$$\mathbf{X} = \begin{pmatrix} X_1 \\ X_2 \\ \dots \\ X_n \end{pmatrix} - \begin{array}{c} n\text{-dimensional} \\ \text{random vector} \\ \end{pmatrix}$$

$$E(\mathbf{X}) = \boldsymbol{\mu}_{\mathbf{X}} = \begin{pmatrix} \boldsymbol{\mu}_1 \\ \boldsymbol{\mu}_2 \\ \dots \\ \boldsymbol{\mu}_n \end{pmatrix}$$

Covariance matrix:

$$\boldsymbol{\Sigma}_{\mathbf{X}} = \begin{pmatrix} \sigma_{11} & \sigma_{12} & \dots & \sigma_{1n} \\ \sigma_{21} & \sigma_{22} & \dots & \sigma_{2n} \\ \dots & \dots & \dots \\ \sigma_{n1} & \sigma_{n2} & \dots & \sigma_{nn} \end{pmatrix} \qquad \begin{aligned} \boldsymbol{\sigma}_{ij} &= \operatorname{Cov}(\mathbf{X}_i, \mathbf{X}_j) \\ \boldsymbol{\Sigma}_{\mathbf{X}} &= \operatorname{E}[(\mathbf{X} - \boldsymbol{\mu})(\mathbf{X} - \boldsymbol{\mu})^{\mathrm{T}}] \\ \boldsymbol{\Sigma}_{\mathbf{X}} - \operatorname{symmetric, nonnegative-definite} \end{aligned}$$

$$n = 1$$

$$n = 2$$

$$\mathbf{X} = (\mathbf{X})$$

$$\mathbf{X} = \begin{pmatrix} \mathbf{X}_1 \\ \mathbf{X}_2 \end{pmatrix}$$

$$\mu_{\mathbf{X}} = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}$$

$$\mu_{\mathbf{X}} = (\mu)$$

$$\Sigma_{\mathbf{X}} = \begin{pmatrix} \operatorname{Var}(\mathbf{X}_1) & \operatorname{Cov}(\mathbf{X}_1, \mathbf{X}_2) \\ \operatorname{Cov}(\mathbf{X}_1, \mathbf{X}_2) & \operatorname{Var}(\mathbf{X}_2) \end{pmatrix} = \begin{pmatrix} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{pmatrix}$$

Let
$$\mathbf{Y} = \mathbf{A} \mathbf{X} + \mathbf{b}$$
, where $\mathbf{A} - m \times n$ (non-random) matrix, $\mathbf{b} \in \mathbf{R}^m$ – (non-random) vector

Then
$$E(Y) = \mu_Y = A \mu_X + b$$
 $\Sigma_Y = A \Sigma_X A^T$

$$\mathbf{a} = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \dots \\ \mathbf{a}_n \end{pmatrix} \qquad \mathbf{a}_1 \mathbf{X}_1 + \mathbf{a}_2 \mathbf{X}_2 + \dots + \mathbf{a}_n \mathbf{X}_n = \mathbf{a}^T \mathbf{X}$$

$$\mathbf{E}(\mathbf{a}^T \mathbf{X}) = \mathbf{a}^T \mathbf{\mu}_{\mathbf{X}} \qquad \mathbf{Var}(\mathbf{a}^T \mathbf{X}) = \mathbf{a}^T \mathbf{\Sigma}_{\mathbf{X}} \mathbf{a}$$

1. Consider a random vector
$$\vec{\mathbf{X}} = \begin{pmatrix} X_1 \\ X_2 \\ X_3 \end{pmatrix}$$
 with mean $E(\vec{\mathbf{X}}) = \vec{\boldsymbol{\mu}} = \begin{pmatrix} 5 \\ 1 \\ 3 \end{pmatrix}$

and variance-covariance matrix Cov($\vec{\mathbf{X}}$) = $\mathbf{\Sigma} = \begin{pmatrix} 9 & 2 & -3 \\ 2 & 4 & -2 \\ -3 & -2 & 16 \end{pmatrix}$.

Then
$$\operatorname{Var}(X_1) = 9$$
, $\operatorname{Var}(X_2) = 4$, $\operatorname{Var}(X_3) = 16$,
$$\rho_{12} = \frac{2}{\sqrt{9} \cdot \sqrt{4}} = \frac{1}{3}, \quad \rho_{13} = \frac{-3}{\sqrt{9} \cdot \sqrt{16}} = -\frac{1}{4}, \quad \rho_{23} = \frac{-2}{\sqrt{4} \cdot \sqrt{16}} = -\frac{1}{4}.$$

Consider
$$2X_1 - 3X_2 - X_3 = (2 -3 -1)\begin{pmatrix} X_1 \\ X_2 \\ X_3 \end{pmatrix}$$
.

