

# Chapter 1

## Fundamental Concepts

**Exercise 1.1:** Consider a sample space  $S$  and assume for a given subset  $B$  that  $P(B) > 0$ . Show that the conditional probability

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad (1.1)$$

satisfies the axioms of probability.

**Exercise 1.2:** Show that

$$P(A \cup B) = P(A) + P(B) - P(A \cap B).$$

(Express  $A \cup B$  as the union of three disjoint sets.)

**Exercise 1.3:** A beam of particles consists of a fraction  $10^{-4}$  electrons and the rest photons. The particles pass through a double-layered detector which gives signals in either zero, one or both layers. The probabilities of these outcomes for electrons ( $e$ ) and photons ( $\gamma$ ) are

$$\begin{array}{ll} P(0|e) = 0.001 & \text{and} \quad P(0|\gamma) = 0.99899 \\ P(1|e) = 0.01 & P(1|\gamma) = 0.001 \\ P(2|e) = 0.989 & P(2|\gamma) = 10^{-5}. \end{array}$$

- (a) What is the probability for a particle detected in one layer only to be a photon?
- (b) What is the probability for a particle detected in both layers to be an electron?

**Exercise 1.4:** Suppose a random variable  $x$  has the p.d.f.  $f(x)$ . Show that the p.d.f. for  $y = x^2$  is

$$g(y) = \frac{1}{2\sqrt{y}}f(\sqrt{y}) + \frac{1}{2\sqrt{y}}f(-\sqrt{y}). \quad (1.2)$$

**Exercise 1.5:** Suppose two independent random variables  $x$  and  $y$  are both uniformly distributed between zero and one, i.e. the p.d.f.  $g(x)$  is given by

$$g(x) = \begin{cases} 1 & 0 < x < 1 \\ 0 & \text{otherwise,} \end{cases} \quad (1.3)$$

and similarly for the p.d.f.  $h(y)$ .

(a) Using SDA equation (1.35), show that the p.d.f.  $f(z)$  for  $z = xy$  is

$$f(z) = \begin{cases} -\log z & 0 < z < 1 \\ 0 & \text{otherwise.} \end{cases} \quad (1.4)$$

(b) Find the same result using SDA equations (1.37) and (1.38) by defining an additional function,  $u = x$ . First, find the joint p.d.f. of  $z$  and  $u$ . Integrate this over  $u$  to find the p.d.f. for  $z$ .

(c) Show that the cumulative distribution of  $z$  is

$$F(z) = z(1 - \log z). \quad (1.5)$$

**Exercise 1.6:** Consider a random variable  $x$  and constants  $\alpha$  and  $\beta$ . Show that

$$\begin{aligned} E[\alpha x + \beta] &= \alpha E[x] + \beta, \\ V[\alpha x + \beta] &= \alpha^2 V[x]. \end{aligned} \quad (1.6)$$

**Exercise 1.7:** Consider two random variables  $x$  and  $y$ .

(a) Show that the variance of  $\alpha x + y$  is given by

$$\begin{aligned} V[\alpha x + y] &= \alpha^2 V[x] + V[y] + 2\alpha \text{cov}[x, y] \\ &= \alpha^2 V[x] + V[y] + 2\alpha \rho \sigma_x \sigma_y, \end{aligned} \quad (1.7)$$

where  $\alpha$  is any constant value,  $\sigma_x^2 = V[x]$ ,  $\sigma_y^2 = V[y]$ , and the correlation coefficient is  $\rho = \text{cov}[x, y]/\sigma_x \sigma_y$ .

(b) Using the result of (a), show that the correlation coefficient always lies in the range  $-1 \leq \rho \leq 1$ . (Use the fact that the variance  $V[\alpha x + y]$  is always greater than or equal to zero and consider the cases  $\alpha = \pm \sigma_y/\sigma_x$ .)

**Exercise 1.8:** Suppose  $\mathbf{x} = (x_1, \dots, x_n)$  is described by the joint p.d.f.  $f(\mathbf{x})$ , and the variables  $\mathbf{y} = (y_1, \dots, y_n)$  are defined by means of a linear transformation,

$$y_i = \sum_{j=1}^n A_{ij} x_j. \quad (1.8)$$

Assume that the inverse transformation  $\mathbf{x} = A^{-1}\mathbf{y}$  exists.

(a) Show that the joint p.d.f. for  $\mathbf{y}$  is given by

$$g(\mathbf{y}) = f(A^{-1}\mathbf{y}) |\det(A^{-1})|. \quad (1.9)$$

(b) Find  $g(\mathbf{y})$  for the case where  $A$  is orthogonal, i.e.  $A^{-1} = A^T$ .