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Complementi di Probabilità e Statistica - Advanced Statistics
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Problems on Sequences of Random Variables with Solutions 2021-12-23

Problem 1 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let X be a uniformly distributed real random variable on the interval $[0, 1]$. In symbols, $X \sim U(0, 1)$. Consider the sequence $(Y_n)_{n \geq 1}$ of real random variables given by

$$Y_n \stackrel{\text{def}}{=} \begin{cases} n, & \text{if } 0 \leq X < \frac{1}{n}, \\ 0, & \text{if } 1/n \leq X \leq 1, \end{cases} \quad \forall n \geq 1.$$

Check whether the sequence $(Y_n)_{n \geq 1}$ converges in distribution, converges in probability, converges almost surely, converges in mean, and converges in mean square in the indicated order.

Hint: to deal with the almost sure convergence consider the event $E_0 \equiv \{\omega \in \Omega : X(\omega) \leq 0\}$ and the complement E_0^c .

Solution. Write $F_{Y_n} : \mathbb{R} \rightarrow \mathbb{R}$ for the distribution function of Y_n . We have

$$F_{Y_n}(y) = \mathbf{P}(Y_n \leq y) = \begin{cases} 0, & \text{if } y < 0, \\ \mathbf{P}(1/n \leq X \leq 1) = 1 - 1/n, & \text{if } 0 \leq y < n, \\ 1, & \text{if } n \leq y. \end{cases}$$

Note that for every $y \geq 0$ there exists $n(y) \in \mathbb{N}$, (e.g. $n(y) \equiv \lceil y \rceil$, where $\lceil \cdot \rceil : \mathbb{R} \rightarrow \mathbb{R}$, is the ceiling function), such that $y < n$ for every $n > n(y)$. Therefore, we have definitively,

$$\mathbf{P}(Y_n \leq y) = 1 - 1/n.$$

It then follows

$$\lim_{n \rightarrow \infty} F_{Y_n}(y) = \begin{cases} 0, & \text{if } y < 0, \\ 1 - \lim_{n \rightarrow \infty} 1/n = 1, & \text{if } 0 \leq y. \end{cases}$$

Considering the Heavside function $H : \mathbb{R} \rightarrow \mathbb{R}$ given by

$$H(y) = \begin{cases} 0, & \text{if } y < 0, \\ 1, & \text{if } 0 \leq y, \end{cases}$$

we clearly have

$$\lim_{n \rightarrow \infty} F_{Y_n}(y) = H(y),$$

at any point $y \in \mathbb{R}$. Hence, the sequence $(Y_n)_{n \geq 1}$ converges in distribution to the standard Dirac real random variable $Dir(0)$.

With regard to the convergence in probability, we know that the convergence in distribution to a Dirac random variables $Dir(y_0)$, concentrated at some $y_0 \in \mathbb{R}$, implies also the convergence in probability to $Dir(y_0)$. However, by a direct approach, according to the definition of Y_n , we have

$$\mathbf{P}(Y_n = n) = \mathbf{P}\left(0 \leq X < \frac{1}{n}\right) = \frac{1}{n} \quad \text{and} \quad \mathbf{P}(Y_n = 0) = \mathbf{P}\left(\frac{1}{n} \leq X \leq 1\right) = 1 - \frac{1}{n},$$

for every $n \geq 1$. Therefore, guessing that $Y_n \rightarrow Dir(0)$, we have definitively,

$$\mathbf{P}(|Y_n - Dir(0)| \leq \varepsilon) = \mathbf{P}(|Y_n| \leq \varepsilon) \geq \mathbf{P}(Y_n = 0) = 1 - \frac{1}{n},$$

for every $\varepsilon > 0$. It follows

$$\lim_{n \rightarrow \infty} \mathbf{P}(|Y_n| \leq \varepsilon) \geq 1 - \lim_{n \rightarrow \infty} \frac{1}{n} = 1,$$

