Omega(f(n))$ .

### Solution:

If f(n) = O(g(n)), then there exist constants c > 0 and  $n_0 > 0$  such that for all  $n \ge n_0$ ,  $f(n) \le c \ g(n)$ . Take  $k = \frac{1}{c}$ . Since c > 0 then k > 0. Note that for all  $n \ge n_0$ ,  $kf(n) \le g(n)$  and so  $g(n) = \Omega(f(n))$ .

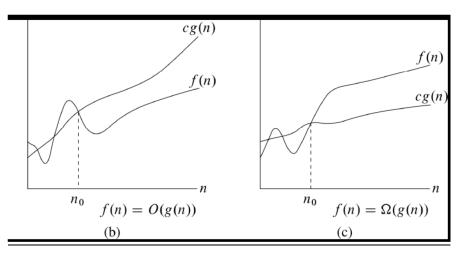
2. [10 points] Prove that if  $g(n) = \Omega(f(n))$  then f(n) = O(g(n)).

#### Solution:

If  $g(n)=\Omega(f(n))$ , then there exist constants k>0 and  $n_0>0$  such that for all  $n\geq n_0$ ,  $g(n)\geq k$  f(n). Take  $c=\frac{1}{k}$ . Since k>0 then c>0. Note that for all  $n\geq n_0$ ,  $f(n)\leq cg(n)$  and so, f(n)=O(g(n)).

3. [5 points] Explain in words and with plots what it means intuitively for a function f(n) to be O(g(n)) or for f(n) to be O(g(n)).

Solution: Graphs taken for the textbook: T.H. Cormen, C.E. Leiserson, R.L. Rivest, and C. Stein. Introduction to Algorithms (Third Edition). MIT Press. 2009. f(n) = O(g(n)) means that for large enough n's, a constant multiple of g(n) is an upper bound for f(n). In other words, the growth rate of g(n) is greater than or equal to that of f(n) as n goes to infinite. Similarly,  $f(n) = \Omega(g(n))$  means



that for large enough n's, a constant multiple of g(n) is a lower bound for f(n). In other words, the growth rate of g(n) is smaller than or equal to that of f(n) when n goes to infinite.

# Problem IV. [35 Points] Asymptotic Growth of Functions.

Consider the following functions:

$$f_1(n) = 3^n$$

$$f_2(n) = 7$$
 (that is,  $f_2(n)$  is a constant function that always returns 7).

$$f_3(n) = n^5$$

1. [5 points] List the above functions in ascending order of growth rate. That is, if function g(n) immediately follows function f(n) in your list, then it should be the case that f(n) = O(g(n)).

Your list: **Solution:** 
$$7 < n^5 < 3^n$$

2. [20 points] Prove in detail that your list is correct. That is, prove that f(n) = O(g(n)) for every pair of functions f(n) and g(n), where g(n) immediately follows f(n) on your list above.

#### **Solutions:**

We provide 2 alternative proofs of each result, but one proof suffices.

Proof: 
$$7 = O(n^5)$$
:

Proof using the definition of Big-Oh:

Take 
$$c=7, n_0=1$$
. Then, for all  $n \ge n_0$ :  $7 \le c \, n^5$ , and so  $7=O(n^5)$ .

Note also that  $n^5 \neq O(7)$  since there are no constants c > 0,  $n_0 > 0$  such that  $n^5 \leq 7c$  for all  $n \geq n_0$ . As a consequence of this,  $n^5 \neq O(7)$ .

Proof using the limit rule:

$$\lim_{n \to \infty} \frac{7}{n^5} = 0$$
. Hence,  $7 = O(n^5)$  and  $n^5 \neq O(7)$ .

Proof: 
$$n^5 = O(3^n)$$
:

Proof using the limit rule:

$$\lim_{n\to\infty} \frac{n^5}{3^n} = \lim_{n\to\infty} \frac{5n^4}{\ln(3)*3^n}$$
 (using de L'Hôpital's rule) =

$$\lim_{n\to\infty}\frac{5*4*3*2*1}{\ln(3)^5*3^n} \text{(using de L'Hôpital's rule 4 more times)} = 0.$$

Hence, 
$$n^5 = O(3^n)$$
 and  $3^n \neq O(n^5)$ . As a consequence of this,  $3^n \neq O(n^5)$ .

## Proof using the definition of Big-Oh:

We need to find constants c>0,  $n_0>0$  such that  $n^5\leq c*3^n$  for all  $n\geq n_0$ .

Take c=1 and  $n_0=27$ . Note that for all  $n\geq n_0$ ,  $5*log_3(n)\leq n$ . Therefore  $3^{5*log_3(n)}\leq 3^n$ , and so  $3^{log_3(n^5)}\leq 3^n$  which implies that  $n^5\leq 3^n$  as we wanted.

3. [10 points ] Are there any pairs of functions f(n) and g(n) from your list above that satisfy  $f(n) = \Theta(g(n))$ ? Prove your answer.

Solution: No, there are no functions f(n) and g(n) from the list above that satisfy  $f(n) = \Theta(g(n))$ . We have already proven this explicitly when using the Limit Rule in our proofs in part 2 above.