

## Lecture #24: Conditional Distributions for the Bivariate Normal

**Reference.** §5.6 pages 127–130

**Recall.** Suppose that  $\mathbf{X} = (X, Y)' \in \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Lambda})$  with

$$\boldsymbol{\mu} = \begin{bmatrix} \mu_x \\ \mu_y \end{bmatrix} \quad \text{and} \quad \boldsymbol{\Lambda} = \begin{bmatrix} \sigma_x^2 & \rho\sigma_x\sigma_y \\ \rho\sigma_x\sigma_y & \sigma_y^2 \end{bmatrix}.$$

In particular,  $X \in \mathcal{N}(\mu_x, \sigma_x^2)$ ,  $Y \in \mathcal{N}(\mu_y, \sigma_y^2)$ , and  $\text{Corr}(X, Y) = \rho$ . Assuming that  $\det[\boldsymbol{\Lambda}] > 0$ , or equivalently, that  $\sigma_x \neq 0$ ,  $\sigma_y \neq 0$ , and  $-1 < \rho < 1$ , then the conditional distribution of  $Y$  given  $X = x$  is

$$Y|X = x \in \mathcal{N}\left(\mu_y + \rho\frac{\sigma_y}{\sigma_x}(x - \mu_x), \sigma_y^2(1 - \rho^2)\right).$$

In particular,

$$\mathbb{E}(Y|X) = \mu_y + \rho\frac{\sigma_y}{\sigma_x}(X - \mu_x)$$

and

$$\text{Var}(Y|X) = \sigma_y^2(1 - \rho^2).$$

**Example.** Suppose that the random vector  $\mathbf{X} = (X_1, X_2)'$  has density

$$f_{\mathbf{X}}(x_1, x_2) = \frac{1}{2\pi} \exp\left\{-\frac{1}{2}(x_1^2 - 2x_1x_2 + 2x_2^2)\right\}.$$

- (a) Determine the distribution of  $\mathbf{X}$ .
- (b) Determine the distribution of  $X_2$  given  $X_1 = x$ .
- (c) Determine the distribution of  $X_1 - X_2$  given  $X_1 + X_2 = 0$ .

**Solution.** (a) It appears that  $\mathbf{X}$  has the density function of a multivariate normal random vector. Thus, we must determine the mean vector  $\boldsymbol{\mu}$ , the covariance matrix  $\boldsymbol{\Lambda}$ , and verify that the distribution is, in fact, MVN. Clearly  $\boldsymbol{\mu} = \bar{\mathbf{0}}$ . Furthermore,

$$x_1^2 - 2x_1x_2 + 2x_2^2 = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

which implies that

$$\boldsymbol{\Lambda}^{-1} = \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix}.$$

It is easy to invert a  $2 \times 2$  matrix, and so we see immediately that

$$\boldsymbol{\Lambda} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

since  $\det[\boldsymbol{\Lambda}] = 1$ .

Thus, matching the general form of a MVN density function with the function given in this problem, we conclude that

$$\mathbf{X} \in \mathcal{N} \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \right).$$

(b) The conditional density of  $X_2$  given  $X_1 = x$  is

$$f_{X_2|X_1=x}(x_2) = \frac{f_{X_1, X_2}(x, x_2)}{f_{X_1}(x)} = \frac{\frac{1}{2\pi} \exp \left\{ -\frac{1}{2}(x^2 - 2xx_2 + 2x_2^2) \right\}}{\frac{1}{\sqrt{2\pi}} \cdot \frac{1}{\sqrt{2}} \exp \left\{ -\frac{1}{2} \cdot \frac{x^2}{2} \right\}}$$

since  $X_1 \in N(0, 2)$ . Therefore,

$$f_{X_2|X=x}(x_2) = \frac{1}{\sqrt{2\pi}} \cdot \frac{1}{\sqrt{1/2}} \exp \left\{ -\frac{1}{2} \frac{(x_2 - x/2)^2}{1/2} \right\}$$

so that  $X_2|X = x \in N\left(\frac{x}{2}, \frac{1}{2}\right)$ .

(c) Consider  $\mathbf{Y} = (Y_1, Y_2)'$  where  $Y_1 = X_1 - X_2$  and  $Y_2 = X_1 + X_2$ . If we let

$$B = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

so that

$$B\mathbf{X} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = \begin{bmatrix} X_1 - X_2 \\ X_1 + X_2 \end{bmatrix} = \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = \mathbf{Y},$$

then by Theorem 5.3.1, we conclude that  $\mathbf{Y} \in \mathcal{N}(B\boldsymbol{\mu}, B\boldsymbol{\Lambda}B')$  where

$$\mathbb{E}(\mathbf{Y}) = B\boldsymbol{\mu} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

and

$$\text{Cov}(\mathbf{Y}) = B\boldsymbol{\Lambda}B' = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 5 \end{bmatrix}.$$

Since  $\det[\text{Cov}(\mathbf{Y})] = 4$ , we deduce that the density of  $\mathbf{Y}$  is given by

$$f_{\mathbf{Y}}(y_1, y_2) = \frac{1}{4\pi} \exp \left\{ -\frac{5}{8} \left( y_1^2 - \frac{2}{5}y_1y_2 + \frac{1}{5}y_2^2 \right) \right\}.$$

We now find

$$\begin{aligned} f_{Y_1|Y_2=0}(y_1) &= \frac{f_{\mathbf{Y}}(y_1, 0)}{f_{Y_2}(0)} = \frac{\frac{1}{4\pi} \exp \left\{ -\frac{5}{8} \left( y_1^2 - \frac{2}{5}y_1(0) + \frac{1}{5}(0)^2 \right) \right\}}{\frac{1}{\sqrt{10\pi}} \exp \left\{ -\frac{1}{10}(0) \right\}} \\ &= \frac{\sqrt{5}}{2\sqrt{2\pi}} \exp \left\{ -\frac{5}{8}y_1^2 \right\}. \end{aligned}$$

In other words, the distribution of  $Y_1|Y_2 = 0$ , or equivalently,  $X_1 - X_2|X_1 + X_2 = 0$ , is  $\mathcal{N}\left(0, \frac{4}{5}\right)$ .