

## Multiple Linear Regression Model

### Expressing the Multiple Linear Regression Model Using Matrix Notation

The multiple linear regression model shown on the previous page:

$$Y = \beta_0 + \beta_1 X_1 + \dots + \beta_k X_k + \varepsilon$$

can be expressed in terms of four matrices:

- $Y$ : the  $n \times 1$  column vector of values of the dependent variable  $Y$
- $X$ : the  $n \times p$  matrix consisting of a column of ones, followed by  $k$  column vectors for the independent variables. Each column of  $X$  contains the values for a particular independent variable
- $\beta$ : the  $p \times 1$  vector of parameters to be estimated, where  $p = k + 1$
- $\varepsilon$ : the  $n \times 1$  vector of random errors

The linear model can now be written in matrix notation as the following:

$$Y = X\beta + \varepsilon$$

where

$$Y = \begin{bmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_n \end{bmatrix}_{(n \times 1)}, \quad X = \begin{bmatrix} 1 & X_{11} & X_{12} & \dots & X_{1k} \\ 1 & X_{21} & X_{22} & \dots & X_{2k} \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & X_{n1} & X_{n2} & & X_{nk} \end{bmatrix}_{(n \times p)}, \quad \beta = \begin{bmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_k \end{bmatrix}_{(p \times 1)}, \quad \varepsilon = \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \vdots \\ \varepsilon_n \end{bmatrix}_{(n \times 1)}$$

In matrix notation, the normal equations are written as the following:

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

The normal equations have a solution, given as the following:

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

If  $X'X$  has an inverse, then the normal equations have a unique solution.

### Testing Model Significance Using the $F$ Statistic

The significance of the model can be tested using the  $F$  statistic shown in the ANOVA table below.

Source of Variation	Degrees of Freedom (df)	Sum of Squares (SS)	Mean Squares (MS=SS/df)	$F$ value	$p$ -value
Due to regression	$k$	$\sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2$	MSM=SSM/dfR	MSM/MSE	If $< \alpha$ (predefined, for example, 0.05), then significant model
Error	$n - k - 1$		MSE=SSE/dfE		

		$\sum_{i=1}^n (Y_i - \hat{Y}_i)^2$			
Total, corrected for mean $\bar{Y}$	$n - 1$	$\sum_{i=1}^n (Y_i - \bar{Y})^2$			

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