## AMS 261: Probability Theory (Fall 2017)

## Homework 5 solutions

- 1. Let X be a random variable defined on a probability space  $(\Omega, \mathcal{F}, P)$  and taking values in a measurable space  $(\Psi, \mathcal{G})$ , where  $\mathcal{G}$  is the  $\sigma$ -field on space  $\Psi$ . Consider the collection  $\mathcal{A}$  of subsets of  $\Omega$  consisting of  $X^{-1}(B)$  for all  $B \in \mathcal{G}$ . Show that  $\mathcal{A}$  is a  $\sigma$ -field on  $\Omega$ .
  - **Solution:** First, note that because X is a random variable, we know that  $A \subseteq \mathcal{F}$ . To show that A is a  $\sigma$ -field, we need to verify the three conditions of the definition of a  $\sigma$ -field. First, because  $\Psi \in \mathcal{G}$ , we have  $X^{-1}(\Psi) = \Omega \in \mathcal{A}$ . Next, consider  $A \in \mathcal{A}$ . We have  $A = X^{-1}(B)$  for some  $B \in \mathcal{G}$ . Using properties of inverse images,  $X^{-1}(B^c) = (X^{-1}(B))^c = A^c$ . Because  $\mathcal{G}$  is a  $\sigma$ -field, we have  $B^c \in \mathcal{G}$ , which implies that  $X^{-1}(B^c) \in \mathcal{A}$ , and therefore  $A^c \in \mathcal{A}$ . Finally, let  $\{A_n : n = 1, 2, ...\}$  be a countable collection of members of  $\mathcal{A}$ . For each n,  $A_n = X^{-1}(B_n)$  for  $B_n \in \mathcal{G}$ . Now,  $\bigcup_{n=1}^{\infty} B_n \in \mathcal{G}$ , since  $\mathcal{G}$  is a  $\sigma$ -field. Hence,  $X^{-1}(\bigcup_{n=1}^{\infty} B_n) \in \mathcal{A}$ , which yields the third condition, since  $X^{-1}(\bigcup_{n=1}^{\infty} B_n) = \bigcup_{n=1}^{\infty} X^{-1}(B_n) = \bigcup_{n=1}^{\infty} A_n$ , using again properties of inverse images.
- 2. For k=1,2,..., consider random variables  $X_k:(\Omega,\mathcal{F},P)\to (\Psi_k,\mathcal{G}_k)$  and measurable functions  $\varphi_k:(\Psi_k,\mathcal{G}_k)\to (\Theta_k,\mathcal{H}_k)$ . Assume that the countable sequence of random variables  $\{X_k:k=1,2,...\}$  is independent. Prove that the sequence  $\{\varphi_k\circ X_k:k=1,2,...\}$  is independent.

**Solution:** We are given that  $\{X_k : k = 1, 2, ...\}$  is independent, i.e.,  $\{\sigma(X_k) : k = 1, 2, ...\}$  is independent, i.e., for any finite index set J (with  $J \subset \{1, 2, ...\}$ ),  $\{\sigma(X_j) : j \in J\}$  is independent, which implies that for any  $B_j \in \mathcal{G}_j$ ,

$$P(\bigcap_{j \in J} X_j^{-1}(B_j)) = \prod_{j \in J} P(X_j^{-1}(B_j)).$$
(2.1)

Consider an arbitrary finite index set J and  $C_j \in \mathcal{H}_j$ . We have

$$P(\bigcap_{j\in J}(\varphi_j\circ X_j)^{-1}(C_j))=P(\bigcap_{j\in J}X_j^{-1}(\varphi_j^{-1}(C_j)))=\prod_{j\in J}P(X_j^{-1}(\varphi_j^{-1}(C_j)))=\prod_{j\in J}P((\varphi_j\circ X_j)^{-1}(C_j))$$

using (2.1) (with  $B_j = \varphi_j^{-1}(C_j)$ ). Hence,  $\{\sigma(\varphi_j \circ X_j) : j \in J\}$  is independent for any finite index set J, and therefore  $\{\sigma(\varphi_k \circ X_k) : k = 1, 2, ...\}$  is independent.

