

# STAT 205A - Homework 9

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## Problem 1. Conditional Expectation

*Proof.* Let  $(\Omega_1, \mathcal{F}, \nu_x)$  be the probability space for  $X$ ,  $(\Omega_2, \mathcal{G}, \nu_y)$  be the probability space for  $Y$ . Let the product:

$$\begin{aligned}\Omega &= \Omega_1 \times \Omega_2 = \{(\omega_1, \omega_2) \mid \omega_1 \in \Omega_1, \omega_2 \in \Omega_2\} \\ \mathcal{S} &= \sigma(\{S_1 \times S_2 \mid S_1 \in \mathcal{F}, S_2 \in \mathcal{G}\}) \\ \nu &= \nu_x \times \nu_y \\ \nu(S_1 \times S_2) &= \nu_x(S_1)\nu_y(S_2), \forall S_1 \in \mathcal{F}, S_2 \in \mathcal{G}\end{aligned}$$

Let  $Z = \mathbb{E}[h(X, Y) \mid \mathcal{G}]$ , then by definition of conditional expectation,  $Z$  is  $\mathcal{G}$ -measurable. Since  $h$  is bounded measurable, it is absolutely integrable so we can interchange integrals. By the definition of conditional expectation we have:

$$\begin{aligned}\mathbb{E}[Z\mathbb{I}_G] &= \mathbb{E}[h(X, Y)\mathbb{I}_G], \forall G \in \mathcal{G} \\ &= \int h(X, Y)\mathbb{I}_G \nu(dx, dy) \\ &= \int \left( \int h(x, Y)\mathbb{I}_G \mu(Y, dx) \right) \nu_y(dy) \\ &= \int \mathbb{I}_G \left( \int h(x, Y)\mu(Y, dx) \right) \nu_y(dy)\end{aligned}$$

So  $Z$  and  $\int h(x, Y)\mu(Y, dx)$  are  $\mathcal{G}$  measurable function, that have the same expectation on all measurable set in  $\mathcal{G}$ . Thus they are equal almost surely.  $\square$

## Problem 2. Conditional Independence Definition

*Proof.* (b)  $\Rightarrow$  (a) Given (b), (a) is true as we can apply (b) for the case  $h_1(X) = \mathbb{I}_{A_1}, h_2(X) = \mathbb{I}_{A_2}$ .

(a)  $\Rightarrow$  (b) Given (a) is true.

First step, since (a) is true, (b) is true for any simple function  $h_1(X_1), h_2(X_2)$  each of the form  $\sum a_i \mathbb{I}_{A_i}$ , by the linearity of expectation.

Second step, from the first step, we have (b) is true for any bounded positive measurable function  $h_i(X_i), i = 1, 2$  by the Monotone Convergence Theorem

Third step, from the second step, for any bounded measurable function, define  $h_i^+$  as the positive part of  $h_i$ , and  $h_i^-$  as the negative part, then from  $h_i^+$  and  $-h_i^-$  are positive bounded measurable function, so (b) is true for both of these function as by second step. Thus (b) is also true for  $h$ .

(c)  $\Rightarrow$  (b) Given (c) we have:

$$\begin{aligned}\mathbb{E}[h_1(X_1)h_2(X_2) \mid \mathcal{G}] &= \mathbb{E}[\mathbb{E}[h_1(X_1)h_2(X_2) \mid \mathcal{G}, X_2] \mid \mathcal{G}] \\ &= \mathbb{E}[h_2(X_2)\mathbb{E}[h_1(X_1) \mid \mathcal{G}, X_2]]; \quad h_2(X_2) \text{ is } \sigma(X_2) - \text{measurable} \\ &= \mathbb{E}[h_2(X_2)\mathbb{E}[h_1(X_1) \mid \mathcal{G}] \mid \mathcal{G}]; \quad \text{because of (c)} \\ &= \mathbb{E}[h_1(X_1) \mid \mathcal{G}]\mathbb{E}[h_2(X_2) \mid \mathcal{G}]; \quad \mathbb{E}[h_1(X_1) \mid \mathcal{G}] \text{ is } \mathcal{G} - \text{measurable}\end{aligned}$$

(b)  $\Rightarrow$  (c) Let  $Y$  be a  $\mathcal{G}$ -measurable r.v. From (b) we have:

$$\begin{aligned}
& \mathbb{E}[h_1(X_1)h_2(X_2)Y] \\
&= \mathbb{E}[\mathbb{E}[h_1(X_1)h_2(X_2)Y \mid \mathcal{G}]] ; \text{ Tower Property} \\
&= \mathbb{E}[\mathbb{E}[h_1(X_1)h_2(X_2) \mid \mathcal{G}]Y] ; Y \text{ is } \mathcal{G}\text{-measurable} \\
&= \mathbb{E}[\mathbb{E}[h_1(X_1) \mid \mathcal{G}]\mathbb{E}[h_2(X_2) \mid \mathcal{G}]Y] ; \text{ from (b)} \\
&=^{(*)} \mathbb{E}[\mathbb{E}[h_1(X_1) \mid \mathcal{G}]h_2(X_2)Y] ; \text{ why?}
\end{aligned}$$

The last statement is true for only  $h_2(X_2)$  that is independent of  $\mathcal{G}$ . So we have to limit the scope of  $h_2(X_2)$ .

