Multivariate Statistics

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Multivariate data and multivariate normal distribution

2D bivariate random vector

$$X = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

Simultaneous pdf $f(x_1,x_2)=P(X_1=x_1,X_2=x_2)$ complete information.

Marginal PDFs

$$f(x_1) = \int f(x_1, x_2) \, \mathrm{d}x_2 \ \ \text{and} \ \ f(x_2) = \int f(x_1, x_2) \, \mathrm{d}x_1$$

Mean vector

$$\mu = \begin{pmatrix} E[x_1] \\ E[x_2] \end{pmatrix} = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}$$

Marginal variances

$$\sigma_1^2 = \sigma_{11} = E[(X_1 - \mu_1)^2] \ge 0$$

$$\sigma_2^2 = \sigma_{22} = E[(X_2 - \mu_2)^2] \ge 0$$

Covariance

$$\sigma_{12} = \mathrm{Cov}(X_1, X_2) = \sigma_{21} = E[(X_1 - \mu_1)(X_2 - \mu_2)]$$

Covariance matrix

$$\sum \begin{bmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{12} & \sigma_2^2 \end{bmatrix}$$

The matrix is symmetric and has positive semi-definite eigenvalues $\lambda_1,\lambda_2\geq 0$

Correlation coefficient

$$\rho_{12} = \frac{\sigma_{12}}{\sigma_1 \sigma_2}$$

Correlation matrix

$$\rho = \begin{bmatrix} 1 & \rho_{12} \\ \rho_{12} & 1 \end{bmatrix}$$

$$-1 \le \rho_{12} \le 1$$

Generalized variance

$$\left|\sum\right|=\det\!\left(\sum\right)=\prod_{i=1}^2\lambda_i$$

Measure of amount of random variation

pD bivariate random vector

$$X = \begin{bmatrix} x_1, x_2, ..., x_p \end{bmatrix}^T$$

Mean vector

$$\mu = \begin{pmatrix} \mu_1 \\ \dots \\ \mu_p \end{pmatrix}$$

Covariance matrix

$$\sum_{p\times p} = \begin{bmatrix} \sigma_1^2 & \sigma_{12} & \sigma_{1p} \\ \dots & \sigma_2 & \dots \\ \dots & \dots & \sigma_p \end{bmatrix} = E\big[(X-\mu)(X-\mu)^T\big]$$

Correlation matrix:

$$\rho_{p\times p} = \begin{bmatrix} 1 & \rho_{12} & \dots & \rho_{1p} \\ \dots & 1 & \dots & \dots \\ \dots & \dots & 1 & \dots \\ \dots & \dots & \dots & 1 \end{bmatrix} - 1 \leq \rho_{ij} \leq 1$$

Sampling

From a random sample we want to estimate the moment of the population (μ, \sum, ρ) .

Data Matrix: (Make = over X)

$$X_{n \times p} = \begin{bmatrix} x_{11} & x_{12} & \dots & x_{1p} \\ x_{21} & x_{22} & \dots & x_{2p} \\ x_{n1} & x_{n2} & \dots & x_{np} \end{bmatrix} = \begin{bmatrix} x_1^T \\ x_2^T \\ \dots \\ x_n^T \end{bmatrix}$$

Columns are the variables and rows are the observations.

Descriptive statistics

Estimation of moments (μ, \sum, ρ)

$$\hat{\mu} = \bar{x} = \frac{1}{n} \sum_{j=1}^{n} X_j$$

% Matlab mean(X)

$$\widehat{\sum}_{p\times p} = S_{p\times p} = \frac{1}{n-1} \sum_{j=1}^n (X_j - \widehat{\mu}) \big(X_j - \widehat{\mu} \big)^T = \begin{pmatrix} s_1^2 & s_{12} & \dots & s_{1p} \\ \dots & s_2^2 & \dots & \dots \\ \dots & \dots & s_3^2 & \dots \\ \dots & \dots & \dots & s_p^2 \end{pmatrix}$$

% Matlab cov(X)

$$\hat{\rho} = R = \begin{pmatrix} 1 & r_{12} & r_{13} & \dots & r_{1p} \\ \dots & 1 & \dots & \dots & \dots \\ \dots & \dots & 1 & \dots & \dots \\ \dots & \dots & \dots & 1 & \dots \\ \dots & \dots & \dots & \dots & \dots & 1 \end{pmatrix}$$

% Matlab
corrcoef(X)

Inference about mean vectors and comparison of multivariate samples

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