

# HOMEWORK SET 1

## *Measure theory background*

Martingale Theory with Applications, 1<sup>st</sup> teaching block, 2019  
School of Mathematics, University of Bristol

Problems with •'s are to be handed in. These are due in the locker with "Martingale Theory" on it in G.90 Fry Building before 12:00pm on Monday, 14<sup>th</sup> October. Please show your work leading to the result, not only the result. Each problem worth the number of •'s you see right next to it. Hence, for example, Problem 1.6 worth three marks.

1.1 Let  $(\Omega, \mathcal{F})$  be a measurable space. Prove that if  $A, B \in \mathcal{F}$ , then

$$A \cap B, \quad A - B \text{ (set-difference),} \quad A \Delta B \text{ (symmetric set-difference)}$$

are also in  $\mathcal{F}$ .

1.2 Is the union of two  $\sigma$ -algebras (on the same set) also a  $\sigma$ -algebra? If yes, prove it, if no, give a counterexample.

1.3 Is the intersection of two  $\sigma$ -algebras (on the same set) also a  $\sigma$ -algebra? If yes, prove it, if no, give a counterexample.

1.4 Define the Borel  $\sigma$ -algebra on  $\mathbb{R}$  as we did in class:

$$\mathfrak{B}(\mathbb{R}) := \sigma \left\{ \bigcup_{i=1}^n (a_i, b_i] : n < \infty, a_1 < b_1 \leq a_2 < b_2 \leq \dots \leq a_n < b_n \text{ in } \mathbb{R} \right\}.$$

Show that each of

$$(a, b), \quad [a, b), \quad [a, b], \quad \{a\}, \quad (a, \infty)$$

are in  $\mathfrak{B}(\mathbb{R})$  for any  $a < b$  in  $\mathbb{R}$ .

1.5 (*Shiryaev.*) Let  $\Omega$  be a countable set and  $\mathcal{F}$  the collection of all its subsets. Put  $\mu(A) = 0$  if  $A$  is finite and  $\mu(A) = \infty$  if  $A$  is infinite. Show that the set function  $\mu$  is finitely additive but not  $\sigma$ -additive.

1.6 •• Consider the example from the exercise class:  $\Omega = \{1, 2, \dots, 12\}$ ,  $\mathcal{F} = \mathcal{P}(\Omega)$ ,  $\mathbb{P}$  is uniform on  $\Omega$ , and the random variables  $X$  and  $Y$  are defined by  $X(\omega) = \lceil \frac{\omega}{2} \rceil$ ,  $Y(\omega) = \lceil \frac{\omega}{4} \rceil$ ,  $\mathcal{G} = \sigma(Y)$ . Show in this example that  $\mathbb{E}(XY | \mathcal{G}) = Y\mathbb{E}(X | \mathcal{G})$ . This is referred to as 'taking out what's known' or 'given  $Y$ ,  $Y$  is not random'.

1.7 ••• (*Monty Hall problem with  $\sigma$ -algebras.*) The famous Monty Hall problem goes like this: We have three doors. Behind one of them is a car, behind the others, goats.

1. You pick a door, let us assume it's door number 1.
2. Monty opens *another door with a goat behind it*.
3. Now you pick one of the two closed doors (repeat your choice, or switch to the other one).
4. Whatever is behind this door is yours.

Make the natural assumptions about the probabilities of the location of the car and the choice of door Monty opens (if he *has* a choice). Would you repeat your choice or switch?

- (a) Write the full probability space of the experiment that involves the first two steps above.
- (b) In this sample space, write the event  $A = \{\text{door 3 has a goat}\}$ , and its generated  $\sigma$ -algebra  $\mathcal{F} = \sigma(A)$ .
- (c) Let  $X = 1, 2, 3$  be the location of the car. Calculate  $\mathbb{E}(X | \mathcal{F})$ .
- (d) Now write the event  $B = \{\text{Monty opens door 3}\}$ , and its generated  $\sigma$ -algebra  $\mathcal{G} = \sigma(B)$ .
- (e) Calculate  $\mathbb{E}(X | \mathcal{G})$ .
- (f) Conclude the optimal strategy for the player in this problem.

