

Math 505  
Some Notes on Continuous Random Variables

**Definition 0.1.** A random variable  $X$  is said to be *continuous* if there exists a function  $f : \mathbb{R} \rightarrow [0, \infty)$  such that, for any real numbers  $a < b$ ,

$$P(a < X < b) = \int_a^b f(x) dx.$$

The function  $f$  is called the *probability density function* (p.d.f.) of  $X$ .

Any p.d.f. must satisfy these properties

- $\int_{-\infty}^{\infty} f(x) dx = 1$
- $f(x) \geq 0$  for (almost) every  $x \in \mathbb{R}$ .<sup>1</sup>

**Example 0.2.** Let  $\theta > 0$ . Then  $X$  has an *exponential distribution* with parameter  $\theta$  if its p.d.f. is

$$f(x) = \begin{cases} \frac{1}{\theta} e^{-x/\theta}, & \text{if } x > 0 \\ 0, & \text{if } x \leq 0. \end{cases}$$

For instance, if  $X$  has an exponential distribution with  $\theta = 5$ , then

$$P(2 < X < 7) = \int_2^7 \frac{1}{5} e^{-x/5} dx = e^{-2/5} - e^{-7/5} = 0.424.$$

**Definition 0.3.** Suppose  $X$  is a continuous random variable with p.d.f.  $f$ . Then  $X$  is *integrable* if

$$\int_{-\infty}^{\infty} |x|f(x) dx \text{ is finite.}$$

If  $X$  is integrable, then its *expected value* is

$$E(X) = \int_{-\infty}^{\infty} xf(x) dx.$$

If  $X$  is *not integrable*, then its expected value does not exist.

**Example 0.4.** If  $X$  has an exponential distribution with parameter  $\theta$ , then

$$\begin{aligned} \int_{-\infty}^{\infty} |x|f(x) dx &= \int_{-\infty}^0 |x|f(x) dx + \int_0^{\infty} |x|f(x) dx \\ &= \int_{-\infty}^0 |x| \cdot 0 dx + \int_0^{\infty} x \cdot \frac{1}{\theta} e^{-x/\theta} dx \\ &= \theta. \end{aligned}$$

In the second integral,  $0 < x < \infty$ , so  $|x| = x$ , and we integrate by parts to finish the calculation. This shows that  $\int_{-\infty}^{\infty} |x|f(x) dx$  is finite, so  $X$  is integrable. Its expected value is

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<sup>1</sup> $f$  could be negative on a set of “measure zero”, such as a finite set or a countably infinite set, but there is always a version of  $f$  that is nonnegative for every  $x \in \mathbb{R}$ , so this technicality is usually not important.

$$E(X) = \int_{-\infty}^{\infty} xf(x) dx = \theta,$$

using the same approach.

**Example 0.5.** Let  $X$  have a *Cauchy distribution*, given by the p.d.f.

$$f(x) = \frac{1}{\pi(1+x^2)}, \text{ for } x \in \mathbb{R}.$$

Then

$$\begin{aligned} \int_{-\infty}^{\infty} |x|f(x) dx &= \int_{-\infty}^{\infty} \frac{|x|}{\pi(1+x^2)} dx \\ &= 2 \int_0^{\infty} \frac{x}{\pi(1+x^2)} dx = \infty. \end{aligned}$$

Here, we are using symmetry of the function to rewrite the integral, which can be calculated with the substitution  $u = 1+x^2$ . Since this integral is infinite, a random variable with a Cauchy distribution is not integrable. Therefore, its expected value does not exist, by definition.

**Definition 0.6.** Suppose  $X$  and  $Y$  are two random variables, and  $f : \mathbb{R}^2 \rightarrow [0, \infty)$ , such that for all real numbers  $a < b$  and  $c < d$ ,

$$P(a < X < b \text{ and } c < Y < d) = \int_a^b \int_c^d f(x, y) dy dx.$$

Then  $f$  is called the *joint probability density function* of  $X$  and  $Y$ .

Any joint p.d.f. must satisfy these properties

- $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dy dx = 1$
- $f(x, y) \geq 0$  for (almost) every pair  $(x, y) \in \mathbb{R}^2$ .

