## Algorithms for Big Data

Fall Semester 2019

Exercise Set 1

In the following we are concerned in designing a (memory/query) efficient algorithm for a following problem: we are given (in an offline form<sup>1</sup>) a binary array A[1..n] where  $\forall_i A[i] \in \{0,1\}$ . Our goal is to estimate (up to some additive error  $\varepsilon$ ) the value of  $Y = \frac{1}{n} \sum_i A[i]$  using only little additional memory.

Exercise 1: (2 pts)

Show that simple random sampling performs well: select independently  $i_1, i_2, \ldots, i_k \in [n]$ . Show that  $\frac{1}{k}(A[i_1] + \ldots + A[i_k])$  is an unbiased estimator of Y. Use Hoeffding bound to bound k, the number of samples necessary, so that the estimation holds:

- with probability 9/10,
- with probability 1 1/n? (So called with high probability.)

Exercise 2: (2 pts)

Use Chebyshev's inequality (instead of Hoeffding bound) to bound k from Exercise 1. How many samples do we need so that the estimation holds:

- with probability 9/10,
- with probability 1 1/n?

Exercise 3: (2 pts)

Consider 9/10 probability estimation from previous exercise. Consider t fully independent repetitions of the same estimation procedure, with values  $Y_1, Y_2, \ldots, Y_t$ . Show that for  $t = \Theta(\log n)$ , the value of  $\operatorname{median}(Y_1, \ldots, Y_t)$  is an  $\pm \varepsilon$  estimation of Y with high probability. What is the total number of samples needed?

Exercise 4: (2 pts)

- Prove Markov's inequality.
- Show that Chebyshev's inequality follows from Markov's inequality.

<sup>&</sup>lt;sup>1</sup>offline: read-only

 $<sup>^{2}</sup>X$  is an unbiased estimator of Y iff  $\mathbb{E}[X] = Y$ .

**Theorem 1 (Markov's inequality)** Let  $X \ge 0$  be a random variable. Then for any  $k \ge 1$ :

$$\Pr(X \ge k \cdot \mathbb{E}[X]) \le \frac{1}{k}.$$

Theorem 2 (Chebyshev's inequality) Let X be a random variable. For any k > 0:

$$\Pr(|X - \mathbb{E}[X]| \ge k \cdot \sqrt{\operatorname{Var}[X]}) \le \frac{1}{k^2}.$$

Theorem 3 (Hoeffding bound) Let  $X_1, X_2, \ldots, X_n \in \{0, 1\}$  be fully independent random variables. Let  $X = \sum_i X_i$ . Then:

$$\Pr(|X - \mathbb{E}[X]| \ge t) \le 2\exp(-\frac{t^2}{n}).$$