for every $\varepsilon > 0$, which eventually shows the convergence in probability of $(Y_n)_{n \geq 1}$ to $Dir(0)$. Now, to check the almost sure convergence to $Dir(0)$ (recall that the almost sure convergence at some random variable implies convergence in probability at the same random variable) consider the event

$$E_0 \equiv \{\omega \in \Omega : X(\omega) \leq 0\}.$$

Since $X \sim U(0, 1)$, we have $\mathbf{P}(E_0) = \mathbf{P}(X \leq 0) = 0$. In addition, for every $\omega \in E_0^c$ we have $X(\omega) > 0$ and it is possible to find $n(\omega)$ such that

$$\frac{1}{n} < X(\omega),$$

for every $n > n(\omega)$. It then follows that

$$Y_n(\omega) = 0,$$

for every $n > n(\omega)$. This implies

$$\lim_{n \rightarrow \infty} Y_n(\omega) = 0,$$

for every $\omega \in E_0^c$, which is the almost sure convergence of the sequence $(Y_n)_{n \geq 1}$ to $Dir(0)$. In the end, to check the convergence in mean to $Dir(0)$ (recall that the convergence in mean at some random variable implies convergence in probability at the same random variable), we consider

$$\mathbf{E}[|Y_n - Dir(0)|] = \mathbf{E}[Y_n] = 0 \cdot \mathbf{P}(Y_n = 0) + n \cdot \mathbf{P}(Y_n = n) = n \cdot \frac{1}{n} = 1.$$

It follows that

$$\lim_{n \rightarrow \infty} \mathbf{E}[|Y_n - Dir(0)|] = 1 \neq 0.$$

Hence, $(Y_n)_{n \geq 1}$ does not converge in mean to $Dir(0)$. This also implies that $(Y_n)_{n \geq 1}$ does not converge in mean at all (recall that convergence in mean at some random variable implies convergence in probability at the same random variable). As a consequence, $(Y_n)_{n \geq 1}$ does not converge in mean square. \square

Problem 2 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let $(\mathbb{R}, \mathcal{B}(\mathbb{R}), \mu_L) \equiv \mathbb{R}$ be the real Borel-Lebesgue state space. Show that the function $f : \mathbb{R} \rightarrow \mathbb{R}$ given by

$$f(x) \stackrel{\text{def}}{=} \frac{\alpha - 1}{x^\alpha} 1_{[1, +\infty)}, \quad \forall x \in \mathbb{R},$$

where $\alpha > 1$, is a density. Thereafter, consider a real random variable X on Ω with density $f_X = f$ and the sequence $(Y_n)_{n \geq 1}$ of real random variables on Ω given by

$$Y_n \stackrel{\text{def}}{=} \frac{X}{n}, \quad \forall n \in \mathbb{N}.$$

1. Study the convergence in distribution, in probability, almost sure, and in p -th mean of the sequence $(Y_n)_{n \geq 1}$, on varying of $\alpha > 1$, in the indicated order.
2. Assume that $(X_n)_{n \geq 1}$ is a sequence of (totally) independent real random variables on Ω with density $f_{X_n} = f$, for every $n \in \mathbb{N}$, and consider the sequence $(Z_n)_{n \geq 1}$ of real random variables on Ω given by

$$Z_n \stackrel{\text{def}}{=} \frac{X_n}{n}, \quad \forall n \in \mathbb{N}.$$

Study the convergence in distribution, in probability, almost sure, and in p -th mean of the sequence $(Z_n)_{n \geq 1}$, on varying of $\alpha > 1$, in the indicated order.

Solution.