3. Let  $\{A_n: n=1,2,...\}$  be a countable independent sequence of events on a probability space  $(\Omega, \mathcal{F}, P)$ . Prove that  $P(\bigcap_{n=1}^{\infty} A_n) = \prod_{n=1}^{\infty} P(A_n)$ . (Note: For a countable sequence of reals,  $\{b_n: n=1,2,...\}$ , the infinite product  $\prod_{n=1}^{\infty} b_n$  is defined by  $\lim_{n\to\infty} \prod_{k=1}^n b_k$ , provided this limit exists.)

Solution: Consider the new sequence of events  $\{B_n: n=1,2,...\}$ , where  $B_n = \bigcap_{k=1}^n A_k$ . This is a decreasing sequence of events with  $\lim_{n\to\infty} B_n = \bigcap_{n=1}^{\infty} B_n = \bigcap_{n=1}^{\infty} \bigcap_{k=1}^n A_k = \bigcap_{n=1}^{\infty} A_n$ . Therefore, using continuity of probability measure and the assumption of independence for  $\{A_n: n=1,2,...\}$ , we have

$$P(\bigcap_{n=1}^{\infty} A_n) = P(\lim_{n \to \infty} B_n) = \lim_{n \to \infty} P(B_n) = \lim_{n \to \infty} P(\bigcap_{k=1}^n A_k) = \lim_{n \to \infty} \prod_{k=1}^n P(A_k) = \prod_{n=1}^{\infty} P(A_n).$$

Note that for the sequence  $q_n = \prod_{k=1}^n P(A_k)$  we have  $1 \ge q_n \ge q_{n+1} \ge ... \ge 0$ , and therefore the infinite product  $\prod_{n=1}^{\infty} P(A_n) = \lim_{n \to \infty} q_n$  exists as either a strictly positive constant or 0.

4. Consider two countable sequences of events,  $\{A_n:n=1,2,...\}$  and  $\{B_n:n=1,2,...\}$ , on the same probability space  $(\Omega,\mathcal{F},P)$ . Assume that, for each n,  $A_n$  and  $B_n$  are independent. Moreover, assume that  $A=\lim_{n\to\infty}A_n$  and  $B=\lim_{n\to\infty}B_n$  exist. Show that A and B are independent. Solution: We have  $\lim_{n\to\infty}1_{A_n}(\omega)=1_A(\omega)$  and  $\lim_{n\to\infty}1_{B_n}(\omega)=1_B(\omega)$ , for each  $\omega\in\Omega$ . Therefore  $1_{A\cap B}(\omega)=1_A(\omega)1_B(\omega)=\lim_{n\to\infty}(1_{A_n}(\omega)1_{B_n}(\omega))=\lim_{n\to\infty}1_{A_n\cap B_n}(\omega)$ , for each  $\omega\in\Omega$ , and thus  $\lim_{n\to\infty}(A_n\cap B_n)=A\cap B$ . Hence,

$$P(A \cap B) = P(\lim_{n \to \infty} (A_n \cap B_n)) = \lim_{n \to \infty} P(A_n \cap B_n) = \lim_{n \to \infty} (P(A_n)P(B_n))$$
  
=  $(\lim_{n \to \infty} P(A_n))(\lim_{n \to \infty} P(B_n)) = P(\lim_{n \to \infty} A_n)P(\lim_{n \to \infty} B_n) = P(A)P(B)$ 

using continuity of probability measure (twice) and the independence of  $A_n$  and  $B_n$ , for each n.

5. A sequence  $\{X_n: n=1,2,...\}$  of  $\mathbb{R}$ -valued random variables, defined on a common probability space  $(\Omega,\mathcal{F},P)$ , is said to converge completely if for any  $k=1,2,...,\sum_{n=1}^{\infty}P(|X_n|>k^{-1})<\infty$ . Show that if  $\{X_n: n=1,2,...\}$  converges completely, then  $\lim_{n\to\infty}X_n=0$  almost surely.

Solution: The assumption of complete convergence yields that

$$P(\limsup_{n\to\infty}\{\omega\in\Omega:|X_n(\omega)|>k^{-1}\})=0, \text{ for } k=1,2,...$$

using the Borel lemma. Now the result follows using one of the equivalent definitions of almost sure convergence proved in class.

