

## 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  $i_1, i_2, \dots, i_k \in [n]$ . Show that  $\frac{1}{k}(A[i_1] + \dots + A[i_k])$  is an unbiased estimator of  $Y$ .<sup>2</sup> 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  $Y_1, Y_2, \dots, Y_t$ . Show that for  $t = \Theta(\log n)$ , the value of  $\text{median}(Y_1, \dots, Y_t)$  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.

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<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  $X \geq 0$  be a random variable. Then for any  $k \geq 1$ :*

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

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

$$\Pr(|X - \mathbb{E}[X]| \geq k \cdot \sqrt{\text{Var}[X]}) \leq \frac{1}{k^2}.$$

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

$$\Pr(|X - \mathbb{E}[X]| \geq t) \leq 2 \exp\left(-\frac{t^2}{n}\right).$$