Foundations of Statistics

Homework 7

Exercise 1 (Order statistic, part I). Let $X_1, X_2, ..., X_n$ be i.i.d. real-valued random variables with CDF $x \mapsto F(x) \in [0, 1]$. Let us consider their maximum and minimum:

$$Y := \max\{X_1, ..., X_n\}, \quad Z := \min\{X_1, ..., X_n\}.$$

The random variables Y and Z are called *largest order statistic* and *smallest order statistic*, respectively.

a Prove that the distribution function of Y is given by

$$F_Y(y) = F(y)^n \quad \forall y \in \mathbb{R}.$$

If, in particular, F has density function f, find the density function f_Y of the random variable Y.

b Prove that the distribution function of Z is given by

$$F_Z(z) = 1 - [1 - F(z)]^n \quad \forall z \in \mathbb{R}.$$

If, in particular, F has density function f, find the density function f_Z of the random variable Z.

- c Find the joint CDF of the random vector $\mathbf{U} := (Z, Y)^{\top}$.
- d If, in particular, F has density function f, find the joint density function of U. Are Z and Y independent?

Exercise 2 (Order statistic, part I, examples).

a Let $U, V \sim \text{Unif}(0, 1)$ be independent. Based on the previous exercise, find density function of $\max\{U, V\}$ and $\min\{U, V\}$. Compare your result with a simulation in R. Generate random samples from the uniform distribution, and for each pair, record both the maximum and minimum values. Finally, plot the histogram of these values.

- b Let $U, V \sim \text{Unif}(0, 1)$ be independent and $p \in (0, 1)$ be a constant. In HW4, Exercise 1, we studied indicator random variables and showed that $\mathbb{I}_{\{U \leq p\}} \sim \text{Ber}(p)$. Now, use order statistic to find the distribution of the random variables $\mathbb{I}_{\{U \leq p\}} \mathbb{I}_{\{V \leq p\}}$ and $\mathbb{I}_{\{U \leq p\}} + \mathbb{I}_{\{V \leq p\}}$.
- c Let $X_1, \dots X_n \sim \text{Unif}(a, b)$ be i.i.d random variables. Find CDF F_Y and F_Z of the largest and smallest order statistic Y and Z, respectively. Do random variables Y and Z converge in distribution as $n \to \infty$?
- d Let $X_1, \dots X_n \sim \exp(\lambda)$ be i.i.d random variables. Find CDF F_Y and F_Z of the largest and smallest order statistic Y and Z, respectively. Do random variables Y and Z converge in distribution as $n \to \infty$?

Exercise 3 (Sample skewness and sample kurtosis).

- a Show with Chebyshev's inequality that for any random variable not more than about 11% of the data can be more than three standard deviations away from the mean.
- b Show that for a $N(\mu, \sigma^2)$ -distributed random variable the proportion calculated in a) is now 0.3%.
- c For a sample $x_1, ..., x_n$ the z-score is defined by

$$z_i := \frac{1}{\tilde{s}}(x_i - \bar{x}), \quad i = 1, ..., n.$$

Here \bar{x} and \tilde{s} are the sample mean and standard deviation (with denominator n not n-1), respectively. Explain what $z_i = 3$ means.

- d Install the package UsingR with the commands install.packages ("UsingR") and require("UsingR"). The dataset exec.pay contains direct compensation data for 199 United States CEOs. Compare the mean, median and quantiles by using the function summary(exec.pay). Draw the boxplot and determine the outliers.
- e Calculate with R the z-score of the data to find out what proportion of the data are more than 3 standard deviations from the mean. Compare your result with the results in a) and b).
- f The sample skewness is defined by

$$\sqrt{n} \cdot \frac{\sum_{i=1}^{n} (x_i - \bar{x})^3}{\left[\sum_{i=1}^{n} (x_i - \bar{x})^2\right]^{3/2}}$$

Show that this is equal to

$$\frac{1}{n}\sum_{i=1}^n z_i^3.$$

- g Calculate the sample skewness of the exec.pay dataset.
- h The *sample kurtosis* is the measure of the tails in a data set. Long tails will lead to larger values, while "normal" data will have kurtosis close to 0. It is defined by the formula

$$n \cdot \frac{\sum_{i=1}^{n} (x_i - \bar{x})^4}{\left[\sum_{i=1}^{n} (x_i - \bar{x})^2\right]^2} - 3.$$

Show that this is equal to

$$\frac{1}{n} \sum_{i=1}^{n} z_i^4 - 3.$$

Guess why we are taking out number 3 here.

- i Calculate the kurtosis of the exec.pay dataset.
- **Exercise 4.** Suppose we have a computer program consisting of n=100 pages of code. Let X_i be the number of errors on the i^{th} page of code. Suppose that the X_i 's are Poisson with mean 1 and that they are independent. Let $Y = \sum_{i=1}^{n} X_i$ be the total number of errors. Use the Central Limit Theorem to approximate $\mathbb{P}(Y \leq 90)$. Check your answer with the exact value of $\mathbb{P}(Y \leq 90)$. (*Hint:* recall HW5, Exercise 4(c)).
- **Exercise 5.** An accountant wants to simplify his bookkeeping by rounding amounts to the nearest integer, for example, rounding ≤ 99.53 and ≤ 100.46 both to ≤ 100 . What is the cumulative effect of this if there are, say, 100 amounts? To study this we model the rounding errors by n = 100 independent U(-0.5, 0.5) random variables $X_1, ..., X_{100}$.
 - a Compute the expectation and the variance of each X_i .
 - b Use Chebyshev's inequality to compute an upper bound for the probability $\mathbb{P}(|X_1 + X_2 + ... + X_{100}| > 10)$ that the cumulative rounding error $X_1 + X_2 + ... + X_{100}$ exceeds $\in 10$.
 - c What can you say about the mean of the absolute error $\frac{1}{n} \sum_{i=1}^{n} |X_i|$, applying the Law of Large Numbers?