

MATH180C: Introduction to Stochastic Processes II

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Today: Conditioning on continuous
random variables

> Q&A: October 30

Next: PK 7.1, Durrett 3.1

This week:

- Homework 3 (due Saturday, October 31, 11:59 PM)

Conditioning on continuous r.v.

Def. Let X and Y be jointly distributed continuous random variables with joint probability density function $f_{X,Y}(x,y)$. We call the function

$$f_{X|Y}(x|y) := \frac{f_{X,Y}(x,y)}{f_Y(y)} \quad \text{if } f_Y(y) > 0$$

the **conditional probability density** function of X given $Y=y$.

The function $F_{X|Y}(x|y) = \int_{-\infty}^x f_{X|Y}(s|y) ds$

is called **conditional distribution** of X given $Y=y$

Conditional expectation

Def. Let X and Y be jointly distributed continuous random variables, let $f_{X|Y}(x|y)$ be a conditional distribution of X given $Y=y$ and let $g: \mathbb{R} \rightarrow \mathbb{R}$ be a function for which $E(|g(X)|) < \infty$.

Then we call

$$E(g(X) | Y=y) := \int_{-\infty}^{+\infty} g(x) f_{X|Y}(x|y) dx \quad \text{if } f_Y(y) > 0$$

the **conditional expectation** of $g(X)$ given $Y=y$.

In particular, if $g(x) = \mathbb{1}_A(x)$ indicator of set A , then

$$E(\mathbb{1}_A(X) | Y=y) = P(X \in A | Y=y) = \int_A g(x) f_{X|Y}(x|y) dx$$

Remark

If Y is a continuous random variable, then

$$P(Y=y) = 0 \text{ for all } y \in \mathbb{R}$$

Therefore, we cannot define $P(X \in A | Y=y)$ as

$$P(X \in A | Y=y) = \frac{P(X \in A, Y=y)}{P(Y=y)}$$

On the other hand consider example:

X, Y i.i.d. r.v., $X, Y \sim \text{Unif}[0,1]$. Define $Z = X - Y$

If $Y = \frac{1}{2}$, $Z \sim \text{Unif}[-\frac{1}{2}, \frac{1}{2}]$ makes perfect sense

Intuitive explanation / derivation

$$P(X \in [x, x + \Delta x], Y \in [y, y + \Delta y]) = f_{X,Y}(x,y) \Delta x \Delta y + o(\Delta x \Delta y)$$

Using the multiplication rule ($f_Y(y) > 0$ on $[y, y+\Delta y]$)

$$P(X \in [x, x+\Delta x], Y \in [y, y+\Delta y]) = P(X \in [x, x+\Delta x] | Y \in [y, y+\Delta y]) P(Y \in [y, y+\Delta y])$$

$$\frac{P(X \in [x, x+\Delta x] | Y \in [y, y+\Delta y])}{\Delta x} \underset{\Delta x \rightarrow 0}{=} \frac{P(X \in [x, x+\Delta x], Y \in [y, y+\Delta y])}{\frac{P(Y \in [y, y+\Delta y])}{\Delta y} \Delta x \Delta y} \underset{\substack{\Delta x \rightarrow 0 \\ \Delta y \rightarrow 0}}{=} \frac{f_{X,Y}(x,y)}{f_Y(y)}$$

Properties of conditional probability/expectation

$$\begin{aligned} 1) \quad P(a < X < b, c < Y < d) &= \int_c^d \left(\int_a^b f_{X|Y}(x|y) dx \right) f_Y(y) dy \\ &= \int_c^d P(X \in (a, b) | Y=y) f_Y(y) dy \end{aligned}$$

$$\begin{aligned} 2) \quad P(a < X < b) &= \int_{-\infty}^{+\infty} \left(\int_a^b f_{X|Y}(x|y) dx \right) f_Y(y) dy \\ &= \int_{-\infty}^{+\infty} P(X \in (a, b) | Y=y) f_Y(y) dy \end{aligned}$$

$$3) \quad E(g(X)) = \int_{-\infty}^{+\infty} E(g(X) | Y=y) f_Y(y) dy$$

Further properties of conditional expectation (PK, p.50)

$$4) E(c_1 g_1(X_1) + c_2 g_2(X_2) | Y=y) = c_1 E(g_1(X_1) | Y=y) + c_2 E(g_2(X_2) | Y=y)$$

$$5) E(v(X, Y) | Y=y) = E(v(X, y) | Y=y)$$

In particular, $E(v(X, Y)) = \int_{-\infty}^{+\infty} E(v(X, y) | Y=y) f_Y(y) dy$

$$\begin{aligned} 6) E(g(X)h(Y)) &= \int_{-\infty}^{+\infty} h(y) E(g(X) | Y=y) f_Y(y) dy \\ &= E(h(Y) E(g(X) | Y)) \end{aligned}$$

$$7) E(g(X) | Y=y) = E(g(X)) \text{ if } X \text{ and } Y \text{ are independent}$$

