

1. Let X be a continuous random variable with pdf f_X . Let $Y = g(X)$, where g is a strictly increasing function. Use the pdf of Y derived in class to show that

$$E[Y] = \int_{-\infty}^{\infty} g(x) f_X(x) dx.$$

2. X is said to be a *standard normal random variable* if its pdf is

$$f_X(x) = (1/\sqrt{2\pi})e^{-x^2/2}.$$

Find the pdf of $Y = X^2$, and find its mean and variance.

3. (a) Let X be a discrete random variable whose range is the nonnegative integers. Show that

$$EX = \sum_{k=0}^{\infty} (1 - F_X(k)),$$

where $F_X(k) = P(X \leq k)$.

- (b) Let X be a continuous, nonnegative random variable, i.e., $f(x) = 0$ for $x < 0$. Show that

$$EX = \int_0^{\infty} (1 - F_X(x)) dx,$$

where $F_X(x)$ is the cdf of X . Compare this with part (a).

- (c) Let X be a continuous random variable. Use part (b) to show that

$$EX = \int_0^{\infty} (1 - F_X(x)) dx - \int_{-\infty}^0 F_X(x) dx,$$

where $F_X(x)$ is the cdf of X .

- (d) Let X and Y be two nonnegative random variables. Show that

$$E[XY] = \int_0^{\infty} \int_0^{\infty} P(X > x, Y > y) dx dy.$$

4. Let (X, Y, Z) be a random vector with the following density:

$$f(x, y, z) = \frac{1 - \sin x \sin y \sin z}{8\pi^3}, \quad 0 \leq x, y, z \leq 2\pi.$$

Show that (X, Y, Z) are pairwise independent but not independent.

5. X and Y are independent exponential random variables with rate λ and μ . Define

$$Z = \min\{X, Y\}, \quad W = \begin{cases} 1 & Z = X \\ 0 & Z = Y \end{cases}.$$

- (a) Find the joint distribution of Z and W .

- (b) Find the marginal distribution of Z and W , respectively.
- (c) Show that Z and W are independent.
6. A random variable X has a beta distribution if $f_X(x) = C(\alpha, \beta)x^{\alpha-1}(1-x)^{\beta-1}$, for $x \in [0, 1]$. C is a constant factor so that $f_X(x)$ integrates to one. We know that $E[X] = \frac{\alpha}{\alpha+\beta}$ and $\text{Var } X = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$.

Now consider the following hierarchy model. Let $(X_1, P_1), \dots, (X_n, P_n)$ be independent random vectors with

$$\begin{aligned} X_i | P_i &\sim \text{Bernoulli}(P_i) \\ P_i &\sim \text{beta}(\alpha, \beta) \end{aligned}$$

This model might be appropriate, for example, if we are measuring the success of a drug on n patients and because the patients are different, their success rates are not constant. A random variable of interest is $Y = \sum_{i=1}^n X_i$, the total number of successes.

- (a) Show that $E[Y] = n\alpha/(\alpha + \beta)$.
- (b) Show that $\text{Var}(Y) = n\alpha\beta/(\alpha + \beta)^2$, and hence Y has the same mean and variance as a $\text{binomial}(n, \alpha/(\alpha + \beta))$. What is the distribution of Y ?
- (c) Suppose now that the model is

$$\begin{aligned} X_i | P_i &\sim \text{binomial}(n_i, P_i), \quad i = 1, \dots, k \\ P_i &\sim \text{beta}(\alpha, \beta) \end{aligned}$$

For $Y = \sum_{i=1}^k X_i$, find $E[Y]$ and $\text{Var}(Y)$.