

## Chapter 5

# Prediction and Association

### Section 5.1 Pearson Correlation Coefficient

#### Description

The Pearson correlation coefficient (sometimes called the *Pearson product-moment correlation coefficient* or simply the *Pearson  $r$* ) determines the strength of the linear relationship between two variables.

#### Assumptions

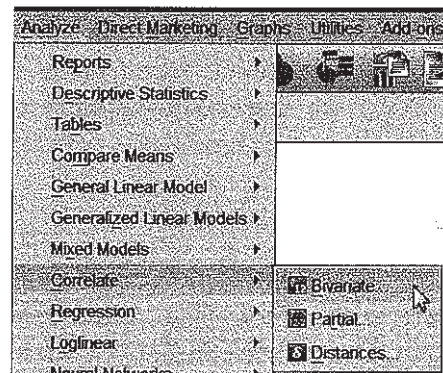
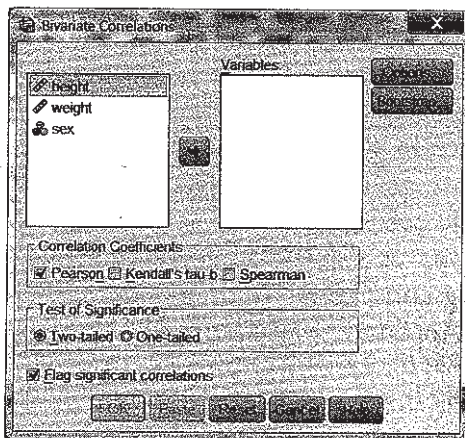
Both variables should be measured on **interval** or **ratio scales** (or as a **dichotomous** nominal variable). If a relationship exists between them, that relationship should be linear. Because the Pearson correlation coefficient is computed with z-scores, both variables should also be normally distributed. If your data do not meet these assumptions, consider using the Spearman *rho* correlation coefficient instead.

#### SPSS Data Format

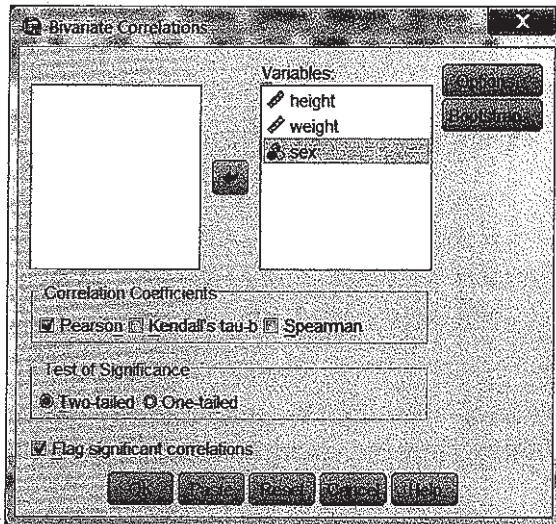
Two variables are required in your SPSS data file. Each subject must have data for both variables.

#### Running the Command

To select the Pearson correlation coefficient, click *Analyze*, then *Correlate*, then *Bivariate* (*bivariate* refers to two variables). This will bring up the Bivariate Correlations **dialog box**. This example uses the HEIGHT.sav data file entered at the start of Chapter 4.



Move at least two variables from the box at the left into the box at the right by using the transfer arrow (or by double-clicking each variable). Make sure that a check is in the *Pearson* box under *Correlation Coefficients*. It is acceptable to move more than two variables.



For our example, we will move all three variables over and click *OK*.

*Reading the Output*

The output consists of a **correlation matrix**. Every variable you entered in the command is represented as both a row and a column. We entered three variables in our command. Therefore, we have a 3 × 3 table. There are also three rows in each cell—the correlation, the **significance** level, and the *N*. If a correlation is significant at less than the .05 level, a single \* will appear next to the correlation. If it is significant at

the .01 level or lower, \*\* will appear next to the correlation. For instance, all of the correlations in the output below have a **significance** level of < .01, so they are flagged with \*\* to indicate that they are less than .01.

To read the correlations, select a row and a column. For instance, the correlation between height and weight is determined through selection of the **WEIGHT** row and the **HEIGHT** column (.806). We get the same answer by selecting the **HEIGHT** row and the **WEIGHT** column. The correlation between a variable and itself is always 1, so there is a diagonal set of 1s.

Correlations

		height	weight	sex
height	Pearson Correlation	1	.806**	-.644**
	Sig. (2-tailed)		.000	.007
	N	16	16	16
weight	Pearson Correlation	.806**	1	-.968**
	Sig. (2-tailed)	.000		.000
	N	16	16	16
sex	Pearson Correlation	-.644**	-.968**	1
	Sig. (2-tailed)	.007	.000	
	N	16	16	16

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

*Drawing Conclusions*

The correlation coefficient will be between -1.0 and +1.0. Coefficients close to 0.0 represent a weak relationship. Coefficients close to 1.0 or -1.0 represent a strong relationship. Generally, correlations with an absolute value greater than 0.7 are considered strong. Correlations with an absolute value less than 0.3 are considered weak. Correlations with an absolute value between 0.3 and 0.7 are considered moderate. Positive correlations indicate that as one variable gets larger, the other variable also gets larger. Negative correlations indicate that as one variable gets larger, the other variable gets smaller.

Significant correlations are flagged with asterisks. A significant correlation indicates a reliable relationship, but not necessarily a strong correlation. With enough participants, a very small correlation can be significant. See Appendix A for a discussion of **effect sizes** for correlations.

### Phrasing Results That Are Significant

In the preceding example, we obtained a correlation of .806 between HEIGHT and WEIGHT. A correlation of .806 is a strong positive correlation, and it is significant at the .001 level. Thus, we could state the following in a results section:

A Pearson correlation coefficient was calculated for the relationship between participants' height and weight. A strong positive correlation was found ( $r(14) = .806, p < .001$ ), indicating a significant linear relationship between the two variables. Taller participants tend to weigh more.

The conclusion states the direction (positive), strength (strong), value (.806), degrees of freedom (14), and **significance** level ( $< .001$ ) of the correlation. In addition, a statement of direction is included (taller is heavier).

Note that the degrees of freedom given in parentheses is 14. The output indicates an  $N$  of 16. While most SPSS procedures give degrees of freedom, the *Correlation* command gives only the  $N$  (the number of pairs). For a correlation, the degrees of freedom is  $N - 2$ .

### Phrasing Results That Are Not Significant

Using our SAMPLE.sav data set from the previous chapters, we can calculate a correlation between ID and GRADE. If we do so, we get the output shown to the right. The correlation has a **significance** level of .783. Thus, we could write the following in a results section (note that the degrees of freedom is  $N - 2$ ):

Correlations

		ID	GRADE
ID	Pearson Correlation	1.000	.217
	Sig. (2-tailed)	.	.783
	N	4	4
GRADE	Pearson Correlation	.217	1.000
	Sig. (2-tailed)	.783	.
	N	4	4

A Pearson correlation was calculated examining the relationship between participants' ID numbers and grades. A weak correlation that was not significant was found ( $r(2) = .217, p > .05$ ). ID number is not related to grade in the course.

### Practice Exercise

Use Practice Data Set 2 in Appendix B. Determine the value of the Pearson correlation coefficient for the relationship between SALARY and YEARS OF EDUCATION, and phrase your results.

## Section 5.2 Spearman Correlation Coefficient

### Description

The Spearman correlation coefficient determines the strength of the relationship between two variables. It is a nonparametric procedure. Therefore, it is weaker than the Pearson correlation coefficient, but it can be used in more situations.