1.8 •• Let  $X$  and  $Y$  be random variables on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ , and  $\mathcal{G} = \sigma(Y)$ . Show that  $X$  is independent of  $\mathcal{G}$  if and only if for any bounded and measurable functions  $f$  and  $g$ , we have  $\mathbb{E}(f(X) \cdot g(Y)) = \mathbb{E}f(X) \cdot \mathbb{E}g(Y)$  (the *Probability 1* definition of independence).

1.9 • Let  $A$  and  $B$  be two events in a probability space,  $B$  of positive probability. Derive the *Probability 1* definition of the conditional probability  $\mathbb{P}\{A | B\}$  from our definition of conditional expectations.

1.10 •• Based on your definition above, show that for any fixed event  $B$  of positive probability in the probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ , the set function  $\mathbb{Q}(\cdot) := \mathbb{P}\{\cdot | B\}$  is a probability measure.

1.11 ••• Continuing the previous problem, show that for any events  $B$  and  $C$  with a positive probability intersection,

$$\mathbb{Q}(\cdot | C) = \mathbb{P}(\cdot | B \cap C).$$

1.12 Give an example of a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ , sub- $\sigma$  algebras  $\mathcal{F}_1 \subset \mathcal{F}, \mathcal{F}_2 \subset \mathcal{F}$ , and a random variable  $X$  such that

$$\mathbb{E}(\mathbb{E}(X | \mathcal{F}_1) | \mathcal{F}_2) \neq \mathbb{E}(\mathbb{E}(X | \mathcal{F}_2) | \mathcal{F}_1).$$

Why is it not a contradiction with the Tower rule?

1.13 Let  $X_1, X_2, \dots, X_n$  be iid. random variables with finite mean, and  $S_n$  their sum. Calculate  $\mathbb{E}(X_1 | S_n)$ .

1.14 ••• (*Shiryayev.*) Let  $\mu$  be the Lebesgue-Stieltjes measure generated by a continuous distribution function. Show that if the set  $A$  is at most countable, then  $\mu(A) = 0$ .

1.15 (*Construction of the Vitali set – an example that cannot be Lebesgue measurable.*) Let  $\Omega := [0, 1)$  and define on  $\Omega$  the following equivalence relation:

$$x \sim y \text{ iff } x - y \in \mathbb{Q} \text{ (the rational numbers)}.$$

Let  $V \subset [0, 1)$  consist of *exactly one representative element from each equivalence class of*  $\sim$ . (Notice: this construction relies on the Axiom of Choice.) For  $q \in \mathbb{Q} \cap [0, 1)$ , denote

$$V_q := \{x + q \pmod{1} : x \in V\}.$$

Prove that

- (a) The sets  $V_q$  are congruent: for any  $q, q' \in \mathbb{Q} \cap [0, 1)$ ,  $V_{q'} = (q' - q) + V_q \pmod{1}$ .

(b) If  $q \neq q'$  in  $\mathbb{Q} \cap [0, 1)$ , then  $V_q \cap V_{q'} = \emptyset$ .

(c)  $\bigcup_{q \in \mathbb{Q} \cap [0, 1)} V_q = [0, 1)$ .

Conclude that the Vitali set  $V$  cannot be Lebesgue measurable.

1.16 Let  $X$  and  $Y$  be random variables with finite mean on a probability space. Prove that if  $\mathbb{E}(X | Y) = Y$  and  $\mathbb{E}(Y | X) = X$ , then  $X = Y$  a.s.

1.17 Let  $X$  and  $Y$  be random variables with finite second moment on the probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ . Let  $\mathcal{G}$  be a sub- $\sigma$  algebra of  $\mathcal{F}$ . Suppose that  $\mathbb{E}(X | \mathcal{G}) = Y$  and  $\mathbb{E}X^2 = \mathbb{E}Y^2$ . Prove that  $X = Y$  a.s.