

## Homework 3: Mathematical Statistics (MATH-UA 234)

Due 10/06 at the beginning of class on Gradescope

**Problem 1.** Let

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \quad \vec{A} = \begin{bmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n,1} & a_{n,2} & \cdots & a_{n,n} \end{bmatrix}$$

where  $a_{i,j}$  are constants and  $x_1, x_2, \dots, x_n \sim N(0, 1)$  are independent and identically distributed standard normal random variables.

- (a) What is the covariance matrix  $\vec{\Sigma}$  for  $\vec{x}$ ? (Recall the  $(i, j)$ -entry of  $\vec{\Sigma}$  is  $\text{CoV}[x_i, x_j]$ .)
- (b) Show that  $\mathbb{E}[\vec{x}^T \vec{A} \vec{x}] = \text{tr}(\vec{A}) := a_{1,1} + a_{2,2} + \cdots + a_{n,n}$  (hint: write out the expression for  $\vec{x}^T \vec{A} \vec{x}$  as a sum over the entries of  $\vec{A}$ )
- (c) Suppose  $\vec{A}$  is diagonal; i.e.  $a_{i,j} = 0$  for all  $i \neq j$ . Compute the variance of  $\vec{x}^T \vec{A} \vec{x}$ . You may use the fact that  $x_i^2$  is a Chi-square random variable with one degree of freedom so that  $\mathbb{V}[x_i^2] = 2$ .

This is an example of “stochastic trace estimation” which is an important algorithmic tool in a number of recent algorithms.

**Problem 2** (Wasserstein 5.5). Suppose  $X_1, X_2, \dots, X_n \sim \text{Ber}(p)$  are independent and identically distributed. Let  $Z_n = (X_1^2 + X_2^2 + \cdots + X_n^2)/n$ . Prove that

- (a)  $Z_n$  converges in probability to the constant random variable  $p$ .
- (b)  $Z_n$  converges in quadratic mean to the constant random variable  $p$ .

**Problem 3** (Wasserstein 6.1). Suppose  $X_1, X_2, \dots, X_n \sim \text{Poisson}(\lambda)$  are independent and identically distributed and let  $\hat{\lambda} = n^{-1} \sum_{i=1}^n X_i$  be our point estimator for  $\lambda$ . Find the bias  $\text{Bias} = \mathbb{E}[\hat{\lambda}] - \lambda$ , standard error  $\text{se}_n = \sqrt{\mathbb{V}[\hat{\lambda}]}$ , and mean squared error  $\text{MSE}_n = \mathbb{E}[(\hat{\lambda} - \lambda)^2]$  of  $\hat{\lambda}$ .

**Problem 4** (Wasserstein 6.2). Suppose  $X_1, X_2, \dots, X_n \sim \text{Unif}(0, \theta)$  are independent and identically distributed and let  $\hat{\theta} = \max\{X_1, X_2, \dots, X_n\}$  be our point estimator for  $\theta$ .

- (a) Write down the distribution for  $\hat{\theta}$  (hint: we’ve already done a similar problem)
- (b) Find the bias  $\text{Bias} = \mathbb{E}[\hat{\theta}] - \theta$ , standard error  $\text{se}_n = \sqrt{\mathbb{V}[\hat{\theta}]}$ , and mean squared error  $\text{MSE}_n = \mathbb{E}[(\hat{\theta} - \theta)^2]$  of  $\hat{\theta}$ .

**Problem 5** (Wasserstein 6.3). Suppose  $X_1, X_2, \dots, X_n \sim \text{Unif}(0, \theta)$  are independent and identically distributed and let  $\hat{\theta} = 2\bar{X}_n$  be our point estimator for  $\theta$ . Find the bias  $\text{Bias} = \mathbb{E}[\hat{\theta}] - \theta$ , standard error  $\text{se}_n = \sqrt{\mathbb{V}[\hat{\theta}]}$ , and mean squared error  $\text{MSE}_n = \mathbb{E}[(\hat{\theta} - \theta)^2]$  of  $\hat{\theta}$ .

**Problem 6.** Suppose  $X_1, X_2, \dots, X_n \sim F_{\mu, \sigma^2}$  are independent and identically distributed samples from some distribution  $F_{\mu, \sigma^2}$  with mean  $\mu$  and variance  $\sigma^2$ .

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problems with a textbook reference are based on, but not identical to, the given reference

Recall that

$$\hat{S}_n^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2, \quad \hat{T}_n^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n)^2$$

are both point estimators for the parameter  $\sigma^2$ .

- (a) Compute  $\mathbb{E}[\hat{S}_n^2]$  and  $\mathbb{E}[\hat{T}_n^2]$ .
- (b) Which point estimator has smaller bias?
- (c) Which point estimator has smaller standard error?

**Problem 7.** Describe of point estimation of a parameter which you noticed in a different part of **your life** (e.g. in other classes, on the subway, at the park, in the dorm, etc.).