## CSE 5717 Big Data Analytics. Fall 2022 Exam IV Solutions

1. A sampling Lemma stated in class states the following. Let X be any set of n arbitrary real numbers and let S be a random sample of X with s elements. If q is an element of S whose rank in S is j and  $r_i$  is the rank of q in X, then:

$$
Prob\left[\left|r_j-j\frac{n}{s}\right|>\sqrt{3\alpha}\frac{n}{\sqrt{s}}\sqrt{\log n}\right]\leq n^{-\alpha}.
$$

The above Lemma suggests the following algorithm: 1) Pick a random sample  $S$  of  $X$  with  $|S| = n^{2/3} \log n$ ; 2) Identify and output the median M of S. This median can be found in  $O(|S|)$  time. This implies that the run time of the algorithm is  $O(n^{2/3} \log n)$ . If r is the rank of  $M$  in  $X$ , then the above Lemma implies that:

$$
Prob\left[\left|r - \frac{n}{2}\right| > \sqrt{3\alpha}n^{2/3}\right] \le n^{-\alpha}.
$$

2. We can employ radix sorting here. Think of each key as a  $log R$ -bit binary number. We sort the keys in stages. In each stage we sort them with respect to  $\log(M/B)$  bits. Sorting is done starting from the LSBs and moving towards the MSBs.

In any stage of sorting we have to sort n keys where each key is an integer in the range  $\left[1, \frac{M}{B}\right]$  $\frac{M}{B}$ . In the core memory we keep  $\frac{M}{B}$  buckets, one for each possible value. We bring one block at a time from the disk and distribute the keys in this block into the buckets based on the values of the keys. A key whose value is i will be put into bucket i (for  $1 \leq i \leq \frac{M}{B}$  $\frac{M}{B}$ . When a bucket gets  $B$  keys, this block is written into the disk and the bucket becomes empty. There will be a run in the disk corresponding to every possible value.

Clearly, there will be  $\frac{\log R}{\log(M/B)}$  stages in the algorithm and hence the total number of I/O operations is  $O\left(\frac{n}{B}\right)$ B  $\log R$  $\frac{\log R}{\log(M/B)}\bigg).$ 

3. Construct a suffix tree  $Q$  for  $S$  in  $O(n)$  time. Followed by this, perform an in-order traversal of Q to label every internal node u of Q with an integer  $c[u]$  such that  $c[u]$  is the number of leaves in the subtree rooted at u.

Now, perform one more traversal through Q to mark every node whose string depth is  $\geq k$ . In one additional traversal through Q identify the node u that is marked and whose  $c[u]$  is the largest. Finally, output any substring of the path label of  $u$  whose length is  $k$ .

Clearly, the total run time of the algorithm is  $O(n)$ .

4. Note that we generate association rules from frequent itemsets. A frequent itemset occurs in at least one transaction. Let  $t$  be any transaction. Since the number of items in  $t$  is no more than c, the number of itemsets we can generate out of t is less than  $2<sup>c</sup>$ . This implies that the

total number of frequent itemsets is  $\langle n2^c \rangle$ . Let X be any such frequent itemset. X has at most  $c$  items in it. Using the result from Homework 3, problem  $4(b)$ , the number of association rules that we can generate from X is  $\leq 3^c$ . As a result, the total number of association rules we can construct from all possible frequent itemsets is  $\langle n2^c 3^c = n6^c = O(n)$ .

5. First compute  $f_i(x) = (x + a_i)^{2^i}$ , for  $1 \le i \le \log n$ . For any  $i, (x + a_i)^{2^i}$  can be computed in time  $O(2<sup>i</sup>)$  (as per Problem 1, Homework 2). Thus all of these polynomials can be computed in a total of  $O\left(\sum_{i=1}^{\log n} 2^i\right) = O(n)$  time.

Followed by this, we do the following:

$$
f(x) = f_1(x);
$$
  
for  $i = 2$  to log n do  

$$
f(x) = f(x) \times f_i(x);
$$

Using the theorem that we can multiply two degree d polynomials in  $O(d \log d)$  time, the total run time will be  $O(\sum_{i=1}^{\log n} 2^i i) = O(n \log n)$ .

6. Note that given the activation values for layer l (for  $1 \leq l \leq (L-1)$ ), we can compute the activation values for layer  $l + 1$  with a matrix-vector multiplication. Specifically,  $\vec{a}^{l+1} =$  $\sigma(W^{l+1}\vec{a}^l + \vec{b}^{l+1})$ . Here  $W^{l+1}$  is the weight matrix (from layer l to layer  $l + 1$ ) and is of dimension  $n \times n$ .  $\vec{a}^{l+1}$  and  $\vec{a}^l$  are activation vectors and  $\vec{b}^{l+1}$  is the bias vector. These are of dimension  $n \times 1$  each.

Consider the computation of  $W^{l+1}\vec{a}^l$ . Each row of  $W^{l+1}$  can be multiplied with  $\vec{a}^l$  in  $O(\log n)$ time using  $\frac{n}{\log n}$  CREW PRAM processors, with the employment of a prefix sums computation. Thus,  $W^{l+1}\vec{a}^l + \vec{b}^{l+1}$  can be computed in  $O(\log n)$  time using  $\frac{n^2}{\log n}$  CREW PRAM processors, given  $\vec{a}^l$ . Given  $W^{l+1}\vec{a}^l + \vec{b}^{l+1}$ , we can compute  $\vec{a}^{l+1}$  in  $O(1)$  time using *n* processors.

In summary, one forward propagation step can be completed in  $O(L \log n)$  time using  $\frac{n^2}{\log n}$  $\overline{\log n}$ CREW PRAM processors.