

# CS624: Analysis of Algorithms

## Midterm exam 2 – practice

Instructor – Nurit Haspel

1. **Medians and Order Statistics:** (30%) Given two *sorted* arrays, A and B, each of size  $n$ . Describe an  $O(\log n)$  worst case algorithm to find the median of all the  $2n$  elements in A and B. Provide a brief but accurate runtime analysis (**Hint 1:** It's not unlike binary search... **Hint2:** Use the medians of A and B to guide you)

First of all - locate the medians of A and B. It can be done in  $O(1)$  since the two arrays are sorted. If  $median[A] = median[B]$  return one of them as the median. If  $median(A) < median(B)$ , then the median of both arrays is certainly not in  $A[1..n/2 - 1]$  (the lower half of A), since all the elements there are smaller than more than half of A and more than half of B (see why?). Similarly, the median of both arrays is certainly not in  $B[n/2 + 1..n]$  (the larger half of B), for the same reason. You can then recursively search the median in the larger half of A and the smaller half of B. It's a logarithmic time algorithm.

2. **Binary search trees:**

- (a) (15%). Let  $x$  be a leaf node in a binary search tree  $T$ . Let  $y$  be  $x$ 's parent. Show that  $y.key$  is either the smallest key in  $T$  larger than  $x.key$  or the largest key in  $T$  smaller than  $x.key$ . This is a special case of the successor-predecessor relationship. Since  $x$  is a leaf, its successor and predecessor (if exist) have to be ancestors. Let's assume, without loss of generality, that  $y$ ,  $x$ 's parent, is larger than  $x$ . Therefore  $x$  is  $y$ 's left child. According to what we saw in class,  $y$ 's predecessor is the largest node on  $y$ 's left subtree, in this case it's  $x$  since  $x$  is a leaf and therefore the only node on  $y$ 's left subtree. Now, assume by contradiction that this were not the case and there is a node  $z$  such that  $x.key < z.key < y.key$ .  $z$  must be on the same subtree as  $x$  and  $y$ , since  $x$  and  $y$  are on the same subtree, and  $z$ 's value is between them. Since  $y$  only has  $x$  as a leaf,  $z$  can only be an ancestor of  $y$ , not a descendant. But this is impossible, because that would put both  $x$  and  $y$  on the same subtree of  $z$ , despite the fact that  $x.key < z.key < y.key$ . Therefore, such  $z$  cannot exist. The same is true in reverse if  $y < x$ .
- (b) (15%) Is the following claim true or false? Explain: in order to determine whether two binary search trees are identical one has to perform an in-order walk on them and compare the results.

False, since inorder gives the keys in a sorted order. So - any two BSTs with the same set of keys but a different structure gives the same inorder.

3. **Dynamic Programming:** Given an array A of  $n$  numbers, the maximum subarray problem is the task of finding the contiguous subarray  $A[i..j]$  of numbers which has the largest sum. For example, if  $A = \{-2, 1, -3, \mathbf{4}, -1, \mathbf{2}, \mathbf{1}, -5, 4\}$  then the subarray that gives the maximum sum is  $\{4, -1, 2, 1\}$  with sum 6 (emphasized in bold font). Let us define  $MS(i)$  as the maximum sum subarray that ends at  $A[i]$  (and must include  $A[i]$ ). For example, in a 1-based index,  $MS(1) = \{-2\}$ .  $MS(2) = \{1\}$  (since concatenating -2 and 1 gives a smaller sum, so  $MS(2)$  includes only  $A[2]$ ). In other words - for  $MS(i)$  we ask ourselves which one is better - for  $A[i]$  to extend  $MS(i-1)$  or be its own subarray.

- (a) Show that the problem has the optimal substructure (Hint:  $A[i]$  either extends the maximum sub-array that ends in  $A[i-1]$  or alternatively, includes only  $A[i]$  itself. Use a cut-and-paste argument for  $MS(i-1)$  with respect to  $MS(i)$ ).

It is a standard cut-paste. If we look at an optimal solution  $MS(i)$ , then either  $A[i]$  extends a maximum sub-array ending in  $A[i-1]$  or not. If there were a better  $MS(i-1)$ , we could have replaced it and get a better  $MS(i)$  or at least not a worse one, in case  $A[i]$  is in itself  $MS(i)$

- (b) Define a recursive algorithm that calculates  $MS(i)$ , that can be used as a basis for a dynamic programming calculation. Remember to also return the overall maximum sum. It doesn't have to be  $MS(n)$  (why?).

The algorithm is:  $MS(i) = \begin{cases} -\infty & \text{if } i < 1 \\ \max(MS(i-1) + A[i], A[i]) & \text{otherwise} \end{cases}$  The  $-\infty$  is to ensure

that  $MS(1) = A[1]$  even if it is negative. The end result is  $\max_i \{MS(i)\}$ , which does not have to be  $MS(n)$ , as in the example here, since the subsequence can end at some point in the middle of the array. We can then convert the recursive algorithm into DP by maintaining an array  $MS[]$  such that  $MS[i] = MS(i)$  for each  $i$ .

- (c) Based on that, calculate  $MS(i)$  for every index in the array above.

$MS = \{-2, 1, -2, 4, 3, 5, 6, 1, 5\}$