
ProbLimI: Problem Set IX

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Abstract

This work contains solutions to the exercises of the problem set IX. The chosen problems are 2,3 and 4.

Question 2.

2. Let X be an r.v. in $(\Omega, \mathcal{F}, \mathbb{P})$ such that $\mathbb{E}X^2 < \infty$ and write $\text{Var}(X | \mathcal{G}) := \mathbb{E}[(X - \mathbb{E}[X | \mathcal{G}])^2 | \mathcal{G}]$.

- (a) Prove that $\mathbb{E}[\text{Var}(X | \mathcal{G})] \leq \mathbb{E}[\text{Var}(X | \mathcal{H})]$ for every σ -fields $\mathcal{H} \subset \mathcal{G} \subset \mathcal{F}$.
- (b) Prove that $\text{Var}(X) = \mathbb{E}[\text{Var}(X | \mathcal{G})] + \text{Var}(\mathbb{E}[X | \mathcal{G}])$ for every σ -field $\mathcal{G} \subset \mathcal{F}$.

Solution.

(a) The result intuitively makes sense, since on average knowing more information should reduce variance. We compute

$$\begin{aligned}\mathbb{E}[\text{Var}(X|H)] &= \mathbb{E}[(X - \mathbb{E}[X|H])^2] = \mathbb{E}[(X - \mathbb{E}[X|G])^2] + \mathbb{E}[(\mathbb{E}[X|G] - \mathbb{E}[X|H])^2] \\ &+ 2\mathbb{E}[(\mathbb{E}[X|G] - \mathbb{E}[X|H])(X - \mathbb{E}[X|G])] \\ &= \mathbb{E}[\text{Var}(X|G)] - \mathbb{E}[(\mathbb{E}[X|G] - \mathbb{E}[X|H])^2] \geq \mathbb{E}[\text{Var}(X|G)]\end{aligned}\tag{1}$$

where (1) holds since $\mathbb{E}[X|G] - \mathbb{E}[X|H]$ is G measurable.

(b) Let $m = \mathbb{E}[Y] = \mathbb{E}[\mathbb{E}[Y|\mathcal{G}]]$. We compute

$$\begin{aligned}\text{Var}[Y] &= \mathbb{E}[(Y - m)^2] = \mathbb{E}[\mathbb{E}[(Y - m)^2|\mathcal{G}]] \\ &= \mathbb{E}[\mathbb{E}[(Y - \mathbb{E}[Y|\mathcal{G}] + \mathbb{E}[Y|\mathcal{G}] - m)^2|\mathcal{G}]] \\ &= \mathbb{E}[\mathbb{E}[(Y - \mathbb{E}[Y|\mathcal{G}])^2|\mathcal{G}]] - \mathbb{E}[\mathbb{E}[(\mathbb{E}[Y|\mathcal{G}] - m)^2|\mathcal{G}]] \\ &+ \mathbb{E}[\mathbb{E}[2(Y - \mathbb{E}[Y|\mathcal{G}])(\mathbb{E}[Y|\mathcal{G}] - m)|\mathcal{G}]]\end{aligned}\tag{2}$$

$$\begin{aligned}&= \mathbb{E}[\text{Var}(Y|\mathcal{G}) + \text{Var}[\mathbb{E}[Y|\mathcal{G}]] \\ &+ \mathbb{E}[\mathbb{E}[2(Y - \mathbb{E}[Y|\mathcal{G}])(\mathbb{E}[Y|\mathcal{G}] - m)|\mathcal{G}]]\end{aligned}\tag{3}$$

where (2) holds by linearity of conditional expectation. Now, from ,

$$\mathbb{E}[(Y - \mathbb{E}[Y|\mathcal{G}])(\mathbb{E}[Y|\mathcal{G}] - m)|\mathcal{G}] = (\mathbb{E}[Y|\mathcal{G}] - m)\mathbb{E}[(Y - \mathbb{E}[Y|\mathcal{G}])|\mathcal{G}]\tag{4}$$

$$= (\mathbb{E}[Y|\mathcal{G}] - m)(\mathbb{E}[Y|\mathcal{G}] - \mathbb{E}[Y|\mathcal{G}]) = 0\tag{5}$$

almost surely, where (4) holds by "taking out what's known." Now, combining (3) and (5),

$$\text{Var}[Y] = \mathbb{E}[\text{Var}(Y|\mathcal{G}) + \text{Var}[\mathbb{E}[Y|\mathcal{G}]]].$$

□

Question 3.