- 6. Construct a sequence  $\{X_n:n=1,2,...\}$  of  $\mathbb{R}^+$ -valued random variables (i.e.,  $X_n\geq 0$ , for all n) that satisfies  $\sum_{n=1}^{\infty}P(X_n>k^{-1})<\infty$ , for any k=1,2,..., but for which  $\lim_{n\to\infty}\mathrm{E}(X_n)\neq 0$ . Solution: For each n=1,2,..., define  $X_n$  so that it takes value  $3^n$  with probability  $2^{-n}$ , and value 0 with probability  $1-2^{-n}$ . (For example,  $X_n$  can be defined on  $\Omega=(0,1]$ , with  $\mathcal F$  the Borel  $\sigma$ -field on (0,1] and P the uniform distribution, such that  $X_n(\omega)=3^n$ , if  $\omega\in(0,2^{-n}]$ , and  $X_n(\omega)=0$ , otherwise.) Then, for any k=1,2,...,  $\sum_{n=1}^{\infty}P(X_n>k^{-1})=\sum_{n=1}^{\infty}2^{-n}=1<\infty$ , but  $\lim_{n\to\infty}\mathrm{E}(X_n)=\lim_{n\to\infty}(3/2)^n=\infty$ .
- 7. Consider a countable sequence  $\{X_n: n=1,2,...\}$  of random variables defined on a common probability space  $(\Omega, \mathcal{F}, P)$ . Assume that each random variable  $X_n$  is uniformly distributed on (0,1), hence,  $P(c < X_n < d) \equiv P(\{\omega \in \Omega: X_n(\omega) \in (c,d)\}) = d-c$ , for any  $0 \le c < d \le 1$ . Show that the sequence  $\{1/(n^2X_n): n=1,2,...\}$  converges almost surely to 0 as  $n \to \infty$ .

Solution: We need to show that  $P\left(\left\{\omega \in \Omega : \forall k, \exists j, \forall n \geq j, \frac{1}{n^2 X_n(\omega)} < \frac{1}{k}\right\}\right) = 1$ , or, equivalently, that

$$P\left(\bigcap_{k=1}^{\infty}\bigcup_{j=1}^{\infty}\bigcap_{n=j}^{\infty}\left\{\omega\in\Omega:\frac{1}{n^{2}X_{n}(\omega)}<\frac{1}{k}\right\}\right)=P\left(\bigcap_{k=1}^{\infty}\liminf_{j\to\infty}A_{j,k}\right)=1,$$

or, equivalently, that

$$0 = P\left(\left(\bigcap_{k=1}^{\infty} \liminf_{j \to \infty} A_{j,k}\right)^{c}\right) = P\left(\bigcup_{k=1}^{\infty} \left(\liminf_{j \to \infty} A_{j,k}\right)^{c}\right) = P\left(\bigcup_{k=1}^{\infty} \limsup_{j \to \infty} A_{j,k}^{c}\right)$$
(7.1)

Here, for each j=1,2,..., k=1,2,...,  $A_{j,k}$  is the event  $\{\omega\in\Omega:\frac{1}{j^2X_j(\omega)}<\frac{1}{k}\}.$ 

Now, if we fix k, there exists some M=M(k) such that  $k/j^2<1$ , for any  $j\geq M$ . Then, for any such  $j\geq M$ ,

$$P\left(A_{j,k}^c\right) = P\left(\left\{\omega \in \Omega: \frac{1}{j^2 X_j(\omega)} \geq \frac{1}{k}\right\}\right) = P\left(\left\{\omega \in \Omega: X_j(\omega) \leq \frac{k}{j^2}\right\}\right) = \frac{k}{j^2},$$

since each  $X_j$  is uniformly distributed on (0,1). Hence, the series  $\sum_{j=1}^{\infty} P(A_{j,k}^c)$  converges, since  $\sum_{j=M}^{\infty} P(A_{j,k}^c) = \sum_{j=1}^{\infty} P(A_{j,k}^c)$ 

 $k \sum_{i=M}^{\infty} j^{-2} < \infty$ . Therefore, the Borel lemma yields that  $P(\limsup_{j\to\infty} A_{j,k}^c) = 0$ , for any k. Finally,

(7.1) is established if we note that 
$$P\left(\bigcup_{k=1}^{\infty} \limsup_{j\to\infty} A_{j,k}^{c}\right) \leq \sum_{k=1}^{\infty} P(\limsup_{j\to\infty} A_{j,k}^{c}) = 0.$$