David S. Moore • George P. McCabe

Introduction to the Practice of Statistics

Fifth Edition

Chapter 10:

Inference for Regression

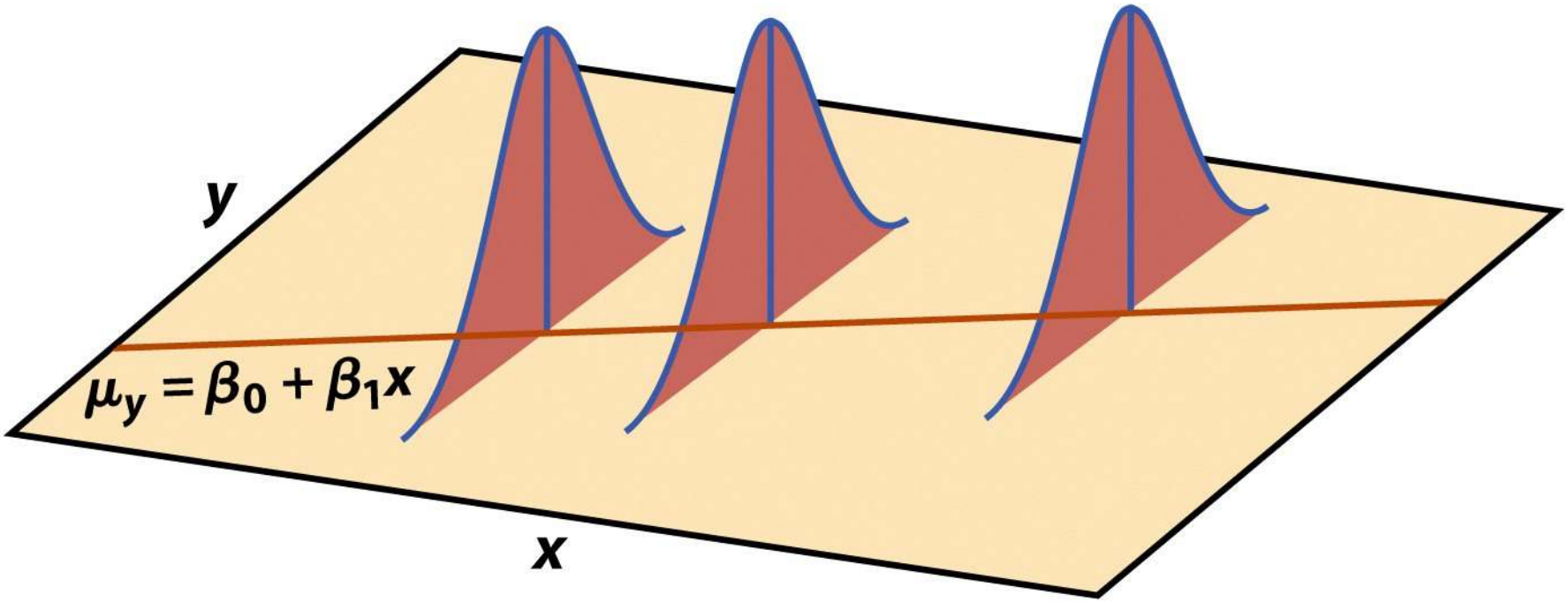

Figure 10-2
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SIMPLE LINEAR REGRESSION MODEL

Given n observations on the explanatory variable x and the response variable y,

$$(x_1, y_1), \quad (x_2, y_2), \quad \ldots, \quad (x_n, y_n)$$

the **statistical model for simple linear regression** states that the observed response y_i when the explanatory variable takes the value x_i is

$$y_i = \beta_0 + \beta_1 x_i + \epsilon_i$$

Here $\beta_0 + \beta_1 x_i$ is the mean response when $x = x_i$. The deviations ϵ_i are assumed to be independent and normally distributed with mean 0 and standard deviation σ.

The parameters of the model are β_0, β_1, and σ.

