

CS 5413 homework assignments

Problem 2-2 in the text

Problem 2-4 in the text

Problem 3-3 in the text

Problem 3-7 in the text

## 2-2 Correctness of bubblesort

Bubblesort is a popular, but inefficient, sorting algorithm. It works by repeatedly swapping adjacent elements that are out of order. The procedure BUBBLESORT sorts array  $A[1 : n]$ .

**BUBBLESORT( $A, n$ )**

```

1  for  $i = 1$  to  $n - 1$ 
2    for  $j = n$  downto  $i + 1$ 
3      if  $A[j] < A[j - 1]$ 
4        exchange  $A[j]$  with  $A[j - 1]$ 
```

- a. Let  $A'$  denote the array  $A$  after  $\text{BUBBLESORT}(A, n)$  is executed. To prove that BUBBLESORT is correct, you need to prove that it terminates and that

$$A'[1] \leq A'[2] \leq \dots \leq A'[n]. \quad (2.5)$$

In order to show that BUBBLESORT actually sorts, what else do you need to prove?

The next two parts prove inequality (2.5).

- b. State precisely a loop invariant for the **for** loop in lines 2–4, and prove that this loop invariant holds. Your proof should use the structure of the loop-invariant proof presented in this chapter.
- c. Using the termination condition of the loop invariant proved in part (b), state a loop invariant for the **for** loop in lines 1–4 that allows you to prove inequality (2.5). Your proof should use the structure of the loop-invariant proof presented in this chapter.
- d. What is the worst-case running time of BUBBLESORT? How does it compare with the running time of INSERTION-SORT?

## 2-3 Correctness of Horner's rule

You are given the coefficients  $a_0, a_1, a_2, \dots, a_n$  of a polynomial

$$P(x) = \sum_{i=0}^n a_i x^i$$

- a. In terms of  $\Theta$ -notation, what is the running time of this procedure?
- b. Write pseudocode to implement the naive polynomial-evaluation algorithm that computes each term of the polynomial from scratch. What is the running time of this algorithm? How does it compare with HORNER?
- c. Consider the following loop invariant for the procedure HORNER:

At the start of each iteration of the **for** loop of lines 2–3,

$$p = \sum_{k=0}^{n-(i+1)} A[k + i + 1] \cdot x^k.$$

Interpret a summation with no terms as equaling 0. Following the structure of the loop-invariant proof presented in this chapter, use this loop invariant to show that, at termination,  $p = \sum_{k=0}^n A[k] \cdot x^k$ .

#### 2-4 Inversions

Let  $A[1 : n]$  be an array of  $n$  distinct numbers. If  $i < j$  and  $A[i] > A[j]$ , then the pair  $(i, j)$  is called an **inversion** of  $A$ .

- a. List the five inversions of the array  $\langle 2, 3, 8, 6, 1 \rangle$ .
- b. What array with elements from the set  $\{1, 2, \dots, n\}$  has the most inversions? How many does it have?
- c. What is the relationship between the running time of insertion sort and the number of inversions in the input array? Justify your answer.
- d. Give an algorithm that determines the number of inversions in any permutation on  $n$  elements in  $\Theta(n \lg n)$  worst-case time. (*Hint:* Modify merge sort.)

The procedure HORNER implements Horner's rule to evaluate  $P(x)$ , given the coefficients  $a_0, a_1, a_2, \dots, a_n$ , in an array  $A[0:n]$  and the value of  $x$ .

HORNER( $A, n, x$ )

```
1   $p = 0$ 
2  for  $i = n$  downto 0
3       $p = A[i] + x \cdot p$ 
4  return  $p$ 
```

- a. In terms of  $\Theta$ -notation, what is the running time of this procedure?
- b. Write pseudocode to implement the naive polynomial-evaluation algorithm that computes each term of the polynomial from scratch. What is the running time of this algorithm? How does it compare with HORNER?
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## Problems

### 3-1 Asymptotic behavior of polynomials

Let

$$p(n) = \sum_{i=0}^d a_i n^i,$$

where  $a_d > 0$ , be a degree- $d$  polynomial in  $n$ , and let  $k$  be a constant. Use the definitions of the asymptotic notations to prove the following properties.

- a. If  $k \geq d$ , then  $p(n) = O(n^k)$ .
- b. If  $k \leq d$ , then  $p(n) = \Omega(n^k)$ .
- c. If  $k = d$ , then  $p(n) = \Theta(n^k)$ .
- d. If  $k > d$ , then  $p(n) = o(n^k)$ .
- e. If  $k < d$ , then  $p(n) = \omega(n^k)$ .

### 3-2 Relative asymptotic growths

Indicate, for each pair of expressions  $(A, B)$  in the table below whether  $A$  is  $O, o, \Omega, \omega$ , or  $\Theta$  of  $B$ . Assume that  $k \geq 1, \epsilon > 0$ , and  $c > 1$  are constants. Write your answer in the form of the table with “yes” or “no” written in each box.

