



KORELASI

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Correlation is primarily concerned with investigating whether a relationship exists and with determining its magnitude and direction. When two variables vary together, such as loneliness and depression, they are said to be correlated

In order to show quantitatively the extent to which two variables are related, it is necessary to calculate a correlation coefficient. There are many types of correlation coefficients, and the decision of which one to employ with a specific set of data depends on the following factors:

- The level of measurement on which each variable is measured
- The nature of the underlying distribution (continuous or discrete)
- The characteristics of the distribution of the scores (linear or nonlinear)

correlational technique the researcher uses, they have the following characteristics in common:

1. Two sets of measurements are obtained on the same individuals or on pairs of individuals who are matched on some basis.
2. The values of the correlation coefficients vary between $+1.00$ and -1.00 . Both of these extremes represent perfect relationships between the variables, and 0.00 represents the absence of a relationship.
3. A *positive relationship* means that individuals obtaining high scores on one variable tend to obtain high scores on a second variable. The converse is also true, that is, individuals scoring low on one variable tend to score low on a second variable.
4. A *negative relationship* means that individuals scoring low on one variable tend to score high on a second variable. Conversely, individuals scoring high on one variable tend to score low on a second variable.

- For each subject in the study, there must be *related pairs of scores*. That is, if a subject has a score on variable X , then the same subject must also receive a score on variable Y .
- The variables should be measured at least at the *ordinal level*.

- **Linearity**—The relationship between the two variables must be *linear*, that is, the relationship can be most accurately represented by a straight line.
- **Homoscedasticity**—The variability of scores along the Y variable should remain constant at all values of the X variable.

Example 1: Pearson Product Moment Correlation Coefficient

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Assume that a researcher wishes to ascertain whether there is a relationship between grade point average (GPA) and the scores on a reading-comprehension (READ) test of 15 first-year students. The researcher recorded the pair of scores below, together with their rankings.

Student	Read	Read_Rank	GPA	GPA_Rank
s1	38	13	2.1	13
s2	54	3	2.9	6
s3	43	10	3.0	5
s4	45	8	2.3	12
s5	50	4	2.6	7.5
s6	61	1	3.7	1
s7	57	2	3.2	4
s8	25	15	1.3	15
s9	36	14	1.8	14
s10	39	11.5	2.5	9.5
s11	48	5.5	3.4	2
s12	46	7	2.6	7.5
s13	44	9	2.4	11
s14	39	11.5	2.5	9.5
s15	48	5.5	3.3	3

Data Entry Format


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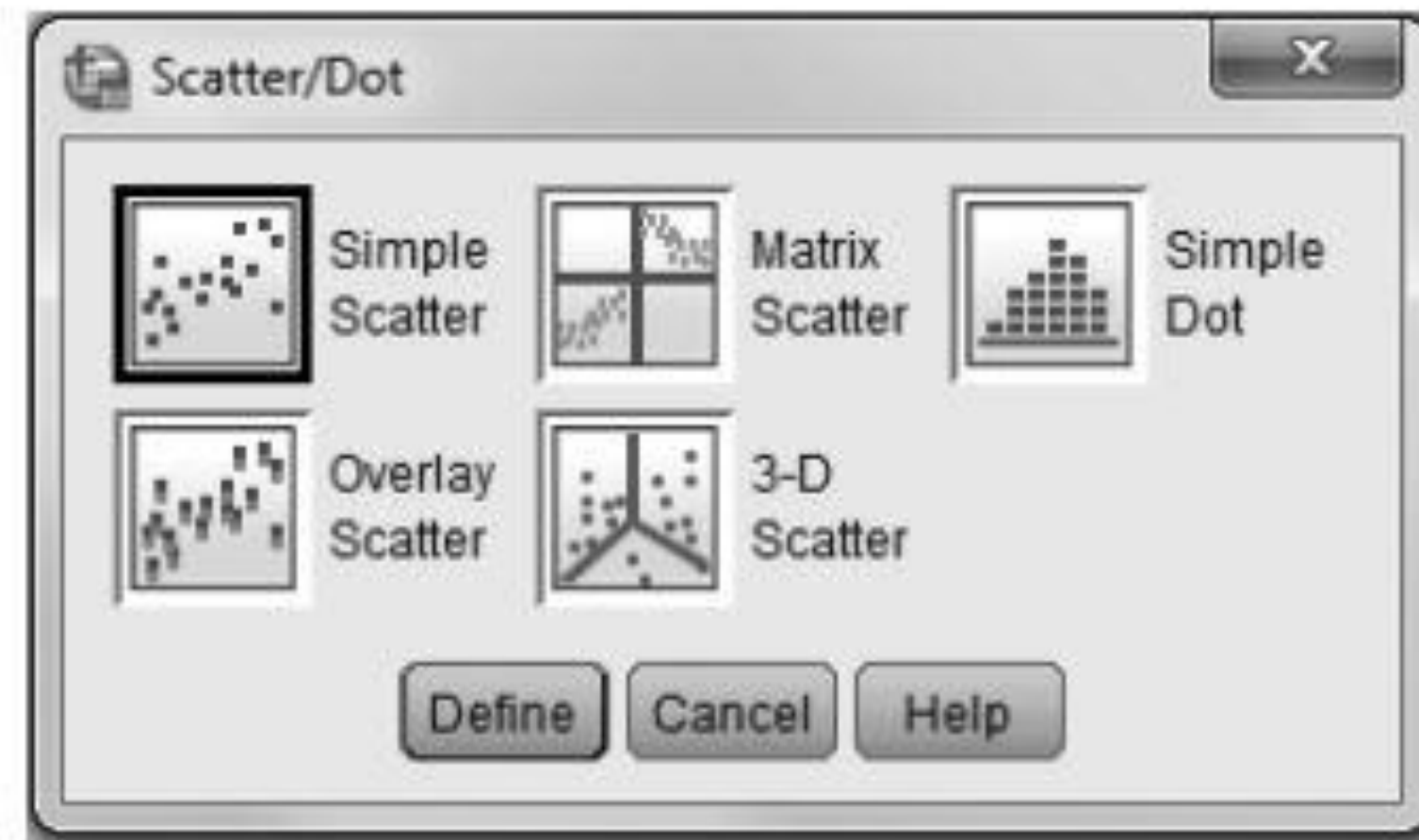
The data set has been saved under the name **CORR.SAV**.