Definition, pg 638
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The regression equation is
MPG = - 7.80 + 7.87 logmph

Predictor         Coef        StDev            T          P
Constant       -7.796        1.155        -6.75      0.000
logmph         7.8742       0.3541        22.24      0.000

S = 0.9995       R-Sq = 89.5%      R-Sq(adj) = 89.3%
```

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	Root MSE	0.99952		R-Square	0.8950	
	Dependent Mean	17.72500		Adj R-Sq	0.8932	
	Coeff Var	5.63902				

Variable	DF	Parameter Estimate	Standard Error	t Value	Pr > |t|	95% Confidence Limits	
Intercept	1	–7.79625	1.15494	–6.75	<.0001	–10.10812	–5.48438
logmph	1	7.87422	0.35411	22.24	<.0001	7.16539	8.58305

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CONFIDENCE INTERVALS AND SIGNIFICANCE TESTS FOR REGRESSION SLOPE AND INTERCEPT

A **level *C* confidence interval for the intercept** β_0 is

$$b_0 \pm t^* \mathrm{SE}_{b_0}$$

A **level *C* confidence interval for the slope** β_1 is

$$b_1 \pm t^* \mathrm{SE}_{b_1}$$

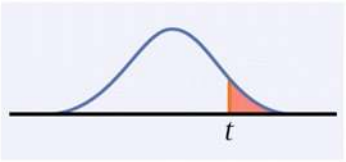
In these expressions t^* is the value for the $t(n-2)$ density curve with area C between $-t^*$ and t^*.

To test the hypothesis $H_0: \beta_1 = 0$, compute the **test statistic**

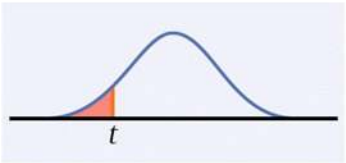
$$t = \frac{b_1}{\mathrm{SE}_{b_1}}$$

The **degrees of freedom** are $n-2$. In terms of a random variable T having the $t(n-2)$ distribution, the P-value for a test of H_0 against

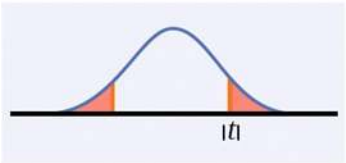
