

# Problem Set 2

## Sublinear Algorithms

Due Thursday, October 9

1. You are given  $n \times n$  matrices  $A, B, C$  whose elements are from  $\mathbb{Z}_2$  (i.e., the integers mod 2). Show a randomized algorithm running in  $O(n^2)$  time which checks whether  $AB = C$ . The algorithm should output YES if  $AB = C$  and output NO with at least  $3/4$  probability if  $AB \neq C$ .
2. We saw a couple different norms for sparse recovery in our study of Count-Min and Count-Sketch, and we will see more in the future.

We say that  $(k, C)$ -approximate  $\ell_p/\ell_q$  recovery of a vector  $x$  finds an  $\bar{x}$  such that

$$\|x - \bar{x}\|_p \leq C \min_{k\text{-sparse } x'} \|x - x'\|_q$$

In this problem we study implications among the various guarantees. We say that  $(k, C)$   $\ell_p/\ell_q$  recovery “implies”  $(k', C')$   $\ell_{p'}/\ell_{q'}$  recovery if, given any vector  $\bar{x}$  satisfying the former, we can construct a vector  $\bar{x}'$  satisfying the latter.

Some of the parts below describe the transformation required to get the implication, while for others you need to identify the transformation. Suppose that  $C > 1$  and  $0 < \epsilon < 1$ .

- (a)  $(k, \epsilon/k)$   $\ell_\infty/\ell_1$  recovery implies  $(k, 1 + O(\epsilon))$   $\ell_1/\ell_1$  recovery by restricting to the largest  $k$  coordinates.
- (b)  $(k, \sqrt{\epsilon/k})$   $\ell_\infty/\ell_2$  recovery implies  $(k, 1 + O(\epsilon))$   $\ell_2/\ell_2$  recovery by restricting to the largest  $2k$  coordinates.
- (c)  $(2k, C)$   $\ell_2/\ell_2$  recovery implies  $(k, C/\sqrt{k})$   $\ell_2/\ell_1$  recovery.
- (d)  $(k, C/\sqrt{k})$   $\ell_2/\ell_1$  recovery implies  $(k, O(C))$   $\ell_1/\ell_1$  recovery.

3. Give a *deterministic* algorithm for the heavy-hitters problem.
- (a) Suppose a stream has  $m$  elements in  $[n]$ , and let  $x_i$  denote the number of elements equal to  $i$ . Construct a deterministic streaming algorithm using  $O(\log(mn))$  bits of space that outputs a single integer  $j$  such that, if there exists an  $i$  such that  $x_i > m/2$ , then the algorithm outputs  $j = i$ .
  - (b) Extend your result to use  $O(k \log(mn))$  bits to output  $k$  numbers such that, for any  $i$  with  $x_i > m/(k+1)$ , then  $i$  is in the list being output.
  - (c) [Optional] Give a deterministic algorithm that supports streaming insertions *and deletions* and gets a result similar to parts (a) or (b). In particular, use  $O(k \log^c(mn))$  space to output  $O(k)$  numbers such that for any  $i$  with  $x_i > \|x\|_1/k$ ,  $i$  is in the output list.
4. The power dissipated by a resistor with resistance  $r$  going between two vertices of voltage  $v_1$  and  $v_2$  is  $(v_1 - v_2)^2/r$ . We can think about a resistor network as a multigraph, where each edge is associated with a resistance  $r_e$ . If we assign a set of voltages  $v_i$  to the vertices, then the total power dissipated is simply the sum over all resistors of the power dissipated by that resistor.

Consider maintaining a resistor network under a stream with two kinds of updates:

- INSERT( $(i, j)$ , “tag”,  $r$ ) which inserts a new resistor labeled “tag” of resistance  $r$  between  $i$  and  $j$ .
  - DELETE( $(i, j)$ , “tag”,  $r$ ) which deletes the resistor labeled “tag” of resistance  $r$  between  $i$  and  $j$ .
- (a) Give a streaming algorithm to maintain a sketch such that, for any set  $S$  of vertices, you can estimate the energy used by the circuit if the nodes of  $S$  are set to 1 volt and the rest are set to 0 volts. You should use  $O(n \frac{1}{\epsilon^2} \log(1/\delta))$  words to get an  $1 \pm \epsilon$  approximation with probability  $1 - \delta$  for each  $S$ .
  - (b) Extend this to estimate the energy used by the circuit for any assignment  $v_1, \dots, v_n$  of voltages to vertices, to error  $1 \pm \epsilon$  with probability  $1 - \delta$ .

- (c) Suppose now that we only allow insertions of resistors. Show how to use  $O(\frac{n}{\epsilon^2} \log^c n)$  bits to have a sketch that with high probability can estimate the energy of *every* assignment of voltages to vertices up to  $1 \pm \epsilon$  error.

**Hint:** You may use the fact that spectral sparsifiers exist. In particular, for any weighted graph  $G$  on  $n$  vertices, there is an efficient offline algorithm to construct a graph  $H$  on those vertices with only  $O(\frac{n}{\epsilon^2} \log^c n)$  edges that matches the energy of *every* assignment of voltages to vertices up to  $1 \pm \epsilon$  error.