1. Since $\alpha > 1$, we have

$$f(x) \geq 0,$$

for every $x \in \mathbb{R}$, and, since $1 - \alpha < 0$, we have

$$\begin{aligned} \int_{\mathbb{R}} f(x) d\mu_L(x) &= \int_{\mathbb{R}} \frac{\alpha-1}{x^\alpha} 1_{[1,+\infty)}(x) d\mu_L(x) = \int_{[1,+\infty)} \frac{\alpha-1}{x^\alpha} d\mu_L(x) \\ &= \int_1^{+\infty} \frac{\alpha-1}{x^\alpha} dx = \lim_{x \rightarrow +\infty} \int_1^x \frac{\alpha-1}{u^\alpha} du = - \lim_{x \rightarrow +\infty} \int_1^x du^{1-\alpha} \\ &= - \lim_{x \rightarrow +\infty} u^{1-\alpha} \Big|_1^x = 1 - \lim_{x \rightarrow +\infty} x^{1-\alpha} = 1. \end{aligned}$$

This shows that $f : \mathbb{R} \rightarrow \mathbb{R}$ is a density.

Write $F_{Y_n} : \mathbb{R} \rightarrow \mathbb{R}$ for the distribution function of Y_n , for every $n \geq 1$. We have

$$F_{Y_n}(y) = \mathbf{P}(Y_n \leq y) = \mathbf{P}(X/n \leq y) = \mathbf{P}(X \leq ny) = \int_{(-\infty, ny]} f(x) d\mu_L(x).$$

for every $y \in \mathbb{R}$. On the other hand,

$$\begin{aligned} \int_{(-\infty, ny]} f(x) d\mu_L(x) &= \int_{(-\infty, ny]} \frac{\alpha-1}{x^\alpha} 1_{[1,+\infty)}(x) d\mu_L(x) \\ &= \int_{(-\infty, ny] \cap [1,+\infty)} \frac{\alpha-1}{x^\alpha} d\mu_L(x) \\ &= \begin{cases} \int_{\emptyset} \frac{\alpha-1}{x^\alpha} d\mu_L(x), & \text{if } ny < 1, \\ \int_{\{ny\}} \frac{\alpha-1}{x^\alpha} d\mu_L(x), & \text{if } ny = 1, \\ \int_{[1,ny]} \frac{\alpha-1}{x^\alpha} d\mu_L(x), & \text{if } 1 < ny, \end{cases} \end{aligned}$$

where

$$\int_{\emptyset} \frac{\alpha-1}{x^\alpha} d\mu_L(x) = \int_{\{ny\}} \frac{\alpha-1}{x^\alpha} d\mu_L(x) = 0$$

and

$$\int_{[1,ny]} \frac{\alpha-1}{x^\alpha} d\mu_L(x) = \int_1^{ny} \frac{\alpha-1}{x^\alpha} dx = - \int_1^{ny} dx^{1-\alpha} = - x^{1-\alpha} \Big|_1^{ny} = 1 - \frac{1}{n^{\alpha-1}y^{\alpha-1}}.$$

Therefore,

$$F_{Y_n}(y) = \begin{cases} 0, & \text{if } y \leq \frac{1}{n}, \\ 1 - \frac{1}{n^{\alpha-1}y^{\alpha-1}}, & \text{if } \frac{1}{n} < y, \end{cases} = \left(1 - \frac{1}{n^{\alpha-1}y^{\alpha-1}}\right) 1_{(1/n, +\infty)}(y),$$

for every $y \in \mathbb{R}$. As a consequence,

$$\lim_{n \rightarrow \infty} F_{Y_n}(y) = \begin{cases} 0, & \text{if } y \leq 0, \\ 1 - \lim_{n \rightarrow \infty} \frac{1}{n^{\alpha-1}y^{\alpha-1}} = 1, & \text{if } 0 < y. \end{cases}$$