$H_a: \beta_1 > 0$ is $P(T \geq t)$



$H_a: \beta_1 < 0$ is $P(T \leq t)$



$H_a: \beta_1 \neq 0$ is $2P(T \geq |t|)$



Definition, pg 644

CONFIDENCE INTERVAL FOR A MEAN RESPONSE

A **level C confidence interval for the mean response** μ_y when x takes the value x^* is

$$\hat{\mu}_y \pm t^* \mathrm{SE}_{\hat{\mu}}$$

where t^* is the value for the $t(n-2)$ density curve with area C between $-t^*$ and t^*.

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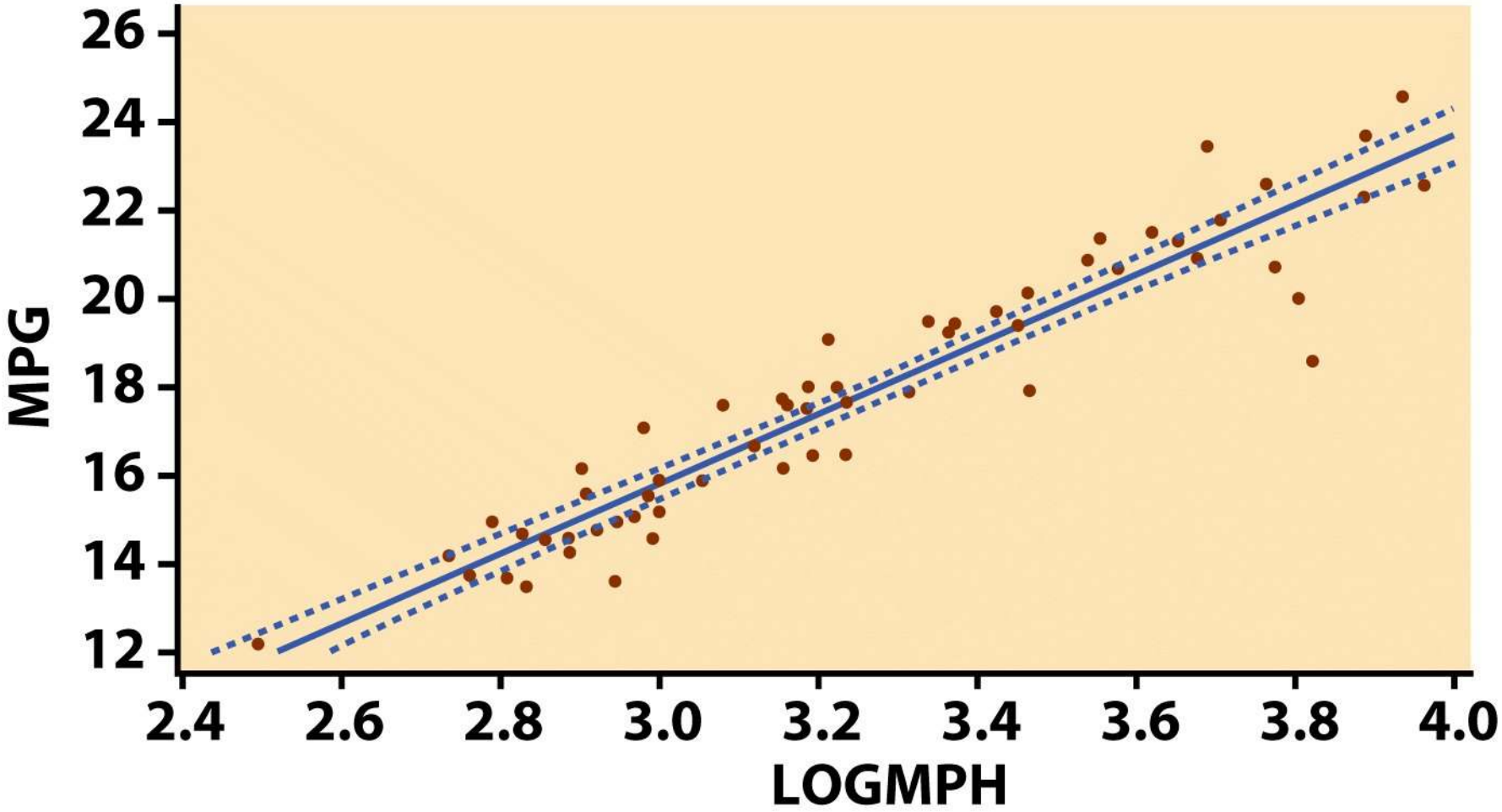

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PREDICTION INTERVAL FOR A FUTURE OBSERVATION

A **level C prediction interval for a future observation** on the response variable y from the subpopulation corresponding to x^* is

$$\hat{y} \pm t^* \mathrm{SE}_{\hat{y}}$$

where t^* is the value for the $t(n-2)$ density curve with area C between $-t^*$ and t^*.

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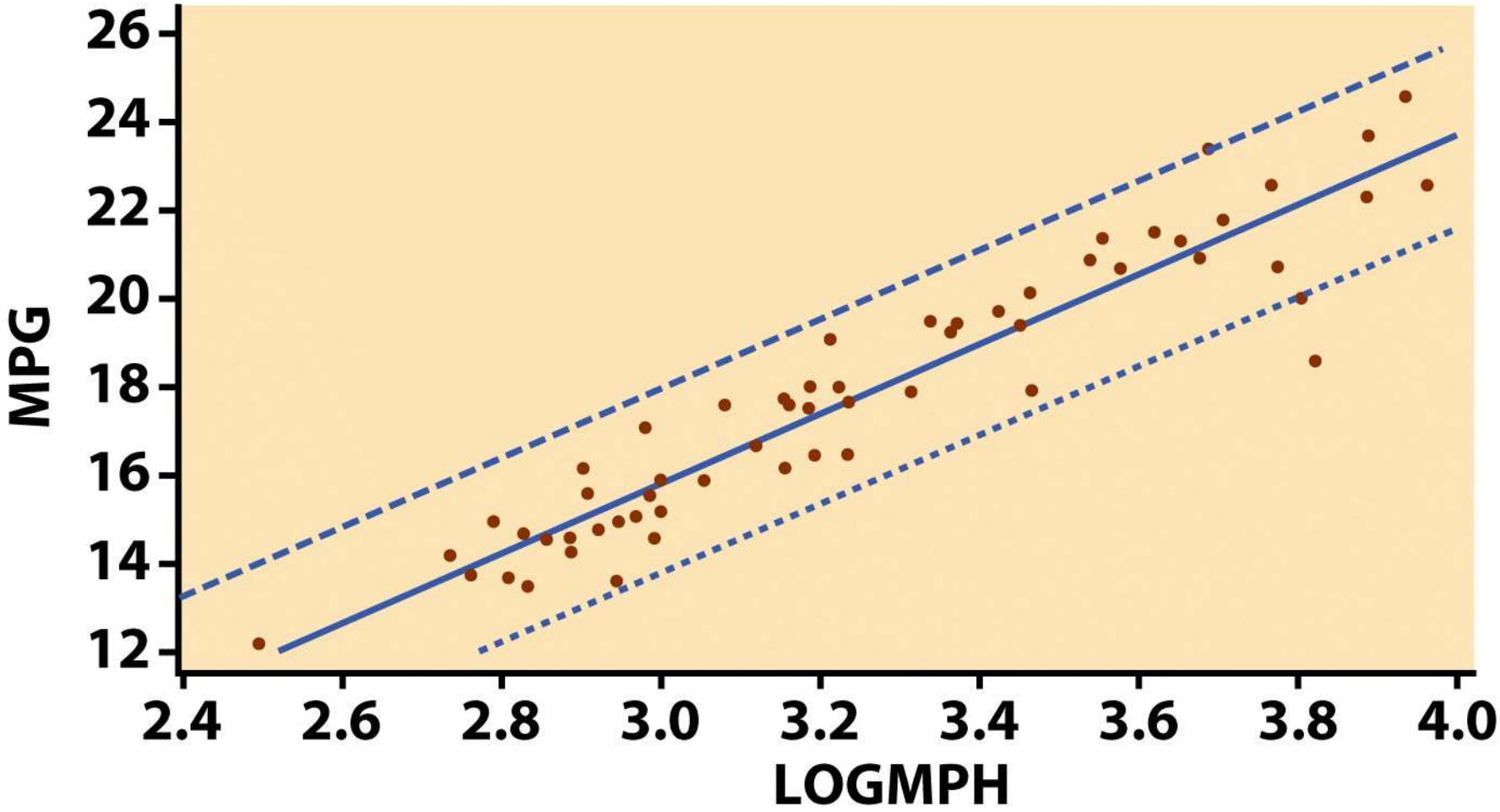

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SUMS OF SQUARES, DEGREES OF FREEDOM, AND MEAN SQUARES

Sums of squares represent variation present in the responses. They are calculated by summing squared deviations. **Analysis of variance** partitions the total variation between two sources.

The sums of squares are related by the formula

$$\text{SST} = \text{SSM} + \text{SSE}$$

That is, the total variation is partitioned into two parts, one due to the model and one due to deviations from the model.

Degrees of freedom are associated with each sum of squares. They are related in the same way:

