

STOR 635 HW 3

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1 Solutions

1. Let (Ω, \mathcal{F}, P) be a probability space. Let \mathcal{G} be a sub σ -field of \mathcal{F} and let X be a square integrable random variable in (Ω, \mathcal{F}, P) , that is, $X \in L^2(\Omega, \mathcal{F}, P)$. Show that there exists a unique $Z \in L^2(\Omega, \mathcal{G}, P)$ such that for every $Y \in L^2(\Omega, \mathcal{G}, P)$, $E(XY) = E(ZY)$. Moreover, if $X_1, X_2 \in L^2(\Omega, \mathcal{F}, P)$ such that $X_1 \geq X_2$ almost surely, then $Z_1, Z_2 \in L^2(\Omega, \mathcal{G}, P)$ obtained by the above recipe for X_1, X_2 respectively should satisfy $Z_1 \geq Z_2$ almost surely.

Show that the Z obtained above (for X) satisfies the *projection property*: for any $W \in L^2(\Omega, \mathcal{G}, P)$, $E[(X - Z)^2] \leq E[(X - W)^2]$.

Proof. Note that $L^2(\Omega, \mathcal{G}, P)$ is a linear subspace of $L^2(\Omega, \mathcal{F}, P)$. Since $L^2(\Omega, \mathcal{G}, P)$ is complete, it is closed. By Hilbert space theory, we have the decomposition

$$L^2(\Omega, \mathcal{F}, P) = L^2(\Omega, \mathcal{G}, P) \oplus L^2(\Omega, \mathcal{G}, P)^\perp.$$

This means that for every $X \in L^2(\Omega, \mathcal{F}, P)$, there exist unique $Z \in L^2(\Omega, \mathcal{G}, P)$, $V \in L^2(\Omega, \mathcal{G}, P)^\perp$ such that

$$X = Z + V.$$

For $X = Z + V$ as above, set $QX = Z$. This defines $Q: L^2(\Omega, \mathcal{F}, P) \rightarrow L^2(\Omega, \mathcal{G}, P)$, the orthogonal projection of $L^2(\Omega, \mathcal{F}, P)$ onto $L^2(\Omega, \mathcal{G}, P)$. For future use, we note that Q is linear and that $\|QX\|_{L^2} \leq \|X\|_{L^2}$ since $\|X\|_{L^2}^2 = \|Z\|_{L^2}^2 + \|V\|_{L^2}^2$.

Let $X \in L^2(\Omega, \mathcal{F}, P)$ be arbitrary. Write $X = Z + V$ as above. Note that the inner product on $L^2(\Omega, \mathcal{F}, P)$ is given by $(f, g) = E(fg)$. Thus for $Y \in L^2(\Omega, \mathcal{G}, P)$,

$$\begin{aligned} E(XY) &= (Z + V, Y) \\ &= (Z, Y) + (V, Y) \\ &= (Z, Y) + 0 \\ &= E(ZY). \end{aligned}$$

Now we verify the projection property. Let $W \in L^2(\Omega, \mathcal{G}, P)$ be arbitrary. By the Pythagorean theorem,

$$\begin{aligned} E((X - W)^2) &= \|X - W\|_{L^2}^2 \\ &= \|Z - W + V\|_{L^2}^2 \\ &= \|Z - W\|_{L^2}^2 + \|V\|_{L^2}^2. \end{aligned}$$

Thus $E((X - W)^2)$ is minimal at $W = Z$. This proves the projection property.

Suppose $X_1, X_2 \in L^2(\Omega, \mathcal{F}, P)$ such that $X_1 \geq X_2$ almost surely. Let $Z_1 = QX_1$, $Z_2 = QX_2$. To show that $Z_1 \geq Z_2$ a.s., it suffices to show that $Z_1 - Z_2 = Q(X_1 - X_2) \geq 0$ a.s. Thus it suffices to show that if $X \in L^2(\Omega, \mathcal{F}, P)$ and $X \geq 0$ a.s., then $QX \geq 0$ a.s.

