CSC373 Fall'19 Tutorial 1 Solution Mon. Sept. 16, 2019

Solution to Q1: Practicing Recurrence Relations

(a) For
$$T(n) \leq 3T(n/2) + O(n \log^3 n)$$
, we have:

•
$$a = 3$$
 and $b = 2$; thus, $n^{\log_b a} = n^{\log_2 3}$.

•
$$f(n) = O(n \log^3 n)$$
.

Hence, by case 1 of the Master theorem, $T(n) = O(n^{\log_2 3})$.

(b) For
$$T(n) \le 4T(n/2) + O(n^2)$$
, we have:

•
$$a = 4$$
 and $b = 2$; thus, $n^{\log_b a} = n^{\log_2 4} = n^2$.

•
$$f(n) = O(n^2)$$
.

Hence, by case 2 of the Master theorem, $T(n) = O(n^2 \log n)$.

(c) For
$$T(n) \leq 2T(n/2) + O(n\log^2 n)$$
, we have

•
$$a = 2$$
 and $b = 2$; thus, $n^{\log_b a} = n^{\log_2 2} = n$.

•
$$f(n) = O(n \log^2 n)$$
.

Hence, again by case 2 of the Master theorem, $T(n) = O(n \log^3 n)$.

(d) For
$$T(n) \leq 2T(n/4) + O(n^{0.5001})$$
, we have

•
$$a = 2$$
 and $b = 4$; thus, $n^{\log_b a} = n^{\log_4 2} = n^{0.5}$.

•
$$f(n) = O(n^{0.5001}).$$

Hence, by case 3 of the Master theorem, $T(n) = O(n^{0.5001})$.

Solution to Q2: Circularly Shifted Sorted Array

Algorithm:

Function LargestInShifted $(A[1 \dots n])$

• If
$$A[n] \ge A[1]$$
 then return $A[n]$.

- Set
$$m \leftarrow \lfloor n/2 \rfloor$$

- If $A[n] \ge A[m+1]$ then return LargestInShifted(A[1...m]).
- Else, return LargestInShifted(A[(m+1)...(n-1)]).

Then, the desired solution is LargestInShifted(A[1...n]).

Correctness: The proof is by induction on n. Let i = k + 1 be the index of the maximum element of A. The following cases are exhaustive:

- Assume i = n. Then clearly $A[n] \ge A[1]$, so the algorithm outputs A[n]. Note that this covers the base case of n = 1.
- Assume $1 \le i \le m$. Then $A[m+1] \le A[n] < A[1]$, so the algorithm outputs LargestInShifted(A[1 ... m]). By induction, this equals the maximum of (A[1], ..., A[m]), which is A[i].
- Assume $m+1 \le i < n$. Then A[n] < A[1] < A[m+1], so the algorithm outputs LargestInShifted(A[(m+1)...(n-1)]). By induction, this equals the maximum of (A[m+1],...,A[n-1]), which is A[i].

Running Time: If T(n) is the running time of the algorithm, then we have T(n) = T(n/2) + O(1) with T(1) = O(1). Using the master theorem, this gives $T(n) = O(\log n)$.

Solution to Q3: Majority Element

Here's a solution with runtime $O(n \log n)$. We recursively solve the following more general problem: output the majority element of A if such an element exists, and output \bot otherwise.

If A has a single element then the solution is to output that element. Otherwise, start by dividing A into two halves, L and R, such that |L| = |R| or |L| = |R| + 1. (Here, |L| denotes the number of elements in L.)

If A has a majority element x, then x is a majority element of either L or R. We prove the contrapositive of this statement as follows. If x is not a majority element of L or R, then x occurs at most |L|/2 times in L and at most |R|/2 times in R. Therefore x occurs at most |L|/2 + |R|/2 = |A|/2 times in A, so x is not a majority element of A.

Recursively determine whether L has a majority element, and what that element is if it exists. Do the same for R. This gives a set S of the majority elements of L and R, where S has cardinality 0, 1 or 2. If A has a majority element, then that element is in S. So for each x in S, count the number of occurrences of x in A, and if this number is greater than n/2 then output x. If this procedure does not output any element, then return \bot .

The runtime is described by the recurrence $T(n) \leq 2T(n/2) + O(n)$, so the master theorem gives $T(n) = O(n \log n)$.

A more sophisticated O(n) time algorithm is possible. (See https://en.wikipedia.org/wiki/Boyer%E2%80%93Moore_majority_vote_algorithm). This cannot be achieved by computing the median in O(n) time, because the question only allows equality checks, and not comparisons.

Solution to Q4: Monotonic Function Evaluation

Let k = 1, and while f(k) > 0, double k. In at most $\lceil \log n \rceil$ iterations, this will terminate as we will have $k \ge n$. Let k^* be the value at which it terminates. Then, we know that $k^*/2 < n \le k^*$. We can binary-search n in this range, as described in more detail below.

The running time for finding k^* is $O(\log n)$, and the running time for the subsequent binary search is also $O(\log n)$. Hence, the overall time is $O(\log n)$.

Function FindFirstNonPositive(A[1 ... r])

- If r = 1 then return A[1].
- Else:
 - Set $m \leftarrow \lfloor r/2 \rfloor$
 - If $A[m] \leq 0$ then return FindFirstNonPositive($A[1 \dots m]$).
 - Else, return FindFirstNonPositive(A[(m+1)...r]).