MA/CSSE 473 – Design and Analysis of Algorithms

Homework 12 (88 points total) Updated for Summer, 2016

When a problem is given by number, it is from the textbook. 1.1.2 means "problem 2 from section 1.1" .

Problems for enlightenment/practice/review (not to turn in, but you should think about them):

How many of them you need to do serious work on depends on you and your background. I do not want to make everyone do one of them for the sake of the (possibly) few who need it. You can hopefully figure out which ones you need to do.

8.4.2 [8.2.2] (Time efficiency of Warshall's Algorithm)
8.4.6 [8.2.6] (Use Warshall to determine whether a digraph is a dag)
8.3.1 (Practice optimal static BST calculation)
8.3.2 (Time and space efficiency of optimal BST calculation)
8.3.5 (Root of Optimal tree)
8.3.9 (Include unsuccessful searches in optimal BST calculation)
8.3.8 (n^2 algorithm for optimalBST. Not for the faint of heart!)
------ For the frequencies of the Day 33 class example (AEIOU), find the optimal tree if we consider only successful searches (set all bi to 0)

Problems to write up and turn in:

1. (5) 8.4.3 [8.2.3] (Warshall with no extra memory use)
2. (10) 8.4.4 [8.2.4] (More efficient Warshall inner loop)
3. (25) Optimal static BST problem: described below.
   Part (d) is extra credit. Not many people have gotten it in past terms.
   In the past, a number of students have said that this problem is long and difficult, especially part a. I have placed on Moodle an excerpt from the original source from which I got this example. Also note that there is some relevant Python code linked from the schedule page (days 28 and 29 in Summer 2016).
4. (10) 8.3.3 (Optimal static BST from root table)
5. (5) 8.3.4 (Sum for optimalBST in constant time).
6. (10) 8.3.6 (optimalBST--successful search only--if all probabilities equal)
7. (5) 8.3.11a [8.3.10a] (Matrix chain multiplication) Also think about parts (b) and (c), which may appear on a later assignment or exam.
8. (10) 8.4.7 [8.2.7] (Floyd example)
9. (8) 8.4.8 [8.2.8] (Floyd: Do we need to save previous matrix?)

Optimal static BST Dynamic Programming Problem (problem #3)

In a binary search tree, the key of each node in the left subtree is smaller than the key of the root, and the key of each node in the right subtree is larger than the root. The same property holds for each subtree. Section 8.3 discusses a dynamic programming algorithm to find an optimal static tree if only successful searches are taken into account. In class (Days 29-30 in Fall, 2014) we discussed a modified algorithm that also takes unsuccessful searches into account. This basis for the approach used in class is from Reingold and Hansen, *Data Structures*. This section of that book is posted on Moodle.

Suppose that we have a static set of N keys $K_1, K_2, \ldots, K_n$ (in increasing order), and that we have collected statistics about the frequency of searches for each key and for items in each "gap" between keys (i.e. each place that an unsuccessful search can end up).

For i = 1 ... n, let a_i be the frequency of searches that end successfully at K_i.

For i = 1 ... n-1, let b_i be the frequency of unsuccessful searches for all "missing keys" that are between K_i and K_{i+1} (also, b_0 is the frequency of searches for keys smaller than K_1, and b_n is the frequency for keys that are larger than K_n).

We build an extended BST T (see Figure 4.5 for a diagram of an extended tree) whose internal nodes contain the N keys, $K_1, \ldots, K_n$. Let x_i be the depth of the node containing K_i, and let y_i be the depth of the "external node" that represents the gap between K_i and K_{i+1} (where y_0 and y_n are the depths of the leftmost and rightmost external nodes of T). Recall that the depth of a tree's root is 0. The optimal tree for the given keys and search frequencies is one that minimizes the *weighted path length* C(T), where C is defined by

$$C = \sum_{i=1}^{N} a_i [1 + x_i] + \sum_{i=0}^{N} b_i y_i$$

For example, in class (in the summer, read the Reingold book excerpt) we considered the following data:

i	K_i	a_i	b_i
0			0
1	A	32	34
2	E	42	38
3	I	26	58
4	O	32	95
5	U	12	21

If we choose to build the BST with I as the root, E and O as I's children, A as E's child, and U as O's child, then C = 948, and the average search length is 948/390 = 2.43 (you should verify this, to check your understanding of the formula for C). It turns out that this tree is not optimal. Note that in this example, a1 (32) is the frequency that the search is for A, while b1 (34) is the frequency that the value being searched for is between A and E

In class we discussed a dynamic programming algorithm that finds a tree that minimizes C for any set of n keys and 2n+1 associated frequencies. It uses an alternate, recursive, formulation of C:

(a) (10) **The recursive formulation:** Let T be a BST. If T is empty, then C(T) = 0. Otherwise T has a root and two subtrees $T_L$and T_R. Then C(T) = C(T_L) + C(T_R) + sum(a_i, i=1..n) + sum(b_i,i=0..n). Show by induction that this recursive definition of C(T) is equivalent to the summation definition given above. [The recursive definition is used in the code provided online (linked from the schedule page)].

(b) (5) The algorithm and a Python implementation are provided, along with a table of final values for the above inputs. Use the information from that table to draw the optimal tree.

(c) (10) What is the big-theta running time for the optimal-tree-generating algorithm (the algorithm that generates the root table from the frequency tables)? Show your computations that lead you to this conclusion

(d) (10) **(Part (d) is extra credit. Not many people have gotten it in past terms** Find a way to improve the given algorithm (the algorithm that generates the root table from the frequency tables) so that it has a smaller big-theta running time, and show that it really is smaller.