So let $X \in L^2(\Omega, \mathcal{F}, P)$ with $X \geq 0$ be arbitrary. Let $Z = QX$. Note that all simple functions are in L^2 . Hence for any $A \in \mathcal{G}$,

$$\begin{aligned} \int_A Z dP &= \int_A X dP \geq 0. \\ \text{Put } A = \{Z < 0\} \in \mathcal{G} \text{ to get} \quad & 0 \geq \int_{\{Z < 0\}} Z dP \geq 0. \end{aligned}$$

Hence $\int_{\{Z < 0\}} Z dP = 0$, so $\int_{\{Z < 0\}} -Z dP = 0$. Since $-Z \geq 0$ on $\{Z < 0\}$, this implies that $-Z = 0$ a.s. on $\{Z < 0\}$. Thus $P(Z < 0) = 0$. Thus $Z \geq 0$ a.s. \square

2. Use the above problem to show that conditional expectation exists and is unique, that is, for each $X \in L^1(\Omega, \mathcal{F}, P)$, there exists a unique $Z \in L^1(\Omega, \mathcal{G}, P)$ such that for any $F \in \mathcal{G}$, $\int_F X dP = \int_F Z dP$.

Proof. Let $X \in L^2(\Omega, \mathcal{F}, P)$ be arbitrary. By the monotonicity proved in problem 1,

$$\begin{aligned} |QX| &= |QX^+ - QX^-| \\ &\leq |QX^+| + |QX^-| \\ &= QX^+ + QX^- \\ &= Q(X^+ + X^-) \\ &= Q|X|. \end{aligned}$$

Thus

$$\begin{aligned} \|QX\|_{L^1} &= \int_{\Omega} |QX| dP \\ &\leq \int_{\Omega} Q|X| dP \\ &= \int_{\Omega} |X| dP \\ &= \|X\|_{L^1}. \end{aligned}$$

We now use this estimate and the density of $L^2(\Omega, \mathcal{F}, P)$ in $L^1(\Omega, \mathcal{F}, P)$ to extend Q to a linear map on $L^1(\Omega, \mathcal{F}, P)$ satisfying the same estimate. Let $X \in L^1(\Omega, \mathcal{F}, P)$ be arbitrary. Pick $X_n \in L^2(\Omega, \mathcal{F}, P)$ with $X_n \rightarrow X$ in L^1 ; for example, take X_n to be simple functions converging to X pointwise with $|X_n| \leq |X|$. Note that

$$\|QX_n - QX_m\|_{L^1} = \|Q(X_n - X_m)\|_{L^1} \leq \|X_n - X_m\|_{L^1} \rightarrow 0 \text{ as } n, m \rightarrow \infty.$$

Thus (QX_n) is a Cauchy sequence in $L^1(\Omega, \mathcal{G}, P)$, so there is an L^1 limit

$$QX := \lim_{n \rightarrow \infty} QX_n \in L^1(\Omega, \mathcal{G}, P).$$

We need to check that QX is independent of the sequence X_n . If $X'_n \in L^2(\Omega, \mathcal{F}, P)$ also converge to X in L^1 , then

$$\left\| QX - \lim_{n \rightarrow \infty} QX_n \right\|_{L^1} = \lim_{n \rightarrow \infty} \|QX_n - QX'_n\|_{L^1} \leq \lim_{n \rightarrow \infty} \|X_n - X'_n\|_{L^1} = \|X - X\|_{L^1} = 0.$$

Thus QX is independent of the sequence X_n . With this in hand, it is easy to check that Q is linear and that $\|QX\|_{L^1} \leq \|X\|_{L^1}$ for all $X \in L^1(\Omega, \mathcal{F}, P)$.

Now let $X \in L^1(\Omega, \mathcal{F}, P)$ be arbitrary. Let $Z = QX \in L^1(\Omega, \mathcal{G}, P)$. Pick $X_n \in L^2(\Omega, \mathcal{F}, P)$ with $X_n \rightarrow X$ in L^1 . Let $F \in \mathcal{G}$ be arbitrary. For each n , we have by problem 1 that

$$\int_F X_n dP = \int_F QX_n dP.$$

By definition, $QX_n \rightarrow Z$ in L^1 . Thus letting $n \rightarrow \infty$ above gives

$$\int_F X dP = \int_F Z dP.$$

This proves existence. Now suppose $Z' \in L^1(\Omega, \mathcal{G}, P)$ is also a conditional expectation of X given \mathcal{G} . Then for every $A \in \mathcal{G}$, $\int_A X dP = \int_A Z dP = \int_A Z' dP$, so

$$\int_A (Z - Z') dP = 0.$$

Putting $A = \{Z - Z' > 0\} \in \mathcal{G}$ yields that $Z - Z' = 0$ a.e. on $\{Z - Z' > 0\}$, which implies that $P(Z - Z' > 0) = 0$. A similar argument yields $P(Z - Z' < 0) = 0$. Thus $Z = Z'$ a.e. This proves uniqueness. \square