Example 1

Let (X, Y) be jointly continuous r.v.s with density $f_{X,Y}(x,y) = \frac{1}{y} e^{-\frac{x}{y}-y}$, $x, y > 0$

Compute the conditional density of X given $Y=y$.

1) Compute the marginal density of Y

$$f_Y(y) = \int_0^{\infty} \frac{1}{y} e^{-\frac{x}{y}-y} dx = e^{-y} \int_0^{\infty} \frac{1}{y} e^{-\frac{x}{y}} dx = e^{-y} \quad (Y \sim \text{Exp}(1))$$

2) Compute the conditional density

$$f_{X|Y}(x|y) = \frac{\frac{1}{y} e^{-\frac{x}{y}-y}}{e^{-y}} = \frac{1}{y} e^{-\frac{x}{y}} \quad \Bigg| \Rightarrow \begin{array}{l} \text{given } Y=y \\ X \sim \text{Exp}(\frac{1}{y}) \end{array}$$

Example 1 (cont.)

Suppose that $Y \sim \text{Exp}(1)$, and X has exponential distribution with parameter $\frac{1}{y}$. Compute $E(X)$

First, $E(X | Y=y) = y$, and using property 3)

$$\begin{aligned} E(X) &= \int_0^{\infty} E(X | Y=y) f_Y(y) dy \\ &= \int_0^{\infty} y e^{-y} dy = 1 \end{aligned}$$

$$\begin{aligned} E(X) &= \int_0^{\infty} \int_0^{\infty} x \frac{1}{y} e^{-\frac{x}{y} - y} dx dy = \int_0^{\infty} \left(\int_0^{\infty} \frac{x}{y} e^{-\frac{x}{y}} dx \right) e^{-y} dy \stackrel{= E(X|Y=y)}{=} \\ &= \int_0^{\infty} y e^{-y} dy = 1 \end{aligned}$$

Example 2: continuous and discrete r.v.s

Let $N \in \mathbb{N}$, $P \sim \text{Unif}[0,1]$, $X \sim \text{Bin}(N, P)$

What is the distribution of X ?

$$P(X=k) = \int_0^1 P(X=k | P=s) f_P(s) ds$$

$$= \int_0^1 P(X=k | P=s) ds$$

$$= \int_0^1 \frac{N!}{k! (N-k)!} \cdot s^k (1-s)^{N-k} ds$$

$$= \frac{N!}{k! (N-k)!} \cdot \frac{k! (N-k)!}{(N+1)!} = \frac{1}{N+1}$$

$\Rightarrow X$ is uniformly distributed $\{0, 1, \dots, N\}$

Example 3

Let X and Y be i.i.d. $\text{Exp}(\lambda)$ r.v.

Define $Z = \frac{X}{Y}$. Compute the density of Z .

- If $X \sim \text{Exp}(\lambda)$, then for $\alpha > 0$ $\alpha X \sim \text{Exp}(\frac{\lambda}{\alpha})$

$$P(\alpha X > t) = P(X > \frac{t}{\alpha}) = e^{-\lambda \frac{t}{\alpha}} \Rightarrow \alpha X \sim \text{Exp}(\frac{\lambda}{\alpha})$$

- $$\begin{aligned} P(Z > t) &= \int_0^{\infty} P(Z > t \mid Y=y) f_Y(y) dy \\ &= \int_0^{\infty} P(\frac{1}{y} X > t) \lambda e^{-\lambda y} dy \\ &= \int_0^{\infty} \lambda e^{-\lambda y t - \lambda y} dy = \lambda \int_0^{\infty} e^{-\lambda(t+1)y} dy = \frac{1}{1+t} \end{aligned}$$

$$f_Z(t) = \frac{1}{(1+t)^2}$$