

## Workshop 3

January 22-28, 2024

For this week's submitted solutions, you are encouraged to try exercises at the level of 4, 6, 7 or 9 here.

### 1 Short and basic exercises about integration

**Exercise 1.** Let  $(X, \mathcal{F}, \mu)$  be a measure space and  $f$  an element of  $m(\mathcal{F})^+$ . Show that if  $\mu(f) = 0$  then  $\mu(\{f > 0\}) = 0$ .

**Exercise 2.** Consider  $f \in (m\mathcal{F})^+$  and  $(f_n)_{n \in \mathbb{N}} \subset (m\mathcal{F})^+$  such that  $f_n \uparrow f$ , except for a  $\mu$ -null set  $N$  (this means  $\mu(N) = 0$ ). Show that  $\mu(f_n) \uparrow \mu(f)$ .

**Exercise 3.** Use the standard machine to prove the linearity of the integral: for any integrable functions  $f, g$  and any  $\alpha, \beta \in \mathbb{R}$ ,

$$\mu(\alpha f + \beta g) = \alpha \mu(f) + \beta \mu(g).$$

### 2 Approximation by simpler functions and an implication of measurability

**Exercise 4.** Let  $(\Omega, \mathcal{F}, P)$  be a probability triple.

1. Let  $X$  be a real-valued, bounded measurable function. Show that for all  $\epsilon > 0$  there exists a finite partition  $(A_i)_{i=1}^N \subset \mathcal{F}$  of  $\Omega$  (which means that these sets are pairwise disjoint and their union is  $\Omega$ ) and real numbers  $(\lambda_i)_{i=1}^N$ , such that

$$\sup_{\omega \in \Omega} \left| X(\omega) - \sum_{i=1}^N \lambda_i I_{A_i}(\omega) \right| < \epsilon. \quad (1)$$

Here  $I_A$  denotes the indicator function of  $A$ :  $I_A(\omega) = 1$  if and only if  $\omega \in A$ .

2. Let  $Y : \Omega \rightarrow \mathbb{R}$  be a random variable. Use the previous result to show that if  $X$  is a bounded  $\sigma(Y)$ -measurable function, then  $X = f(Y)$  for some Borel function  $f : \mathbb{R} \rightarrow \mathbb{R}$ . (Hint: Apply the insights from the standard machine.)

### 3 Convergence of integrals

Read first the statements and proofs of Fatou's lemma and Lebesgue's dominated convergence theorem.

**Exercise 5** (cf. Williams E5.1). Consider the triple  $([0, 1], \mathcal{B}([0, 1]), \lambda)$ , where  $\lambda$  is the Lebesgue measure. Define  $f_n = nI_{[0, 1/n]}$ . Prove that  $f_n \rightarrow f$  almost surely (a.s.) on  $[0, 1]$ , but that  $\lambda(f_n) = 1$  for every  $n \in \mathbb{N}$ . Draw a picture of  $g = \sup_n |f_n|$  and show that  $g \notin \mathcal{L}^1([0, 1], \mathcal{B}([0, 1]), \lambda)$ .

**Exercise 6.** Given a probability triple  $(\Omega, \mathcal{F}, P)$  and a random variable  $X$  on it, we introduce the function<sup>1</sup>

$$\phi(t) := P(e^{itX}) = \int_{\Omega} e^{itx} dP(x). \quad (2)$$

Prove that  $\phi$  is continuous on  $\mathbb{R}$ . (Hint: You may need that, for functions between metric spaces, continuity is equivalent to sequential continuity.)

## 4 Relative entropy a.k.a. Kullback-Leibler divergence

Let  $(X, \mathcal{F})$  be a measurable space and  $\mu, \nu$  two  $(\sigma)$ -finite measures on  $(X, \mathcal{F})$ . If there exists an  $f \in \mathcal{L}^1(X, \mathcal{F}, \nu)$  such that  $\mu(A) = \int_A f d\nu$  for every  $A \in \mathcal{F}$ , we say that  $f$  is the Radon-Nikodym derivative of  $\mu$  with respect to  $\nu$ ; this is denoted  $f = \frac{d\mu}{d\nu}$ .

