

ADVANCED MATHEMATICS AND STATISTICS

MODULE 2 - ADVANCED STATISTICAL METHODS

Exercise 1. If the joint probability density function of (X, Y) is defined by

$$f(x, y) = \begin{cases} x(y-x)e^{-y} & 0 < x < y < +\infty \\ 0 & \text{elsewhere} \end{cases}$$

- (a) Determine the marginal distributions of X and Y .
- (b) How can you simulate realizations of the random variable Y ?
- (c) Identify the probability distribution of $U = X/Y$.

Solution.

(a)

$$\begin{aligned} f_X(x) &= x \int_x^{+\infty} (y-x) e^{-y} dy \mathbf{1}_{(0,+\infty)}(x) = x e^{-x} \int_0^{+\infty} w e^{-w} dw \mathbf{1}_{(0,+\infty)}(x) \\ &= x e^{-x} \mathbf{1}_{(0,+\infty)}(x) \end{aligned}$$

$$\begin{aligned} f_Y(y) &= e^{-y} \int_0^y x(y-x) dx \mathbf{1}_{(0,+\infty)}(y) = e^{-y} \int_0^1 y^3 w(1-w) dw \mathbf{1}_{(0,+\infty)}(y) \\ &= \frac{1}{\Gamma(4)} y^3 e^{-y} \mathbf{1}_{(0,+\infty)}(y) \end{aligned}$$

(b) Note that $Y \sim \text{Ga}(4, 1)$. Hence, Y equals in distribution the sum of 4 independent and identically distributed random variables having a negative-exponential distribution with parameter 1, i.e.

$$Y \stackrel{d}{=} X_1 + X_2 + X_3 + X_4$$

where $X_i \stackrel{\text{iid}}{\sim} \text{E}(1)$. If U_1, \dots, U_4 are iid from a $\text{Unif}(0, 1)$, then $X_i = -\log(1 - U_i)$ for each $i = 1, 2, 3, 4$. Hence to simulate Y , one may proceed as follows:

(1) For $i = 1, 2, 3, 4$

(1.1) generate $U_i \sim \text{Unif}(0, 1)$

(1.1) set $X_i = -\log(1 - U_i)$

(2) Set $Y = X_1 + X_2 + X_3 + X_4$

(c) Set $U = g_1(X, Y) = X/Y$ and $V = g_2(X, Y) = Y$ and note that

$$g_1^{-1}(u, v) = uv, \quad g_2^{-1}(u, v) = v.$$

The Jacobian of the transformation is

$$J = \begin{pmatrix} v & u \\ 0 & 1 \end{pmatrix}$$

and $|\det(J)| = |v|$. Hence

$$\begin{aligned} f_{U,V}(u, v) &= f_{X,Y}(uv, v) |v| = uv(v - uv)|v| e^{-v} \mathbb{1}_{(0,+\infty)}(v) \mathbb{1}_{(0,v)}(uv) \\ &= v^3 e^{-v} u(1 - u) \mathbb{1}_{(0,+\infty)}(v) \mathbb{1}_{(0,1)}(u) \\ &= \frac{1}{\Gamma(4)} v^3 e^{-v} \mathbb{1}_{(0,+\infty)}(v) \frac{\Gamma(4)}{\Gamma(2)\Gamma(2)} u(1 - u) \mathbb{1}_{(0,1)}(u) \\ &= f_V(v) f_U(u) \end{aligned}$$

which entails that $V \perp U$ and $U \sim \text{Beta}(2, 2)$.

Exercise 2. Let X and Y be independent random variables such that

$$f_X(x) = 2e^{-2x} \mathbf{1}_{(0,+\infty)}(x), \quad f_Y(y) = 4ye^{-2y} \mathbf{1}_{(0,+\infty)}(y).$$

- (a) Determine $\mathbb{P}[X < Y]$
- (b) Show that the moment generating function of Y exists and determine it.
- (c) Using the result in (b) evaluate the first two moments of Y , i.e. $\mathbb{E}Y$ and $\mathbb{E}Y^2$

Solution.