Considering the Heavside function $H : \mathbb{R} \rightarrow \mathbb{R}$ given by

$$H(y) = \begin{cases} 0, & \text{if } y < 0, \\ 1, & \text{if } 0 \leq y, \end{cases}$$

we then have

$$\lim_{n \rightarrow \infty} F_{Y_n}(y) = H(y),$$

at any point $y \in \mathbb{R} - \{0\}$, where the Heavside function is continuous. Hence, the sequence $(Y_n)_{n \geq 1}$ converges in distribution to the standard Dirac real random variable $Dir(0)$ on Ω . Regarding to the convergence in probability, we know that the convergence in distribution to a Dirac random variables $Dir(y_0)$, concentrated at some $y_0 \in \mathbb{R}$, implies also the convergence in probability to $Dir(y_0)$. Thus, we can state that the sequence $(Y_n)_{n \geq 1}$ converges in distribution to $Dir(0)$. However, by a direct approach, observe that

$$F'_{Y_n}(y) = \frac{1-\alpha}{n^{\alpha-1}y^\alpha} 1_{(1/n, +\infty)}(y),$$

for every $y \in \mathbb{R} - \{1/n\}$, and

$$\int_{(-\infty, y]} \frac{1-\alpha}{n^{\alpha-1}u^\alpha} 1_{(1/n, +\infty)}(u) d\mu_L(u) = \int_{(-\infty, y] \cap (1/n, +\infty)} \frac{1-\alpha}{n^{\alpha-1}u^\alpha} d\mu_L(u) = \begin{cases} \int_{\emptyset} \frac{1-\alpha}{n^{\alpha-1}u^\alpha} d\mu_L(u), & \text{if } y \leq 1/n \\ \int_{[1/n, y]} \frac{1-\alpha}{n^{\alpha-1}u^\alpha} d\mu_L(u), & \text{if } y > 1/n \end{cases}$$

where

$$\int_{\emptyset} \frac{1-\alpha}{n^{\alpha-1}u^\alpha} d\mu_L(u) = 0$$

and

$$\begin{aligned} \int_{[1/n, y]} \frac{1-\alpha}{n^{\alpha-1}u^\alpha} d\mu_L(u) &= \frac{1}{n^{\alpha-1}} \int_{1/n}^y \frac{\alpha-1}{u^\alpha} du = -\frac{1}{n^{\alpha-1}} \int_{1/n}^y du^{1-\alpha} = -\frac{1}{n^{\alpha-1}} u^{1-\alpha} \Big|_{1/n}^y \\ &= \frac{1}{n^{\alpha-1}} \left(\frac{1}{n^{1-\alpha}} - \frac{1}{y^{\alpha-1}} \right) = 1 - \frac{1}{n^{\alpha-1}y^{\alpha-1}}. \end{aligned}$$

Therefore, we can write

$$\int_{(-\infty, y]} \frac{1-\alpha}{n^{\alpha-1}u^\alpha} 1_{(1/n, +\infty)}(u) d\mu_L(u) = \left(1 - \frac{1}{n^{\alpha-1}y^{\alpha-1}} \right) 1_{(1/n, +\infty)}(y).$$

This shows that Y_n is absolutely continuous with density $f_{Y_n} : \mathbb{R} \rightarrow \mathbb{R}$, given by

$$f_{Y_n}(y) \stackrel{\text{def}}{=} \frac{1-\alpha}{n^{\alpha-1}y^\alpha} 1_{(1/n, +\infty)}(y), \quad \forall y \in \mathbb{R}.$$

As a consequence, provided n is sufficiently large,

$$\begin{aligned}
\mathbf{P}(Y_n \geq \varepsilon) &= \int_{[\varepsilon, +\infty)} f_{Y_n}(y) d\mu_L(y) \\
&= \int_{[\varepsilon, +\infty)} \frac{1-\alpha}{n^{\alpha-1}y^\alpha} 1_{(1/n, +\infty)}(y) d\mu_L(y) \\
&= \int_{[\varepsilon, +\infty) \cap (1/n, +\infty)} \frac{1-\alpha}{n^{\alpha-1}y^\alpha} d\mu_L(y) \\
&= \int_{[\varepsilon, +\infty)} \frac{1-\alpha}{n^{\alpha-1}y^\alpha} d\mu_L(y) \\
&= \int_{\varepsilon}^{+\infty} \frac{1-\alpha}{n^{\alpha-1}y^\alpha} dy \\
&= \frac{1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} \int_{\varepsilon}^y \frac{\alpha-1}{u^\alpha} du \\
&= -\frac{1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} \int_{\varepsilon}^y du u^{\alpha-1} \\
&= -\frac{1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} u^{1-\alpha} \Big|_{\varepsilon}^y \\
&= -\frac{1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} \left(\frac{1}{y^{\alpha-1}} - \frac{1}{\varepsilon^{\alpha-1}} \right) \\
&= \frac{1}{\varepsilon^{\alpha-1} n^{\alpha-1}}.
\end{aligned}$$

