## Chapter 6

## Exercises

1. Let  $X_1, X_2 \in \mathcal{L}^2$  and define  $Cov(X_1, X_2) = E([X_1 - E(X_1)][X_2 - E(X_2)])$ . Show that  $Cov(X_1, X_2) = E(X_1X_2) - E(X_1)E(X_2)$  and that if  $X_1$  and  $X_2$  are independent  $Cov(X_1, X_2) = 0$ .

**Answer:** From the definition of  $Cov(X_1, X_2)$  and linearity of expectations

$$Cov(X_1, X_2) = E(X_1 X_2 - X_1 E(X_2) - X_2 E(X_1) + E(X_1) E(X_2)) = E(X_1 X_2) - E(X_1) E(X_2).$$

Independence of  $X_1$  and  $X_2$  implies that  $E(X_1X_2) = E(X_1)E(X_2)$ . Hence,  $Cov(X_1, X_2) = 0$ .

2. Let  $\{X_n\}_{n\in\mathbb{N}}$  be a sequence of random variables that are independent and share the same continuous distribution. Let p be a permutation of  $\{1, \dots, n\}$  for  $n \in \mathbb{N}$ . Show that  $(X_1, \dots, X_n)$  and  $(X_{p(1)}, \dots, X_{p(n)})$  have the same distribution.

**Answer:** Since the random variables are independent and have the same distribution, say F,

$$P(X_1 \le x_1, \cdots, X_n \le x_n) = \prod_{i=1}^n F(x_i)$$

for all  $x_i \in \mathbb{R}$ . If  $\{p(i)\}_{i=1}^n$  is a permutation of  $\{1, \dots, n\}$ , then

$$P(X_{p(1)} \le x_1, \dots, X_{p(n)} \le x_n) = \prod_{i=1}^n F(x_i).$$

Hence,  $(X_1, \dots, X_n)$  and  $X_{p(1)}, \dots, X_{p(n)}$  have the same distribution.

3. Let I be a finite index set and consider the collection of  $\sigma$ -algebras  $\{\mathcal{B}_i\}_{i\in I}$ . Show that this collection is independent if, and only if, for every choice of non-negative  $\mathcal{B}_i$ -measurable random variable  $X_i$ , we have  $E\left(\prod_{i\in I}X_i\right)=\prod_{i\in I}E(X_i)$ .

**Answer:** If  $E(\prod_{i\in I} X_i) = \prod_{i\in I} E(X_i)$  whenever  $X_i \in \mathcal{B}_i$ , then for any  $A_i \in \mathcal{B}_i$ , take  $X_i = I_{A_i}$  and

$$E\left(\prod_{i\in I} X_i\right) = P\left(\bigcap_{i\in I} A_i\right) = \prod_{i\in I} P(A_i) = \prod_{i\in I} E(X_i)$$

and consequently  $\{A_i\}_{i\in I}$  are independent and  $\{\mathcal{B}_i\}_{i\in I}$  are independent  $\sigma$ -algebras.

Now, suppose  $\{\mathcal{B}_i\}_{i\in I}$  are independent  $\sigma$ -algebras. For  $A_i \in \mathcal{B}_i$  and  $X_i = I_{A_i}$  we have

$$E\left(\prod_{i\in I}X_i\right) = \prod_{i\in I}E(X_i).$$

Now, if  $\{X_i\}_{i\in I}$  are simple, then write  $X_i = \sum_j x_{ij} I_{A_{ij}}$  for  $A_{ij} \in \mathcal{B}_i$ . Then, we have

$$E\left(\prod_{i \in I} X_{i}\right) = E\left(\prod_{i \in I} \sum_{j(i)} x_{ij(i)} I_{A_{ij(i)}}\right) = E\left(\sum_{j(i), i \in I} \prod_{i \in I} x_{ij(i)} I_{\bigcap_{i \in I} A_{ij(i)}}\right)$$

$$= \sum_{j(i), i \in I} \prod_{i \in I} x_{ij(i)} P\left(\bigcap_{i \in I} A_{ij(i)}\right) = \sum_{j(i), i \in I} \prod_{i \in I} x_{ij(i)} P(A_{ij(i)})$$

$$= \prod_{i \in I} \sum_{j(i)} x_{ij(i)} P(A_{ij(i)}) = \prod_{i \in I} E(X_{i}).$$

If  $X_i$  is a non-negative  $\mathcal{B}_i$ -measurable function, there exists  $X_i^{(n)}$  such that  $X_i^{(n)}$  is simple and  $0 < X_i^{(n)} \uparrow X_i$ . Then, it follows that  $\prod_{i \in I} X_i^{(n)} \uparrow \prod_{i \in I} X_i$  and by the monotone convergence theorem  $E\left(\prod_{i \in I} X_i^{(n)}\right) \uparrow E\left(\prod_{i \in I} X_i\right)$  and from the previous argument, the left side is  $\prod_{i \in I} E\left(X_i^{(n)}\right) \uparrow \prod_{i \in I} E\left(X_i\right)$  again using the monotone convergence theorem.

4. If E is an event that is independent of the  $\pi$ -system P and  $E \in \sigma(P)$ , then P(E) is either 0 or 1.

**Answer:** Set  $C_1 = E$  and  $C_2 = P$  and it follows that  $C_1$  is independent of  $C_2$ . This implies that  $\sigma(C_1)$  is independent of  $\sigma(C_2)$ . Therefore, E is independent of E and P(E) = 0 or 1.

5. Let  $\{A_i\}_{i=1}^n$  be independent events. Show that  $P(\bigcup_{i=1}^n A_i) = 1 - \prod_{i=1}^n P(A_i^c)$ .

**Answer:** By De Morgan's Law  $(\bigcup_{i=1}^n A_i)^c = \bigcap_{i=1}^n A_i^c$ . Hence,

$$P\left(\left(\bigcup_{i=1}^{n} A_{i}\right)^{c}\right) = 1 - P\left(\bigcup_{i=1}^{n} A_{i}\right) = P\left(\bigcap_{i=1}^{n} A_{i}^{c}\right)$$

which implies

$$P\left(\bigcup_{i=1}^{n} A_{i}\right) = 1 - P\left(\bigcap_{i=1}^{n} A_{i}^{c}\right) = 1 - \bigcap_{i=1}^{n} P(A_{i}^{c})$$

where the last equality follows from the fact that if  $\{A_i\}_{i=1}^n$  are independent events, so are  $\{A_i^c\}_{i=1}^n$ .

6. We have proved that if X and Y are independent, then f(X) and g(Y) are independent if f and g are measurable. Is it possible to have X and Y be dependent and f(X) and g(Y) be independent? If so, give an example, if not, prove.

**Answer:** Yes, it is possible. Consider two independent random variables  $X_1$  and  $X_2$  and another random variable W that is independent of  $X_1$  and  $X_2$  and takes on the values 1 and -1 with probability 1/2 each. Now, define two new random variables  $X = WX_1$  and  $Y = WX_2$ . X and Y are functionally connected and cannot be independent. However,  $X^2 = X_1^2$  and  $Y^2 = X_2^2$ , which are independent since  $X_1$  and  $X_2$  are independent.