

# Video 4 Summary

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## 1 Overview

Firstly , note that here we wish to express the  $\mathbf{3*1}$  vector  $\mathbf{Y} = \begin{bmatrix} 9.8 \\ 9.1 \\ 7.0 \end{bmatrix}$  approximately as a linear combination of the two  $\mathbf{3*1}$  column vectors  $\begin{bmatrix} 3 \\ 4 \\ 2 \end{bmatrix}$  and  $\begin{bmatrix} 4 \\ 1 \\ 3 \end{bmatrix}$ , i.e. we want to find  $\mathbf{A}$  and  $\mathbf{B}$  such that the following holds approximately :-

$$\begin{bmatrix} 3 \\ 4 \\ 2 \end{bmatrix} * \mathbf{A} + \begin{bmatrix} 4 \\ 1 \\ 3 \end{bmatrix} * \mathbf{B} \approx \begin{bmatrix} 9.8 \\ 9.1 \\ 7.0 \end{bmatrix}$$

This is of the form  $\mathbf{Y} \approx \mathbf{X}\beta$ , where  $\mathbf{Y} = \begin{bmatrix} 9.8 \\ 9.1 \\ 7.0 \end{bmatrix}$ ,  $\mathbf{X} = \begin{bmatrix} 3 & 4 \\ 4 & 1 \\ 2 & 3 \end{bmatrix}$  and  $\beta = \begin{bmatrix} A \\ B \end{bmatrix}$ .

Now , we can also write this approximate model as the following:-  
 $\bar{\mathbf{Y}} = \mathbf{X}\bar{\beta} + \varepsilon$  , where  $\bar{\mathbf{Y}}$  is a  $\mathbf{n*1}$  vector ,  $\mathbf{X}$  is a  $\mathbf{n*p}$ ,  $\bar{\beta}$  is a  $\mathbf{p*1}$  and  $\varepsilon$  is a  $\mathbf{n*1}$  vector .

The above form  $\bar{\mathbf{Y}} = \mathbf{X}\bar{\beta} + \varepsilon$  is called the general form of a **Linear Model**, which is nothing but an approximate system of Linear Equations !  
 [ Note :- We are going to put some statistical assumptions over  $\varepsilon$  later on , so that the Model shall then be called a **Linear Statistical Model** ]

## 2 Solving for $\beta$

Now , we try to find  $\beta$  so that  $\mathbf{Y} \approx \mathbf{X}\beta$  holds approximately . Using some Linear Algebra , we shall see that to solve for the required  $\hat{\beta}$  we shall need the orthogonal projection of vector  $\mathbf{Y}$  onto the column space of matrix  $\mathbf{X}$  !

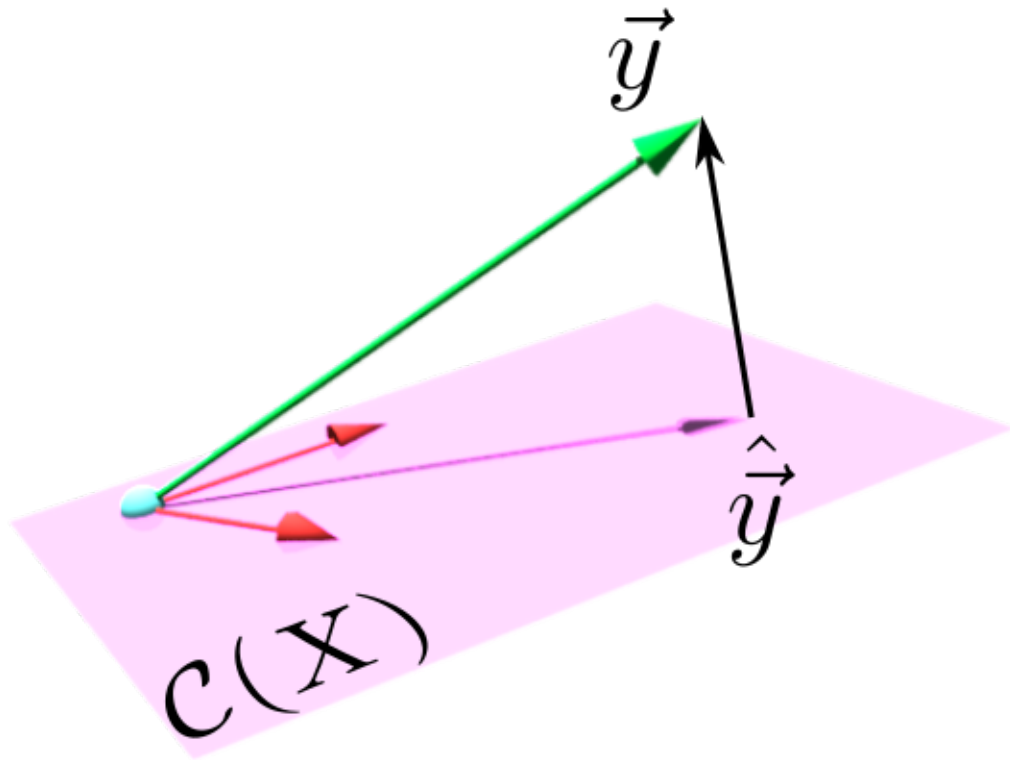


Figure 1: **Projection onto Column Space**

There is an easier technique using Linear Algebra to solve for this :-

1 . Given Equation

$$Y \approx X * \beta$$

2. Multiplying both sides by X transpose and considering equality , we have

$$X^T Y = X^T X \beta \text{ (.... i)}$$

Solving the above equation shall give us our required  $\beta$  and hence the above equation is also the **NORMAL EQUATION** ! Also , this equation is always consistent thus guaranteeing atleast one solution for *beta* .

3. Now if  $X^T X$  is non-singular , then we have a unique solution for  $\beta$  else we shall infinitely many solutions . So , considering  $X^T X$  to be non-singular i.e. invertible , we multiply  $X^T X^{-1}$  on both sides of equation ( ..... i ) to get:-

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

The  $\hat{\beta}$  thus obtained through the process by solving the **NORMAL EQUATION** is called the Least Squares Estimator !