

# Multivariate Normal Distribution

Chao Cheng

July 22, 2022

For a multivariate normal distribution denoted by

$$\mathbf{x} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma}),$$

where  $\mathbf{x} = (x_1, \dots, x_n)^T \in \mathcal{R}^n$ ,  $\boldsymbol{\mu} = (\mu_1, \dots, \mu_n)^T$  is the mean vector and

$$\boldsymbol{\Sigma} = (\rho_{ij}\sigma_i\sigma_j)_{n \times n} = \begin{pmatrix} \sigma_1^2 & \cdots & \rho_{1n}\sigma_1\sigma_n \\ \vdots & \rho_{ij}\sigma_i\sigma_j & \vdots \\ \rho_{n1}\sigma_n\sigma_1 & \cdots & \sigma_n^2 \end{pmatrix}$$

is the  $n \times n$  covariance matrix. Here  $\sigma_1^2, \dots, \sigma_n^2$  is the variance of  $x_1, \dots, x_n$  and  $\rho_{ij}$  is the correlation between  $x_i$  and  $x_j$ . Since it's symmetric,  $\rho_{ij} = \rho_{ji}$  and  $\rho_{ii} = 1$  for  $i = 1, \dots, n$ .

The density function is

$$f(\mathbf{x}) = (2\pi)^{-n/2} |\boldsymbol{\Sigma}|^{-1/2} \exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu})\right).$$

## 1 Bivariate normal distribution

For a bivariate normal variable  $(x, y)$  following

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

# A R codes

R codes