

## Multivariate distributions

### Calculating conditional PDF

Let  $f(x, y) = 15x^2y$  for  $0 \leq x \leq y \leq 1$ . Find  $f(x|y)$ .<sup>1</sup>

**Solution:**

$$\begin{aligned} f(y) &= \int_0^y f(x, y) dx \\ &= \int_0^y 15x^2y dx \\ &= 15y \int_0^y x^2 dx \\ &= 15y \left. \frac{x^3}{3} \right|_0^y \\ &= \frac{15y^4}{3} \\ f(x|y) &= \frac{f(x, y)}{f(y)} \\ &= \frac{15x^2y}{15y^4/3} \\ &= \frac{3x^2}{y^3} \end{aligned}$$

### Properties of a joint PDF

Continuous random variables  $X$  and  $Y$  have the following joint probability density function (PDF):<sup>2</sup>

$$f_{XY}(x, y) = \begin{cases} kx^2y^3 & \text{where } 0 < x, y < 6 \\ 0 & \text{otherwise} \end{cases}$$

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<sup>1</sup>Grimmer HW12.4

<sup>2</sup>Grimmer HW12.1

Note:  $0 < x, y < 6$  means that both  $x$  and  $y$  are between 0 and 6; it does not mean that  $x$  is greater than 0 and  $y$  is less than 6.

a. Find  $k$ .

**Solution:**

$$\begin{aligned}
 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) \, dx \, dy &= 1 \\
 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} kx^2y^3 \, dx \, dy &= 1 \\
 \int_0^6 \int_0^6 kx^2y^3 \, dx \, dy &= 1 \\
 k \int_0^6 \int_0^6 x^2y^3 \, dx \, dy &= 1 \\
 k \int_0^6 y^3 \cdot \frac{x^3}{3} \Big|_0^6 \, dy &= 1 \\
 k \int_0^6 y^3 \cdot \left( \frac{6^3}{3} - 0 \right) \, dy &= 1 \\
 72k \int_0^6 y^3 \, dy &= 1 \\
 72k \cdot \frac{y^4}{4} \Big|_0^6 &= 1 \\
 72k \left( \frac{6^4}{4} - 0 \right) &= 1 \\
 72k \cdot 324 &= 1 \\
 23328k &= 1 \\
 k &= \frac{1}{23328}
 \end{aligned}$$

b. Find the marginal PDF of  $X$ ,  $f_X(x)$ .

**Solution:**

$$\begin{aligned}
f_X(x) &= \int_{-\infty}^{\infty} f_{XY}(x, y) dy \\
&= \int_{-\infty}^{\infty} \frac{x^2 y^3}{23328} dy \\
&= \int_0^6 \frac{x^2 y^3}{23328} dy \\
&= \frac{x^2}{23328} \int_0^6 y^3 dy \\
&= \frac{x^2}{23328} \cdot \frac{y^4}{4} \Big|_0^6 \\
&= \frac{x^2}{23328} \cdot \frac{6^4}{4} \\
&= \frac{x^2}{72}
\end{aligned}$$

c. Find the marginal PDF of  $Y$ ,  $f_Y(y)$ .

**Solution:**

$$\begin{aligned}
f_Y(y) &= \int_{-\infty}^{\infty} f(x, y) dx \\
&= \frac{1}{23328} \int_0^6 x^2 y^3 dx \\
&= \frac{1}{23328} \cdot \frac{x^3}{3} \Big|_0^6 y^3 \\
&= \frac{1}{23328} \left( \frac{216}{3} - 0 \right) y^3 \\
&= \frac{1}{23328} (72) y^3 \\
&= \frac{y^3}{324}
\end{aligned}$$

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d. Find  $E[X]$ .

**Solution:**

$$\begin{aligned} E[X] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f(x, y) dx dy \\ &= \frac{1}{23328} \int_0^6 \int_0^6 x \cdot x^2 y^3 dx dy \\ &= \frac{1}{23328} \times \int_0^6 x^3 dx \times \int_0^6 y^3 dy \\ &= \frac{1}{23328} \times \left. \frac{x^4}{4} \right|_0^6 \times \left. \frac{y^4}{4} \right|_0^6 \\ &= \frac{1}{23328} \left( \frac{1296}{4} - 0 \right) \left( \frac{1296}{4} - 0 \right) \\ &= \frac{1}{23328} (324) (324) \\ &= 4.5 \end{aligned}$$

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e. Find  $E[Y]$ .

