## II Università di Roma, Tor Vergata Dipartimento d'Ingegneria Civile e Ingegneria Informatica

## LM in Ingegneria dell'Informazione e dell'Automazione Complementi di Probabilità e Statistica - Advanced Statistics

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**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

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

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

**Exercise 2** Hint: to deal with the almost sure convergence consider the event  $E_0 \equiv \{\omega \in \Omega : X(\omega) = 0\}$  and the complement  $E_0^c$ .

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

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

On the other hand, for every  $y \ge 0$  there exists  $n_y \in \mathbb{N}$ , (e.g.  $n_y = \lceil y \rceil$ , where  $\lceil \cdot \rceil : \mathbb{R} \to \mathbb{R}$ , is the ceiling function), such that y < n for every  $n > n_y$ . Therefore, definitively,

$$P(Y_n \le y) = 1 - 1/n.$$

It then follows

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

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

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

we clearly have

$$\lim_{n\to\infty}F_{Y_{n}}\left( y\right) =H\left( y\right) ,$$

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, according to the definition, we have

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

Therefore, definitively,

$$\mathbf{P}(|Y_n| \le \varepsilon) \ge \mathbf{P}(Y_n = 0) = 1 - \frac{1}{n},$$

for every  $\varepsilon > 0$ . It follows

$$\lim_{n \to \infty} \mathbf{P}(|Y_n| \le \varepsilon) \ge \lim_{n \to \infty} \left(1 - \frac{1}{n}\right) = 1,$$

for every  $\varepsilon > 0$ , which is the convergence in probability to Dir(0). Now, to check the convergence in mean, we consider

$$\mathbf{E}[Y_n] = n\mathbf{P}(Y_n = n) + 0\mathbf{P}(Y_n = 0) = 1.$$

It follows that

$$\lim_{n\to\infty} \mathbf{E}\left[|Y_n - 0|\right] = \lim_{n\to\infty} \mathbf{E}\left[Y_n\right] = 1 \neq 0.$$

Hence,  $(Y_n)_{n\geq 1}$  does not converge in mean. In the end, consider the event

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

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

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

for every  $n > n_{\omega}$ . It then follows that

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

for every  $n > n_{\omega}$ . This implies

$$\lim_{n\to\infty} Y_n\left(\omega\right) = 0,$$

for every  $\omega \in E_0^c$ , which is the almost sure convergence of the sequence  $(Y_n)_{n\geq 1}$  to Dir(0).

**Problem 3** 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  $\sigma$ -algebra. Prove that the function  $f : \mathbb{R} \to \mathbb{R}_+$  given by

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

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

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

**Exercise 4** Study the convergence in distribution, in probability and in p-th mean of the sequence  $(Y_n)_{n\geq 1}$  on varying of  $\alpha>1$ .

**Solution.** Since  $\alpha > 1$ , that is  $\alpha - 1 > 0$  and  $1 - \alpha < 0$ , we have

$$\int_{\mathbb{R}} f(x) d\mu_{L}(x) = \int_{\mathbb{R}} \frac{\alpha - 1}{x^{\alpha}} 1_{[1, +\infty)}(x) d\mu_{L}(x) = \int_{[1, +\infty)} \frac{\alpha - 1}{x^{\alpha}} d\mu_{L}(x) 
= \int_{1}^{+\infty} \frac{\alpha - 1}{x^{\alpha}} dx = \lim_{x \to +\infty} \int_{1}^{x} \frac{\alpha - 1}{u^{\alpha}} du = \lim_{x \to +\infty} -\int_{1}^{x} du^{1-\alpha} 
= \lim_{x \to +\infty} -u^{1-\alpha} \Big|_{1}^{x} = 1 - \lim_{x \to +\infty} x^{1-\alpha} = 1.$$

This proves that  $f: \mathbb{R} \to \mathbb{R}_+$  is a density.

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

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

On the other hand,

$$\int_{(-\infty,ny]} f(x) d\mu_L(x) = \int_{(-\infty,ny]} \frac{\alpha - 1}{x^{\alpha}} 1_{[1,+\infty)}(x) d\mu_L(x)$$

$$= \int_{(-\infty,ny] \cap [1,+\infty)} \frac{\alpha - 1}{x^{\alpha}} d\mu_L(x)$$

$$= \begin{cases} \int_{\varnothing} \frac{\alpha - 1}{x^{\alpha}} d\mu_L(x), & \text{if } ny < 1, \\ \int_{\{ny\}} \frac{\alpha - 1}{x^{\alpha}} d\mu_L(x), & \text{if } ny = 1, \\ \int_{[1,ny]} \frac{\alpha - 1}{x^{\alpha}} d\mu_L(x), & \text{if } 1 < ny, \end{cases}$$

where

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

and

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

Therefore,

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

As a consequence,

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

Considering the Heaviside function  $H: \mathbb{R} \to \mathbb{R}_+$  given by

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

we clearly have

$$\lim_{n\to\infty} F_{Y_n}\left(y\right) = H\left(y\right),\,$$

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). 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, since

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

for every  $y \in \mathbb{R}$ , we have that the random variables of the sequence  $(Y_n)_{n \geq 1}$  are absolutely continuous with density  $f_{Y_n} : \mathbb{R} \to \mathbb{R}_+$  given by

$$f_{Y_n}\left(y\right) = \frac{1-\alpha}{n^{\alpha-1}v^{\alpha}} 1_{\left(1/n,+\infty\right)}\left(y\right).$$

Note that

$$F_{Y_{n}}'\left(y\right)=f_{Y_{n}}\left(y\right),$$

for every  $y \neq 1/n$ . As a consequence, provided n is sufficiently large,

$$\mathbf{P}(|Y_n| > \varepsilon) = \mathbf{P}(Y_n > \varepsilon) = \int_{(\varepsilon, +\infty)} f_{Y_n}(y) \, d\mu_L(y)$$

$$= \int_{(\varepsilon, +\infty)} \frac{1 - \alpha}{n^{\alpha - 1} y^{\alpha}} 1_{(1/n, +\infty)}(y) \, d\mu_L(y)$$

$$= \int_{(\varepsilon, +\infty) \cap (1/n, +\infty)} \frac{1 - \alpha}{n^{\alpha - 1} y^{\alpha}} d\mu_L(y)$$

$$= \int_{\varepsilon} \frac{1 - \alpha}{n^{\alpha - 1} y^{\alpha}} d\mu_L(y)$$

$$= \int_{\varepsilon}^{+\infty} \frac{1 - \alpha}{n^{\alpha - 1} y^{\alpha}} dy$$

$$= \frac{1}{n^{\alpha - 1}} \lim_{y \to +\infty} \int_{\varepsilon}^{y} \frac{\alpha - 1}{u^{\alpha}} du$$

$$= -\frac{1}{n^{\alpha - 1}} \lim_{y \to +\infty} \int_{\varepsilon}^{y} du^{\alpha - 1}$$

$$= -\frac{1}{n^{\alpha - 1}} \lim_{y \to +\infty} u^{1 - \alpha} \Big|_{\varepsilon}^{y}$$

$$= -\frac{1}{n^{\alpha - 1}} \lim_{y \to +\infty} \left(\frac{1}{y^{\alpha - 1}} - \frac{1}{\varepsilon^{\alpha - 1}}\right)$$

$$= \frac{1}{n^{\alpha - 1} \varepsilon^{\alpha - 1}}.$$

It follows

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

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

By virtue of what shown above, to study the convergence in p-th mean of the sequence  $(Y_n)_{n\geq 1}$  it is sufficient to consider

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

Alternatively,

$$\mathbf{E}\left[Y_{n}^{p}\right] = \mathbf{E}\left[\left(\frac{X}{n}\right)^{p}\right] = \int_{\mathbb{R}} \frac{x^{p}}{n^{p}} f_{X}\left(x\right) d\mu_{L}\left(x\right) = \int_{\mathbb{R}} \frac{x^{p}}{n^{p}} \frac{\alpha - 1}{x^{\alpha}} \mathbf{1}_{[1, +\infty)}\left(x\right) d\mu_{L}\left(x\right)$$

$$= \frac{\alpha - 1}{n^{p}} \int_{[1, +\infty)} \frac{1}{x^{\alpha - p}} d\mu_{L}\left(x\right) = \frac{\alpha - 1}{n^{p}} \int_{1}^{+\infty} \frac{1}{x^{\alpha - p}} dx = \frac{\alpha - 1}{n^{p}} \lim_{x \to +\infty} \int_{1}^{x} \frac{1}{u^{\alpha - p}} du$$

$$= \begin{cases} \frac{\alpha - 1}{p - \alpha + 1} \frac{1}{n^{p}} \lim_{x \to +\infty} \int_{1}^{x} d\ln\left(u\right) = \frac{\alpha - 1}{n^{p}} \lim_{x \to +\infty} \ln\left(u\right)\Big|_{1}^{x}, & \text{if } p \neq \alpha - 1, \\ \frac{\alpha - 1}{n^{p}} \lim_{x \to +\infty} \int_{1}^{x} d\ln\left(u\right) = \frac{\alpha - 1}{n^{p}} \lim_{x \to +\infty} \ln\left(u\right)\Big|_{1}^{x}, & \text{if } p = \alpha - 1. \end{cases}$$

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

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

The sequence converges in p-th mean to the standard Dirac random variable.

**Problem 5** Let  $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$  be a probability space and let  $(X_n)_{\geq n}$  be a sequence of real random variables on  $\Omega$ . Assume that  $(X_n)_{\geq n}$  are identically distributed and let  $f_X : \mathbb{R} \to \mathbb{R}_+$  their common density function given by

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

Set

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

where  $\alpha > 0$ .

- 1. Study the convergence in distribution, probability, and  $L^p$  of the sequence  $(Y_n)_{n\geq 1}$  on varying of  $\alpha>0$ .
- 2. Under the additional assumption of independence of the random variables of the sequence  $(X_n)_{\geq n}$ , compute  $\limsup_{n\to\infty} Y_n$  and  $\liminf_{n\to\infty} Y_n$  on varying of  $\alpha>0$ . Does the sequence  $(Y_n)_{n\geq 1}$  converge almost surely?

Solution.  $\Box$ 

**Problem 6** 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  $X_n \sim Ber(1/n^{\alpha})$  for some  $\alpha > 0$ . Consider the sequence  $(Y_n)_{n\geq 1}$  of real random variables on  $\Omega$  given by

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

- 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.

1. We clearly have

$$Y_n(\omega) = \begin{cases} 1 & \Leftrightarrow X_1(\omega) = \cdots = X_n(\omega) = 1 \\ 0 & \text{otherwise} \end{cases}$$

Hence, by virtue of the independence of the random variables of the sequence  $(X_n)_{n>1}$ , we have

$$\mathbf{P}(Y_n = 1) = \mathbf{P}(X_1 = 1, ..., X_n = 1) = \mathbf{P}(X_1 = 1) \cdots \mathbf{P}(X_n = 1) = \prod_{k=1}^{n} \frac{1}{k^{\alpha}} = \frac{1}{n!^{\alpha}}$$

and

$$\mathbf{P}(Y_n = 0) = 1 - \mathbf{P}(Y_n = 1) = 1 - \frac{1}{n!^{\alpha}}.$$

In other words,  $(Y_n)_{n\geq 1}$  is a sequence of standard Bernoulli random variables with succes probability  $\frac{1}{n!^{\alpha}}$ . Considering the distribution functions  $F_{X_n}: \mathbb{R} \to \mathbb{R}_+$  and  $F_{Y_n}: \mathbb{R} \to \mathbb{R}_+$  of  $X_n$  and  $Y_n$ , respectively, we have

$$F_{X_n}(x) \stackrel{\text{def}}{=} \mathbf{P}(X_n \le x) = \begin{cases} 0, & \text{if } x < 0, \\ 1 - \frac{1}{n^{\alpha}}, & \text{if } 0 \le x < 1, \\ 1, & \text{if } 1 \le x, \end{cases}$$

and

$$F_{Y_n}(x) \stackrel{\text{def}}{=} \mathbf{P}(Y_n \le x) = \begin{cases} 0, & \text{if } x < 0, \\ 1 - \frac{1}{n!^{\alpha}}, & \text{if } 0 \le x < 1, \\ 1, & \text{if } 1 \le x. \end{cases}$$

Therefore, considering the Heaviside function  $H: \mathbb{R} \to \mathbb{R}_+$  given by

$$H(x) \stackrel{\text{def}}{=} \begin{cases} 0, & \text{if } x < 0, \\ 1, & \text{if } 0 \le x, \end{cases}$$

we have

$$\lim_{n\to\infty} F_{X_n}(x) = \lim_{n\to\infty} F_{Y_n}(x) = H(x),$$

for every  $x \in \mathbb{R}$ . Thus, both the sequences  $(F_{X_n})_{n\geq 1}$  and  $(F_{Y_n})_{n\geq 1}$  converge pointwise to H. It follows that both the sequences  $(X_n)_{n\geq 1}$  and  $(Y_n)_{n\geq 1}$  converge 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, according to the definition, we have definitively

$$\mathbf{P}\left(\left|X_{n}-Dir\left(0\right)\right|<\varepsilon\right)=\mathbf{P}\left(X_{n}<\varepsilon\right)=\mathbf{P}\left(X_{n}=0\right)=1-\frac{1}{n^{\alpha}}$$

and

$$\mathbf{P}\left(\left|Y_{n}-Dir\left(0\right)\right|<\varepsilon\right)=\mathbf{P}\left(Y_{n}<\varepsilon\right)=\mathbf{P}\left(Y_{n}=0\right)=1-\frac{1}{n!^{\alpha}},$$

for every  $\varepsilon > 0$ . Therefore,

$$\lim_{n \to \infty} \mathbf{P}\left(|X_n - Dir\left(0\right)| < \varepsilon\right) = \lim_{n \to \infty} \left(1 - \frac{1}{n^{\alpha}}\right) = 1$$

and

$$\lim_{n \to \infty} \mathbf{P}\left(|Y_n - Dir\left(0\right)| < \varepsilon\right) = \lim_{n \to \infty} \left(1 - \frac{1}{n!^{\alpha}}\right) = 1,$$

which is the convergence in probability of both  $(X_n)_{n\geq 1}$  and  $(X_n)_{n\geq 1}$  to Dir(0). To check the convergence in  $L^p(\Omega;\mathbb{R})$ , we need to consider  $\|X_n - Dir(0)\|_p$  and  $\|Y_n - Dir(0)\|_p$ , because in case of convergence the limit has to be Dir(0). We then have

$$||X_n - Dir(0)||_p = \left(\int_{\Omega} |X_n - Dir(0)|^p d\mathbf{P}\right)^{1/p} = \left(\int_{\Omega} X_n^p d\mathbf{P}\right)^{1/p} = \mathbf{P}(X_n = 1)^{1/p} = \frac{1}{n^{\frac{\alpha}{p}}}$$

and

$$\|Y_n - Dir(0)\|_p = \left(\int_{\Omega} |Y_n - Dir(0)|^p d\mathbf{P}\right)^{1/p} = \left(\int_{\Omega} Y_n^p d\mathbf{P}\right)^{1/p} = \mathbf{P}(Y_n = 1)^{1/p} = \frac{1}{n!^{\frac{\alpha}{p}}},$$

for every  $\alpha > 0$ . Hence, we obtain

$$\lim_{n \to \infty} ||X_n - X||_p = \lim_{n \to \infty} ||Y_n - X||_p = 0,$$

which proves the convergence of both  $(X_n)_{n\geq 1}$  and  $(X_n)_{n\geq 1}$  to Dir(0) in  $L^p(\Omega;\mathbb{R})$ .

2. With regard to the almost sure convergence, note that also in this case, if the sequences  $(X_n)_{n\geq 1}$  and  $(Y_n)_{n\geq 1}$  converge almost surely, the limit has to be Dir(0). Hence, for any fixed  $\varepsilon > 0$  consider the events  $E_0 \equiv \{|X_n - Dir(0)| \geq \varepsilon\}$  and  $F_0 \equiv \{|Y_n - Dir(0)| \geq \varepsilon\}$  we have

$$\left\{\left|X_{n}-Dir\left(0\right)\right|\geq\varepsilon\right\}=\left\{X_{n}\geq\varepsilon\right\}\quad\text{and}\quad\left\{\left|Y_{n}-Dir\left(0\right)\right|\geq\varepsilon\right\}=\left\{Y_{n}\geq\varepsilon\right\}.$$

Hence,

$$\mathbf{P}\left(\left|X_{n}-Dir\left(0\right)\right|\geq\varepsilon\right)=\left\{\begin{array}{ll}\mathbf{P}\left(X_{n}=1\right)=\frac{1}{n^{\alpha}}, & \text{if } 0<\varepsilon\leq1,\\ 0, & \text{if } \varepsilon>1,\end{array}\right.$$

and

$$\mathbf{P}(|Y_n - Dir(0)| \ge \varepsilon) = \begin{cases} \mathbf{P}(Y_n = 1) = \frac{1}{n!^{\alpha}}, & \text{if } 0 < \varepsilon \le 1, \\ 0, & \text{if } \varepsilon > 1. \end{cases}$$

As a consequence,

$$\sum_{n=1}^{\infty} \mathbf{P}\left(\left|X_{n} - Dir\left(0\right)\right| \ge \varepsilon\right) = \begin{cases} \sum_{n=1}^{\infty} \frac{1}{n^{\alpha}}, & \text{if } 0 < \varepsilon \le 1, \\ 0, & \text{if } \varepsilon > 1, \end{cases}$$

and

$$\sum_{n=1}^{\infty} \mathbf{P}\left(\left|Y_{n} - Dir\left(0\right)\right| \ge \varepsilon\right) = \begin{cases} \sum_{n=1}^{\infty} \frac{1}{n!^{\alpha}}, & \text{if } 0 < \varepsilon \le 1, \\ 0, & \text{if } \varepsilon > 1. \end{cases}$$

It then follows that  $\sum_{n=1}^{\infty} \mathbf{P}(|X_n - Dir(0)| \ge \varepsilon)$  converges for every  $\alpha > 1$  and  $\sum_{n=1}^{\infty} \mathbf{P}(|X_n - Dir(0)| \ge \varepsilon)$  converges for every  $\alpha > 0$ . This yields the almost sure convergence of  $(X_n)_{n\ge 1}$  to Dir(0) for every  $\alpha > 1$  and the almost sure convergence of  $(Y_n)_{n\ge 1}$  to Dir(0) for every  $\alpha > 0$ . In fact, the convergence of the series implies that

$$\lim_{m \to \infty} \mathbf{P} \left( \bigcup_{n \ge m} \{ |Z_n - Z| \ge \varepsilon \} \right) \le \sum_{n = m}^{\infty} \mathbf{P} \left( |Z_n - Z| \ge \varepsilon \right) = 0.$$