

# Multi Linear regression

← Predict

$x_1$	$x_2$	$y$
Cgpa	iq	placement
$\beta_0$ 8 ( $\beta_1$ )	80 ( $\beta_2$ )	8
7	70	7
5	120	15

$$\hat{y}_1 = \beta_0 + \beta_1 x_{11} + \beta_2 x_{12} + \dots + \beta_m x_{1m}$$

$$\hat{y}_2 = \beta_0 + \beta_1 x_{21} + \beta_2 x_{22} + \dots + \beta_m x_{2m}$$

$$\hat{y}_3 = \beta_0 + \beta_1 x_{31} + \beta_2 x_{32} + \dots + \beta_m x_{3m}$$

⋮

$$\hat{y}_n = \beta_0 + \beta_1 x_{n1} + \beta_2 x_{n2} + \dots + \beta_m x_{nm}$$

$$y_1 = 8 \quad y_2 = 7 \quad y_3 = 15$$

$$\hat{y}_1 = 7 \quad \hat{y}_2 = 2 \quad \hat{y}_3 = 2$$

$$\hat{y}_1 = \beta_0 + \beta_1 8 + \beta_2 80$$

$$\hat{y}_2 = \beta_0 + \beta_1 7 + \beta_2 70$$

$$\hat{y}_3 = \beta_0 + \beta_1 5 + \beta_2 120$$

$$\hat{y} = \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} = \begin{bmatrix} \beta_0 + \beta_1 x_{11} + \dots + \beta_m x_{1m} \\ \beta_0 + \beta_1 x_{21} + \dots + \beta_m x_{2m} \\ \beta_0 + \beta_1 x_{31} + \dots + \beta_m x_{3m} \\ \vdots \\ \beta_0 + \beta_1 x_{n1} + \dots + \beta_m x_{nm} \end{bmatrix}$$

$n \times 1$

$n \times 1$

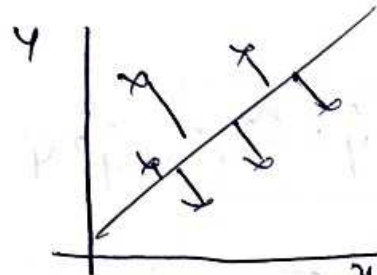
$$= \begin{bmatrix} 1 & x_{11} & x_{12} & \dots & x_{1m} \\ 1 & x_{21} & x_{22} & \dots & x_{2m} \\ 1 & x_{31} & x_{32} & \dots & x_{3m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n1} & x_{n2} & \dots & x_{nm} \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \\ \vdots \\ \beta_m \end{bmatrix}$$

$$\hat{y} = X\beta$$

$$n \times (m+1) \leftrightarrow (m+1) \times 1$$

Shape →  $n \times 1$  → proved

derived and proved



$$d_1^2 + d_2^2 + d_3^2 + \dots + d_n^2$$

min

$$E = \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

minimize  
matrix form

$$y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_n \end{bmatrix}$$

$$\hat{y} = \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \hat{y}_3 \\ \vdots \\ \hat{y}_n \end{bmatrix}$$

$$e = y - \hat{y} = \begin{bmatrix} y_1 - \hat{y}_1 \\ y_2 - \hat{y}_2 \\ \vdots \\ y_n - \hat{y}_n \end{bmatrix} \rightarrow [n \times 1]$$

$$e^T e = \begin{bmatrix} y_1 - \hat{y}_1 & y_2 - \hat{y}_2 & \dots & y_n - \hat{y}_n \end{bmatrix} \begin{bmatrix} y_1 - \hat{y}_1 \\ y_2 - \hat{y}_2 \\ \vdots \\ y_n - \hat{y}_n \end{bmatrix}$$

