# ML LECTURE-5

BY
Dr. Ramesh Kumar Thakur
Assistant Professor (II)
School Of Computer Engineering

### Multiple linear regression

- $\diamondsuit$  We assume that there are N independent variables  $x_1, x_2, \cdots, x_N$ . Let the dependent variable be y.
- Let there also be n observed values of these variables:

Variables	Values (examples)				
(features)	Example 1	Example 2		Example n	
$x_1$	$x_{11}$	$x_{12}$	•••	$x_{1n}$	
$x_2$	$x_{21}$	$x_{22}$	•••	$x_{2n}$	
$x_N$	$x_{N1}$	$x_{N2}$	•••	$x_{Nn}$	
y (outcomes)	$y_1$	$y_2$	•••	$y_n$	

Table 7.3: Data for multiple linear regression

The multiple linear regression model defines the relationship between the N independent variables and the dependent variable by an equation of the following form:

$$y = \beta_0 + \beta_1 x_1 + \cdots + \beta_N x_N$$

#### Multiple linear regression

As in simple linear regression, here also we use the ordinary least squares (OLS) method to obtain the optimal estimates of  $\beta_0$ ,  $\beta_1$ , ...,  $\beta_N$ . The method yields the following procedure for the computation of these optimal estimates. Let

$$X = \begin{bmatrix} 1 & x_{11} & x_{21} & \cdots & x_{N1} \\ 1 & x_{12} & x_{22} & \cdots & x_{N2} \\ \vdots & & & & & \\ 1 & x_{1n} & x_{2n} & \cdots & x_{Nn} \end{bmatrix}, \quad Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \quad B = \begin{bmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_N \end{bmatrix}$$

❖ Then it can be shown that the regression coefficients are given by

$$\mathbf{B} = (\mathbf{X}^{\mathrm{T}} \mathbf{X})^{-1} \mathbf{X}^{\mathrm{T}} \mathbf{Y}$$

# Multiple linear regression Example

- Example:
- ❖ Fit a multiple linear regression model to the following data:

$x_1$	1	1	2	0
$x_2$	1	2	2	1
y	3.25	6.5	3.5	5.0

Table 7.4: Example data for multi-linear regression

- **❖** Solution:
- ❖ In this problem, there are two independent variables and four sets of values of the variables. Thus, in the notations used above, we have n = 2 and N = 4. The multiple linear regression model for this problem has the form

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2$$
.

The computations are shown below.

$$X = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \\ 1 & 2 & 2 \\ 1 & 0 & 1 \end{bmatrix}, \quad Y = \begin{bmatrix} 3.25 \\ 6.5 \\ 3.5 \\ 5.0 \end{bmatrix}, \quad B = \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{bmatrix}$$

### Multiple linear regression Example

$$X^{T}X = \begin{bmatrix} 4 & 4 & 6 \\ 4 & 6 & 7 \\ 6 & 7 & 10 \end{bmatrix}$$
$$(X^{T}X)^{-1} = \begin{bmatrix} \frac{11}{4} & \frac{1}{2} & -2 \\ \frac{1}{2} & 1 & -1 \\ -2 & -1 & 2 \end{bmatrix}$$
$$B = (X^{T}X)^{-1}X^{T}Y$$
$$= \begin{bmatrix} 2.0625 \\ -2.3750 \\ 3.2500 \end{bmatrix}$$

The required model is

## Multiple linear regression Example

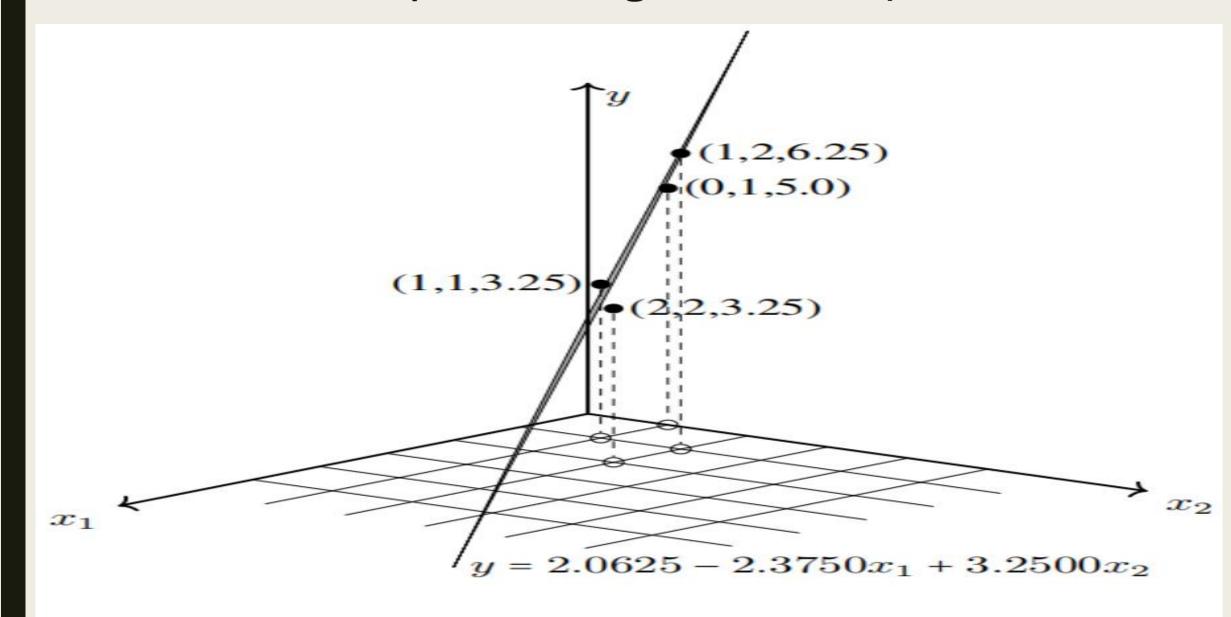


Figure 7.4: The regression plane for the data in Table 7.4

