## CSE 4502/5717 Big Data Analytics Fall 2024; Homework 2 Solutions

1. Dijkstra's algorithm can be described as follows:

Algorithm 1: Dijkstra $(V, E, s)$ **Data:**  $(V, E)$ : a graph; s: a source node; let  $w(u, v)$  be the weight of edge  $(u, v)$ ; **Result:** array d where  $d_u$  is the length of the shortest path from s to u; begin for u in V do  $\bigsqcup d_u := \infty;$  $d_s := 0;$ Create a priority queue  $Q$  to store pairs of the form (node, distance); Insert the pair  $(s, 0)$  into  $Q$ ; while Q not empty do  $(u, r) :=$  ExtractMin $(Q);$ for every child  $c$  of  $u$  do if  $d_c > d_u + w(u, c)$  then  $d_c := d_u + w(u, c);$ Insert $(Q, (c, d_c))$ ; // update distance if c present

We assume that we can store the priority queue in memory  $(O(|V|))$ . The algorithm will read the neighbors of each node at most once. Therefore, the total number of I/Os is  $\sum_{u \in E} \left\lceil \frac{deg_u}{B} \right\rceil$  $\left\lfloor \frac{eg_u}{B} \right\rfloor = O\left(\frac{|E|}{B} + |V|\right).$ 

2. We apply the LMM algorithm with  $l = m =$ √ M. We assume known that we can merge  $\sqrt{M}$  sequences of length M each in 3 passes through the data. The pseudocode of the algorithm is given below:

## Algorithm 2:  $Sort(X, N)$

## Data:

X: array of elements;

 $N = M^2$ : number of elements in X;

**Result:** sorted array  $X$ ;

## begin

// First Pass; Split the input into  $M$  runs of length  $M$  each; Sort each run and unshuffle it into  $m =$  $\sqrt{M}$  sequences of length  $\sqrt{M}$  each; // Second Pass; Merge groups of  $l =$ √ M unshuffled sequences (in memory); // Third Pass; Shuffle groups of  $m =$ √  $M$  merged sequences of length  $M$  each; At the same time clean up the dirty regions; At this point we have  $\sqrt{M}$  sorted runs of length M √  $M$  each; // Third Pass (can be done with the previous pass); Unshuffle each run of length M √  $M$  into  $m =$ √ M sequences of length M each ; // Fourth, Fifth and Sixth Pass; Merge groups of  $l =$ √ M unshuffled sequences of length M each; // Seventh Pass; Shuffle groups of  $m =$ √ M merged sequences of length M √ M each; Clean up dirty regions;

For an arbitrary N, the general principle is to first merge  $\sqrt{M}$  sequences of length  $M$ each, then merge  $\sqrt{M}$  sequences of length M √  $M$  each and so on. Let  $K$  stand for √ M and let  $T(u, v)$  be the number of passes required to merge u sorted sequences of length  $v$  each. Then we have the familiar formulas:

$$
T(K, M) = 3
$$
  
\n
$$
T(K, K^{i}M) = 2 + T(K, K^{i-1}M) = 2i + 3
$$
  
\n
$$
T(K^{c}, M) = T(K, M) + T(K, KM) + T(K, K^{2}M) + ... + T(K, K^{c-1})
$$
  
\n
$$
= \sum_{i=0}^{c-1} (2i + 3) = c^{2} + 2c
$$

However, as we saw in the previous pseudocode, when we compute  $T(K^c, M)$  we can

overlap the unshuffling at the beginning of a  $T(K, K<sup>i</sup>M)$  computation with the shuffling done at the end of the previous  $T(K, K^{i-1}M)$  computation. Therefore, the last equation becomes:

$$
T(K^c, M) = T(K, M) + \ldots + T(K, K^{c-1}) - (c - 1) = c^2 + c + 1
$$

Therefore the number of passes for  $M^2$  and  $M^3$  elements are:

$$
T(M^{2}) = T(M, M) = T(K^{2}, M) = 2^{2} + 2 + 1 = 7
$$
  

$$
T(M^{3}) = T(M^{2}, M) = T(K^{4}, M) = 4^{2} + 4 + 1 = 21 \square
$$

In general, for a given N, if  $K^c = N/M$  it means that  $c = 2 \frac{\log N/M}{\log M}$  and the number of passes to sort N elements is:

$$
T(N) = T(K^{c}, M) = 4\left(\frac{\log N/M}{\log M}\right)^{2} + 2\frac{\log N/M}{\log M} + 1.
$$

