

## High Dimensional Probability with Applications to Big Data Sciences

Summer term 2020

## Assignment 1: Appetizers Due: Tuesday, May 19, 2020

**Problem 1** (Integral Inequalities). The aim of this exercise is to prove useful integral identities that are used to bound the expectation and  $p^{th}$  moments by tail probabilities.

1. Let X be a non-negative random variable. Show that

$$\mathbb{E}X = \int_0^\infty \mathbb{P}(X > t) \, dt.$$

The two sides of the identity are either finite or infinite simultaneously.

2. Show that for any random variable X (not necessarily non-negative):

$$\mathbb{E}X = \int_0^\infty \mathbb{P}(X > t) \, dt - \int_{-\infty}^0 \mathbb{P}(X < t) \, dt.$$

3. Let X be a random variable and  $p \in (0, \infty)$ . Show that

$$\mathbb{E}|X|^p = \int_0^\infty pt^{p-1}\mathbb{P}(|X| > t) \ dt$$

whenever the right-hand side exists.

**Problem 2** (Chebyshev's Inequality). Let X be a random variable with mean  $\mu$  and variance  $\sigma^2$ . Prove Chebyshev's inequality: for any t > 0,

$$\mathbb{P}(|X - \mu| \ge t) \le \frac{\sigma^2}{t^2}.$$

Please turn the page.

**Problem 3** (Weak Law of Large Numbers). Let  $X_1, X_2, \ldots$  be a sequence of i.i.d. random variables with finite mean  $\mu$  and variance  $\sigma^2$ . Show that for any t > 0,

$$S_n := \frac{1}{n} \sum_{i=1}^n X_i \xrightarrow[n \to \infty]{} \mu, \text{ in Probability.}$$

**Problem 4** (Strong Law of Large Numbers). Let  $X_1, X_2, \ldots$  be a sequence of i.i.d. random variables with finite mean  $\mu$  and variance  $\sigma^2$ . Then the Strong Law of Large Numbers (SLLN) states that: for any t > 0,

$$S_n := \frac{1}{n} \sum_{i=1}^n X_i \xrightarrow[n \to \infty]{} \mu \qquad a.s.$$

Prove the SLLN under the additional assumption that  $\mathbb{E}|X_i|^4 < \infty$ . *Hint: Show that*  $\mathbb{E}(S_n - \mu)^4 = \mathcal{O}(n^{-2})$ .

**Problem 5** (Central Limit Theorem). Prove the CLT: Let  $X_1, X_2, \ldots$  be a sequence of i.i.d. random variables with finite mean  $\mu$  and variance  $\sigma^2$ . Then

$$S_n := \frac{1}{\sigma \sqrt{n}} \sum_{i=1}^n (X_i - \mu) \xrightarrow[n \to \infty]{\mathcal{D}} \mathcal{N}(0, 1).$$