It follows

$$\lim_{n \rightarrow \infty} \mathbf{P}(|Y_n - Dir(0)| \geq \varepsilon) = \lim_{n \rightarrow \infty} \mathbf{P}(Y_n \geq \varepsilon) = \lim_{n \rightarrow \infty} \frac{1}{\varepsilon^{\alpha-1} n^{\alpha-1}} = 0,$$

for every $\varepsilon > 0$. This proves directly that $(Y_n)_{n \geq 1}$ converges in probability to $Dir(0)$. In addition, we have

$$\lim_{n \rightarrow \infty} Y_n(\omega) = \lim_{n \rightarrow \infty} \frac{X(\omega)}{n} = 0,$$

for every $\omega \in \Omega$, which proves that $(Y_n)_{n \geq 1}$ converges almost surely to $Dir(0)$.

By virtue of what shown above, to study the convergence in p -th mean of the sequence $(Y_n)_{n \geq 1}$, we have to study the convergence in p -th mean of the sequence $(Y_n)_{n \geq 1}$ to $Dir(0)$. Hence, it is sufficient to show whether

$$\lim_{n \rightarrow \infty} \mathbf{E}[Y_n^p] = 0.$$

We have

$$\begin{aligned}
\mathbf{E}[Y_n^p] &= \int_{\mathbb{R}} y^p f_{Y_n}(y) d\mu_L(u) = \int_{\mathbb{R}} \frac{1-\alpha}{n^{\alpha-1}y^{\alpha-p}} 1_{(1/n, +\infty)}(y) d\mu_L(u) = \frac{1-\alpha}{n^{\alpha-1}} \int_{(1/n, +\infty)} \frac{1}{y^{\alpha-p}} d\mu_L(u) \\
&= \frac{1-\alpha}{n^{\alpha-1}} \int_{1/n}^{+\infty} \frac{1}{y^{\alpha-p}} dy = \frac{\alpha-1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} \int_{1/n}^y \frac{1}{u^{\alpha-p}} du \\
&= \begin{cases} \frac{\alpha-1}{p-\alpha+1} \frac{1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} \int_{1/n}^y du u^{p-\alpha+1} = \frac{\alpha-1}{p-\alpha+1} \frac{1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} u^{p-\alpha+1} \Big|_{1/n}^y, & \text{if } p \neq \alpha-1, \\ \frac{\alpha-1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} \int_{1/n}^y d \ln(u) = \frac{\alpha-1}{n^{\alpha-1}} \lim_{y \rightarrow +\infty} \ln(u) \Big|_{1/n}^y, & \text{if } p = \alpha-1. \end{cases}
\end{aligned}$$