$$E(2X_1 - 3X_2 - X_3) = (2 - 3 - 1)\vec{\mu} = 2\mu_1 - 3\mu_2 - \mu_3 = 2 \cdot 5 - 3 \cdot 1 - 1 \cdot 3 = 4.$$

$$Var(2X_{1}-3X_{2}-X_{3}) = \begin{pmatrix} 2 & -3 & -1 \end{pmatrix} \begin{pmatrix} 9 & 2 & -3 \\ 2 & 4 & -2 \\ -3 & -2 & 16 \end{pmatrix} \begin{pmatrix} 2 \\ -3 \\ -1 \end{pmatrix}$$
$$= \begin{pmatrix} 15 & -6 & -16 \end{pmatrix} \begin{pmatrix} 2 \\ -3 \\ -1 \end{pmatrix} = 64.$$

OR

$$Var(2X_1 - 3X_2 - X_3) = 4 Var(X_1) + 9 Var(X_2) + Var(X_3)$$
$$-12 Cov(X_1, X_2) - 4 Cov(X_1, X_3) + 6 Cov(X_2, X_3)$$
$$= 36 + 36 + 16 - 24 + 12 - 12 = 64.$$

Multivariate Normal Distribution:

$$\mathbf{X} \sim N_n(\boldsymbol{\mu}, \boldsymbol{\Sigma})$$

$$f_{\mathbf{X}}(\mathbf{x}) = \frac{1}{(2\pi)^{n/2} |\mathbf{\Sigma}|^{1/2}} \exp\left\{-\frac{1}{2} (\mathbf{x} - \mathbf{\mu})' \mathbf{\Sigma}^{-1} (\mathbf{x} - \mathbf{\mu})\right\}$$

$$\mathbf{Z} \sim N_n(\mathbf{0}, \mathbf{I}_n)$$

$$f_{\mathbf{Z}}(\mathbf{z}) = \frac{1}{(2\pi)^{n/2}} \exp\left\{-\frac{1}{2}\mathbf{z}'\mathbf{z}\right\} = \prod_{i=1}^{n} \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{1}{2}z_{i}^{2}\right\}$$

$$\boldsymbol{X} = \boldsymbol{\Sigma}^{1/2} \; \boldsymbol{Z} + \boldsymbol{\mu}$$

$$M_{\mathbf{X}}(\mathbf{t}) = \exp\left\{\mathbf{t'}\boldsymbol{\mu} + \frac{1}{2}\mathbf{t'}\boldsymbol{\Sigma}\mathbf{t}\right\}$$

$$\mathbf{t} = \begin{pmatrix} t_1 \\ t_2 \\ \dots \\ t_n \end{pmatrix} \in \mathbf{R}^n$$

$$Y = AX + b$$

$$\mathbf{A} - m \times n$$

$$\mathbf{b} \in \mathbf{R}^m$$

$$\Rightarrow$$
 $\mathbf{Y} \sim N_m(\mathbf{A}\,\mathbf{\mu} + \mathbf{b}, \mathbf{A}\,\mathbf{\Sigma}\,\mathbf{A}')$

$$\mathbf{X} = \begin{pmatrix} \mathbf{X}_1 \\ \mathbf{X}_2 \end{pmatrix}$$

 $\mathbf{X} = \begin{pmatrix} \mathbf{X}_1 \\ \mathbf{X}_2 \end{pmatrix}$ \mathbf{X}_1 is of dimension m < n \mathbf{X}_2 is of dimension n - m

$$\boldsymbol{\mu} = \begin{pmatrix} \boldsymbol{\mu}_1 \\ \boldsymbol{\mu}_2 \end{pmatrix} \qquad ($$

$$\boldsymbol{\mu} = \begin{pmatrix} \boldsymbol{\mu}_1 \\ \boldsymbol{\mu}_2 \end{pmatrix} \qquad \boldsymbol{\Sigma} = \begin{pmatrix} \boldsymbol{\Sigma}_{11} & \boldsymbol{\Sigma}_{12} \\ \boldsymbol{\Sigma}_{21} & \boldsymbol{\Sigma}_{22} \end{pmatrix}$$

$$\Rightarrow \mathbf{X}_{1} | \mathbf{X}_{2} \sim N_{m} \Big(\mathbf{\mu}_{1} + \mathbf{\Sigma}_{12} \mathbf{\Sigma}_{22}^{-1} (\mathbf{X}_{2} - \mathbf{\mu}_{2}), \mathbf{\Sigma}_{11} - \mathbf{\Sigma}_{12} \mathbf{\Sigma}_{22}^{-1} \mathbf{\Sigma}_{21} \Big)$$

2.
$$\mathbf{X} \sim N_3(\mathbf{\mu}, \mathbf{\Sigma})$$
 $\mathbf{\mu} = \begin{pmatrix} 5 \\ 3 \\ 7 \end{pmatrix}$ $\mathbf{\Sigma} = \begin{pmatrix} 4 & -1 & 0 \\ -1 & 4 & 2 \\ 0 & 2 & 9 \end{pmatrix}$

a) Find $P(X_1 > 6)$.

$$X_1 \sim N(5,4)$$

$$P(X_1 > 6) = P(Z > \frac{6-5}{\sqrt{4}}) = P(Z > 0.5) = 0.3085.$$

b) Find P($5X_2 + 4X_3 > 70$).

