CSE 431/531: Algorithm Analysis and Design

Fall 2021

## Homework 3

Instructor: Shi Li Deadline: 11/4/2021

Your Name: \_\_\_\_\_ Your Student ID: \_\_\_\_\_

Problems	1	2	3	Total
Max. Score	20	30	30	80
Your Score				

**Problem 1.** For each of the following recurrences, use the master theorem to give the tight asymptotic upper bound.

(a) 
$$T(n) = 4T(n/3) + O(n)$$
.  $T(n) = O(\frac{n^{\log_3 4}}{n})$ .

(b) 
$$T(n) = 3T(n/3) + O(n)$$
.  $T(n) = O(n \log n)$ .

(c) 
$$T(n) = 4T(n/2) + O(n^2\sqrt{n}).$$
  $T(n) = O(n^2\sqrt{n}).$ 

(d) 
$$T(n) = 8T(n/2) + O(n^3)$$
.  $T(n) = O(\frac{n^3 \log n}{n})$ .

**Problem 2.** We consider the following problem of counting stronger inversions. Given an array A of n positive integers, a pair  $i, j \in \{1, 2, 3, \dots, n\}$  of indices is called a *strong inversion* if i < j and A[i] > 2A[j]. The goal of the problem is to count the number of strong inversions for a given array A.

Give a divide-and-conquer algorithm that runs in  $O(n \log n)$  time to solve the problem. Write down the recurrence for the running time, and use the master theorem to show that the running time is indeed  $O(n \log n)$ .

We modify the divide-and-conquer algorithm for counting inversions slightly. The only thing that needs to be changed is the procedure merge-and-count  $(B, C, n_1, n_2)$ . As in the algorithm for counting inversions, we are given two sorted arrays: B of length  $n_1$ , and C of length  $n_2$ . But now we need to count the number of strong inversions between B and C, and merge B and C. The two tasks are performed using two different while loops: in the first while loop, we count the number of strong inversions between B and C, that is, the number of pairs (i, j) such that  $B[i] > 2 \times C[j]$ . In the second while loop, we merge B and C into a sorted array. The procedure is given by the following pseudo-code:

## **Algorithm 1** merge-and-count $(B, C, n_1, n_2)$

```
1: count \leftarrow 0, A \leftarrow []
 2: i \leftarrow 1; j \leftarrow 1
 3: while i \leq n_1 or j \leq n_2 do
         if j > n_2 or (i \le n_1 \text{ and } B[i] \le 2 \times C[j]) then
 4:
              i \leftarrow i + 1
 5:
 6:
              count \leftarrow count + (j-1)
 7:
         else
 8:
              j \leftarrow j + 1
 9: i \leftarrow 1; j \leftarrow 1
10: while i \leq n_1 or j \leq n_2 do
         if j > n_2 or (i \le n_1 \text{ and } B[i] \le C[j]) then
11:
              append B[i] to A; i \leftarrow i+1
12:
         else
13:
              append C[j] to A; j \leftarrow j+1
14:
15: return (A, count)
```

The recurrence for the running time is still T(n) = 2T(n/2) + O(n). So, the running time of the algorithm is  $O(n \log n)$ .

**Problem 3.** Given two sorted arrays A and B with total size n, and a positive integer  $k \leq n$ , you need to design an  $O(\log n)$ -time algorithm that outputs the k-th smallest number in the union of A and B. You need to prove that the running time of your algorithm is indeed  $O(\log n)$ .

For example, if A = [3, 5, 12, 18, 50], B = [2, 7, 11, 30], and k = 4 then you need to output 7 since the union of A and B is [2, 3, 5, 7, 11, 12, 18, 30, 50] after sorting.

In the algorithm, we maintain four indices lA, rA, lB, rB and an integer k' between 1 and rA - lA + 2 + rB - lB. We guarantee that our goal is to output the k'-th smallest number in  $A[lA..rA] \uplus B[lB..rB]$ . The pseudo-code is given in Algorithm 2.

We explain Step 6 and Step 8 of the algorithm.

- As  $A[mA] \leq B[mB]$  we know elements in  $A[lA..mA] \uplus B[lB..mB]$  are smaller than or equal to elements in B[mB+1..rB].  $A[lA..mA] \uplus B[lB..mB]$  contains mA-lA+2+mB-lB elements. So, when  $k' \leq mA-lA+2+mB-lB$ , any element in B[mB+1..rB] is too big to be the k'-th smallest element. So, Step 6 reduced the problem correctly.
- Equivalently, our goal is to find the (rA lA + 3 + rB lB k')-th largest number in  $A[lA..rA] \uplus B[lB..rB]$ . If k' > mA lA + 2 + mB lB, then  $rA lA + 3 + rB lB k' \le rA mA + rB mB$ . If  $A[mA] \le B[mB]$ , then elements in  $A[mA + 1..rA] \uplus B[mB + 1..rB]$  are larger than or equal to elements in A[lA..mA]. So, any element in A[lA..mA] is too small to be the k'-th smallest element. So, Step 8 reduced the problem correctly.

So, in each iteration of Loop 2, we either reduced the size of A[lA..rA] by a factor of 2, or the size of B[lB..rB] by a factor of 2. In at most  $O(\log n)$  iterations, one of the two sub-arrays will become of size 1. The running time of the algorithm is  $O(\log n)$ .

```
Algorithm 2 kth-smallest-number (A, B, k)
```

```
1: lA \leftarrow 1, rA \leftarrow \text{size of } A, lB \leftarrow 1, rB \leftarrow \text{size of } B, k' \leftarrow k
 2: while rA > lA and rB > lB do
          mA \leftarrow \left\lfloor \frac{lA+rA}{2} \right\rfloor \text{ and } mB \leftarrow \left\lfloor \frac{lB+rB}{2} \right\rfloor
          if A[mA] \stackrel{\perp}{\leq} B[mB] then
 4:
               if k \leq mA - lA + 2 + mB - rB then
 5:
                    rB \leftarrow mB
 6:
 7:
               else
                                                                                                                               \triangleright
                    lA \leftarrow mA + 1, k' \leftarrow k' - (mA - lA + 1)
 8:
                                                                                                      \triangleright A[mA] > B[mB]
          else
 9:
               handle this case in a symmetric way
10:
11: if lA = rA then
12:
          return min{A[lA], B[lB + k' - 1]}
                                                                                                                  \triangleright lB = rB
13: else
          \textbf{return} \, \min\{B[lB], A[lA+k'-1]\}
14:
```