## Homework 5

Due: 8:00pm (PDT) Thursday, Nov 10th, 2022

For Problems 1 and 2, review your undergrad probbility for pairs of random variables and/or read Chapter 4 of Prof. Kim's note

- 1. Problem 4.5 of Prof. Kim's note.
- 2. Problem 4.11 of Prof. Kim's note.
- 3. Fair vs Unfair Coin:
  - (a) Consider a coin with the probability of head p = 0.3. Provide an estimate for the probability that after 1000 flipping of this coin we observe between 250 and 350 heads.
  - (b) Repeat the above part for a fair coin with the probability of head p = 0.5, i.e., provide an estimate for the probability that after 1000 flipping of this coin we observe between 250 and 350 heads.
  - (c) If a third party mixes the two coins, how can you distinguish between the two coins?
- 4. Consider the simple pandemics model  $X_{k+1} = W_k X_k$  where  $\{W_k\}$  is an i.i.d. sequence that is uniformly distributed over [0.4, 3].
  - (a) Find the CDF and PDF of  $\log W_k$ .
  - (b) Find  $\mathbb{E}[\log W_k]$ . What can you say about  $\lim_{k\to\infty} X_k$ ?
  - (c) Let  $X_1 = 1$ . Generate 100 sample paths  $\{X_k\}$  for  $1 \le k \le 200$  and plot them using the regular linear plot (command:plot) and log-plot (command:semilogy) and store all the 100 sample paths for a later use.
  - (d) Theoretically explain your observation.
  - (e) For a level  $\alpha \geq 1$ , define the random variable  $T_{\alpha} = \min\{k \geq 1 \mid X_k \geq \alpha\}$ . Such random variables are called *stopping times*. In this case, it is the first time that the pandemics hits  $\alpha$  people. Show that  $T_{\alpha}$  is indeed a random variable.
  - (f) Let  $\alpha = 10^4$ . Familiarize yourself with the command histogram. Plot the histogram of  $T_{10^4}$  for the 100 sample paths.
  - (g) Based on the observed data, provide an estimate for  $\mathbb{E}[T_{10^4}]$ .
- 5. For a random vector  $\mathbf{X} = (X_1, \dots, X_n)$ , we define its covariance matrix to be the  $n \times n$  matrix C with

$$C_{ij} = \mathbb{E}[(X_i - \bar{X}_i)(X_j - \bar{X}_j)],$$

where  $\bar{X}_i = \mathbb{E}[X_i]$ . Show that a covariance matrix C is always positive semi-definite. Can we say that it is always positive definite?

**hint**: An  $n \times n$  symmetric matrix A is called positive semi-definite if  $x^T A x \ge 0$  for all  $x \in \mathbb{R}^n$ . If the inequality holds as a strict inequality for all non-zero x, it is called positive definite.

6. In the class, we looked at the LMMSE estimator of X given Y = X + Z, where  $X, Z \sim \mathcal{N}(0, 1)$  are independent. We found out that in this case the LMMSE estimator is  $\hat{X} = \frac{1}{2}Y$ . Now, suppose that we have k independent measurements  $Y_i = X + Z_i$  for i = 1, ..., k of X where  $X, Z_1, ..., Z_k \sim \mathcal{N}(0, 1)$  are all independent. Find the LMMSE estimator of X given  $Y_1, ..., Y_k$  using the provided formula

$$\hat{X} = \mathbf{Cov}(X, Y)\mathbf{Cov}^{-1}(Y, Y)(Y - \bar{Y}) + \bar{X}.$$

You don't need to prove the above identity.

7. Suppose that  $\{X_k\}$  is an i.i.d. random process with finite mean  $\mathbb{E}[X_k] = \mu$ . By the strong law of large numbers we know that

$$\lim_{n \to \infty} \frac{S_n}{n} = \mu \quad \text{a.s.},$$

where  $S_n = \sum_{k=1}^n X_k$ . Show that if we further have a finite variance, i.e.,  $\mathbf{Var}(X_k) = \sigma^2 < \infty$ , then  $\lim_{n\to\infty} \frac{S_n}{n} \stackrel{L_2}{\to} \mu$ . In other words,

$$\lim_{n\to\infty}\|\frac{S_n}{n}-\mu\|^2=\lim_{n\to\infty}\mathbf{Var}(\frac{S_n}{n})=0.$$

This is known as the weak law of large numbers.