

**Note**

See page 242 for a quick review of how to obtain the sample estimates for the linear regression model. The estimates can also be obtained using technology.

In order to perform inference on the linear model, some assumptions about the nature of the error terms are required.

**Assumptions about the Error Term in the Linear Model**

1. The  $\varepsilon_i$  are presumed to be normally distributed with a mean of 0 and a variance of  $\sigma_e^2$ .
2. The  $\varepsilon_i$  are presumed to be independent of each other.

**PROPERTIES****Technology**

The instructions for calculating the coefficients for the simple linear regression model using various technologies can be found on [stat.hawkeslearning.com](http://stat.hawkeslearning.com) under

**Discovering Statistics and Data, Fourth Edition > Technology Instructions > Regression > Simple Linear Regression.**

With the addition of the error term, the model's parameters are  $\beta_0$ ,  $\beta_1$ , and  $\sigma_e^2$ . The estimation of these quantities was discussed in Chapter 5. The actual verification of these assumptions cannot be made prior to a regression analysis, but they can be validated by doing an analysis of the **residuals** (errors). A residual analysis is beyond the scope of this book, therefore, we will assume that the error terms satisfy the necessary assumptions in order to proceed with inference in the regression analysis.

In addition to the formal assumptions stated above, a linear model should be used to fit data that appears to be reasonably linearly related. Because of the wide availability of computer programs that calculate least squares estimates, you will not need to manually calculate estimates very often.

## 13.1 Exercises

### Basic Concepts

1. Why is an error term incorporated in the simple linear model?
2. What does the error term represent?
3. What assumptions are made about the error term in the simple linear model?
4. What are the parameters of the simple linear regression model? Identify their estimates from the sample.

### Exercises

5. Consider the following data and regression output relating a student's grade to the number of absences from class.
  - a. Determine the independent and dependent variables and write the equation of the model we desire to estimate.
  - b. Determine the estimated model (regression equation) for predicting a student's grade based on the number of absences from class.

Class Absences and Grade	
Number of Absences	Grade
3	3.9
5	3.8
6	2.9
6	2.7
6	2.4
7	2.3
8	1.9

## SUMMARY OUTPUT

*Regression Statistics*

Multiple R	0.917589
R Square	0.84197
Adjusted R Square	0.810364
Standard Error	0.329598
Observations	7

## ANOVA

	<i>df</i>	<i>SS</i>
Regression	1	2.89396978
Residual	5	0.543173077
Total	6	3.437142857

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>
Intercept	5.427885	0.51610437	10.51703
Number of Absences	-0.44135	0.085509913	-5.16134

- c. Determine the mean square error (MSE). (See Section 5.3.)
6. The following table of data gives the weeks of gestation and corresponding birth weights for a sample of ten babies.

Gestation Period and Birth Weight	
Weeks of Gestation	Birth Weight
34	5.9
34	5.7
35	6.2
36	6.6
36	6.8
37	7.0
38	7.2
38	7.5
39	8.0
40	8.2

- a. Determine the independent and dependent variables and write the equation of the model we desire to estimate.
- b. Determine the estimated model (regression equation) for predicting a baby's birth weight based on the number of weeks of gestation.
- c. Determine the mean square error (MSE). (See Section 5.3.)
- d. Determine the coefficient of determination and explain its meaning in terms of the problem.

### Data

The data set can be found on [stat.hawkeslearning.com](http://stat.hawkeslearning.com) under **Discovering Statistics and Data, Fourth Edition > Data Sets > Global Statistics by Country.**

7. Use the Global Statistics by Country data set to answer the following questions.
  - a. Determine the regression equation for predicting birth rate based on female literacy rate.
  - b. Determine the coefficient of determination and explain its meaning in terms of the problem.
  - c. Find the predicted birth rate for a female literacy rate of 70.
  - d. Find the predicted birth rate for a female literacy rate of 90.
  - e. As the female literacy rate rises what happens to the birth rate? Does this make sense? Explain your answer.

## 13.2 Inference Concerning $\beta_1$

Since  $\beta_1$  specifies the rate of change between  $x$  and  $y$ , in most linear models the parameter of interest is  $\beta_1$ . Two inferential techniques are useful in evaluating the estimate of  $\beta_1$ . Confidence intervals, similar in structure to those used for means and proportions, will be developed. In addition, a hypothesis testing procedure will be presented to test whether  $\beta_1$  is equal to some particular value.

### The Confidence Interval for $\beta_1$

Developing a confidence interval for  $\beta_1$  requires thinking about the estimate  $b_1$  as a random variable. Each random sample from the population will produce different data and hence different least squares estimates of  $b_0$  and  $b_1$ . The confidence interval will serve two purposes, to place bounds on the location of  $\beta_1$  as well as to provide information about the quality of the point estimate  $b_1$ . The form of the confidence interval is familiar.

$$\text{Sample estimate of parameter} \pm \left( \begin{array}{c} \text{A certain number of standard} \\ \text{deviations units depending on} \\ \text{the desired confidence} \end{array} \right) \cdot \left( \begin{array}{c} \text{The standard} \\ \text{deviation of the} \\ \text{sample estimate} \end{array} \right)$$

The **sample estimate** of  $\beta_1$  is  $b_1$ . The variance of  $b_1$  is given by

$$\sigma_{b_1}^2 = \frac{\sigma_e^2}{\sum (x_i - \bar{x})^2}$$

but like all population measurements,  $\sigma_{b_1}^2$  usually has to be estimated from the data. The sample estimate of the variance of  $b_1$  is given by

$$s_{b_1}^2 = \frac{s_e^2}{\sum (x_i - \bar{x})^2}$$

The only difference in the computation of  $\sigma_{b_1}^2$  and  $s_{b_1}^2$  is the replacement of the population variance of the error terms,  $\sigma_e^2$ , with the corresponding sample statistic,  $s_e^2$ . The **standard deviation (standard error) of the sample estimate  $b_1$**  is

$$s_{b_1} = \sqrt{\frac{s_e^2}{\sum (x_i - \bar{x})^2}}$$