

- If  $X_1, X_2$  are independent  $N(0, 1)$  random variables, then  $Y_1 = X_1 - 3X_2 + 2$  is normal with mean  $E(Y_1) = E(X_1) - 3E(X_2) + 2 = 2$  and variance  $\text{var}(Y_1) = \text{var}(X_1 - 3X_2 + 2) = \text{var}(X_1) + 9\text{var}(X_2) - 6\text{cov}(X_1, X_2) = 1 + 9 - 0 = 10$ , and  $Y_2 = 2X_1 - X_2 - 1$  is normal with mean  $E(Y_2) = 2E(X_1) - E(X_2) - 1 = -1$  and variance  $\text{var}(Y_2) = \text{var}(2X_1 - X_2 - 1) = 4\text{var}(X_1) + \text{var}(X_2) - 4\text{cov}(X_1, X_2) = 4 + 1 - 0 = 5$ . Since  $\text{cov}(Y_1, Y_2) = \text{cov}(X_1 - 3X_2 + 2, 2X_1 - X_2 - 1) = 2\text{var}(X_1) - 7\text{cov}(X_1, X_2) + 3\text{var}(X_2) = 2 - 0 + 3 = 5$ , we conclude that  $\mathbf{Y} = (Y_1, Y_2)'$  is multivariate normal  $N(\boldsymbol{\mu}, \boldsymbol{\Lambda})$  where

$$\boldsymbol{\mu} = \begin{pmatrix} 2 \\ -1 \end{pmatrix} \quad \text{and} \quad \boldsymbol{\Lambda} = \begin{pmatrix} 10 & 5 \\ 5 & 5 \end{pmatrix}.$$

- Let

$$B = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \end{pmatrix}$$

so that  $\mathbf{Y} = B\mathbf{X}$ . By Theorem 3.1,  $\mathbf{Y}$  is MVN with mean

$$B\boldsymbol{\mu} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \end{pmatrix} \begin{pmatrix} 3 \\ 4 \\ -3 \end{pmatrix} = \begin{pmatrix} 0 \\ 8 \end{pmatrix}$$

and covariance matrix

$$B\boldsymbol{\Lambda}B' = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \end{pmatrix} \begin{pmatrix} 2 & 1 & 3 \\ 1 & 4 & -2 \\ 3 & -2 & 8 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 2 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 16 & -2 \\ -2 & 16 \end{pmatrix}.$$

- Let

$$B = \begin{pmatrix} 1 & 0 & -1 \\ 2 & 1 & -2 \\ -2 & 0 & 3 \end{pmatrix}$$

so that  $\mathbf{Y} = B\mathbf{X}$ . By Theorem 3.1,  $\mathbf{Y}$  is MVN with mean

$$B\boldsymbol{\mu} = \begin{pmatrix} 1 & 0 & -1 \\ 2 & 1 & -2 \\ -2 & 0 & 3 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

and covariance matrix

$$B\boldsymbol{\Lambda}B' = \begin{pmatrix} 1 & 0 & -1 \\ 2 & 1 & -2 \\ -2 & 0 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & -2 \\ 0 & 1 & 0 \\ -1 & -2 & 3 \end{pmatrix} = \begin{pmatrix} 2 & 4 & -5 \\ 4 & 9 & -10 \\ -5 & -10 & 13 \end{pmatrix}.$$

- Let

$$B = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}$$

so that  $\mathbf{Y} = B\mathbf{X}$ . By Theorem 3.1,  $\mathbf{Y}$  is MVN with mean

$$B\boldsymbol{\mu} = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

and covariance matrix

$$B\Lambda B' = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix}.$$

**Exercise 4.2, page 127:** If  $\phi(t, u) = \exp\{it - 2t^2 - u^2 - tu\} = \exp\{it - \frac{1}{2}(4t^2 + 2tu + 2u^2)\}$  then we recognize this as the characteristic function of a normal random variable

$$\mathbf{X} = (X_1, X_2)' \in N\left(\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 4 & 1 \\ 1 & 2 \end{pmatrix}\right).$$

Therefore, by Definition I,  $X_1 + X_2$  is normal with mean  $E(X_1) + E(X_2) = 0 + 1 = 1$  and variance  $\text{var}(X_1 + X_2) = \text{var}(X_1) + \text{var}(X_2) + 2\text{cov}(X_1, X_2) = 4 + 2 + 2 \cdot 1 = 8$ . That is,

$$X_1 + X_2 \in N(1, 8).$$

**Exercise 5.2, page 129:** If  $\psi(t, u) = \exp\{t^2 + 3tu + 4u^2\} = \exp\{\frac{1}{2}(2t^2 + 6tu + 8u^2)\}$  then we recognize this as the moment generating function of a normal random variable

$$\mathbf{X} = (X_1, X_2)' \in N\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 2 & 3 \\ 3 & 8 \end{pmatrix}\right).$$

Since

$$\rho = \text{corr}(X, Y) = \frac{\text{cov}(X, Y)}{\sqrt{\text{var}(X) \cdot \text{var}(Y)}} = \frac{3}{\sqrt{2 \cdot 8}} = \frac{3}{4},$$

we conclude that the density function of  $\mathbf{X}$  is given by

$$f_{\mathbf{X}}(\mathbf{x}) = \frac{1}{2\pi \cdot 2 \cdot 8 \cdot \sqrt{1 - (3/4)^2}} \exp\left\{-\frac{1}{2(1 - (3/4)^2)} \left(\frac{x_1^2}{4} - 2\frac{3x_1x_2}{4 \cdot 2 \cdot 8} + \frac{x_2^2}{64}\right)\right\}.$$

**Exercise 7.1, page 134:** By Definition I, we know that  $X$  and  $Y - \rho X$  are normally distributed. Therefore, by Theorem 7.1,  $X$  and  $Y - \rho X$  are independent if and only if  $\text{cov}(X, Y - \rho X) = 0$ . We compute

$$\begin{aligned} \text{cov}(X, Y - \rho X) &= \text{cov}(X, Y) - \text{cov}(X, \rho X) = \text{cov}(X, Y) - \rho \text{var}(X) = \rho \text{SD}(X) \text{SD}(Y) - \rho \text{var}(X) \\ &= \rho \text{var}(X) - \rho \text{var}(X) = 0 \end{aligned}$$

since  $\text{SD}(X) \cdot \text{SD}(Y) = \text{SD}(X) \cdot \text{SD}(X) = \text{var}(X)$  by the assumption that  $\text{var}(X) = \text{var}(Y)$ . Hence,  $X$  and  $Y - \rho X$  are in fact independent.