

Week 10: Bivariate continuous random variables

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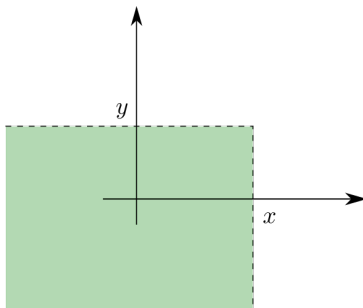
Thursday class

- Let X, Y be two random variables (do not have to be discrete) defined on the same set of outcomes S .

Definition

The joint cdf of X, Y is called the following function

$$F(x, y) = P(X \leq x, Y \leq y).$$



Definition

Bivariate random variable (X, Y) is called **continuous** if there exists a function $f : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ such that

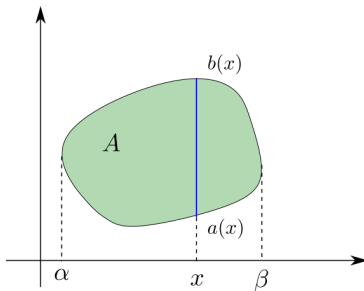
$$F(x, y) = \int_{-\infty}^x \int_{-\infty}^y f(\xi, \eta) \, d\xi \, d\eta.$$

$f(x, y)$ is called the **joint pdf** of X and Y .

Theorem

For any set $A \subset \mathbb{R} \times \mathbb{R}$,

$$P[(X, Y) \in A] = \iint_A f(x, y) \, dx \, dy.$$



Here

$$\iint_A f(x, y) \, dx \, dy = \int_{\alpha}^{\beta} \int_{a(x)}^{b(x)} f(x, y) \, dx \, dy$$

Theorem

A function $f : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ is the joint pdf of a bivariate continuous random variable if and only if

1. $f(x, y) \geq 0$,
2. $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \, dx \, dy = 1$

Definition

For any $x \in \mathbb{R}$ and $y \in \mathbb{R}$ define

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) \, dy, \quad f_Y(y) = \int_{-\infty}^{\infty} f(x, y) \, dx.$$

$f_X(x)$ and $f_Y(y)$ are called the **marginal pdf** of X and Y correspondingly.

► $f_X(x) \geq 0, f_Y(y) \geq 0$



$$\int_{-\infty}^{\infty} f_X(x) \, dx = 1, \quad \int_{-\infty}^{\infty} f_Y(y) \, dy = 1.$$



$$P(X \leq x) = \int_{-\infty}^x f_X(\xi) \, d\xi, \quad P(Y \leq y) = \int_{-\infty}^y f_Y(\eta) \, d\eta.$$

Definition

Let (X, Y) be a continuous bivariate random variable. X and Y are called independent if

$$f(x, y) = f_X(x)f_Y(y).$$

- If X and Y are independent then for any sets $A \subset \mathbb{R}$ and $B \subset \mathbb{R}$,

$$P(X \in A, Y \in B) = P(X \in A) \cdot P(Y \in B).$$

Exercise 1

Problem (4.4-1 in the textbook)

Let $f(x, y) = (3/16)xy^2$, $0 \leq x \leq 2, 0 \leq y \leq 2$, be the joint pdf of X and Y .

- (a) Find $f_X(x)$ and $f_Y(y)$, the marginal probability density functions.
- (b) Are the two random variables independent? Why or why not?
- (c) Compute the means and variances of X and Y .
- (d) Find $P(X \leq Y)$.

Solution

- (a) From definition

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_0^2 \frac{3}{16} xy^2 dy = \frac{x}{16} 2^3 - \frac{x}{16} 0^3 = \frac{x}{2}.$$

$$f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_0^2 \frac{3}{16} xy^2 dx = \frac{3y^2}{16} \frac{2^2}{2} - \frac{3y^2}{16} \frac{0^2}{2} = \frac{3y^2}{8}.$$

Solution

(b) *Independent, because*

$$f_X(x)f_Y(y) = \frac{3xy^2}{16} = f(x, y).$$

(c)

$$E[X] = \int_{-\infty}^{\infty} xf_X(x) \, dx = \int_0^2 \frac{x^2}{2} \, dx = \frac{2^3}{6} - \frac{0^3}{6} \boxed{= \frac{4}{3}}.$$

$$E[X^2] = \int_{-\infty}^{\infty} x^2 f_X(x) \, dx = \int_0^2 \frac{x^3}{2} \, dx = \frac{2^4}{8} - \frac{0^4}{8} = 2.$$

$$\text{Var}(X) = E[X^2] - E[X]^2 = 2 - \frac{16}{9} \boxed{= \frac{2}{9}}.$$

Solution

(d)

$$\begin{aligned} P(X \leq Y) &= \iint_{x \leq y} f(x, y) \, dx \, dy = \int_{-\infty}^{\infty} \int_{-\infty}^y f(x, y) \, dx \, dy \\ &= \int_0^2 \int_0^y \frac{3}{16} xy^2 \, dx \, dy = \int_0^2 \left[\frac{3y^2}{16} \frac{y^2}{2} - \frac{3y^2}{16} \frac{0^2}{2} \right] dy \\ &= \int_0^2 \frac{3y^4}{32} \, dy = \frac{3 \cdot 2^5}{32 \cdot 5} - \frac{3 \cdot 0^5}{32 \cdot 5} \\ &= \frac{3}{5}. \end{aligned}$$

- For any $u : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$,

$$E[u(X, Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} u(x, y) f(x, y) dx dy.$$