**Example 0.7.** Let  $X$  and  $Y$  have joint p.d.f.

$$f(x, y) = \begin{cases} \frac{1}{2}e^{-2x} & \text{if } 0 < x < \infty, -e^x < y < e^x \\ 0 & \text{otherwise.} \end{cases}$$

This function is nonnegative, so to verify that it's a p.d.f., we perform the following calculation

$$\begin{aligned} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dy dx &= \int_0^{\infty} \int_{-e^x}^{e^x} \frac{1}{2}e^{-2x} dy dx \\ &= \int_0^{\infty} \left[ \frac{1}{2}e^{-2x}y \right]_{-e^x}^{e^x} dx \\ &= \int_0^{\infty} \frac{1}{2}[e^{-2x}e^x - (-e^{-2x}e^x)] dx \\ &= \int_0^{\infty} e^{-x} dx = 1. \end{aligned}$$

**Proposition 0.8.** If  $X$  and  $Y$  have a joint p.d.f.  $f$ , then the (marginal) p.d.f. of  $X$  is

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

**Example 0.9.** Continuing the above example, the marginal p.d.f. of  $X$  is

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

If  $x \leq 0$ , then  $f(x, y) = 0$ , the above integral will be zero, and  $f_X(x) = 0$ . If  $x > 0$ , then this integral is

$$f_X(x) = \int_{-e^x}^{e^x} \frac{1}{2} e^{-2x} dy = e^{-x}.$$

Therefore, the marginal p.d.f. of  $X$  is

$$f_X(x) = \begin{cases} e^{-x}, & \text{if } x > 0 \\ 0, & \text{if } x \leq 0, \end{cases}$$

so  $X$  has an exponential distribution with parameter  $\theta = 1$ .

**Definition 0.10.** Suppose  $X$  and  $Y$  have joint p.d.f.  $f$  and  $X$  has marginal p.d.f.  $f_X$ . Then the conditional p.d.f. of  $Y$  given  $X = x$  is

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

The conditional expected value of  $Y$  given  $X = x$  is then

$$E(Y | X = x) = \int_{-\infty}^{\infty} y f(y | x) dy,$$

assuming that the integral *converges absolutely*, meaning

$$\int_{-\infty}^{\infty} |y| f(y | x) dy \text{ is finite.}$$

**Example 0.11.** For the above example,

$$(0.1) \quad f(y | x) = \frac{f(x, y)}{f_X(x)} = \frac{\frac{1}{2} e^{-2x}}{e^{-x}} = \frac{1}{2} e^{-x}, \text{ for } 0 < x < \infty, -e^x < y < e^x$$

$$E(Y | X = x) = \int_{-\infty}^{\infty} y f(y | x) dy = \int_{-e^x}^{e^x} y \cdot \frac{1}{2} e^{-x} dy = 0, \text{ for } 0 < x < \infty.$$

It's easy to check that this integral converges absolutely, so the conditional expected value does indeed exist. Let's take a moment to consider what this means.

The p.d.f. of  $X$  is only positive when  $x > 0$ , so with probability 1,  $X$  is positive. Assuming  $X$  has taken a particular value  $x$ , which is positive, the p.d.f. for  $Y$  is given in equation (0.1). Now, this p.d.f. does not depend at all on  $y$ . In other words, the conditional distribution of  $Y$  given  $X = x$  is uniform on the interval  $(-e^x, e^x)$ . Since an interval of this form is symmetric about 0, it makes sense that the conditional expected value of  $Y$  given  $X = x$  is 0, regardless of what  $x$  is.

What about  $E(Y)$ , the unconditional expected value of  $Y$ ? This only exists if  $Y$  is integrable, meaning that the following integral is finite.

$$\begin{aligned}
 \int_{-\infty}^{\infty} |y| f_Y(y) dy &= \int_{-\infty}^{\infty} |y| \left[ \int_{-\infty}^{\infty} f(x, y) dx \right] dy \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |y| f(x, y) dy dx \\
 &= \int_0^{\infty} \int_{-e^x}^{e^x} |y| \cdot \frac{1}{2} e^{-2x} dy dx \\
 &= \int_0^{\infty} \frac{1}{2} e^{-2x} \int_{-e^x}^{e^x} |y| dy dx \\
 &= \int_0^{\infty} \frac{1}{2} e^{-2x} \cdot 2 \int_0^{e^x} y dy dx \\
 &= \int_0^{\infty} \frac{1}{2} e^{-2x} e^{2x} dx \\
 &= \int_0^{\infty} \frac{1}{2} dx = \infty.
 \end{aligned}$$

Therefore, we have just seen an example where  $E(Y | X) = 0$  for all possible values of  $X$ , but  $Y$  is not integrable, so  $E(Y)$  does not exist.