**Solution:**

$$\begin{aligned} E[Y] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(x, y) dx dy \\ &= \frac{1}{23328} \int_0^6 \int_0^6 y \cdot x^2 y^3 dx dy \\ &= \frac{1}{23328} \times \int_0^6 x^2 dx \times \int_0^6 y^4 dy \\ &= \frac{1}{23328} \times \left. \frac{x^3}{3} \right|_0^6 \times \left. \frac{y^5}{5} \right|_0^6 \\ &= \frac{1}{23328} \left( \frac{216}{3} - 0 \right) \left( \frac{6^5}{5} - 0 \right) \\ &= \frac{1}{23328} (72) \left( \frac{7776}{5} \right) \\ &= 4.8 \end{aligned}$$

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f. Find  $Var(X)$ .

**Solution:** To find  $Var(X)$ , we first need to find  $E[X^2]$ .

$$\begin{aligned} E[X^2] &= \frac{1}{23328} \int_0^6 \int_0^6 x^2 \cdot x^2 y^3 \, dx \, dy \\ &= \frac{1}{23328} \times \int_0^6 \int_0^6 x^4 y^3 \, dx \, dy \\ &= \frac{1}{23328} \times \int_0^6 x^4 \, dx \times \int_0^6 y^3 \, dy \\ &= \frac{1}{23328} \times \left. \frac{x^5}{5} \right|_0^6 \times \left. \frac{y^4}{4} \right|_0^6 \\ &= \frac{1}{23328} \left( \frac{7776}{5} - 0 \right) \left( \frac{1296}{4} - 0 \right) \\ &= \frac{1}{23328} \left( \frac{7776}{5} \right) (324) \\ &= 21.6 \end{aligned}$$

With  $E[X^2]$  determined, we can now calculate  $Var(X)$ .

$$\begin{aligned} Var(X) &= E[X^2] - E[X]^2 \\ &= 21.6 - (4.5)^2 \\ &= 21.6 - 20.25 \\ &= 1.35 \end{aligned}$$

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a. Find  $Var(Y)$ .

**Solution:** Again, to find  $Var(Y)$ , we first need  $E[Y^2]$ .

$$\begin{aligned} E[Y^2] &= \frac{1}{23328} \int_0^6 \int_0^6 y^2 \cdot x^2 y^3 dx dy \\ &= \frac{1}{23328} \times \int_0^6 \int_0^6 x^2 y^5 dx dy \\ &= \frac{1}{23328} \times \int_0^6 x^2 dx \times \int_0^6 y^5 dy \\ &= \frac{1}{23328} \times \frac{x^3}{3} \Big|_0^6 \times \frac{y^6}{6} \Big|_0^6 \\ &= \frac{1}{23328} \left( \frac{216}{3} - 0 \right) (7776 - 0) \\ &= \frac{1}{23328} (72)(7776) \\ &= 24 \end{aligned}$$

With  $E[Y^2]$  in hand, we can now calculate  $Var(Y)$ .

$$\begin{aligned} Var(Y) &= E[Y^2] - E[Y]^2 \\ &= 24 - (4.8)^2 \\ &= 24 - 23.04 \\ &= 0.96 \end{aligned}$$

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b. Find  $Cov(X, Y)$ .

**Solution:** To find  $Cov(X, Y)$ , we first need  $E[XY]$ .

$$\begin{aligned}
E[XY] &= \frac{1}{23328} \int_0^6 \int_0^6 xy \cdot x^2 y^3 dx dy \\
&= \frac{1}{23328} \times \int_0^6 \int_0^6 x^3 y^4 dx dy \\
&= \frac{1}{23328} \times \int_0^6 x^3 dx \times \int_0^6 y^4 dy \\
&= \frac{1}{23328} \times \frac{x^4}{4} \Big|_0^6 \times \frac{y^5}{5} \Big|_0^6 \\
&= \frac{1}{23328} \left( \frac{1296}{4} - 0 \right) \left( \frac{7776}{5} - 0 \right) \\
&= \frac{1}{23328} (324) \left( \frac{7776}{5} \right) \\
&= 21.6
\end{aligned}$$

Now, we calculate  $Cov(X, Y)$ .

$$\begin{aligned}
Cov(X, Y) &= E[XY] - E[X]E[Y] \\
&= 21.6 - (4.5)(4.8) \\
&= 21.6 - 21.6 \\
&= 0
\end{aligned}$$

c. Are  $X$  and  $Y$  independent? Explain your reasoning using mathematical concepts from the course.