3. Let $X \in L^1(\Omega, \mathcal{F}, \mathbb{P})$ and $Y = \mathbb{E}[X | \mathcal{G}]$ for a σ -field $\mathcal{G} \subset \mathcal{F}$.

- (a) Prove that if $\mathbb{E}Y^2 = \mathbb{E}X^2 < \infty$ then $Y = X$ a.s.
- (b) Prove that if $Y \stackrel{d}{=} X$ then $Y = X$ a.s., even in the case $\mathbb{E}X^2 = \infty$.

Solution.

(a) We compute

$$\begin{aligned} \mathbb{E}[(X - Y)^2] &= \mathbb{E}[X^2] + \mathbb{E}[Y^2] - 2\mathbb{E}[XY] \\ &= \mathbb{E}[X^2] + \mathbb{E}[Y^2] - 2\mathbb{E}[\mathbb{E}[XY|\mathcal{G}]] \end{aligned} \tag{6}$$

$$\begin{aligned} &= \mathbb{E}[X^2] + \mathbb{E}[Y^2] - 2\mathbb{E}[Y\mathbb{E}[X|\mathcal{G}]] \\ &= \mathbb{E}[X^2] + \mathbb{E}[Y^2] - 2\mathbb{E}[Y^2] = 0 \end{aligned} \tag{7}$$

where (6) holds as the expectation of the conditional expectation of any L^1 random variable is the expectation of the random variable, and (7) holds as Y is \mathcal{G} measurable. Therefore, $X = Y$ almost surely.

(b) We proceed by a standard truncation argument. Let $a > 0$ and $b < 0$. By conditional Jensen,

$$\mathbb{E}[X \wedge a | \mathcal{G}] \leq \mathbb{E}[X | \mathcal{G}] \wedge a = Y \wedge a \text{ almost surely}$$

Since $X = Y$ in distribution,

$$\mathbb{E}[X \wedge a] = \mathbb{E}[Y \wedge a].$$

Therefore, the above inequality cannot be strict, which implies

$$\mathbb{E}[X \wedge a | \mathcal{G}] = Y \wedge a \text{ almost surely}$$

as otherwise by taking the expectation both sides, we get a contradiction. Similarly,

$$\mathbb{E}[(X \wedge a) \vee b | \mathcal{G}] = (Y \wedge a) \vee b \text{ almost surely} \tag{8}$$

From $X = Y$ in distribution,

$$(X \wedge a) \vee b = (Y \wedge a) \vee b \text{ in distribution}$$

and hence

$$\mathbb{E}[(X \wedge a) \vee b] = \mathbb{E}[(Y \wedge a) \vee b].$$

Therefore, from part (a) and (8),

$$(X \wedge a) \vee b = (Y \wedge a) \vee b \text{ almost surely}$$

for any $a > 0$ and $b < 0$. Taking $a \rightarrow \infty$ and $b \rightarrow -\infty$,

$$X = Y \text{ almost surely.}$$

□

Question 4.

