

# Math 275B: Homework 1

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B.21.18

*Claim:* Suppose  $\mathcal{F} \subset \mathcal{G}$  are two  $\sigma$ -fields and  $X$  and  $Y$  are bounded  $\mathcal{G}$  measurable random variables. Prove that

$$\mathbb{E}[X\mathbb{E}[Y|\mathcal{F}]] = \mathbb{E}[Y\mathbb{E}[X|\mathcal{F}]].$$

*Proof:* The functions  $X$ ,  $Y$ ,  $\mathbb{E}[X|\mathcal{F}]$ , and  $\mathbb{E}[Y|\mathcal{F}]$  are all  $\mathcal{G}$  measurable and bounded meaning that their products are  $\mathcal{G}$  measurable and bounded (hence integrable w.r.t.  $\mathcal{G}$ ) as well. Thus

$$\mathbb{E}[X\mathbb{E}[Y|\mathcal{F}]] = \mathbb{E}[\mathbb{E}[X\mathbb{E}[Y|\mathcal{F}]]|\mathcal{F}] = \mathbb{E}[\mathbb{E}[X|\mathcal{F}]\mathbb{E}[Y|\mathcal{F}]] = \mathbb{E}[\mathbb{E}[Y\mathbb{E}[X|\mathcal{F}]]|\mathcal{F}] = \mathbb{E}[Y\mathbb{E}[X|\mathcal{F}]]$$

where the first and last equalities follow from the definition of conditional expectation (applied to the entire probability space  $\Omega$ ) while the second and third equalities follow from Proposition 21.20 (Bass), taking into account that  $\mathbb{E}[X|\mathcal{F}]$  and  $\mathbb{E}[Y|\mathcal{F}]$  are both  $\mathcal{F}$  measurable.

B.21.19

*Claim:* Let  $\mathcal{F} \subset \mathcal{G}$  be two  $\sigma$ -fields and let  $X$  be a bounded  $\mathcal{G}$  measurable random variable. Prove that if

$$\mathbb{E}[XY] = \mathbb{E}[X\mathbb{E}[Y|\mathcal{F}]]$$

for all bounded  $\mathcal{G}$  measurable random variables  $Y$ , then  $X$  is  $\mathcal{F}$  measurable.

*Proof:* From the previous problem we can infer that

$$\mathbb{E}[XY] = \mathbb{E}[X\mathbb{E}[Y|\mathcal{F}]] = \mathbb{E}[Y\mathbb{E}[X|\mathcal{F}]]$$

for any bounded  $Y \in \mathcal{G}$ . Then for any set  $A \in \mathcal{G}$ ,

$$\mathbb{E}[1_A X] - \mathbb{E}[1_A \mathbb{E}[X|\mathcal{F}]] = \mathbb{E}[1_A (X - \mathbb{E}[X|\mathcal{F}])] = 0$$

Thus if  $A = \{X > \mathbb{E}[X|\mathcal{F}]\}$  it follows that  $A$  must be a set of measure zero. Else it would contain some set  $A'$  over which  $\mathbb{E}[1_{A'}(X - \mathbb{E}[X|\mathcal{F}])] > 0$ . A similar argument can be made for  $A = \{X < \mathbb{E}[X|\mathcal{F}]\}$  such that  $X = \mathbb{E}[X|\mathcal{F}]$  a.e. except possibly on a  $\mathcal{G}$  measurable set of measure zero. In particular the equality must hold on every positive measure set in  $\mathcal{F}$ .

Since  $\mathbb{E}[X|\mathcal{F}]$  is by definition  $\mathcal{F}$  measurable, it follows that  $X$  is almost  $\mathcal{F}$  measurable as well. Let  $N_{\mathcal{G}} \in \mathcal{G}$  be the measure zero set where  $\mathbb{E}[X|\mathcal{F}]$  and  $X$  differ. It follows that there exists a measure zero set  $N_{\mathcal{F}} \in \mathcal{F}$  containing the points where the functions differ, as they are equal on every positive measure set in  $\mathcal{F}$ . Thus

$$X^{-1}((c, \infty]) = (X^{-1}((c, \infty]) \cap N_{\mathcal{F}}^c) \cup (X^{-1}((c, \infty]) \cap N_{\mathcal{F}}) = (\mathbb{E}[X|\mathcal{F}]^{-1}((c, \infty]) \cap N_{\mathcal{F}}^c) \cup (X^{-1}((c, \infty]) \cap N_{\mathcal{F}})$$

The first set intersection on the right-hand side obviously lies in  $\mathcal{F}$ . Thus there must be an  $\mathcal{F}$  measure zero set over which we can modify  $X$  so as to render the second intersection (and thus  $X$ )  $\mathcal{F}$  measurable. In lieu of a modification,  $X$  may only be  $L^1(\mathcal{F})$  equivalent to  $\mathbb{E}[X|\mathcal{F}]$ .

