

Multivariate Normal Distribution

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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 ρ_{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 Conditional distribution

2 Bivariate normal distribution

For a bivariate normal variable (x, y) following

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

and the density function is

$$\begin{aligned} f_{X,Y}(x, y) &= \frac{1}{2\pi} (\sigma_1^2\sigma_2^2 - \rho^2\sigma_1^2\sigma_2^2)^{-1/2} \exp\left(-\frac{(x - \mu_1 \quad y - \mu_2) \begin{pmatrix} \sigma_2^2 & -\rho\sigma_1\sigma_2 \\ -\rho\sigma_1\sigma_2 & \sigma_1^2 \end{pmatrix} \begin{pmatrix} x - \mu_1 \\ y - \mu_2 \end{pmatrix}}{2(\sigma_1^2\sigma_2^2 - \rho^2\sigma_1^2\sigma_2^2)}\right) \\ &= \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1 - \rho^2}} \exp\left(\frac{\sigma_2^2(x - \mu_1)^2 - 2\rho\sigma_1\sigma_2(x - \mu_1)(y - \mu_2) + \sigma_1^2(y - \mu_2)^2}{2\sigma_1^2\sigma_2^2(1 - \rho^2)}\right). \end{aligned}$$

The probability function $F_{X,Y}(x,y) = P(X \leq x, Y \leq y)$ is

$$\begin{aligned} F_{X,Y}(x,y) &= \int_{-\infty}^x \int_{-\infty}^y f_{X,Y}(s,t) \, ds dt \\ &= \int_{-\infty}^x \int_{-\infty}^y f_{Y|X}(s|t) f_X(t) \, ds dt \\ &= \int_{-\infty}^x f_X(t) \left(\int_{-\infty}^y f_{Y|X}(s|t) \, ds \right) dt. \end{aligned}$$

For the conditional distribution $Y|X$, the density is

$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)}$$

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