Alternatively,

$$\begin{aligned}
\mathbf{E}[Y_n^p] &= \mathbf{E}\left[\left(\frac{X}{n}\right)^p\right] = \int_{\mathbb{R}} \frac{x^p}{n^p} f_X(x) d\mu_L(x) = \int_{\mathbb{R}} \frac{x^p}{n^p} \frac{\alpha-1}{x^\alpha} 1_{[1,+\infty)}(x) d\mu_L(x) \\
&= \frac{\alpha-1}{n^p} \int_{[1,+\infty)} \frac{1}{x^{\alpha-p}} d\mu_L(x) = \frac{\alpha-1}{n^p} \int_1^{+\infty} \frac{1}{x^{\alpha-p}} dx = \frac{\alpha-1}{n^p} \lim_{x \rightarrow +\infty} \int_1^x \frac{1}{u^{\alpha-p}} du \\
&= \begin{cases} \frac{\alpha-1}{p-\alpha+1} \frac{1}{n^p} \lim_{x \rightarrow +\infty} \int_1^x du^{p-\alpha+1} = \frac{\alpha-1}{p-\alpha+1} \frac{1}{n^p} \lim_{x \rightarrow +\infty} u^{p-\alpha+1} \Big|_1^x, & \text{if } p \neq \alpha-1, \\ \frac{\alpha-1}{n^p} \lim_{x \rightarrow +\infty} \int_1^x d \ln(u) = \frac{\alpha-1}{n^p} \lim_{x \rightarrow +\infty} \ln(u) \Big|_1^x, & \text{if } p = \alpha-1. \end{cases}
\end{aligned}$$

Now, if $p \geq \alpha-1$ we have that $\mathbf{E}[Y_n^p]$ is not finite. Therefore, the sequence $(Y_n)_{n \geq 1}$ cannot converge in p -th mean. If $1 \leq p < \alpha-1$, we have

$$\mathbf{E}[Y_n^p] = -\frac{\alpha-1}{p-\alpha+1} \frac{1}{n^{\alpha-1}} \frac{1}{n^{p-\alpha+1}} = -\frac{\alpha-1}{p-\alpha+1} \frac{1}{n^p}.$$

Hence,

$$\lim_{n \rightarrow \infty} \mathbf{E}[Y_n^p] = -\lim_{n \rightarrow \infty} \frac{\alpha-1}{p-\alpha+1} \frac{1}{n^p} = 0.$$

In this case, the sequence $(Y_n)_{n \geq 1}$ converges in p -th mean to $Dir(0)$.

- Clearly, the random variable Z_n has the same distribution function of the random variable Y_n for every $n \geq 1$. It follows that

$$Z_n \xrightarrow{\mathbf{w}} Dir(0) \quad \text{and} \quad Z_n \xrightarrow{\mathbf{P}} Dir(0).$$

Regarding to the almost sure convergence, observe that, with the same computations as above, we obtain

$$\mathbf{P}(|Z_n| \geq \varepsilon) = \frac{1}{\varepsilon^{\alpha-1} n^{\alpha-1}}.$$

Therefore,

$$\sum_{n=1}^{\infty} \mathbf{P}(|Z_n| \geq \varepsilon) = \frac{1}{\varepsilon^{\alpha-1}} \sum_{n=1}^{\infty} \frac{1}{n^{\alpha-1}}.$$

This converges for $\alpha > 2$, which implies the almost sure convergence of the sequence $(Z_n)_{n \geq 1}$ to $Dir(0)$, for $\alpha > 2$. In case $1 < \alpha \leq 2$, we have

$$\mathbf{P}\left(\bigcap_{n \geq m} \{|Z_n| < \varepsilon\}\right) \leq \mathbf{P}\left(\bigcap_{n \geq m}^s \{|Z_n| < \varepsilon\}\right)$$

for every $s \geq m$. On the other hand, thanks to the independence of the random variables in $(X_n)_{n \geq 1}$, we can write

$$\mathbf{P}\left(\bigcap_{n \geq m}^s \{|Z_n| < \varepsilon\}\right) = \prod_{n=m}^s \mathbf{P}(|Z_n| < \varepsilon) = \prod_{n=m}^s \left(1 - \frac{1}{\varepsilon^{\alpha-1} n^{\alpha-1}}\right).$$