3. Let (Ω, \mathcal{F}, P) be a probability space. Let \mathcal{G} be a sub σ -field of \mathcal{F} and let X be an integrable random variable. Let U be a \mathcal{G} -measurable random variable such that $E(|UX|) < \infty$. Show that

$$E(UX) = E(UE(X|\mathcal{G})).$$

Proof. Suppose $Y : \Omega \rightarrow [0, \infty)$ is \mathcal{F} -measurable and integrable (actually, what follows still makes sense when Y is not integrable since $E(Y|\mathcal{G})$ still makes sense). For all $A \in \mathcal{G}$, we have

$$\int 1_A Y dP = \int 1_A E(Y|\mathcal{G}) dP.$$

By linearity, it follows that for nonnegative \mathcal{G} -measurable simple functions f ,

$$\int fY dP = \int fE(Y|\mathcal{G}) dP.$$

By the monotone convergence theorem, the above holds for all \mathcal{G} -measurable $f : \Omega \rightarrow [0, \infty]$. In particular, if $fY \in L^1$, then $fE(Y|\mathcal{G}) \in L^1$. Note that $|U^\pm X^\pm| \leq |U| |X| \in L^1$. Thus $U^\pm E(X^\pm|\mathcal{G}) \in L^1$ as well. Thus by linearity of the integral on L^1 ,

$$\begin{aligned} \int UX &= \int (U^+ - U^-)(X^+ - X^-) \\ &= \int (U^+ X^+ - U^+ X^- - U^- X^+ + U^- X^-) \\ &= \int (U^+ E(X^+|\mathcal{G}) - U^+ E(X^-|\mathcal{G}) - U^- E(X^+|\mathcal{G}) + U^- E(X^-|\mathcal{G})) \\ &= \int (U^+ E(X|\mathcal{G}) - U^- E(X|\mathcal{G})) \\ &= \int UE(X|\mathcal{G}). \end{aligned}$$

□

4. Let (Ω, \mathcal{F}, P) be a probability space. Let \mathcal{G}, \mathcal{H} be sub σ -fields of \mathcal{F} such that $\mathcal{H} \subset \mathcal{G}$. Let X be an integrable random variable. Show

$$E(E(X|\mathcal{G})|\mathcal{H}) = E(X|\mathcal{H}), \quad E(E(X|\mathcal{H})|\mathcal{G}) = E(X|\mathcal{H}).$$

Proof. Let $A \in \mathcal{H}$. Since $A \in \mathcal{G}$,

$$E(E(X|\mathcal{G})1_A) = E(X1_A) = E(E(X|\mathcal{H})1_A).$$

Thus $E(E(X|\mathcal{G})|\mathcal{H}) = E(X|\mathcal{H})$. Since $E(X|\mathcal{H})$ is \mathcal{H} -measurable, it is also \mathcal{G} measurable. Thus trivially, $E(E(X|\mathcal{H})|\mathcal{G}) = E(X|\mathcal{H})$. □

5. Let (Ω, \mathcal{F}, P) be a probability space and let \mathcal{G} be a sub σ -field of \mathcal{F} . Show that if $\{X_n\}$ converges to X in L^p ($1 \leq p < \infty$) then $E(X_n|\mathcal{G}) \rightarrow E(X|\mathcal{G})$ in L^p .

Proof. Since conditional expectation is linear, it suffices to show that $\|E(X|\mathcal{G})\|_{L^p} \leq \|X\|_{L^p}$ for all $X \in L^p(\Omega, \mathcal{F}, P)$. So let $X \in L^p(\Omega, \mathcal{F}, P)$ be arbitrary. Note that $\varphi: [0, \infty) \rightarrow [0, \infty)$ defined by $\varphi(x) = x^p$ is convex since $\varphi''(x) = p(p-1)x^{p-2} \geq 0$ for all $x \geq 0$. Thus by monotonicity and Jensen's inequality,

$$|E(X|\mathcal{G})|^p \leq E(|X|^p|\mathcal{G}) \leq E(|X|^p|\mathcal{G}).$$

Integrating gives

$$\|E(X|\mathcal{G})\|_{L^p}^p \leq E(E(|X|^p|\mathcal{G})) = E(|X|^p) = \|X\|_{L^p}^p. \quad \square$$

6. Let $\{X_i\}_{i \in \mathcal{I}}$ be a uniformly integrable family and $\{\mathcal{G}_j\}_{j \in \mathcal{J}}$ be a collection of sub σ -fields of \mathcal{F} . Show that the collection $\mathcal{U} \doteq \{E(X_i|\mathcal{G}_j) : (i, j) \in \mathcal{I} \times \mathcal{J}\}$ is a u.i. family.