**Exercise 7.** Let  $(\Omega, \mathcal{F}, \mu)$  be a probability triple. For every probability measure  $\rho$  on  $(\Omega, \mathcal{F})$  absolutely continuous with respect to  $\mu$ , we introduce the quantity

$$D(\rho|\mu) = \int_{\Omega} \log \frac{d\rho}{d\mu} d\rho, \quad (3)$$

called *the relative entropy of  $\rho$  with respect to  $\mu$* . Here we interpret  $\log$  as an extended function from  $[0, \infty)$  to  $[-\infty, \infty)$ , such that  $\log 0 = -\infty$ . Remember that  $0 \leq \frac{d\rho}{d\mu} < \infty$  almost surely.

1. Show that

$$\int_{\Omega} \left| \log \frac{d\rho}{d\mu} \right| d\rho = \int_{\{\frac{d\rho}{d\mu} > 0\}} \left| \log \frac{d\rho}{d\mu} \right| d\rho \quad (4)$$

and that if any of these integrals converge, then

$$D(\rho|\mu) = \int_{\{\frac{d\rho}{d\mu} > 0\}} \log \frac{d\rho}{d\mu} d\rho. \quad (5)$$

2. (possibly more difficult) Show that if  $\mu \ll \nu \ll \xi$ , then

$$\frac{d\mu}{d\xi} = \frac{d\mu}{d\nu} \frac{d\nu}{d\xi}. \quad (6)$$

Two measures  $\mu$  and  $\rho$  are called *equivalent* (denoted  $\mu \sim \rho$ ) if  $\mu \ll \rho$  and  $\rho \ll \mu$ . Use (6) to prove that if  $\mu \sim \rho$ , then

$$\frac{d\rho}{d\mu} = \left( \frac{d\mu}{d\rho} \right)^{-1} \quad (7)$$

3. Show that if  $\rho \sim \mu$  and the integrals in (4) are finite,  $D(\rho|\mu) \geq 0$ . (Hint:  $\log$  is concave, so...; you may use (7); be explicit with the necessary conditions of inequalities.)

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<sup>1</sup>By definition, a function  $h : \Omega \rightarrow \mathbb{C}$  is integrable (with respect to a measure  $\mu$ ) if its real and imaginary parts are integrable (as real-valued functions); in this case,

$$\mu(h) = \mu(\operatorname{Re} h) + i\mu(\operatorname{Im} h).$$

## 5 Independence again

**Exercise 8.** (an easy one, but good warm-up) Let  $X$  and  $Y$  be independent Bernoulli random variables, both with parameter  $1/2$ .

1. Compute the probability mass functions of  $X + Y$  and  $|X - Y|$ .
2. Show that  $X + Y$  and  $|X - Y|$  are dependent.
3. Show that  $X + Y$  and  $|X - Y|$  are uncorrelated. (This means that  $\mathbb{E}(XY) = \mathbb{E}(X)\mathbb{E}(Y)$ .)

**Exercise 9.**

1. Let  $X$  be a real-valued random variable with law  $\Lambda_X$  and cumulative distribution function (c.d.f.)  $F_X$ , and  $Y$  a real-valued random variable with law  $\Lambda_Y$  and c.d.f.  $F_Y$ . We denote by  $\Lambda_{X,Y}$  the law of the joint variable  $(X, Y)$  and by  $F_{X,Y}$  the corresponding c.d.f.:

$$F_{X,Y} := \Lambda_{X,Y}((-\infty, x] \times (-\infty, y]). \quad (8)$$

Prove that the following statements are equivalent:

- (a)  $X$  and  $Y$  are independent.
  - (b)  $\Lambda_{X,Y} = \Lambda_X \otimes \Lambda_Y$ .
  - (c)  $F_{X,Y}(x, y) = F_X(x)F_Y(y)$ .
2. Show that if  $(X, Y)$  has a joint probability density function  $f_{X,Y}$  (with respect to the Lebesgue measure  $\lambda^{\otimes 2}$ ), then some of the statements above (therefore each of them) is equivalent to

- (a)  $f_{X,Y}(x, y) = f_X(x)f_Y(y)$  for  $\lambda^{\otimes 2}$ -almost every  $(x, y)$ .

Remember that  $f_X(x) = \int_{\mathbb{R}} f_{X,Y}(x, y) dy$ , etc.  
(Hint: Uniqueness lemma!)