$[1 \times n]$        $[n \times 1]$

$$e^T e = (y_1 - \hat{y}_1)^2 + (y_2 - \hat{y}_2)^2 + \dots + (y_n - \hat{y}_n)^2$$

$$= \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

$$E = e^T e \rightarrow \text{II equation}$$

↑  
minimize

$$E = e^T e$$

$$E = (y - \hat{y})^T (y - \hat{y}) = y^T y - \boxed{y^T \hat{y} - \hat{y}^T y} + \hat{y}^T \hat{y}$$

$$\boxed{y^T \hat{y} = \hat{y}^T y} \leftarrow \text{Prove}$$

$$y = A \quad \hat{y} = B$$

$$\boxed{(AB)^T = B^T A^T}$$

$$(A^T)^T = A$$

$$A^T B = B^T A$$

↳ Took this and transpose eqn

$$A^T B \rightarrow (A^T B)^T = B^T A$$

Now want want to prove  $A^T B$  is equals to  $(A^T B)^T$

$$A^T B = (A^T B)^T$$

Let  $A^T B = C \rightarrow$  symmetric

then Prove

$$C = C^T$$

$A = Y \quad B = \hat{Y}$  we know that

$$y^T \hat{y} = \begin{matrix} y^T & \times & \hat{y} \\ \downarrow & & \downarrow \\ 1 \times n & & n \times (m+1) \end{matrix} \rightarrow \begin{matrix} (m+1) \times 1 \\ \downarrow \\ n \times 1 \end{matrix}$$

$$1 \times 1 = \text{Scalar} \quad [1] \quad [2]$$

Proved

$A^T B$  is symmetric

$$\text{So, } \boxed{y^T \hat{y} = \hat{y}^T y}$$



$$E = Y^T Y - 2Y^T \hat{Y} + \hat{Y}^T \hat{Y} \rightarrow \text{III eqn}$$

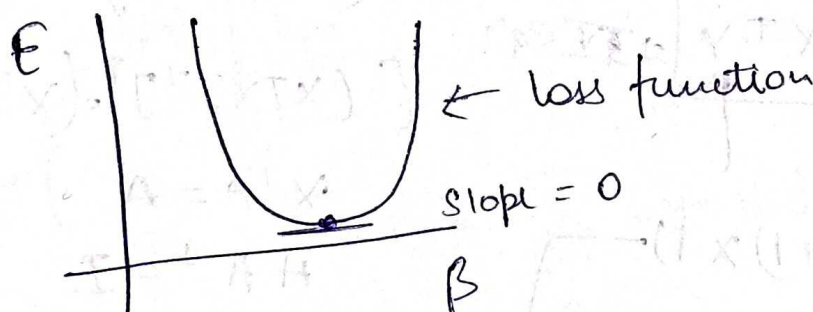
$$E = Y^T Y - 2Y^T X\beta + (X\beta)^T (X\beta)$$

$$\hat{Y} = X\beta$$

$$E = Y^T Y - 2Y^T X\beta + \beta^T X^T X\beta$$

$Y = f(m)$   
 change change change

find such value of  $\beta$  matrix  
 for which  $E$  is min



$$\frac{dE}{d\beta} = 0$$

$$E = Y^T Y - 2Y^T X\beta + (X\beta)^T (X\beta) \rightarrow \text{differentiate with respect to } \beta$$

$$\frac{dE}{d\beta} = 0 - 2Y^T X + 2\beta^T X^T X$$

$$\frac{d}{d\beta} \beta^T X^T X \beta = X^T A X$$

if  $A$  is symmetric

$$\beta^T X^T X = Y^T Y$$

$$\beta^T X^T X = Y^T Y$$

product both side  $(X^T X)^{-1}$

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

$$\beta^T I = Y^T X (X^T X)^{-1}$$

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

$$(\beta^T)^T = \left[ \underbrace{Y^T X}_A \underbrace{(X^T X)^{-1}}_B \right]^T$$

$$\beta = \left[ (X^T X)^{-1} \right]^T (Y^T X)^T$$

$$\beta = \left[ X^T X \right]^{-1} X^T Y$$

$$\boxed{\beta = (X^T X)^{-1} X^T Y} \rightarrow \text{IV eqn}$$

$\beta = \text{values}$

$((m+1) \times 1)$

$[X^T X]^{-1}$

$m+1 \times n$   $n \times (m+1)$

$(m+1) \times m+1$   $(m+1) \times n$

$(m+1) \times n$   $n \times 1$

$(m+1) \times 1$

$(X^T X)^{-1}$  symmetric

$$[ (X^T X)^{-1} ]^T = (X^T X)^{-1}$$

$$X^T X = A$$

$$A A^{-1} = I$$

$$(A A^{-1})^T = I^T$$

$$(A^{-1})^T A^T = I$$

$$(A^{-1})^T I = A^{-1}$$

$$\boxed{(A^{-1})^T = A^{-1}}$$

Proved