**Solution:**  $X$  and  $Y$  are independent because  $f_{XY}(x, y) = f_X(x)f_Y(y)$  (definition of independence), or in other words, the product of the marginal densities of  $X$  and  $Y$  is equal to the joint density of  $X$  and  $Y$ :

$$f_X(x)f_Y(y) = \frac{x^2}{72} \left( \frac{y^3}{324} \right) = \frac{x^2 y^3}{23328} = f_{XY}(x, y)$$

We **cannot** say that  $X$  and  $Y$  are independent because the covariance is zero. While it is true that independent variables have a covariance of zero, it is not necessarily true that variables with a covariance of zero are independent.

d. What is the PDF of  $X$  conditional on  $Y$ ,  $f_{X|Y}(x|y)$ ?

**Solution:** We've previously shown that  $X$  and  $Y$  are independent. This implies that  $f(x) = f(x|y)$  so the answer is the same as the marginal distribution of  $x$  from part (b),

$$\begin{aligned} f(x|y) &= f(x) \\ &= \frac{x^2}{72} \end{aligned}$$



- a. What is the PDF of  $Y$  conditional on  $X$ ,  $f_{Y|X}(y|x)$ ?

**Solution:** Again since we've already shown that  $X$  and  $Y$  are independent we can just refer back to the answer to (c).

$$\begin{aligned} f(y|x) &= f(y) \\ &= \frac{y^3}{324} \end{aligned}$$

## Properties of joint random variables<sup>3</sup>

Suppose the following:

- $E[D] = 10$
- $E[F] = 4$
- $E[DF] = 8$
- $Var(D) = 60$
- $Var(F) = 60$

- a. What is  $Cov(D, F)$ ?

$$\begin{aligned} Cov(D, F) &= E[DF] - E[D]E[F] \\ &= 8 - (4 \times 10) \\ &= -32 \end{aligned}$$

- b. What is the correlation between  $D$  and  $F$ ?

$$\begin{aligned} Cor(D, F) &= \frac{Cov(D, F)}{\sqrt{Var(F)Var(D)}} \\ &= \frac{-32}{\sqrt{60 \times 60}} \\ &= -0.5333333 \end{aligned}$$

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<sup>3</sup>Grimmer HW12.3

c. Suppose you multiplied  $F$  by 2 to generate a new variable,  $H$ . What is  $Cov(D, H)$ ?

**Solution:** Multiplying  $F$  by 2 increases the magnitude of the covariance between  $D$  and  $H$ .

$$Cov(D, H) = E[DH] - E[D]E[H]$$

$$E[DH] = E[D \times 2F] = \int 2DFf(D, F)d(D, F) = 2 \int DFf(D, F)d(D, F) = 2E[DF] = 16$$

$$E[H] = E[2F] = \int 2Ff(F)dF = 2 \int Ff(F)dF = 2E[F] = 8$$

$$Cov(D, H) = 16 - (8 \times 10) = -64$$

d. What is  $Cor(D, H)$ ? How does this compare to your answer to Part (b) of this question?

$$Var(H) = Var(2F) = 2^2 Var(F) = 4 \times 60 = 240$$

$$\begin{aligned} Cor(D, H) &= \frac{Cov(D, H)}{\sqrt{Var(F)Var(H)}} \\ &= \frac{-64}{\sqrt{60 \times 240}} \\ &= -0.5333333 \end{aligned}$$

This is the same as  $Cor(D, F)$ . In other words, multiplying one of the variables by a constant leaves the correlation between the two variables unchanged. This occurs despite the covariance changing.

- e. Suppose instead that  $Var(D) = 30$ . How would this change  $Cor(D, F)$ ?

