

## Homework # 4: 5463: Real Analysis 2

1. Let  $(\Omega, \mathcal{M}, P)$  be a probability space (= measure space with  $P(\Omega) = 1$ ), let  $X \in L^1(\Omega)$  be a (real valued) random variable, and let  $\mathcal{A} \subset \mathcal{M}$  be a  $\sigma$ -subalgebra of  $\mathcal{M}$ .

(a) Prove that there exists an  $(\mathcal{A}, \mathcal{B}_{\mathbb{R}})$ -measurable function  $f : \Omega \rightarrow \mathbb{R}$  so that

$$\int_A f(\omega) dP(\omega) = \int_A X(\omega) dP(\omega) \quad \text{for all } A \in \mathcal{A}. \quad (1)$$

Moreover, if  $g$  is another such function, then  $f = g$  almost surely (=  $P$ -almost everywhere).

This function is usually denoted by  $E(X|\mathcal{A})$  and called the *conditional expectation* of  $X$ , given  $\mathcal{A}$ .

*Hint:* The right-hand side of (1) defines a (signed) measure on  $(\Omega, \mathcal{A})$ . Use the Radon-Nikodym Theorem.

(b) If  $Y : \Omega \rightarrow \mathbb{R}$  is another random variable, then

$$\mathcal{A}_Y = \{Y^{-1}(B) : B \in \mathcal{B}_{\mathbb{R}}\}$$

defines a  $\sigma$ -algebra on  $\Omega$ , and  $\mathcal{A}_Y \subset \mathcal{M}$ . We can therefore define the *conditional expectation of  $X$ , given  $Y$*  by

$$E(X|Y) = E(X|\mathcal{A}_Y).$$

(c) Prove that if  $f : \Omega \rightarrow \mathbb{R}$  is  $\mathcal{A}_Y$ -measurable, then  $f$  is constant on every set of the form  $Y^{-1}(\{c\})$ . Then show that this implies that  $f$  can be written as a function of  $Y$ :  $f = \varphi \circ Y$  for some  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$ . (With more effort, one can show that  $\varphi$  can be chosen as a Borel function.)

So  $E(X|Y)$  can be thought of as that function of  $Y$  that produces the correct partial averages of  $X$ .

(d) Find  $\mathcal{A}_Y$  and  $E(X|Y)$  in the following example:  $\Omega = \{(00), (01), (10), (11)\}$ ,  $\mathcal{M} = \mathcal{P}(\Omega)$ ,  $P(\{\omega\}) = 1/4$  for all  $\omega \in \Omega$ ,  $X((ab)) = a$ ,  $Y((ab)) = a + b$

(e) Random variables  $X, Y$  are called *independent* if

$$P(X \in B, Y \in B') = P(X \in B)P(Y \in B') \quad \text{for all } B, B' \in \mathcal{B}_{\mathbb{R}}.$$

Show that  $X, Y$  are independent if and only if the *joint distribution*  $P_{X,Y}$  is the product (in the sense of Section 2.5)  $P_X \otimes P_Y$ . Here,

$$P_{X,Y}(B_2) = P((X, Y) \in B_2) \quad (B_2 \in \mathcal{B}_{\mathbb{R}^2}), \quad P_X(B) = P(X \in B) \quad (B \in \mathcal{B}_{\mathbb{R}}).$$

(See also pg. 316.) Then use Fubini-Tonelli to show that if  $X, Y \in L^1(\Omega)$  are independent, then

$$E(XY) = E(X)E(Y) \quad (E(Z) := \int_{\Omega} Z(\omega) dP(\omega)).$$

(f) What is  $E(X|Y)$  if  $X, Y$  are independent? What is  $E(X|X)$ ?

2. Exercise 43 from Section 6.5