$$\text{DFT} = \text{DFM} + \text{DFE}$$

To calculate **mean squares,** use the formula

$$\text{MS} = \frac{\text{sum of squares}}{\text{degrees of freedom}}$$

Definition, pg 654

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ANALYSIS OF VARIANCE *F* TEST

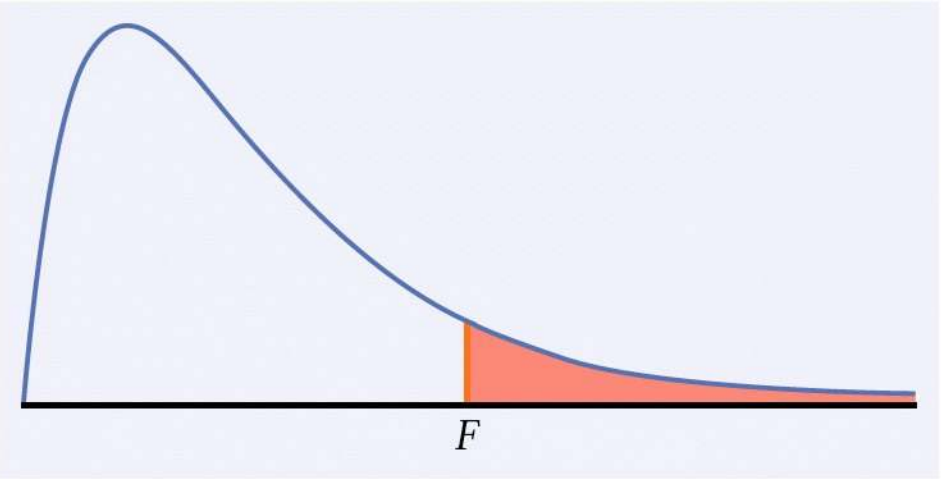
In the simple linear regression model, the hypotheses

$$H_0: \beta_1 = 0$$

$$H_a: \beta_1 \neq 0$$

are tested by the ***F* statistic**

$$F = \frac{\text{MSM}}{\text{MSE}}$$



The *P*-value is the probability that a random variable having the $F(1, n-2)$ distribution is greater than or equal to the calculated value of the F statistic.

Definition, pg 655

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ANOVA[b]

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	493.989	1	493.989	494.467	.000[a]
	Residual	57.944	58	.999		
	Total	551.932	59			

a. Predictors: (Constant), LOGMPH

b. Dependent Variable: MPG

Model Summary

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate
1	.946[a]	.895	.893	.9995

a. Predictors: (Constant), LOGMPH

Coefficients[a]

Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.
		B	Std. Error	Beta		
1	(Constant)	-7.796	1.155		-6.750	.000
	LOGMPH	7.874	.354	.946	22.237	.000

a. Dependent Variable: MPG

Figure 10-12
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STANDARD ERRORS FOR ESTIMATED REGRESSION COEFFICIENTS

The standard error of the slope b_1 of the least-squares regression line is

$$\mathrm{SE}_{b_1} = \frac{s}{\sqrt{\sum (x_i - \bar{x})^2}}$$

