### MULTIVARIATE STATISTICAL ANALYSIS

# Lecture 3 Multivariate Linear Regression

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# 3.1. Purpose of LR

Regression analysis is the statistical methodology for predicting values of one or more response (**dependent**) variables from a collect of predictor (**independent**) variable values..



## 3.2. Problem statement

Let  $z_1, z_2, ..., z_r$  be r predictor variables to be related to a response variable Y .

The linear regression model with a single response takes the form:

$$Y = \beta_0 + \beta_1 z_1 + \dots + \beta_r z_r + \varepsilon$$



## 3.2. Problem statement

Let  $z_{i1}, z_{i2}, ..., z_{ir}$  be r predictor variables to be related to a response variable  $Y_i, i = 1..n$ 

The linear regression model with n response takes the form:

$$Y_{1} = \beta_{0} + \beta_{1}z_{11} + \dots + \beta_{r}z_{1r} + \varepsilon_{1}$$

$$Y_{2} = \beta_{0} + \beta_{1}z_{21} + \dots + \beta_{r}z_{2r} + \varepsilon_{2}$$

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$$Y_n = \beta_0 + \beta_1 z_{n1} + ... + \beta_r z_{nr} + \varepsilon_n$$

Where  $\mathcal{E}_i$  are assumed:

1. 
$$E(\varepsilon_i) = 0$$
; 2.  $Var(\varepsilon_i) = \sigma^2$ ; 3.  $Cov(\varepsilon_i, \varepsilon_j) = 0, i \neq j$ 



## 3.2. Problem statement

#### fit@hcmus

Let  $z_{i1}, z_{i2}, ..., z_{ir}$  be r predictor variables to be related to a response variable  $Y_i, i = 1..n$ 

The linear regression model with n response in the form of the

matrix:

$$\begin{bmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_n \end{bmatrix} = \begin{bmatrix} 1 & z_{11} & z_{12} \cdots & z_{1r} \\ 1 & z_{21} & z_{22} \cdots & z_{2r} \\ \vdots & \vdots & & \vdots \\ 1 & z_{n1} & z_{n2} \cdots & z_{nr} \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_r \end{bmatrix} + \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \vdots \\ \varepsilon_n \end{bmatrix}$$

$$Y = Z \cdot \beta + \varepsilon$$

$$\varepsilon$$
 are assumed: 1.  $E(\varepsilon) = 0$ ; 2.  $Cov(\varepsilon) = E(\varepsilon \varepsilon') = \sigma^2 I$ 



## 3.3. Method

Let  $z_1, z_2, ..., z_r$  be r predictor variables to be related to a response variable Y .

The linear regression model with a single response takes the form:

$$Y = \beta_0 + \beta_1 z_1 + \dots + \beta_r z_r + \varepsilon$$

Estimate  $\{\beta_i\}$ , i=1..r based on the samples having the corresponding between  $\{z_{jr}\}$ , j=1..n and  $\{y_j\}$ , j=1..n

Estimate  $\{\beta_i\}$ , i=1..r to minimize sum of squares of the differences between the predicting values and groundtruth values.



# 3.3. Phương pháp

Sum of squares of the differences between the prediting values and groundtruth values :

$$S(\beta) = \sum_{j=1}^{n} (y_i - \beta_0 - \beta_1 z_{j1} - \dots - \beta_r z_{jr})^2 = (y - Z\beta)^T \cdot (y - Z\beta)$$

The necessary condition for the above quantity to be minimized is:

$$\frac{\partial S(\beta)}{\partial \beta} = 0$$

$$\Rightarrow -2\frac{\partial (\beta^T Z^T y)}{\partial \beta} + \frac{\partial (\beta^T Z^T Z \beta)}{\partial \beta} = 0$$

$$\Rightarrow -2Z^T y + 2Z^T Z \beta = 0$$

$$\Rightarrow \beta = (Z^T Z)^{-1} Z^T y$$



## 3.4. Geometrical Explanation

Discuss in class



# 3.5. Model Checking

Discuss in class



## 3.6. Case study

#### Discuss in class