Proof. In homework 1, it was proved as a lemma that if μ is a finite measure, a collection $\mathcal{C} \subset L^1(\mu)$ is uniformly integrable if and only if

$$\limsup_{a \rightarrow \infty} \int (|f| - a)^+ d\mu = 0.$$

Let $a > 0$ be arbitrary. Note that the map $\varphi: \mathbb{R} \rightarrow [0, \infty)$ given by $\varphi(x) = x^+ = \max(x, 0)$ is convex since it is the maximum of two convex functions. Note also that φ is increasing. For $i \in \mathcal{I}, j \in \mathcal{J}$, we have by monotonicity and Jensen's inequality that

$$(|E(X_i|\mathcal{G}_j)| - a)^+ \leq (E(|X_i|\mathcal{G}_j) - a)^+ = E(|X_i| - a|\mathcal{G}_j)^+ \leq E((|X_i| - a)^+|\mathcal{G}_j).$$

Taking expectations gives

$$E((|E(X_i|\mathcal{G}_j)| - a)^+) \leq E(E((|X_i| - a)^+|\mathcal{G}_j)) = E((|X_i| - a)^+).$$

Thus

$$\sup_{i \in \mathcal{I}, j \in \mathcal{J}} E((|E(X_i|\mathcal{G}_j)| - a)^+) \leq \sup_{i \in \mathcal{I}} E((|X_i| - a)^+) \rightarrow 0 \text{ as } a \rightarrow \infty.$$

Thus \mathcal{U} is uniformly integrable. □

7. Let \mathbf{P} and \mathbf{Q} be two probability measures on (Ω, \mathcal{F}) such that $\mathbf{Q} \ll \mathbf{P}$. Let $\{\mathcal{G}_j : j \in \mathcal{J}\}$ be a collection of sub σ -fields of \mathcal{F} . Let $\mathbf{Q}_j \doteq \mathbf{Q}|_{\mathcal{G}_j}$ and $\mathbf{P}_j = \mathbf{P}|_{\mathcal{G}_j}$. Regarding $\mathbf{Q}_j, \mathbf{P}_j$ as probability measures on (Ω, \mathcal{G}_j) , let X_j be the \mathcal{G}_j measurable random variable such that $X_j = \frac{d\mathbf{Q}_j}{d\mathbf{P}_j}$. Show $\{X_j : j \in \mathcal{J}\}$ is u.i. on $(\Omega, \mathcal{F}, \mathbf{P})$.

Proof. Let $X = \frac{d\mathbf{Q}}{d\mathbf{P}}$. For all $A \in \mathcal{G}_j$, we have

$$\begin{aligned} \int_A X d\mathbf{P} &= \mathbf{Q}(A) \\ &= \mathbf{Q}_j(A) \\ &= \int_A X_j d\mathbf{P}_j \\ &= \int_A X_j d\mathbf{P}. \end{aligned}$$

Thus $X_j = E(X | \mathcal{G}_j)$. Since $\{X\}$ is uniformly integrable, problem 6 implies $\{E(X | \mathcal{G}_j) : j \in \mathcal{J}\} = \{X_j : j \in \mathcal{J}\}$ is uniformly integrable. \square

8. Let $X, Y \in L^2$. Suppose $E(X | Y) = Y$ and $E(Y | X) = X$. Show $X = Y$ a.s.

Proof. We have

$$\begin{aligned} E((X - Y)^2) &= E(E(X^2 + Y^2 - 2XY | X)) \\ &= E(X^2 + E(Y^2 | X) - 2E(XY | X)). \end{aligned}$$

Note that in the setting of problem 3, replacing U with $1_A U$ for $A \in \mathcal{G}$ yields $E(UX | \mathcal{G}) = U E(X | \mathcal{G})$. Thus since $XY \in L^1$, $E(XY | X) = X E(Y | X) = X^2$. Thus

$$\begin{aligned} E((X - Y)^2) &= E(X^2 + E(Y^2 | X) - 2X^2) \\ &= E(E(Y^2 | X)) - E(X^2) \\ &= E(Y^2) - E(X^2). \end{aligned}$$

Swapping X and Y gives

$$E((X - Y)^2) = E(X^2) - E(Y^2) = -E((X - Y)^2).$$

Thus $E((X - Y)^2) = 0$, so $X = Y$. \square