$$E(5X_2 + 4X_3) = 5 \cdot 3 + 4 \cdot 7 = 43.$$

$$Var(5X_2 + 4X_3) = 25 Var(X_2) + 40 Cov(X_2, X_3) + 16 Var(X_3)$$
$$= 25 \cdot 4 + 40 \cdot 2 + 16 \cdot 9 = 324.$$

OR

$$Var(5X_2 + 4X_3) = \begin{pmatrix} 0 & 5 & 4 \end{pmatrix} \begin{pmatrix} 4 & -1 & 0 \\ -1 & 4 & 2 \\ 0 & 2 & 9 \end{pmatrix} \begin{pmatrix} 0 \\ 5 \\ 4 \end{pmatrix} = \begin{pmatrix} -5 & 28 & 46 \end{pmatrix} \begin{pmatrix} 0 \\ 5 \\ 4 \end{pmatrix} = 324.$$

$$P(5X_2 + 4X_3 > 70) = P(Z > \frac{70 - 43}{\sqrt{324}}) = P(Z > 1.50) = 0.0668.$$

c) Find P($3X_3 - 5X_2 < 17$).

$$E(3X_3 - 5X_2) = 3 \cdot 7 - 5 \cdot 3 = 6.$$

$$Var(3X_3 - 5X_2) = 9 Var(X_3) - 30 Cov(X_2, X_3) + 25 Var(X_2)$$
$$= 9 \cdot 9 - 30 \cdot 2 + 25 \cdot 4 = 121.$$

$$Var(3X_3 - 5X_2) = \begin{pmatrix} 0 & -5 & 3 \end{pmatrix} \begin{pmatrix} 4 & -1 & 0 \\ -1 & 4 & 2 \\ 0 & 2 & 9 \end{pmatrix} \begin{pmatrix} 0 \\ -5 \\ 3 \end{pmatrix} = \begin{pmatrix} 5 & -14 & 17 \end{pmatrix} \begin{pmatrix} 0 \\ -5 \\ 3 \end{pmatrix} = 121.$$

$$P(3X_3 - 5X_2 < 17) = P(Z < \frac{17 - 6}{\sqrt{121}}) = P(Z < 1.00) = 0.8413.$$

d) Find
$$P(4X_1 - 3X_2 + 5X_3 < 80)$$
.

$$E(4X_1 - 3X_2 + 5X_3) = 4\mu_1 - 3\mu_2 + 5\mu_3 = 4 \cdot 5 - 3 \cdot 3 + 5 \cdot 7 = 46.$$

$$\operatorname{Var}(4X_{1} - 3X_{2} + 5X_{3}) = \begin{pmatrix} 4 & -3 & 5 \end{pmatrix} \begin{pmatrix} 4 & -1 & 0 \\ -1 & 4 & 2 \\ 0 & 2 & 9 \end{pmatrix} \begin{pmatrix} 4 \\ -3 \\ 5 \end{pmatrix} = \begin{pmatrix} 19 & -6 & 39 \end{pmatrix} \begin{pmatrix} 4 \\ -3 \\ 5 \end{pmatrix}$$
$$= 289.$$

$$4X_1 - 3X_2 + 5X_3 \sim N(46, 289)$$

$$P(4X_1 - 3X_2 + 5X_3 < 80) = P(Z < \frac{80 - 46}{\sqrt{289}}) = P(Z < 2.00) = 0.9772.$$

e)* Find $P(X_1 > 8 | X_2 = 1, X_3 = 10)$.

$$\Sigma_{22} = \begin{pmatrix} 4 & 2 \\ 2 & 9 \end{pmatrix}$$
 $\Sigma_{22}^{-1} = \frac{1}{32} \begin{pmatrix} 9 & -2 \\ -2 & 4 \end{pmatrix}$

$$\Sigma_{12} \Sigma_{22}^{-1} = \frac{1}{32} \begin{pmatrix} -1 & 0 \end{pmatrix} \begin{pmatrix} 9 & -2 \\ -2 & 4 \end{pmatrix} = \frac{1}{32} \begin{pmatrix} -9 & 2 \end{pmatrix}$$

$$\mu_1 + \Sigma_{12} \Sigma_{22}^{-1} (\mathbf{X}_2 - \mu_2) = 5 + \frac{1}{32} (-9 \quad 2) \begin{pmatrix} 1-3 \\ 10-7 \end{pmatrix} = 5.75.$$

$$\Sigma_{11} - \Sigma_{12} \Sigma_{22}^{-1} \Sigma_{21} = 4 - \frac{1}{32} (-9 \ 2) \begin{pmatrix} -1 \\ 0 \end{pmatrix} = 3.71875.$$

$$X_1 \mid X_2 = 1, X_3 = 10 \sim N(5.75, 3.71875)$$

$$P(X_1 > 8 | X_2 = 1, X_3 = 10) = P\left(Z > \frac{8 - 5.75}{\sqrt{3.71875}}\right) = P(Z > 1.17) = 0.1210.$$