

**Stage: Second  
Subject: Linear Algebra  
Application**

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## **Chapter five**

### **Some application of Linear Algebra:**

Using matrices for the Linear Regression model(Regression Analysis): Matrix Formulation of Linear Regression:

Where X is the input data and each column is a data feature, b is a vector of coefficients and y is a vector of output variables for each row in X. Reformulated, the problem becomes a system of linear equations where the b vector values are unknown.

$$Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \dots + \beta_p X_{pi} + \varepsilon_i$$

Where:

$$i = 1, 2, 3, \dots, n$$

$\underline{Y}$ : is the dependent variable(the response vector).

$X$ 's : are the independent variables (Information matrix) (design matrix).

$\underline{\beta}$ : is the vector of parameter.

$\varepsilon_i$ : error vector

The Design Matrix:

$$\underline{Y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \quad X = \begin{bmatrix} 1 & x_{11} & x_{21} & \cdots & x_{p1} \\ 1 & x_{12} & x_{22} & \cdots & x_{p2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{1n} & x_{2n} & \cdots & x_{pn} \end{bmatrix}$$

$$\underline{\beta} = \begin{bmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_n \end{bmatrix}, \quad \underline{\varepsilon} = \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \vdots \\ \varepsilon_n \end{bmatrix}$$

**Then:**  $Y = X\beta + \varepsilon_i$

Parameter Estimation of this model by using matrix is:

$$\hat{\underline{\beta}} = (X'X)^{-1} X'Y$$

### **Simple Linear Regression model :**

$$Y_i = \beta_0 + \beta_1 X_{1i} + \varepsilon_i$$

Then  $\hat{\underline{\beta}} = (X'X)^{-1} X'Y$

Where:

$$\underline{Y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \quad X = \begin{bmatrix} 1 & x_{11} \\ 1 & x_{12} \\ \vdots & \vdots \\ 1 & x_{1n} \end{bmatrix}$$

$$X'X = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_{11} & x_{12} & \cdots & x_{1n} \end{bmatrix} \begin{bmatrix} 1 & x_{11} \\ 1 & x_{12} \\ \vdots & \vdots \\ 1 & x_{1n} \end{bmatrix} = \begin{bmatrix} n & \sum_{i=1}^n x_i \\ \sum_{i=1}^n x_i & \sum_{i=1}^n x_i^2 \end{bmatrix}$$

$$X'Y = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_{11} & x_{12} & \cdots & x_{1n} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^n y_i \\ \sum_{i=1}^n x_i y_i \end{bmatrix}$$

then:

$$\hat{\underline{\beta}} = \begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \end{bmatrix} = (X'X)^{-1} X'Y$$

**Ex1: From the following data:**

|   |      |      |     |     |      |      |      |     |
|---|------|------|-----|-----|------|------|------|-----|
| X | -2   | 0.6  | 1.4 | 1.3 | 0    | -1.6 | -1.7 | 0.7 |
| Y | -6.1 | -0.5 | 7.2 | 6.9 | -0.2 | -2.1 | -3.9 | 3.8 |

**Estimate the Linear Regression Model.**

**Solution:**

|                |      |      |       |      |      |      |      |      | Sum   |
|----------------|------|------|-------|------|------|------|------|------|-------|
| X              | -2   | 0.6  | 1.4   | 1.3  | 0    | -1.6 | -1.7 | 0.7  | -1.3  |
| Y              | -6.1 | -0.5 | 7.2   | 6.9  | -0.2 | -2.1 | -3.9 | 3.8  | 5.1   |
| XY             | 12.2 | -0.3 | 10.08 | 8.97 | 0    | 3.36 | 6.63 | 2.66 | 43.6  |
| X <sup>2</sup> | 4    | 0.36 | 1.96  | 1.69 | 0    | 2.56 | 2.89 | 0.49 | 13.95 |

$$X'X = \begin{bmatrix} n & \sum_{i=1}^n x_i \\ \sum_{i=1}^n x_i & \sum_{i=1}^n x_i^2 \end{bmatrix} = \begin{bmatrix} 8 & -1.3 \\ -1.3 & 13.95 \end{bmatrix}$$

$$(X'X)^{-1} = \frac{\text{adj}(X'X)}{|X'X|}$$

$$|X'X| = \begin{vmatrix} 8 & -1.3 \\ -1.3 & 13.95 \end{vmatrix} = 111.6 - 1.69 = 109.91$$

$$\text{adj}(X'X) = \begin{bmatrix} 13.95 & 1.3 \\ 1.3 & 8 \end{bmatrix}$$

$$(X'X)^{-1} = \frac{1}{109.91} \begin{bmatrix} 13.95 & 1.3 \\ 1.3 & 8 \end{bmatrix}$$

$$X'Y = \begin{bmatrix} \sum_{i=1}^n y_i \\ \sum_{i=1}^n x_i y_i \end{bmatrix} = \begin{bmatrix} 5.1 \\ 43.6 \end{bmatrix}$$

$$\hat{\beta} = (X'X)^{-1} X'Y$$

$$\begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \end{bmatrix} = \frac{1}{109.91} \begin{bmatrix} 13.95 & 1.3 \\ 1.3 & 8 \end{bmatrix} \begin{bmatrix} 5.1 \\ 43.6 \end{bmatrix}$$

$$= \frac{1}{109.91} \begin{bmatrix} 71.145 + 56.68 \\ 6.63 + 348.8 \end{bmatrix} = \frac{1}{109.91} \begin{bmatrix} 127.825 \\ 355.43 \end{bmatrix}$$

$$\begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \end{bmatrix} = \begin{bmatrix} 1.16 \\ 3.23 \end{bmatrix}$$

$$\begin{aligned} \therefore \hat{Y}_i &= \hat{\beta}_0 + \hat{\beta}_1 X_i \\ &= 1.16 + 3.23 X_i \end{aligned}$$

Ex2:

If we have this information in 6 observations:

$$\sum_{i=1}^n x_i = 153.8 \quad \sum_{i=1}^n x_i^2 = 5859.26 \quad , \quad \sum_{i=1}^n y_i = 18.7 \quad , \quad \sum_{i=1}^n x_i y_i = 682.77$$

Find the estimated equation of the regression line.

Ex3:

From the following data

|       |   |   |   |   |   |   |   |   |   |    |
|-------|---|---|---|---|---|---|---|---|---|----|
| $X_i$ | 4 | 2 | 3 | 6 | 7 | 5 | 9 | 8 | 4 | 11 |
| $Y_i$ | 3 | 3 | 2 | 5 | 7 | 6 | 6 | 4 | 2 | 7  |

**Explain the relation between ( $X_i$ ) and ( $Y_i$ ).**

Solution: