

## Probability and Measure

### Example sheet 3

Unless otherwise specified, let  $(E, \mathcal{E}, \mu)$  be a measure space and  $(\Omega, \mathcal{A}, \mathbb{P})$  be a probability space.

**3.1** Prove that the space  $L^\infty(E, \mathcal{E}, \mu)$  is complete.

**3.2** Let  $p \in [1, \infty]$  and let  $f_n, f \in L^p(E, \mathcal{E}, \mu)$  for  $n \in \mathbb{N}$ . Show that:

$$f_n \rightarrow f \text{ in } L^p \Rightarrow f_n \rightarrow f \text{ in measure, but the converse is not true.}$$

**3.3** Read hand-out 2 carefully. Find examples which show that the reverse implications, concerning the concepts of convergence on page 1, are in general false.

**3.4** Let  $X$  be a random variable in  $\mathbb{R}$  and let  $1 \leq p < q < \infty$ . Show that

$$\mathbb{E}(|X|^p) = \int_0^\infty p \lambda^{p-1} \mathbb{P}(|X| \geq \lambda) d\lambda$$

$$\text{and deduce: } X \in L^q(\mathbb{P}) \Rightarrow \mathbb{P}(|X| \geq \lambda) = O(\lambda^{-q}) \Rightarrow X \in L^p(\mathbb{P}).$$

Remark on questions 3.5 to 3.7, part (a): Start with an indicator function and extend your argument to the general case, analogous to the proof of Lemma 4.10(ii).

**3.5** A *stepfunction*  $g : \mathbb{R} \rightarrow \mathbb{R}$  is any finite linear combination of indicator functions of finite intervals.

(a) Show that the set of stepfunctions  $\mathcal{I}$  is dense in  $L^p(\mathbb{R})$  for all  $p \in [1, \infty)$ , i.e. for all  $f \in L^p(\mathbb{R})$  and every  $\epsilon > 0$  there exists  $g \in \mathcal{I}$  such that  $\|f - g\|_p < \epsilon$ .  
[Use the result of question 1.9.]

(b) Using (a), argue that the set of continuous functions  $C(\mathbb{R})$  is dense in  $L^p(\mathbb{R})$ ,  $p \in [1, \infty)$ .

**3.6** Let  $f \in L^1(E, \mathcal{E}, \mu)$  be non-negative and define  $\nu(A) := \int_E f \mathbb{1}_A d\mu$ .

(a) Show that  $\nu$  is a measure on  $(E, \mathcal{E})$  and that  $\int_E g d\nu = \int_E gf d\mu$ ,  
for all measurable functions  $g : E \rightarrow [0, \infty]$ .

(b) Let  $X$  be a random variable in  $\mathbb{R}$  with measurable *density function*  $f_X : \mathbb{R} \rightarrow [0, \infty]$ , i.e.

$$\mathbb{P}(X \in A) = \int_A f_X(x) dx \quad \text{for all } A \in \mathcal{B}.$$

Use (a) to show that  $\mathbb{E}(g(X)) = \int_{\mathbb{R}} g(x) f(x) dx$  for all measurable  $g : \mathbb{R} \rightarrow [0, \infty]$ .

**3.7** (a) Show that, if  $X$  and  $Y$  are independent random variables, then  $\|XY\|_1 = \|X\|_1 \|Y\|_1$ , but that the converse is in general not true.

(b) Show that, if  $X$  and  $Y$  are independent and integrable, then  $\mathbb{E}(XY) = \mathbb{E}(X)\mathbb{E}(Y)$ .

**3.8** Let  $(X_n)_{n \in \mathbb{N}}$  be iid random variables with moment generating function  $\phi(\theta) = \mathbb{E}(e^{\theta X_1})$  and expected value  $m$ . Assume that  $\phi$  is finite on some neighbourhood of the origin.

(a) Show that  $\mathbb{P}(X > x) \leq \Lambda(x)$  for all  $x > m$ , where  $\Lambda(x) = \inf \{e^{-\theta x} \phi(\theta) : \theta \in \mathbb{R}\}$ .

(b) Show that  $S_n = \sum_{i=1}^n X_i$  satisfies  $\mathbb{P}(S_n > nx) \leq \Lambda(x)^n$  for all  $x > m$ .

(c) Compute  $\Lambda(x)$  for  $x > 1$ , when the  $X_n$  have exponential distribution with parameter 1.

**3.9** For  $f, g \in L^2$  show Pythagoras' rule and the parallelogram law,

$$\|f + g\|_2^2 = \|f\|_2^2 + 2\langle f, g \rangle + \|g\|_2^2, \quad \|f + g\|_2^2 + \|f - g\|_2^2 = 2(\|f\|_2^2 + \|g\|_2^2).$$

**3.10** Show that every covariance matrix is non-negative definite.

**3.11** Let  $V_1 \subseteq V_2 \subseteq \dots$  be an increasing sequence of closed subspaces of  $L^2 = L^2(E, \mathcal{E}, \mu)$ .

For  $f \in L^2$ , denote by  $f_n$  the orthogonal projection of  $f$  on  $V_n$ . Show that  $f_n$  converges in  $L^2$ .

**3.12** Given a countable family of disjoint events  $(G_i)_{i \in I}$ ,  $G_i \in \mathcal{A}$ , with  $\bigcup_{i \in I} G_i = \Omega$ .

Set  $\mathcal{G} = \sigma(G_n : n \in \mathbb{N})$  and  $V = L^2(\Omega, \mathcal{G}, \mathbb{P})$ .

Show that, for  $X \in L^2(\Omega, \mathcal{A}, \mathbb{P})$ , the conditional expectation  $\mathbb{E}(X | \mathcal{G})$  is a version of the orthogonal projection of  $X$  on  $V$ .

**3.13** (a) Find a sequence of random variables  $(X_n)_{n \in \mathbb{N}}$  which is not bounded in  $L^1$ , but satisfies the other condition for uniform integrability, i.e.

$$\forall \epsilon > 0 \exists \delta > 0 \forall A \in \mathcal{A} \forall i \in I : \mathbb{P}(A) < \delta \Rightarrow \mathbb{E}(|X_i| \mathbb{1}_A) < \epsilon.$$

(b) Find a uniformly integrable sequence of random variables  $(X_n)_{n \in \mathbb{N}}$  such that

$$X_n \rightarrow 0 \text{ a.s. and } \mathbb{E}\left(\sup_n |X_n|\right) = \infty.$$

**3.14** Let  $(X_n)_{n \in \mathbb{N}}$  be a sequence of iid random variables in  $L^2(\mathbb{P})$ . Show that, as  $n \rightarrow \infty$ ,

(a) for all  $\epsilon > 0$ ,  $n \mathbb{P}(|X_1| > \epsilon \sqrt{n}) \rightarrow 0$ ,

(b)  $n^{-1/2} \max_{k \leq n} |X_k| \rightarrow 0$  in probability,

(c)  $n^{-1/2} \max_{k \leq n} |X_k| \rightarrow 0$  in  $L^1$ .