

# Bi-variate Gaussian RVs (2-D Gaussian)

①

two random variables

$$X \sim N(\mu_x, \sigma_x^2) \text{ then } f_x(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} e^{-\frac{(x-\mu_x)^2}{2\sigma_x^2}}$$

$\uparrow$   $\uparrow$   
 $E[X]$   $\text{Var}(X)$

(X, Y) are said to be "jointly Gaussian" if their pdf is of the following form:

$$\begin{pmatrix} X \\ Y \end{pmatrix} \sim N \left( \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}, \begin{bmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{bmatrix} \right)$$

Annotations:  
 -  $\begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}$  is the mean vector, with  $E[X]$  and  $E[Y]$  pointing to  $\mu_1$  and  $\mu_2$  respectively.  
 - The covariance matrix contains  $\sigma_1^2$  (labeled Var(X)),  $\sigma_2^2$  (labeled Var(Y)), and  $\rho\sigma_1\sigma_2$  (labeled cov(X, Y) and cov(Y, X)).  
 -  $\rho$  is the correlation coefficient between X and Y.

Continuous RVs with pdf:

$$f_{XY}(x, y) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} e^{-\frac{\left[ \left(\frac{x-\mu_1}{\sigma_1}\right)^2 + \frac{2\rho(x-\mu_1)(y-\mu_2)}{\sigma_1\sigma_2} + \left(\frac{y-\mu_2}{\sigma_2}\right)^2 \right]}{2(1-\rho^2)}}$$

$$\begin{bmatrix} \mu_1 = E[X] \\ \mu_2 = E[Y] \end{bmatrix} = E \begin{bmatrix} X \\ Y \end{bmatrix}$$

$$\text{Cov}(X, Y) = \begin{bmatrix} \text{Var}(X) & \text{Cov}(X, Y) \\ \text{Cov}(Y, X) & \text{Var}(Y) \end{bmatrix}$$

Need  $-1 \leq \rho \leq +1$ ,  $\sigma_1 > 0$ ,  $\sigma_2 > 0$ .

## Properties:

(2)

1) Marginals of  $f_{XY}(x,y)$  are  $f_X(x) \rightarrow \text{Gaussian } \mathcal{N}(\mu_1, \sigma_1^2)$   
 $f_Y(y) \rightarrow \text{Gaussian } \mathcal{N}(\mu_2, \sigma_2^2)$

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}}, \quad f_Y(y) = \frac{1}{\sqrt{2\pi\sigma_2^2}} e^{-\frac{(y-\mu_2)^2}{2\sigma_2^2}}$$

2) The pdf of  $Y$  given  $X$  is also Gaussian!

$$f_{Y|X}(y|x) = \frac{1}{\sqrt{2\pi\tilde{\sigma}_2^2}} e^{-\frac{(y-\tilde{\mu}_2)^2}{2\tilde{\sigma}_2^2}} \quad \text{where } \tilde{\mu}_2 = \mu_2 + \frac{\rho\sigma_2}{\sigma_1}(x-\mu_1)$$
$$\tilde{\sigma}_2^2 = \sigma_2^2(1-\rho^2)$$

Get it by  $f_{XY}(x,y) = f_X(x) f_{Y|X}(y|x)$ .

3) The correlation coefficient is  $\rho_{XY} = \rho$ .

4)  $X, Y$  uncorrelated  $\stackrel{\text{iff.}}{\iff}$   $X, Y$  independent  $(f_{XY}(x,y) = f_X(x) f_Y(y))$

NOT TRUE IN GENERAL, HOLDS FOR GAUSSIAN  
RVs!!

EX: Textbook 4.11.6

(3)

Under what conditions on the constants  $a, b, c, d$  is

$$f_{xy}(x, y) = d e^{-(a^2 x^2 + bxy + c^2 y^2)} \quad (*)$$

jointly Gaussian?

Sol:  $d = \frac{1}{2\pi \sigma_1 \sigma_2 \sqrt{1-\rho^2}}$ ,  $a^2 = \frac{1}{2\sigma_1^2(1-\rho^2)}$

$$c^2 = \frac{1}{2\sigma_2^2(1-\rho^2)}, \quad b = \frac{\rho}{\sigma_1 \sigma_2 (1-\rho^2)}$$

Then the above (\*) could be a zero-mean Gaussian if  $a, b, c, d$  satisfy:

$$-1 \leq \rho \leq +1, \quad \sigma_1 > 0, \quad \sigma_2 > 0.$$

What do these conditions imply about  $a, b, c, d$ ?

Solve for  $\sigma_1, \sigma_2$ :  $\sigma_1 = \frac{1}{a\sqrt{2(1-\rho^2)}}$ ,  $\sigma_2 = \frac{1}{c\sqrt{2(1-\rho^2)}}$

Plug  $\sigma_1, \sigma_2$  into equation for  $b$ :

$$b = \frac{\rho \cdot \frac{1}{2(1-\rho^2)} \cdot c}{(1-\rho^2) \sqrt{2(1-\rho^2)} \sqrt{2(1-\rho^2)}} \\ = \frac{2\rho a c}{2(1-\rho^2)}$$

or  $\rho = \frac{b}{2ac}$

conditions.

Since we need  $|\rho| \leq 1$ , we need  
Then, for any  $a, b, c$  that satisfy we can find

$|b| \leq 2ac.$   
 $d^2 = a^2 c^2 - b^2/4.$