Solution Set 11

Problem 1: Properties

Let $(\Omega, \mathbb{F}, \mathbb{P})$ be a probability space, X be an integrable random variable defined on this space and let \mathcal{G} be a sub- σ -field of \mathbb{F} . Relying only on the definition of conditional expectation, show the following properties:

- a) $\mathbb{E}(\mathbb{E}(X|\mathcal{G})) = \mathbb{E}(X)$.
- b) If X is independent of \mathcal{G} , then $\mathbb{E}(X|\mathcal{G}) = \mathbb{E}(X)$ a.s.
- c) If X is \mathcal{G} -measurable, then $\mathbb{E}(X|\mathcal{G}) = X$ a.s.
- d) If Y is \mathcal{G} -measurable and bounded, then $\mathbb{E}(XY|\mathcal{G}) = \mathbb{E}(X|\mathcal{G})Y$ a.s.
- e) If \mathcal{H} is a sub- σ -field of \mathcal{G} , then $\mathbb{E}(\mathbb{E}(X|\mathcal{H})|\mathcal{G}) = \mathbb{E}(X|\mathcal{H}) = \mathbb{E}(\mathbb{E}(X|\mathcal{G})|\mathcal{H})$ a.s.

Hint for parts b) to e): According to the course definition, in order to check that some candidate random variable Z is the conditional expectation of X given \mathcal{G} , you should check the following two conditions:

- (i) Z is \mathcal{G} -measurable;
- (ii) Z satisfies $\mathbb{E}((Z-X)U)=0$ for every U \mathcal{G} -measurable and bounded.

Solution a) Use part (ii) of the definition with $U \equiv 1$ (such a U belongs to G).

- b) (i) $Z = \mathbb{E}(X)$ is constant and therefore \mathcal{G} -measurable; (ii) Let $U \in G$: $\mathbb{E}(XU) = \mathbb{E}(X)\mathbb{E}(U) = \mathbb{E}(X)U$ (using the independence of X and U and the linearity of expectation).
- c) (i) Z = X is \mathcal{G} -measurable by assumption; (ii) Let $U \in G$: $\mathbb{E}(XU) = \mathbb{E}(ZU)$!
- d) (i) $Z = \mathbb{E}(X|\mathcal{G}) Y$ is \mathcal{G} -measurable; (ii) Let $U \in G$: $\mathbb{E}(XYU) = \mathbb{E}(\mathbb{E}(X|\mathcal{G}) YU)$, because part (ii) of the definition of $\mathbb{E}(X|\mathcal{G})$ implies the previous equality (indeed, $YU \in G$). Therefore, $\mathbb{E}(XYU) = \mathbb{E}(ZU)$.
- e) Let us first check the left-hand side equality: $\mathbb{E}(X|\mathcal{H})$ is \mathcal{H} -measurable, therefore \mathcal{G} -measurable, so one can apply property c).

For the right-hand side equality, one has: (i) $Z = \mathbb{E}(X|\mathcal{H})$ is \mathcal{H} -measurable; (ii) Let $U \in \mathcal{H}$:

$$\mathbb{E}(\mathbb{E}(X|\mathcal{G})\,U) = \mathbb{E}(\mathbb{E}(XU|\mathcal{G})) = \mathbb{E}(XU) = \mathbb{E}(\mathbb{E}(X|\mathcal{H})\,U) = \mathbb{E}(ZU)$$

using successively d), a) and the definition of $\mathbb{E}(X|\mathcal{H})$.

Problem 2: Discrete conditional expectation

Let X, Y be two discrete random variables (with values in a countable set C). Let us moreover assume that X is integrable.

a) Show that the random variable $\psi(Y)$, where ψ is defined as

$$\psi(y) = \sum_{x \in C} x \, \mathbb{P}(\{X=x\} | \{Y=y\})$$

matches the definition of conditional expectation $\mathbb{E}(X|Y)$ given in the lectures.

b) Application: One rolls two independent and balanced dice (say Y and Z), each with four faces. What is the conditional expectation of the maximum of the two, given the value of one of them?