Now, the function $g : \mathbb{R}_+ \rightarrow \mathbb{R}$, given by

$$g(x) \stackrel{\text{def}}{=} x^{\alpha-1}, \quad \forall x \in \mathbb{R}_+$$

is increasing. Therefore

$$\varepsilon^{\alpha-1} n^{\alpha-1} \leq \varepsilon^{\alpha-1} s^{\alpha-1}$$

for every $n \leq s$. In addition, we have

$$\varepsilon^{\alpha-1} s^{\alpha-1} \leq \varepsilon^{\alpha-1} s$$

for every $s \in \mathbb{N}$ and every α such that $1 < \alpha \leq 2$. We can then write

$$1 - \frac{1}{\varepsilon^{\alpha-1} n^{\alpha-1}} \leq 1 - \frac{1}{\varepsilon^{\alpha-1} s^{\alpha-1}} \leq 1 - \frac{1}{\varepsilon^{\alpha-1} s},$$

for every $n \leq s$. As a consequence, in case $1 < \alpha \leq 2$, we can write

$$\prod_{n=m}^s \left(1 - \frac{1}{\varepsilon^{\alpha-1} n^{\alpha-1}}\right) \leq \prod_{n=m}^s \left(1 - \frac{1}{\varepsilon^{\alpha-1} s}\right).$$

In particular, setting $s = 2m$, we obtain

$$\mathbf{P} \left(\bigcap_{n \geq m} \{|Z_n| < \varepsilon\} \right) \leq \mathbf{P} \left(\bigcap_{n \geq m}^m \{|Z_n| < \varepsilon\} \right) \leq \prod_{n=m}^{2m} \left(1 - \frac{1}{2\varepsilon^{\alpha-1} m}\right) = \left(1 - \frac{1}{2\varepsilon^{\alpha-1} m}\right)^m.$$

In the end, since

$$\lim_{m \rightarrow \infty} \left(1 - \frac{1}{2\varepsilon^{\alpha-1} m}\right)^m = \frac{1}{e^{\frac{1}{2}\varepsilon^{1-\alpha}}} < 1,$$

for every $\varepsilon > 0$, we cannot have the almost sure convergence of the sequence $(Z_n)_{n \geq 1}$ to $Dir(0)$. Alternatively, we obtain the same result by observing that, since the random variables of the sequence $(Z_n)_{n \geq 1}$ are independent, we have

$$Z_n \xrightarrow{\text{a.s.}} Dir(0) \Leftrightarrow \sum_{n=1}^{\infty} \mathbf{P}(|Y_n| \geq \varepsilon) < \infty, \quad \forall \varepsilon > 0$$

(see Rohatgi, 1976, p. 265) and the series

$$\sum_{n=1}^{\infty} \mathbf{P}(|Z_n| \geq \varepsilon) = \frac{1}{\varepsilon^{\alpha-1}} \sum_{n=1}^{\infty} \frac{1}{n^{\alpha-1}}$$

does not converge in case $1 < \alpha \leq 2$.

Problem 3 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let $(X_n)_{n \geq 1}$ be a sequence of real random variables on Ω . Assume that $(X_n)_{n \geq 1}$ are identically distributed and let $f_X : \mathbb{R} \rightarrow \mathbb{R}$ their common density function given by

$$f_X(x) \stackrel{\text{def}}{=} \frac{2}{x^3} 1_{[1, +\infty)}(x), \quad \forall x \in \mathbb{R}.$$

Consider the random variable X with density $f_X : \mathbb{R} \rightarrow \mathbb{R}$ and set

$$Y_n \stackrel{\text{def}}{=} \frac{X}{n^\alpha}, \quad \forall n \geq 1,$$

where $\alpha > 0$.

1. Study the convergence in distribution, probability, almost surely, and L^p of the sequence $(Y_n)_{n \geq 1}$, on varying of $\alpha > 0$, in the indicated order.

2. Under the additional assumption of independence of the random variables of the sequence $(X_n)_{n \geq 1}$, consider the sequence

$$Y_n \equiv \frac{X_n}{n^\alpha}, \quad \forall n \geq 1,$$

Does the sequence $(Y_n)_{n \geq 1}$ converge almost surely?