3. The input striping is good for accessing the rows of the matrix in a disk parallel manner. However, if we want to access the columns, this striping is not good. To multiply A and  $C$  we need the transpose of  $C$ . To get this, we first restripe the matrix  $C$  as follows. Let  $R_i$  be the *i*th row of C. We read  $R_i$  into core memory in  $\frac{n}{DB}$  parallel I/Os. We then rewrite row  $R_i$  starting from disk i mod  $D$  (with one block per disk). This is done for every  $1 \leq i \leq n$ . After this restriping, we read one column at a time into the core memory and write it back to the disks one block per disk (starting from the first disk). Note that a column can be read in  $\frac{n}{D}$  parallel I/O operations. Thus the matrix C can be transposed in  $\frac{n^2}{D} < \frac{n^3}{DB}$  parallel I/O operations.

We then use the following algorithm. Let  $E = AC$ .

for  $i := 1$  to n do

Read row *i* of A into core memory. Let this row be called  $A_i$ .

for  $j := 1$  to n do

Read column j of C into core memory. Let this column be  $C_j$ .

 $E_{ij} = \sum_{k=1}^{n} A_i[k] * C_j[k].$ 

Write row  $i$  of  $E$  into the disks, striping the data in a row-major order.

Each row or column of A or C can be read in  $O\left(\frac{n}{DB}\right)$  parallel I/Os. Also, each row of E can be written in  $O\left(\frac{n}{DB}\right)$  I/Os. Thus the total number of parallel I/Os is  $O\left(\frac{n^3}{DB}\right)$ .

4. Let the input strings be  $S_1, S_2, \ldots, S_k$  with  $\sum_{i=1}^k |S_i| = M$ . Build a generalized suffix tree for these strings in  $O(M)$  time. Let the suffixes be labelled with  $(i, j)$  where i refers to  $S_i$  and j refers to the j<sup>th</sup> suffix in  $S_i$ . Perform a depth first traversal in this tree.

When we reach a leaf labelled  $(i, 1)$  for some i, this leaf corresponds to the entire string  $S_i$ . This leaf might have more than one labels. Let these labels (in addition to  $(i, 1)$ ) be  $(i_1, l_1), (i_2, l_2), \ldots, (i_q, l_q)$ . Clearly, all the strings  $S_{i_1}, S_{i_2}, \ldots, S_{i_q}$  have  $S_i$  as a substring. Output all of these strings as those that contain  $S_i$ . Check if the edge to this leaf's parent is labeled with  $\$ . If not, proceed with the traversal. If yes, let x be the parent of this leaf. Also, let  $c_1, c_2, \ldots, c_r$  be the other children of x. Traverse through all the subtrees rooted at these children. All the leaves in these subtrees also correspond to strings that have  $S_i$  as a substring. Output these strings as well (as those that contain  $S_i$ ) and proceed with the traversal.

The entire algorithm can be implemented to run in time  $O(M + k^2)$ .

- 5. Let  $S_1, S_2, \ldots, S_k$  be the given input strings. Let  $|S_i| = n_i$ , for  $1 \le i \le k$ . For any two strings  $S_i$  and  $S_j$  we can compute the longest common substring between them in  $O(n_i + n_j)$  time, for  $1 \leq i, j \leq k$ . Use this algorithm to compute the longest common substring between every pair of strings. The total run time is  $O(\sum_{i=1}^{k} \sum_{j=1}^{k} (n_i + n_j)) =$  $O(kM)$ .
- 6. Note that on a common CRCW PRAM we can compute the minimum or maximum of *n* integers (in the range  $[1, n^{O(1)}]$ ) in  $O(1)$  time using *n* processors.

Let T be the text and P be the pattern with  $|T| = m$  and  $|P| = n$ . We can use binary search on the suffix array. In any iteration of binary search, we have to compare the pattern  $P$  with a suffix  $T_i$  of the text. This comparison involves the identification of the smallest integer q such that  $P[q] \neq T_i[q]$ . This can be done in  $O(1)$  time using the above algorithm. Thus the entire binary search takes  $O(\log m)$  time.