2 Data Gathering

Exercise 2.1

- 1. Assume that X_1, \ldots, X_n are n samples from a cdf F(x). For a given value $x \in \mathbb{R}$, show that $n\hat{F}(x)$ follows a binomial distribution with parameters n and success probability F(x).
- 2. Calculate $\mathbb{E}(\hat{F}(x))$ and $\operatorname{var}(\hat{F}(x))$.

Exercise 2.2

- 1. Simulate n > 0 independent samples of a random variable following a normal distribution with mean 0.3 and standard-deviation 1.4.
- 2. Compute and plot the ecdf of these samples.
- 3. Compare with a figure the ecdf with the true distribution.
- 4. Propose an algorithm to verify (approximatively) the Dvoretzky-Kiefer-Wolfowitz inequality. Code this algorithm.

Hints: the verification should be based on M sequences of n samples. Each sequence is used to compute an estimate of the maximum gap between the ecdf and the cdf. Finally, the M sequences allow you to estimate the probability that the maximum gap exceeds a given value ε .

- 5. Discuss the accuracy of the Dvoretzky-Kiefer-Wolfowitz inequality with respect to n.
- 6. Is the empirical mean close to the true mean? Discuss the result with respect to n.
- 7. Conclude.

Exercise 2.3

Assume that r_1, r_2, \ldots, r_n are n samples from a uniform distribution (0, 1). Use the Inverse-transform technique to calculate n samples x_1, x_2, \ldots, x_n from the Triangular distribution. The probability density function of the Triangular distribution is:

$$f_a(x) = \frac{2}{a} \left(1 - \frac{x}{a} \right), \ 0 \le x \le a,$$

where a > 0 is known, and $f_a(x) = 0$ otherwise.

- 1. Verify that $f_a(x)$ is a probability density function. Draw $f_a(x)$. Justify the name of this distribution.
- 2. Calcultate the cdf $F_a(x)$.
- 3. Calculate $F_a^{-1}(x)$.
- 4. Describe the algorithm to calculate the samples x_1, x_2, \ldots, x_n .

Exercise 2.4

We want to generate n i.i.d. samples x_1, \ldots, x_n from a discrete random variable X following the distribution:

i	1	2	3	4
u_i	1	3	5	7
$p_i = \Pr(X = u_i)$	0.1	0.3	0.4	0.2

1. Propose an algorithm which transforms n samples r_1, r_2, \ldots, r_n generated from a uniform distribution (0, 1) into the n samples x_1, \ldots, x_n .

- 2. Implement the algorithm with R.
- 3. Verify the quality of the algorithm by plotting the histogram of n = 1000 samples.
- 4. Let Y_i be the fraction of generated samples taking on the value u_i . What is the mean and the variance of Y_i ? For n = 1000, is the histogram an accurate estimate of the target discrete distribution? Is the accuracy the same for all the Y_i 's?

Exercise 2.5

Let us consider a random vector (X,Y) composed of two random variables. A random vector is said to be a normal random vector if all linear combinations of X and Y is a normal random variable, i.e., if the random variable T = aX + bY is a normal random variable for all $(a,b) \in \mathbb{R}^2$, $(a,b) \neq (0,0)$.

- 1. Show that a couple of two independent normal random variables (X, Y), with $\mathbb{E}(X) = m_X$, $\mathbb{E}(Y) = m_Y$, $\text{var}(X) = \sigma_X^2$ and $\text{var}(Y) = \sigma_Y^2$, is a normal random vector.
- 2. Let U, V and W three normal random variables with zeros means and unit variances. These variables are used to create the random variables X and Y:

$$X = \sqrt{1 - \varrho} \, U + \sqrt{\varrho} \, W \tag{1}$$

$$Y = \sqrt{1 - \varrho} \, V + \sqrt{\varrho} \, W \tag{2}$$

where $0 \le \varrho \le 1$ is a given parameter.

- (a) Show that (X, Y) is a normal random vector.
- (b) Calculate $\mathbb{E}(X)$, $\mathbb{E}(Y)$, var(X), var(Y), cov(X,Y) and corr(X,Y).
- (c) Propose an algorithm to generate a couple of correlated normal random variables with zero means, unit variances and a correlation $\varrho \geq 0$ from some realizations of three independent normal random variables.
- (d) How to modify this algorithm if $\varrho < 0$?

Hints in R

1. Useful R commands:

R-function	Package
ecdf	stats

2. To install and use a package (example with "stats"):

install.packages("stats")
library(stats)