	$A$	$B$	$O$	$o$	$\Omega$	$\omega$	$\Theta$
a.	$\lg^k n$	$n^\epsilon$					
b.	$n^k$	$c^n$					
c.	$\sqrt{n}$	$n^{\sin n}$					
d.	$2^n$	$2^{n/2}$					
e.	$n^{\lg c}$	$c^{\lg n}$					
f.	$\lg(n!)$	$\lg(n^n)$					

### 3-3 Ordering by asymptotic growth rates

- a. Rank the following functions by order of growth. That is, find an arrangement  $g_1, g_2, \dots, g_{30}$  of the functions satisfying  $g_1 = \Omega(g_2), g_2 = \Omega(g_3), \dots, g_{29} = \Omega(g_{30})$ . Partition your list into equivalence classes such that functions  $f(n)$  and  $g(n)$  belong to the same class if and only if  $f(n) = \Theta(g(n))$ .

$$\begin{array}{ccccccc}
 \lg(\lg^* n) & 2^{\lg^* n} & (\sqrt{2})^{\lg n} & n^2 & n! & (\lg n)! \\
 (3/2)^n & n^3 & \lg^2 n & \lg(n!) & 2^{2^n} & n^{1/\lg n} \\
 \ln \ln n & \lg^* n & n \cdot 2^n & n^{\lg \lg n} & \ln n & 1 \\
 2^{\lg n} & (\lg n)^{\lg n} & e^n & 4^{\lg n} & (n+1)! & \sqrt{\lg n} \\
 \lg^*(\lg n) & 2^{\sqrt{2 \lg n}} & n & 2^n & n \lg n & 2^{2^n+1}
 \end{array}$$

- b. Give an example of a single nonnegative function  $f(n)$  such that for all functions  $g_i(n)$  in part (a),  $f(n)$  is neither  $O(g_i(n))$  nor  $\Omega(g_i(n))$ .

### 3-4 Asymptotic notation properties

Let  $f(n)$  and  $g(n)$  be asymptotically positive functions. Prove or disprove each of the following conjectures.

- a.  $f(n) = O(g(n))$  implies  $g(n) = O(f(n))$ .
- b.  $f(n) + g(n) = \Theta(\min\{f(n), g(n)\})$ .
- c.  $f(n) = O(g(n))$  implies  $\lg f(n) = O(\lg g(n))$ , where  $\lg g(n) \geq 1$  and  $f(n) \geq 1$  for all sufficiently large  $n$ .
- d.  $f(n) = O(g(n))$  implies  $2^{f(n)} = O(2^{g(n)})$ .
- e.  $f(n) = O((f(n))^2)$ .
- f.  $f(n) = O(g(n))$  implies  $g(n) = \Omega(f(n))$ .
- g.  $f(n) = \Theta(f(n/2))$ .
- h.  $f(n) + o(f(n)) = \Theta(f(n))$ .

### 3-5 Manipulating asymptotic notation

Let  $f(n)$  and  $g(n)$  be asymptotically positive functions. Prove the following identities:

- a.  $\Theta(\Theta(f(n))) = \Theta(f(n))$ .
- b.  $\Theta(f(n)) + O(f(n)) = \Theta(f(n))$ .
- c.  $\Theta(f(n)) + \Theta(g(n)) = \Theta(f(n) + g(n))$ .
- d.  $\Theta(f(n)) \cdot \Theta(g(n)) = \Theta(f(n) \cdot g(n))$ .

$\widetilde{O}(g(n)) = \{f(n) : \text{there exist positive constants } c, k, \text{ and } n_0 \text{ such that}$   
 $0 \leq f(n) \leq cg(n) \lg^k(n) \text{ for all } n \geq n_0\}.$

- e. Define  $\widetilde{\Omega}$  and  $\widetilde{\Theta}$  in a similar manner. Prove the corresponding analog to Theorem 3.1.

### 3-7 Iterated functions

We can apply the iteration operator  $*$  used in the  $\lg^*$  function to any monotonically increasing function  $f(n)$  over the reals. For a given constant  $c \in \mathbb{R}$ , we define the iterated function  $f_c^*$  by

$$f_c^*(n) = \min \{i \geq 0 : f^{(i)}(n) \leq c\},$$

which need not be well defined in all cases. In other words, the quantity  $f_c^*(n)$  is the minimum number of iterated applications of the function  $f$  required to reduce its argument down to  $c$  or less.

For each of the functions  $f(n)$  and constants  $c$  in the table below, give as tight a bound as possible on  $f_c^*(n)$ . If there is no  $i$  such that  $f^{(i)}(n) \leq c$ , write “undefined” as your answer.

	$f(n)$	$c$	$f_c^*(n)$
a.	$n - 1$	0	
b.	$\lg n$	1	
c.	$n/2$	1	
d.	$n/2$	2	
e.	$\sqrt{n}$	2	
f.	$\sqrt{n}$	1	
g.	$n^{1/3}$	2	

## Chapter notes

Knuth [259] traces the origin of the  $O$ -notation to a number-theory text by P. Bachmann in 1892. The  $o$ -notation was invented by E. Landau in 1909 for his discussion of the distribution of prime numbers. The  $\Omega$  and  $\Theta$  notations were advocated by Knuth [265] to correct the popular, but technically sloppy, practice in the literature of using  $O$ -notation for both upper and lower bounds. As noted earlier in this chapter, many people continue to use the  $O$ -notation where the  $\Theta$ -notation is more technically precise. The soft-oh notation  $\widetilde{O}$  in Problem 3-6 was introduced