The standard error of the intercept b_0 is

$$\mathrm{SE}_{b_0} = s\sqrt{\frac{1}{n} + \frac{\bar{x}^2}{\sum (x_i - \bar{x})^2}}$$

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STANDARD ERRORS FOR $\hat{\mu}$ AND $\hat{y}$

The standard error of $\hat{\mu}$ is

$$\text{SE}_{\hat{\mu}} = s\sqrt{\frac{1}{n} + \frac{(x^* - \overline{x})^2}{\sum(x_i - \overline{x})^2}}$$

The standard error for predicting an individual response $\hat{y}$ is[4]

$$\text{SE}_{\hat{y}} = s\sqrt{1 + \frac{1}{n} + \frac{(x^* - \overline{x})^2}{\sum(x_i - \overline{x})^2}}$$

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TEST FOR A ZERO POPULATION CORRELATION

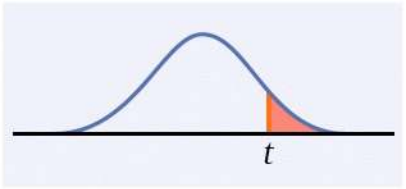
To test the hypothesis $H_0: \rho = 0$, compute the t statistic:

$$t = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}}$$

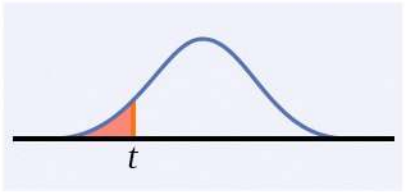
where n is the sample size and r is the sample correlation.

In terms of a random variable T having the $t(n-2)$ distribution, the P-value for a test of H_0 against

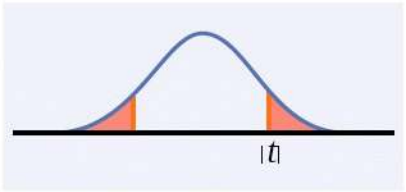
$H_a: \rho > 0$ is $P(T \geq t)$



$H_a: \rho < 0$ is $P(T \leq t)$



$H_a: \rho \neq 0$ is $2P(T \geq |t|)$



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Correlations

		LOGMPH	MPG
LOGMPH	Pearson Correlation	1	.946**
	Sig. (2-tailed)	.	.000
	N	60	60
MPG	Pearson Correlation	.946**	1
	Sig. (2-tailed)	.000	.
	N	60	60

**. Correlation is significant at the 0.01 level (2-tailed).

Figure 10-14
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