# Problem Set 4

### **Submission:**

Thursday, 03/03/2022, until 1 PM, to be uploaded on the NYU Brightspace course homepage.

## 1. A brief reminder on conditional distributions

[4 Points]

In this problem we recall the notion of **conditional distributions** for two random variables X and Y. This is in particular relevant for the notion of sufficient statistics.

• If X and Y are discrete random variables with values in  $\Omega_X$  and  $\Omega_Y$  respectively, then for any  $y \in \Omega_Y$  with  $\mathbf{P}[Y=y] > 0$ , the conditional distribution of X given Y=y is characterized by the conditional probability mass function

$$p_{X|Y=y}(x) = \frac{\mathbf{P}[X=x, Y=y]}{\mathbf{P}[Y=y]}, \quad x \in \Omega_X.$$

This is the distribution of X under the probability measure  $\mathbf{P}[\cdot|Y=y]$ .

• If X and Y are continuous real random variables, then for any  $y \in \mathbb{R}$  with  $f_Y(y) > 0$ , the conditional distribution of X given Y = y is characterized by the conditional probability density function

$$f_{X|Y=y}(x) = \frac{f_{X,Y}(x,y)}{f_Y(y)}, \qquad x \in \mathbb{R},$$

where  $f_{X,Y}$  is the joint probability density function of (X,Y) and  $f_Y$  is the probability density function of Y.

- (a) A fair coin is tossed 4 times. Let X denote the number of times that heads comes up and Y=1 if heads comes up on the first toss and Y=0 otherwise. Determine the conditional distribution of X given Y=1 and the conditional distribution of Y given X=3.
- (b) Consider jointly continuous random variables X, Y with density

$$f_{X,Y}(x,y) = 4ye^{-2y(x+1)} \mathbb{1}_{\{x,y>0\}}$$

Determine the conditional probability density function  $f_{X|Y=y}$ . What is the conditional distribution of X given Y=y?

#### 2. Sufficient statistics

[4 Points]

- (a) Let  $X_1,...,X_n \overset{\text{i.i.d.}}{\sim} Pois(\lambda)$  with unknown parameter  $\lambda > 0$ . Show that  $T(\mathbf{X}) = \sum_{j=1}^n X_j$  is a sufficient statistic for  $\lambda$ 
  - (i) ...directly, using the definition of sufficiency.
  - (ii) ...using the Neyman-characterization of sufficiency.

*Hint:* For (i), use the result of Problem 2 (a) on Problem set 1. You need to calculate  $\mathbf{P}[X_1 = x_1, ..., X_n = x_n | T(\mathbf{X}) = t]$  for all possible values of  $x_1, ..., x_n, t \in \mathbb{N}_0$ .

(b) Let  $X_1,...,X_n \overset{\text{i.i.d.}}{\sim} Pareto(\lambda,a)$  with  $known\ a>0$  and unknown  $\lambda>0$ , where we say that  $X\sim Pareto(\lambda,a)$  if

$$f_X(x) = \frac{\lambda a^{\lambda}}{x^{\lambda+1}} \mathbb{1}_{(a,\infty)}(x).$$

Find a sufficient statistic  $T(\mathbf{X}) \in \mathbb{R}$  for  $\lambda$ , using the Neyman-characterization.

- (a) Consider  $X_1,...,X_n \overset{\text{i.i.d.}}{\sim} \mathcal{U}([0,\theta])$  with unknown  $\theta>0$ . Calculate an estimator  $\widehat{\theta}_n$  for  $\theta$  based on the method of moments. Check this estimator for consistency and unbiasedness.
- (b) Consider  $X_1,...,X_n \overset{\text{i.i.d.}}{\sim} Bin(N,p)$ , where  $\theta = (N,p) \in \mathbb{N} \times (0,1)$  is unknown (meaning both N and p are unknown). Determine an estimator  $(\widehat{N}_n,\widehat{p}_n)$  for (N,p) based on the method of moments.

*Hint:* You need both  $\mathbf{E}_{(N,p)}[X_1]$  and  $\mathbf{E}_{(N,p)}[X_1^2]$ .

## 4. (R exercise) The empirical distribution

[4 Points]

Note: Please provide your source code and images obtained with your solution.

Suppose that  $X_1, ..., X_n$  are i.i.d. real random variables such that  $X_1$  has cumulative distribution function F. Given a realization of these random variables, we can consider the *empirical cumulative distribution function* 

$$\widehat{F}_n(x) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{X_i \le x\}}.$$

- (a) Explain why  $\widehat{F}_n(x)$  fulfills the following:
  - (i) For every  $x \in \mathbb{R}$ , one has  $\mathbf{E}[\widehat{F}_n(x)] = F(x)$  and  $\mathrm{Var}[\widehat{F}_n(x)] = \frac{1}{n}F(x)(1 F(x))$ .
  - (ii) For every  $x \in \mathbb{R}$ , one has  $\widehat{F}_n(x) \xrightarrow[n \to \infty]{\mathbf{P}} F(x)$ .

*Hint:* For any event A in some probability space, one has  $\mathbf{E}[\mathbbm{1}_A] = \mathbf{P}[A]$ .

- (b) Use R to generate  $n \in \{5, 20, 500\}$  samples of i.i.d. random variables  $X_1, ..., X_n$  following a  $\mathcal{U}([0,1])$ -distribution or a  $\mathcal{E}(3)$ -distribution. Plot the empirical cumulative distribution function together with the graph of the cumulative distribution function  $F_{X_1}$ . What do you observe?
  - *Hint:* The R-command ecdf() calculates the empirical distribution function of a vector and plot(ecdf()) plots the respective graph.
- (c) Data on the magnitudes of earthquakes near Fiji are available from R, using the command quakes.  $^1$  For help on this dataset type ?quakes. Plot a histogram and the empirical cumulative distribution function for the *magnitudes*. Calculate the average  $\overline{X}_n$  and sample variance  $S_n^2$  for the magnitude.
  - *Hint:* The data set quakes is a data frame containing information on 5 observations (i.e. a table with 5 columns). To obtain a vector *only* containing the data in column 1, use quakes [, 4].
- (d) Suppose it is suggested that the data for the magnitudes  $X_1,...,X_n$  can be modelled by a  $\Gamma(\alpha,\beta)$  distribution. Find consistent estimators  $\widehat{\alpha}_n$  and  $\widehat{\beta}_n$  for  $\alpha$  and  $\beta$ , and calculate the estimates using the data from quakes. Plot the cumulative distribution function of  $\Gamma(\widehat{\alpha}_n,\widehat{\beta}_n)$  together with the empirical distribution function of the data. What do you observe?

*Hint:* Recall that for  $X \sim \Gamma(\alpha, \beta)$ , we have  $\mathbf{E}[X] = \frac{\alpha}{\beta}$  and  $\mathrm{Var}[X] = \frac{\alpha}{\beta^2}$ .

 $<sup>^{1}</sup> alternatively, you can find this data on {\it https://www.stat.cmu.edu/larry/all-of-statistics/=data/fijiquakes.dat}$