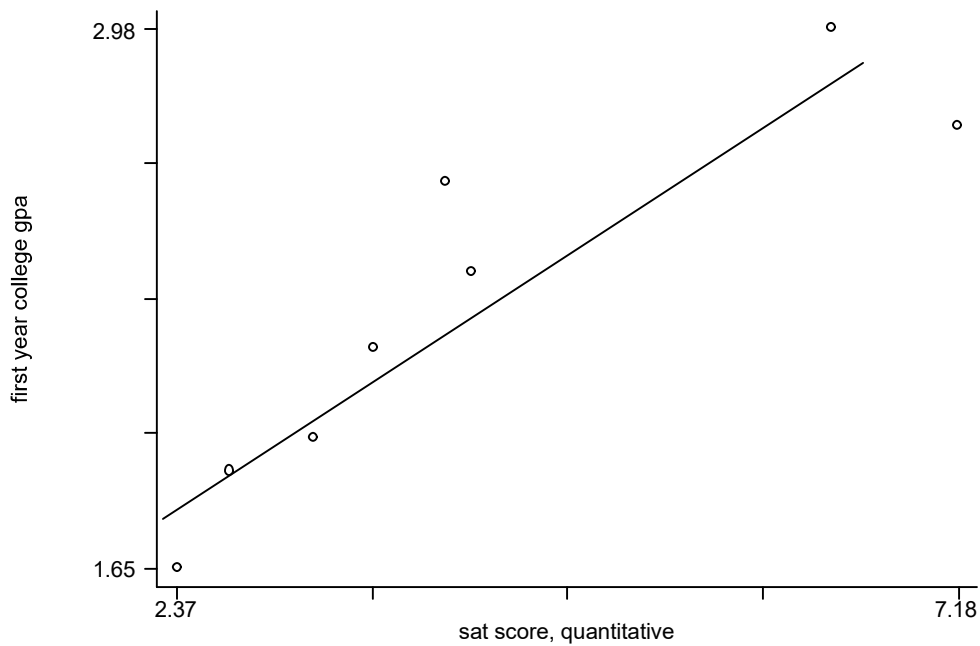


Chapter 1

A Review of the Linear Regression Model

1. a. Construct a scatterplot with GPA on the y -axis and SAT-Quant. on the x -axis. Fit by hand (and straight edge) the estimated linear regression line. Comment on the relationship between these two variables. (p.6)



There is a positive linear relationship between GPA and SAT-Quant.

- b. Using the formulas for a two-variable OLS regression model, compute the slope and intercept for the following model: $GPA = \alpha + \beta_1(SAT-Quant.)$. (p.7)

$$\text{Slope (1.5): } \beta_1 = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2}$$

$$\begin{aligned} \beta_1 &= \text{COV}\{XY\}^a \div \text{SS}(x)^b \\ \beta_1 &= 4.954 \div 20.71279 \\ \beta_1 &= \mathbf{0.239176} \end{aligned}$$

Intercept (1.6): $\alpha = \bar{Y} - \beta_1 \bar{X}$

" " mean of Y & (Slope \times the mean of X)

" " 2.3 & (0.239176 \times 4.20625)

" " **1.293966**

- c. Compute the predicted values, the residuals, the Sum of Squared Errors (SSE), and the R^2 for the model. (pp.5-8)

As an example of the predicted and residual values we will use the data point (3.21, 1.97). See the table on page 5 for a complete list.

Predicted Value(1.3): $\hat{Y} = \alpha + \beta_1 X$

\hat{Y} " intercept + (slope \times an X value)

\hat{Y} " 1.29 + (.239 \times 3.21)

\hat{Y} " **2.06**

Residual: $e = Y - \hat{Y}$

e " Y value & predicted Y value

e " 1.97 & 2.06

e " **-.09**

Sum of Squared Errors^c (1.4): $\text{SSE} = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$

^a The covariance of X and Y .

^b The Sum of Squares of X .

^c Also termed the Residual Sum of Squares.