**Solution:** The magnitude of the correlation between the variables increases as  $Var(D)$  decreases:

$$\begin{aligned} Cor(D, F) &= \frac{Cov(D, F)}{\sqrt{Var(F)Var(D)}} \\ &= \frac{-32}{\sqrt{60 * 30}} \\ &= -0.7542472 \end{aligned}$$

## Deriving a joint PDF

We start with a stick of length  $l$ . We break it at a point which is chosen according to a uniform distribution and keep the piece, of length  $Y$ , that contains the left end of the stick. We then repeat the same process on the piece that we were left with, and let  $X$  be the length of the remaining piece after breaking for the second time.<sup>4</sup>

- a. Find the joint PDF of  $Y$  and  $X$

**Solution:** We have

$$f_Y(y) = \frac{1}{l}, \forall 0 \leq y \leq l$$

Furthermore, given the value  $y$  of  $Y$ , the random variable  $X$  is uniform in the interval  $[0, y]$ . Therefore

$$f_{X|Y}(x|y) = \frac{1}{y}, \forall 0 \leq x \leq y$$

We conclude that

$$f_{X,Y}(x, y) = f_Y(y)f_{X|Y}(x|y) = \begin{cases} \frac{1}{l} \times \frac{1}{y} & , 0 \leq x \leq y \leq l, \\ 0 & , \text{otherwise} \end{cases}$$

- b. Find the marginal PDF of  $X$

**Solution:** We have

$$f_X(x) = \int f_{X,Y}(x, y) dy = \int_x^l \frac{1}{ly} dy = \frac{1}{l} \log \left( \frac{l}{x} \right), \forall 0 \leq x \leq l$$

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<sup>4</sup>BT 3.21

- c. Use the PDF of  $X$  to evaluate  $E[X]$

**Solution:** We have

$$E[X] = \int_0^l x f_X(x) dx = \int_0^l \frac{x}{l} \log\left(\frac{l}{x}\right) dx = \frac{l}{4}$$

- d. Evaluate  $E[X]$ , by exploiting the relation  $X = Y \times \frac{X}{Y}$

**Solution:** The fraction  $\frac{Y}{l}$  of the stick that is left after the first break, and the further fraction  $\frac{X}{Y}$  of the stick that is left after the second break are independent. Furthermore, the random variables  $Y$  and  $\frac{X}{Y}$  are uniformly distributed over the sets  $[0, l]$  and  $[0, 1]$ , respectively, so that  $E[Y] = \frac{l}{2}$  and  $E\left[\frac{X}{Y}\right] = \frac{1}{2}$ . Thus,

$$E[X] = E[Y]E\left[\frac{X}{Y}\right] = \frac{l}{2} \times \frac{1}{2} = \frac{l}{4}$$

## Continuous Bayes' theorem

Previously, we used Bayes' theorem to link the conditional probability of discrete events  $A$  given  $B$  to the probability of  $B$  given  $A$ . There is an analogous Bayes' theorem that relates the conditional densities of random variables  $X$  and  $\theta$  (below) Prove the continuous Bayes' theorem.<sup>5</sup>

$$f(\theta | X) = \frac{f(X | \theta)f(\theta)}{\int f(X | \theta)f(\theta)d\theta}$$

**Solution:**

Recall the definition of the conditional distribution of two random variables:

$$f_{\theta|X}(\theta | X) = \frac{f(\theta, X)}{f_X(X)}$$

Remember via the “chain rule” of probability that  $f(\theta, X) = f(X | \theta)f_{\theta}(\theta)$ , and via our rule for marginalization,  $f_X(X) = \int f_{X|\theta}(X | \theta)f_{\theta}(\theta)d\theta$ . Substitute these equalities in and we have proven the statement:

$$\begin{aligned} f_{\theta|X}(\theta | X) &= \frac{f(\theta, X)}{f_X(X)} \\ &= \frac{f(X | \theta)f_{\theta}(\theta)}{\int f_{X|\theta}(X | \theta)f_{\theta}(\theta)d\theta} \end{aligned}$$

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<sup>5</sup>Grimmer HW12.5

## **Submission of practice questions**

Submit practice questions for the final exam here: <https://forms.gle/CPo9FMQgQRPePDfN7> Note that we need at least 10 people to submit before there's enough to circulate!