Solution a) We must check that $\mathbb{E}(\psi(Y) g(Y)) = \mathbb{E}(X g(Y))$ for any continuous and bounded function g. The computation gives indeed:

$$\mathbb{E}(\psi(Y)\,g(Y)) = \sum_{y \in C} \psi(y)\;g(y)\;\mathbb{P}(\{Y = y\}) = \sum_{x,y \in C} x\;g(y)\;\mathbb{P}(\{X = x, Y = y\}) = \mathbb{E}(X\,g(Y))$$

b) Let Y and Z be the two independent dice rolls: $\mathbb{P}(\{Y=i\}) = \mathbb{P}(\{Z=j\}) = 0.25$ and $\mathbb{P}(\{Y=i\}) = \mathbb{P}(\{Y=i\}) = \mathbb{P}(\{Y=i\}) = \mathbb{P}(\{Z=j\})$. We therefore have $\mathbb{E}(\max(Y,Z)|Y) = \psi(Y)$, where

$$\begin{split} \psi(i) &=& \sum_{j=i}^{4} \max(i,j) \, \mathbb{P}(\{\max(Y,Z)=j\} | \{Y=i\}) = \sum_{j=i}^{4} \max(i,j) \, \frac{\mathbb{P}(\{\max(Y,Z)=j,Y=i\})}{\mathbb{P}(\{Y=i\})} \\ &=& i \, \frac{\mathbb{P}(\{Z \leq i,Y=i\})}{\mathbb{P}(\{Y=i\})} + \sum_{j=i+1}^{4} j \, \frac{\mathbb{P}(\{Z=j,Y=i\})}{\mathbb{P}(\{Y=i\})} = i \, \mathbb{P}(\{Z \leq i\}) + \sum_{j=i+1}^{4} j \, \mathbb{P}(\{Z=j\}) \end{split}$$

So $\psi(1) = 2.5$, $\psi(2) = 2.75$, $\psi(3) = 3.25$ and $\psi(4) = 4$.

Problem 3: Zombie apocalypse

A zombie bites N people during its lifetime (un-death time?), where N has the Poisson distribution with parameter λ . Once a person is bitten, they become a zombie with probability p or their brain is eaten with probability q = 1 - p, independently of all other bitten individuals. Let Z be the number of zombie offsprings created by a zombie.

- a) Find $\mathbb{E}(Z|N)$ and $\mathbb{E}(Z)$
- b) Find $\mathbb{E}(N|Z)$
- c) Optional: Assume that the initial zombie infestation is small (relative to the population of the planet). According to the above model, for which values of p and λ would you expect the infestation to turn into a full blown apocalypse and for which values would it die down locally?

Solution a) From the previous Exercise,

$$\psi(n) = \sum_{k>0} k \mathbb{P}(\{Z=k\} | \{N=n\}) = \sum_{k>1}^{n} k \binom{n}{k} p^k (1-p)^{n-k} = pn$$

and so $\mathbb{E}(Z|N) = pN$. We also have that

$$\mathbb{E}(Z) = \mathbb{E}(\mathbb{E}(Z|N)) = \mathbb{E}(pN) = p\lambda.$$

b) We have that

$$\mathbb{P}(\{N=k\}|\{Z=z\}) = \frac{\mathbb{P}(\{N=k,Z=z\})}{\{Z=z\})} = \frac{\binom{k}{z}p^zq^{k-z}(\lambda^k/k!)e^{-\lambda}}{\sum_{m>z}\binom{m}{z}p^zq^{m-z}(\lambda^m/m!)e^{-\lambda}} = \frac{(q\lambda)^{k-z}}{(k-z)!}e^{-q\lambda}$$

From the previous Exercise,

$$\psi(z) = \sum_{k \ge 0} k \mathbb{P}(\{N = k\} | \{Z = z\}) = \sum_{k \ge z}^{n} k \frac{(q\lambda)^{k-z}}{(k-z)!} e^{-q\lambda} = z + q\lambda$$

and so $\mathbb{E}(N|Z) = Z + q\lambda$.

c) Intuitively, we would like each zombie to produce less than one offspring, so the infestation will die down if $p\lambda < 1$. However, even if this condition is met, we will also need λ to be small (relative to the population of the planet). Otherwise, we risk having a few very hungry zombies eating the brains of some significant percentage of world's population before disappearing themselves.