- **Covariance:**

$$\text{Cov}(X, Y) = E[(X - \mu_X)(Y - \mu_Y)] = E[XY] - \mu_X \mu_Y.$$

- **Correlation:**

$$\rho = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}.$$

- **Least squares line:**

$$y = \rho \frac{\sigma_Y}{\sigma_X} (y - \mu_X) + \mu_Y.$$

- It minimizes the

$$g(a, b) = E[(Y - aX - b)^2].$$

Definition

Conditional pdf of Y given $X = x$ is called the pdf defined by

$$h(y|x) = \frac{f(x, y)}{f_X(x)}.$$

Definition

Conditional pdf of X given $Y = y$ is called the pdf defined by

$$g(x|y) = \frac{f(x, y)}{f_Y(y)}.$$

Definition

Conditional mean and variance of Y given $X = x$ are defined

$$E[Y|x] = \int_{-\infty}^{\infty} y h(y|x) dy, \quad \text{Var}[Y|x] = \int_{-\infty}^{\infty} (y - E[Y|x])^2 h(y|x) dy.$$

- ▶ $E[X|y], \text{Var}(X|y)$ are defined similarly.
- ▶ $m(x) = E[Y|x]$ minimizes (among all functions $m : \mathbb{R} \rightarrow \mathbb{R}$)

$$E[(Y - m(X))^2].$$

- ▶ If $E[Y|x]$ is a linear function, then

$$E[Y|x] = \rho \frac{\sigma_Y}{\sigma_X} (y - \mu_X) + \mu_Y.$$

Exercise 2

Problem (4.4-17, modified)

Let $f(x, y) = c$, $0 \leq x \leq 4$, $x^2 - x/2 \leq y \leq x^2 + x/2$, be the joint pdf of X and Y .

- (a) Find c .
- (b) Sketch the region for which $f(x, y) > 0$.
- (c) Find $f_X(x)$, the marginal pdf of X .
- (d) Calculate and plot the least squares line.
- (e) Determine $h(y|x)$, the conditional pdf of Y , given that $X = x$.
- (f) Calculate and plot $E(Y|x)$, the conditional mean of Y , given that $X = x$.

Solution

(a)

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_{x^2 - x/2}^{x^2 + x/2} c dy = c(x^2 + x/2) - c(x^2 - x/2) = cx$$

$$1 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dy dx = \int_{-\infty}^{\infty} f_X(x) dx = \int_0^4 cx dx = c \frac{4^2}{2} - c \frac{0^2}{2} = 8c \Rightarrow \boxed{c = \frac{1}{8}}$$

(b) See below.

(c) $f_X(x) = \frac{x}{8}$.

(d)

$$\mu_X = \int_{-\infty}^{\infty} x f_X(x) dx = \int_0^4 x \frac{x}{8} dx = \frac{1}{8} \frac{4^3}{3} \boxed{= \frac{8}{3}}.$$

$$\begin{aligned} \mu_Y &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(x, y) dx dy = \int_0^4 \int_{x^2 - x/2}^{x^2 + x/2} y \frac{1}{8} dy dx \\ &= \frac{1}{8} \int_0^4 \left[\frac{1}{2} (x^2 + x/2)^2 - \frac{1}{2} (x^2 - x/2)^2 \right] dy dx \\ &= \frac{1}{8} \int_0^4 x^3 dy dx = \frac{1}{8} \frac{4^4}{4} \boxed{= 8} \end{aligned}$$

(d)

$$E[X^2] = \int_{-\infty}^{\infty} x^2 f_X(x) dx = \int_0^4 x^2 \frac{x}{8} dx = \frac{1}{8} \frac{4^4}{4} = 8.$$

$$\text{Var}(X) = E[X^2] - \mu_X^2 = 8 - \frac{64}{9} \boxed{= \frac{8}{9}}$$

$$\begin{aligned} E[XY] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x, y) dx dy = \int_0^4 \int_{x^2-x/2}^{x^2+x/2} xy \frac{1}{8} dy dx \\ &= \frac{1}{8} \int_0^4 x \left[\frac{1}{2} (x^2 + x/2)^2 - \frac{1}{2} (x^2 - x/2)^2 \right] dy dx \\ &= \frac{1}{8} \int_0^4 x^4 dy dx = \frac{1}{8} \frac{4^5}{5} = \frac{128}{5} \end{aligned}$$

$$\text{Cov}(X, Y) = E[XY] - \mu_X \mu_Y = \frac{128}{5} - \frac{64}{3} \boxed{= \frac{64}{15}}$$

(d)

$$\boxed{y = \frac{\text{Cov}(X, Y)}{\text{Var}(X)}(x - \mu_X) + \mu_Y} = \frac{128}{5} \frac{9}{8} \left(x - \frac{8}{3}\right) + 8 = \frac{144}{5} \left(x - \frac{8}{3}\right) + 8$$

(e)

$$h(y|x) = \frac{f(x, y)}{f_X(x)} = \frac{c}{cx} = \frac{1}{x}.$$

(f)

$$\begin{aligned} E[Y|x] &= \int_{-\infty}^{\infty} y h(y|x) dy = \int_{x^2-x/2}^{x^2+x/2} y \frac{1}{x} dy \\ &= \frac{1}{x} \left[\frac{1}{2} (x^2 + x/2)^2 - \frac{1}{2} (x^2 - x/2)^2 \right] \\ &= \frac{1}{x} x^3 \\ &= x^2. \end{aligned}$$

Fitting least squares line and conditional mean to a distribution