D.4.1.3

*Claim:* Imitate the proof in the remark after Theorem 1.5.2 to prove the Cauchy-Schwarz inequality.

$$E(XY|\mathcal{G})^2 \leq E(X^2|\mathcal{G})E(Y^2|\mathcal{G})$$

*Proof:* For any set  $A \in \mathcal{G}$  and real value  $\theta$  we can write

$$\int_A (X + \theta Y)^2 dP = \int_A X^2 dP + 2\theta \int_A XY dP + \theta^2 \int_A Y^2 dP.$$

As is mentioned in the referenced remark, the fact that the quadratic expression on the right-hand side is positive for all  $\theta$  implies that it must have either one or zero real roots. In particular, its discriminant must be less than or equal to zero such that

$$\left( \int_A XY dP \right)^2 \leq \int_A X^2 dP \int_A Y^2 dP$$

always holds. But this relation implies the claim up to equivalence on  $\mathcal{G}$ . To see this let

$$B = \{E(XY|\mathcal{G})^2 > E(X^2|\mathcal{G})E(Y^2|\mathcal{G}) \geq \epsilon > 0\}.$$

The integral inequality that we derived implies that  $B$  must have measure zero for all  $\epsilon$ . Thus the claim can only be violated on a set of measure zero.

D.4.1.6

*Claim:* Show that if  $\mathcal{G} \subset \mathcal{F}$  and  $EX^2 \leq \infty$ , then

$$E(\{X - E(X|\mathcal{F})\}^2) + E(\{E(X|\mathcal{F}) - E(X|\mathcal{G})\}^2) = E(\{X - E(X|\mathcal{G})\}^2).$$

*Proof:* The proof of Theorem 4.1.15 (Durrett) demonstrates that  $E(X|\mathcal{F})$  and  $E(X|\mathcal{G})$  can be viewed as orthogonal projections of  $X \in L^2$  onto  $L^2(\mathcal{F})$  and  $L^2(\mathcal{G})$ . It shows that  $E((X - E(X|\mathcal{F}))Z) = 0$  for any  $Z \in L^2(\mathcal{F})$ . Because  $\mathcal{G} \subset \mathcal{F}$ ,  $L^2(\mathcal{G})$  is a subspace of  $L^2(\mathcal{F})$  and we can write

$$\begin{aligned} E(\{X - E(X|\mathcal{G})\}^2) &= E(\{X - E(X|\mathcal{F}) + E(X|\mathcal{F}) - E(X|\mathcal{G})\}^2) \\ &= E(\{X - E(X|\mathcal{F})\}^2 + 2E(\{X - E(X|\mathcal{F})\} \{E(X|\mathcal{F}) - E(X|\mathcal{G})\}) \\ &\quad + \{E(X|\mathcal{F}) - E(X|\mathcal{G})\}^2) \\ &= E(\{X - E(X|\mathcal{F})\}^2 + 2E(\{X - E(X|\mathcal{F})\} E(X|\mathcal{F})) \\ &\quad - 2E(\{X - E(X|\mathcal{F})\} E(X|\mathcal{G})) + \{E(X|\mathcal{F}) - E(X|\mathcal{G})\}^2) \\ &= E(\{X - E(X|\mathcal{F})\}^2) + E(\{E(X|\mathcal{F}) - E(X|\mathcal{G})\}^2) \end{aligned}$$

where the cancellations of the middle terms of the second-to-last equation follow from  $(X - E(X|\mathcal{F}))$  being orthogonal to both  $E(X|\mathcal{F})$  and  $E(X|\mathcal{G})$ .

D.4.1.7

*Claim:* An important special case of the previous result occurs when  $\mathcal{G} = \{\emptyset, \Omega\}$ . Let  $\text{var}(X|\mathcal{F}) = E(X^2|\mathcal{F}) - E(X|\mathcal{F})^2$ . Show that

$$\text{var}(X) = E(\text{var}(X|\mathcal{F})) + \text{var}(E(X|\mathcal{F})).$$

*Proof:* When  $\mathcal{G}$  is the trivial sub-algebra we know that  $E(X|\mathcal{G}) = EX$ . Thus the statement proved in the previous problem becomes

$$\text{var}(E) = E(\{X - EX\}^2) = E(\{X - E(X|\mathcal{F})\}^2) + E(\{E(X|\mathcal{F}) - EX\}^2)$$

Expanding the right hand expression above gives us

$$\begin{aligned}
\text{var}(E) &= E(X^2) - 2E(XE(X|\mathcal{F})) + E(E(X|\mathcal{F})^2) + E(E(X|\mathcal{F})^2) - 2EX \cdot E(E(X|\mathcal{F})) + E((EX)^2) \\
&= E(E(X^2|\mathcal{F})) - 2E(E(X|\mathcal{F})E(X|\mathcal{F})) + E(E(X|\mathcal{F})^2) + E(E(X|\mathcal{F})^2) - 2(EX)^2 + (EX)^2 \\
&= E(E(X^2|\mathcal{F})) - 2E(E(X|\mathcal{F})^2) + E(E(X|\mathcal{F})^2) + E(E(X|\mathcal{F})^2) - (EX)^2 \\
&= E(E(X^2|\mathcal{F}) - E(X|\mathcal{F})^2) + E(E(X|\mathcal{F})^2) - E(E(X|\mathcal{F}))^2 \\
&= E(\text{var}(X|\mathcal{F})) + \text{var}(E(X|\mathcal{F}))
\end{aligned}$$

where I make use of Theorem 4.1.14 (Durrett) and the fact that  $E(X|\mathcal{F}) \in \mathcal{F}$  to determine

$$E(XE(X|\mathcal{F})) = E(E(XE(X|\mathcal{F})|\mathcal{F})) = E(E(X|\mathcal{F})E(X|\mathcal{F})).$$

I also exploit the equivalence  $EX = E(E(X|\mathcal{F}))$  several times.