Solution. .

Problem 4 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a complete probability space and let $(X_n)_{n \geq 1}$ be a sequence of independent real random variables such that $X_n \sim \text{Ber}(1/n^\alpha)$ for some $\alpha > 0$. Consider the sequence $(Y_n)_{n \geq 1}$ of real random variables on Ω given by

$$Y_n \stackrel{\text{def}}{=} \min \{X_1, \dots, X_n\}.$$

1. study the convergence in distribution, in probability and in $L^p(\Omega; \mathbb{R})$ of $(X_n)_{n \geq 1}$ and $(Y_n)_{n \geq 1}$ on varying of $\alpha > 0$;
2. study the almost sure convergence of $(X_n)_{n \geq 1}$ and $(Y_n)_{n \geq 1}$ on varying of $\alpha > 0$.

Solution. .

Problem 5 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let $(X_n)_{n \geq 1}$ be a sequence of real random variables on Ω . Prove that

$$X_n \xrightarrow{\mathbf{P}} 0 \Leftrightarrow \lim_{n \rightarrow \infty} \mathbf{E} \left[\frac{|X_n|}{1 + |X_n|} \right] = 0.$$

Solution. .

Problem 6 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let $(\mathbb{R}, \mathcal{B}(\mathbb{R})) \equiv \mathbb{R}$ be the Euclidean real line endowed with the Borel σ -algebra. Assume that Ω is countable and $\mathcal{E} = \mathcal{P}(\Omega)$. Let $(X_n)_{n \geq 1}$ be a sequence of real random variables on Ω . Show that if there exists a real random variable X on Ω such that $X_n \xrightarrow{\mathbf{P}} X$, then $X_n \xrightarrow{\text{a.s.}} X$.

Solution. .

Problem 7 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let $(\mathbb{R}, \mathcal{B}(\mathbb{R})) \equiv \mathbb{R}$ be the Euclidean real line endowed with the Borel σ -algebra. Let $(X_n)_{n \geq 1}$ be a sequence of independent standard Bernoulli random variables on Ω each X_n of which has success probability p_n . Prove that

Exercise 8 1. $X_n \xrightarrow{\mathbf{P}} 0$ if and only if $\lim_{n \rightarrow \infty} p_n = 0$.

2. $X_n \xrightarrow{\text{a.s.}} 0$ if and only if $\sum_{n=1}^{\infty} p_n < \infty$.

Solution. .

Exercise 9 (Sheldon M. Ross - 4.2 - 4.3) Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let $(X_n)_{n \geq 1}$ be a sequence of independent Bernoulli random variables. Recall that for a Bernoulli random variable X we have

$$X = \begin{cases} 1 & \mathbf{P}(X = 1) = p \\ 0 & \mathbf{P}(X = 0) = q \end{cases},$$

where $p \in (0, 1)$ and $q \equiv 1 - p$. Consider the sequence $(Z_n)_{n \geq 1}$ of random variables given by

$$Z_n \stackrel{\text{def}}{=} \sum_{k=1}^n X_k, \quad \forall n \geq 1$$

and let $(H_n)_{n \geq 1}$ be the sequence of random variables given by

$$H_n \stackrel{\text{def}}{=} 2Z_n - n, \quad \forall n \geq 1.$$

1. Assume that X_n is the random variable which represents the toss of a coin with the convention that “success” [resp. “failure”] is for the outcome “head” [resp. “tail”] represented, in turn, by the outcome 1 [resp. 0]. Give an interpretation of the random variables Z_n and H_n and compute their mean and variance.
2. Assume that $p = 1/2$. Prove that we have

$$\lim_{n \rightarrow \infty} \mathbf{P} \left(\frac{H_n}{\sqrt{n}} < z \right) = \Phi(z),$$

for every $z \in \mathbb{R}$, where Φ is the standard notation for the distribution function of the standard normal random variable.

Solution.