4. (a) Fix $p > 0$ and let X be an r.v. in $(\Omega, \mathcal{F}, \mathbb{P})$ such that $\mathbb{E}|X|^p < \infty$. Show that for every σ -field $\mathcal{G} \subset \mathcal{F}$, a.s.

$$\mathbb{E}[|X|^p | \mathcal{G}] = \int_0^\infty px^{p-1} \mathbb{P}(|X| > x | \mathcal{G}) dx,$$

and conclude that for every $a > 0$,

$$\mathbb{P}(|X| \geq a | \mathcal{G}) \leq a^{-p} \mathbb{E}[|X|^p | \mathcal{G}].$$

- (b) Let X, Y be r.v.'s on $(\Omega, \mathcal{F}, \mathbb{P})$ such that $\mathbb{E}|X|^p < \infty$, $\mathbb{E}|Y|^q < \infty$ for $p, q > 1$ with $\frac{1}{p} + \frac{1}{q} = 1$. Prove that for every σ -field $\mathcal{G} \subset \mathcal{F}$,

$$\mathbb{E}[|XY| | \mathcal{G}] \leq \mathbb{E}[|X|^p | \mathcal{G}]^{1/p} \mathbb{E}[|Y|^q | \mathcal{G}]^{1/q}.$$

Solution.

(a) By Fubini,

$$\begin{aligned} \mathbb{E}[|X|^p | \mathcal{G}] &= \int_\Omega |x|^p \mathbb{P}(dx | \mathcal{G}) = \int_\Omega \int_0^\infty px^{p-1} 1_{\{|X| > x\}} dx \mathbb{P}(dx | \mathcal{G}) \\ &= \int_0^\infty px^{p-1} 1_{\{|X| > x\}} dx \mathbb{P}(dx | \mathcal{G}) = \int_0^\infty px^{p-1} \mathbb{P}(|X| > x | \mathcal{G}) dx \quad \text{a.s.} \end{aligned}$$

for any $\mathcal{G} \subset \mathcal{F}$. Let $a > 0$. Then,

$$\begin{aligned} a^{-p} \mathbb{E}[|X|^p | \mathcal{G}] &= a^{-p} \int_0^a px^{p-1} \mathbb{P}(|X| > x | \mathcal{G}) dx + a^{-p} \int_a^\infty px^{p-1} \mathbb{P}(|X| > x | \mathcal{G}) dx \\ &\leq a^{-p} \int_0^a px^{p-1} \mathbb{P}(|X| > x | \mathcal{G}) dx + \mathbb{P}(|X| = a | \mathcal{G}) = \mathbb{P}(|X| \geq a | \mathcal{G}) \end{aligned}$$

(b) Let $A = (\mathbb{E}[|X|^p | \mathcal{G}])^{1/p}$ and $B = (\mathbb{E}[|Y|^q | \mathcal{G}])^{1/q}$. We compute

$$\begin{aligned} \mathbb{E}[|X|^p 1_{\{A=0\}}] &= \mathbb{E}[\mathbb{E}[|X|^p 1_{\{A=0\}}]] \\ &= \mathbb{E}[1_{\{A=0\}} \mathbb{E}[|X|^p | \mathcal{G}]] = \mathbb{E}[1_{\{A=0\}} A^p] = 0 \end{aligned}$$

and hence $|X| = 0$ a.s. on $\{A = 0\}$. By the same computation, $|Y| = 0$ a.s. on $\{B = 0\}$, which implies

$$\mathbb{E}[|XY| | \mathcal{G}] = 0 \quad \text{a.s. on } \{A = 0\} \cup \{B = 0\}.$$

Hence, it suffices to show the inequality on $\Omega_0 = \{A \neq 0\} \cap \{B \neq 0\}$. We compute

$$\begin{aligned} \mathbb{E}\left[\frac{\mathbb{E}[|XY| | \mathcal{G}]}{AB} 1_G\right] &= \mathbb{E}\left[\frac{|X|}{A} 1_G \frac{|Y|}{B} 1_G\right] \\ &\leq (\mathbb{E}\left[\frac{|X|^p}{A^p} 1_G\right])^{1/p} (\mathbb{E}\left[\frac{|Y|^q}{B^q} 1_G\right])^{1/q} \\ &= \mathbb{E}[1_G] \end{aligned} \tag{9}$$

for any $G \in \mathcal{G}$ where (9) holds by Holder. Therefore, the inequality holds on Ω_0 , so we are done.

□