(a)

$$\begin{aligned} \mathbb{P}[X < Y] &= \int_{-\infty}^{+\infty} dy \int_{-\infty}^y dx f_{X,Y}(x, y) = \int_{-\infty}^{+\infty} f_Y(y) \left\{ \int_0^y f_X(x) dx \right\} dy \\ &= 4 \int_0^{+\infty} ye^{-2y} \left\{ \int_0^y 2e^{-2x} dx \right\} dy \\ &= 4 \int_0^{+\infty} ye^{-2y} (1 - e^{-2y}) dy \\ &= 4 \left(\frac{1}{4} - \frac{1}{16} \right) = \frac{3}{4}. \end{aligned}$$

(b)

$$\begin{aligned} m_Y(t) &= \int_{-\infty}^{+\infty} e^{ty} f_Y(y) dy = \int_0^{+\infty} e^{ty} 4ye^{-2y} dy \\ &= 4 \int_0^{+\infty} ye^{-(2-t)y} dy \end{aligned}$$

and it is apparent that $m_Y(t) < +\infty$ for any $t < 2$. Hence, there exists a neighbourhood of the origin where m_Y is defined. We can conclude that, for any $t < 2$,

$$m_Y(t) = \frac{4}{(2-t)^2}.$$

(c) Note that since $m'_Y(t) = 8(2-t)^{-3}$, one has

$$\mathbb{E}Y = m'_Y(0) = 1.$$

On the other hand, $m''_Y(t) = 24(2-t)^{-4}$ and

$$\mathbb{E}Y^2 = m''_Y(0) = \frac{3}{2}.$$

Exercise 3. Let X be a random variable whose distribution is negative exponential with parameter 1, i.e. $f_X(x) = e^{-x} \mathbb{1}_{(0,+\infty)}(x)$.

- (a) What is the distribution of $Y = 1 - e^{-X}$? Provide an answer and motivate it, without actually determining f_Y through the change of variable formula.
- (b) If X_1, X_2, \dots are random variables that are independent and identically distributed, with the same law as X , let $X_{(n)} = \max\{X_1, \dots, X_n\}$ be the n -th order statistics. Evaluate the cumulative distribution function of $W_n = X_{(n)} - \log n$. (*Optional question*: Does it admit a limit as $n \rightarrow \infty$?)
- (c) If X_1, X_2, \dots are random variables that are independent and identically distributed, with the same law as X , show that $\bar{X}_n/2$ converges in quadratic mean to $1/2$? Does the convergence hold true also almost surely?

Solution.

- (a) Since $F_X(x) = (1 - e^{-x}) \mathbb{1}_{(0,+\infty)}(x)$, then

$$Y = 1 - e^{-X} = F_X(X)$$

and $Y \sim \text{Unif}(0, 1)$.

- (b) Since $F(x) = (1 - e^{-x}) \mathbb{1}_{(0,+\infty)}(x)$, for any $x > 0$ one has

$$F_{X_{(n)}}(x) = (F(x))^n = (1 - e^{-x})^n.$$

From this

$$F_{W_n}(x) = \mathbb{P}[X_{(n)} \leq x + \log n] = (1 - e^{-x - \log n})^n = \left(1 - \frac{e^{-x}}{n}\right)^n$$

for any $x > 0$. It can be seen that as $n \rightarrow \infty$.

$$F_{W_n}(x) \longrightarrow e^{-e^{-x}}$$

- (c) Since $\mathbb{E}(\bar{X}_n/2) = 1/2$, one has

$$\mathbb{E}\left|\frac{1}{2} \bar{X}_n - \frac{1}{2}\right|^2 = \text{Var}(\bar{X}_n/2) = \frac{1}{4n} \rightarrow 0$$

as $n \rightarrow \infty$. If we let $X'_n = X_n/2$, the sequence of random variables $(X'_n)_{n \geq 1}$ satisfies the assumption of the SLLN, which states the following

Let $(X'_n)_{n \geq 1}$ be a sequence of iid random variables. Then $\bar{X}'_n \xrightarrow{\text{a.s.}} \mu$ if and only if $\mathbb{E}X'_1 = \mu < \infty$

According to this, one also has a stronger convergence result and $\bar{X}_n/2 \xrightarrow{\text{a.s.}} 1/2$, as $n \rightarrow \infty$.