Let  $Z = \mathbb{E}[h_1(X_1) \mid \mathcal{G}, X_2]$ , since  $h_2(X_2)Y$  is  $\sigma(\mathcal{G}, X_2)$ -measurable, we have:

$$\begin{aligned}
& \mathbb{E}[Zh_2(X_2)Y] \\
&= \mathbb{E}[\mathbb{E}[h_1(X_1) \mid \mathcal{G}, X_2]h_2(X_2)Y] ; \text{ definition of } Z \\
&= \mathbb{E}[\mathbb{E}[h_1(X_1)h_2(X_2)Y \mid \mathcal{G}, X_2]] ; h_2(X_2)Y \text{ is } \mathcal{G}\text{-measurable} \\
&= \mathbb{E}[h_1(X_1)h_2(X_2)Y] ; \text{ Tower Property} \\
&= \mathbb{E}[\mathbb{E}[h_1(X_1) \mid \mathcal{G}]h_2(X_2)Y] ; \text{ because of } (*)
\end{aligned}$$

Thus  $\mathbb{E}[(Z - \mathbb{E}[h_1(X_1) \mid \mathcal{G}])h_2(X_2)Y] = 0$ , by the  $\pi - \lambda$  theorem, we have:

$$\mathbb{E}[(Z - \mathbb{E}[h_1(X_1) \mid \mathcal{G}])X] = 0$$

for all  $X$  that is  $\sigma(\mathcal{G}, X_2)$ -measurable. Take  $X = Z - \mathbb{E}[h_1(X_1) \mid \mathcal{G}]$  then we have  $Z = \mathbb{E}[h_1(X_1) \mid \mathcal{G}]$ .  $\square$

**Problem 3.** Conditional Independence with respect to different  $\sigma$ -algebra.

- (a)  $X, Y$  conditional independent given  $Z$
- (b)  $X, Z$  conditional independent given  $\mathcal{F}$

*Proof.* For  $h$  bounded and measurable, we have:

$$\begin{aligned}
& \mathbb{E}[h(X) \mid Y, Z] = \mathbb{E}[h(X) \mid Z] \text{ by (a) and 2(c)} & (1) \\
& \mathbb{E}[h(Y) \mid X, Z] = \mathbb{E}[h(Y) \mid Z] \text{ by (a) 2(c)} & (2) \\
& \mathbb{E}[h(X) \mid Z, \mathcal{F}] = \mathbb{E}[h(X) \mid \mathcal{F}] \text{ by (b) and 2(c)} & (3) \\
& \mathbb{E}[h(X) \mid Z, \mathcal{F}] = \mathbb{E}[h(X) \mid Z] \text{ because } \mathcal{F} \subset \sigma(Z) & (4) \\
& \Rightarrow \mathbb{E}[h(X) \mid \mathcal{F}] = \mathbb{E}[h(X) \mid Z] & (5)
\end{aligned}$$

So

$$\begin{aligned}
& \mathbb{E}[h(X) \mid Y, \mathcal{F}] \\
&= \mathbb{E}[\mathbb{E}[h(X) \mid Y, Z] \mid Y, \mathcal{F}] \text{ Tower Property} \\
&= \mathbb{E}[\mathbb{E}[h(X) \mid Z] \mid Y, \mathcal{F}] \text{ by (1)} \\
&= \mathbb{E}[\mathbb{E}[h(X) \mid \mathcal{F}] \mid Y, \mathcal{F}] \text{ by (5)} \\
&= \mathbb{E}[h(X) \mid \mathcal{F}] \text{ Tower Property}
\end{aligned}$$

So by 2(c) we have  $X, Y$  are conditionally independent given  $\mathcal{F}$ .  $\square$

**Problem 4.** Super martingale

*Proof.* Given  $X_n, Y_n$  submartingale, we have:

$$\begin{aligned}\mathbb{E}[X_{n+1} + Y_{n+1} \mid \mathcal{F}_n] &= \mathbb{E}[X_{n+1} \mid \mathcal{F}_n] + \mathbb{E}[Y_{n+1} \mid \mathcal{F}_n] \\ &\geq X_n + Y_n\end{aligned}$$

So  $X_n + Y_n$  is a submartingale.

$$\begin{aligned}\mathbb{E}[\max(X_{n+1}, Y_{n+1}) \mid \mathcal{F}_n] &\geq \mathbb{E}[X_{n+1} \mid \mathcal{F}_n] \geq X_n \\ \mathbb{E}[\max(X_{n+1}, Y_{n+1}) \mid \mathcal{F}_n] &\geq \mathbb{E}[Y_{n+1} \mid \mathcal{F}_n] \geq Y_n \\ \Rightarrow \mathbb{E}[\max(X_{n+1}, Y_{n+1}) \mid \mathcal{F}_n] &\geq \max(X_n, Y_n)\end{aligned}$$

So  $\max(X_n, Y_n)$  is a submartingale. □

**Problem 5.** Counter Example

*Proof.* Let  $W_i$  be i.i.d with  $\mathbb{P}[W_i = 1] = \mathbb{P}[W_i = -1] = 1/2$ . Let  $X_n = -\sum_{i=1}^n W_i, Y_n = \sum_{i=1}^{n+1} W_i$ . Then  $X_n$  is a martingale w.r.t.  $\sigma(W_1, \dots, W_n)$ , thus it is a submartingale.  $Y_n$  is a martingale w.r.t.  $\sigma(W_1, \dots, W_{n+1})$ , thus it is a submartingale.

We will prove by contradiction, assuming that  $X_n + Y_n = W_{n+1}$  is a submartingale. Since  $\sup_{n \geq 0} \mathbb{E}[W_n^+] = 1 < \infty$ , by the Doob's first martingale convergence theorem, we have  $W_n$  converges pointwise to a random variable. But  $W_n$  are i.i.d. non constant. Thus we have contradiction. So  $W_{n+1}$  is not a submartingale. □