Variables	Column(s)	Code
READ	1	Reading score
READ_RANK	2	Ranking
GPA	3	Grade point average
GPA_RANK	4	Ranking

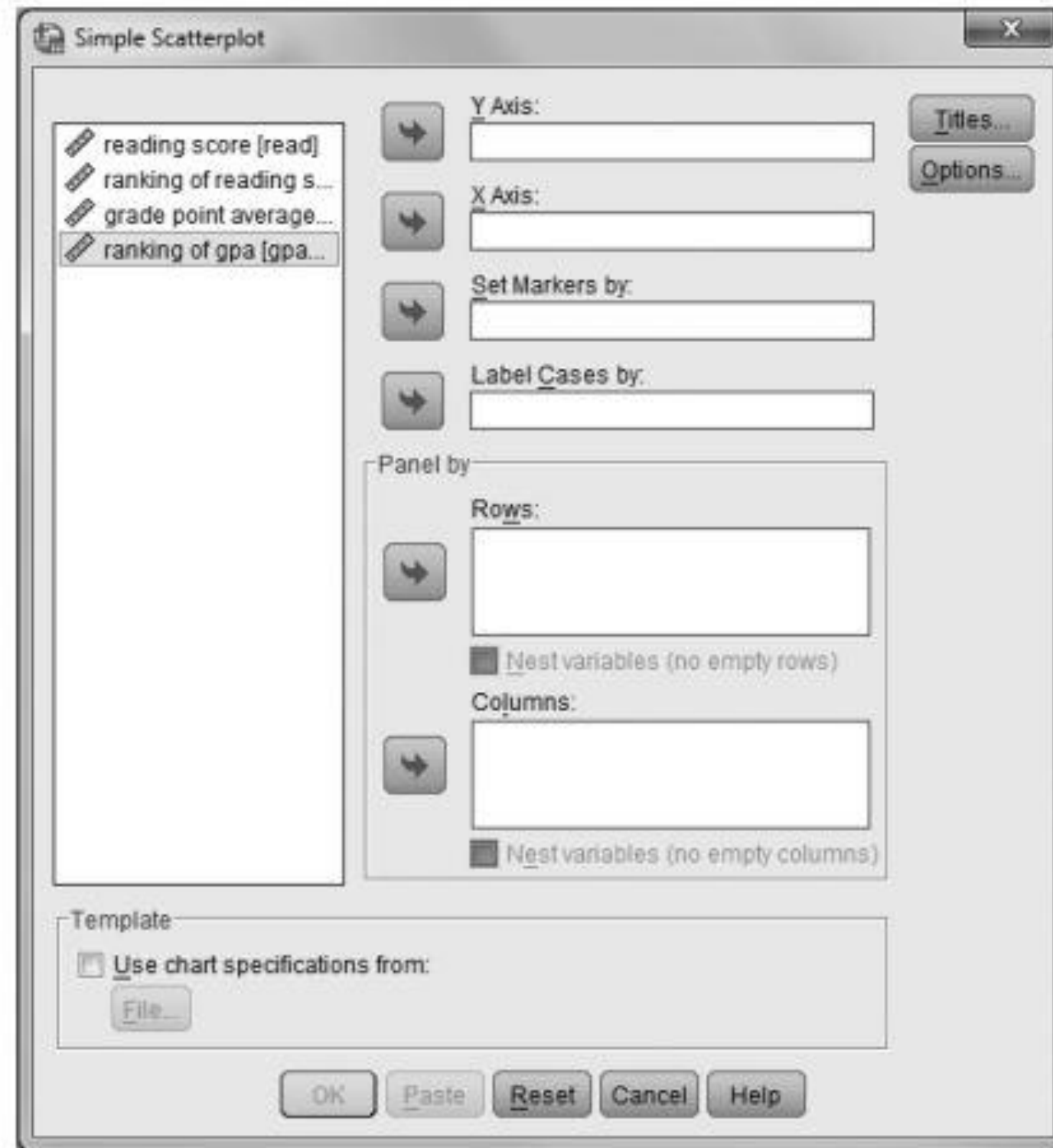
Testing Assumptions

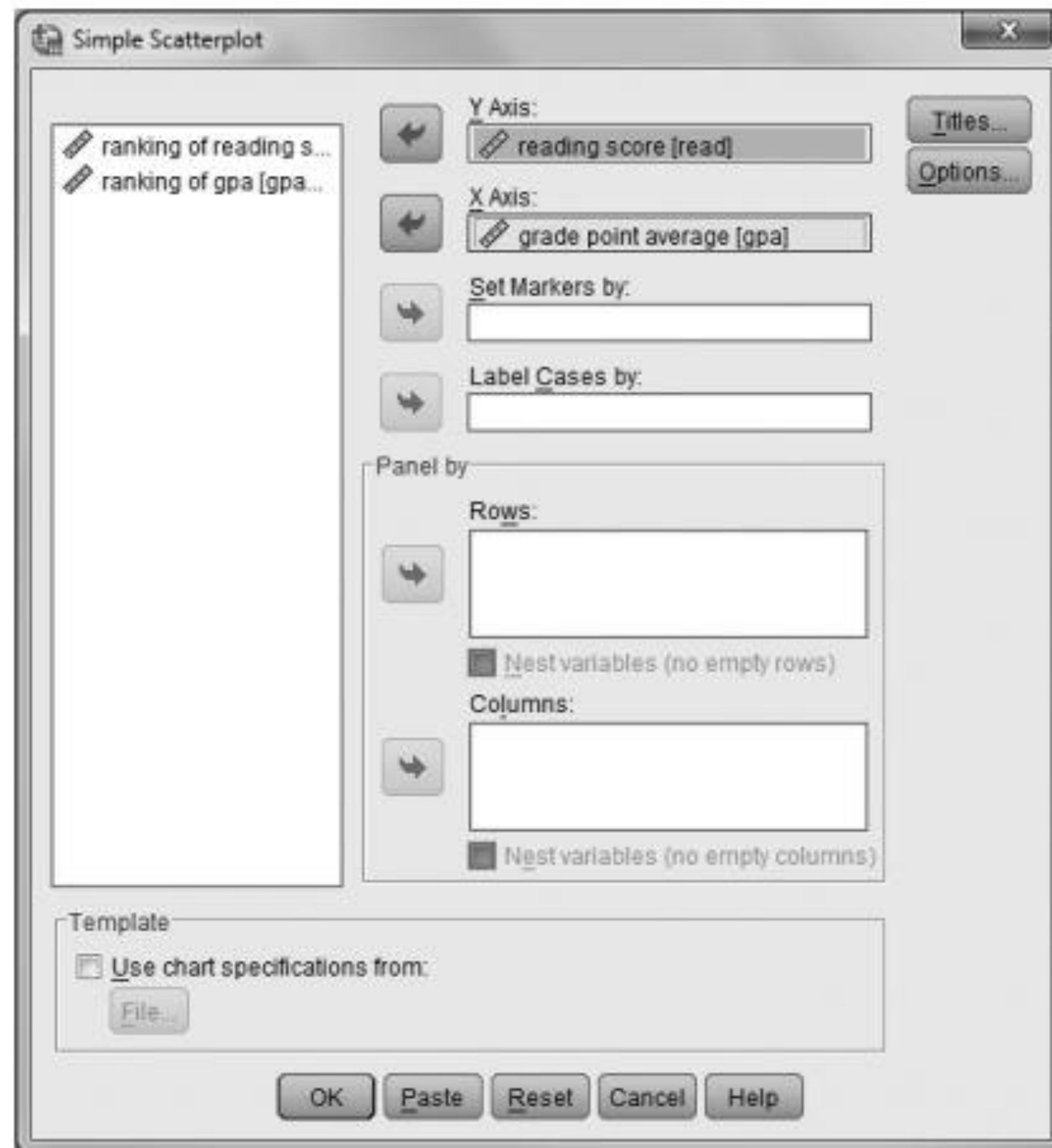
10

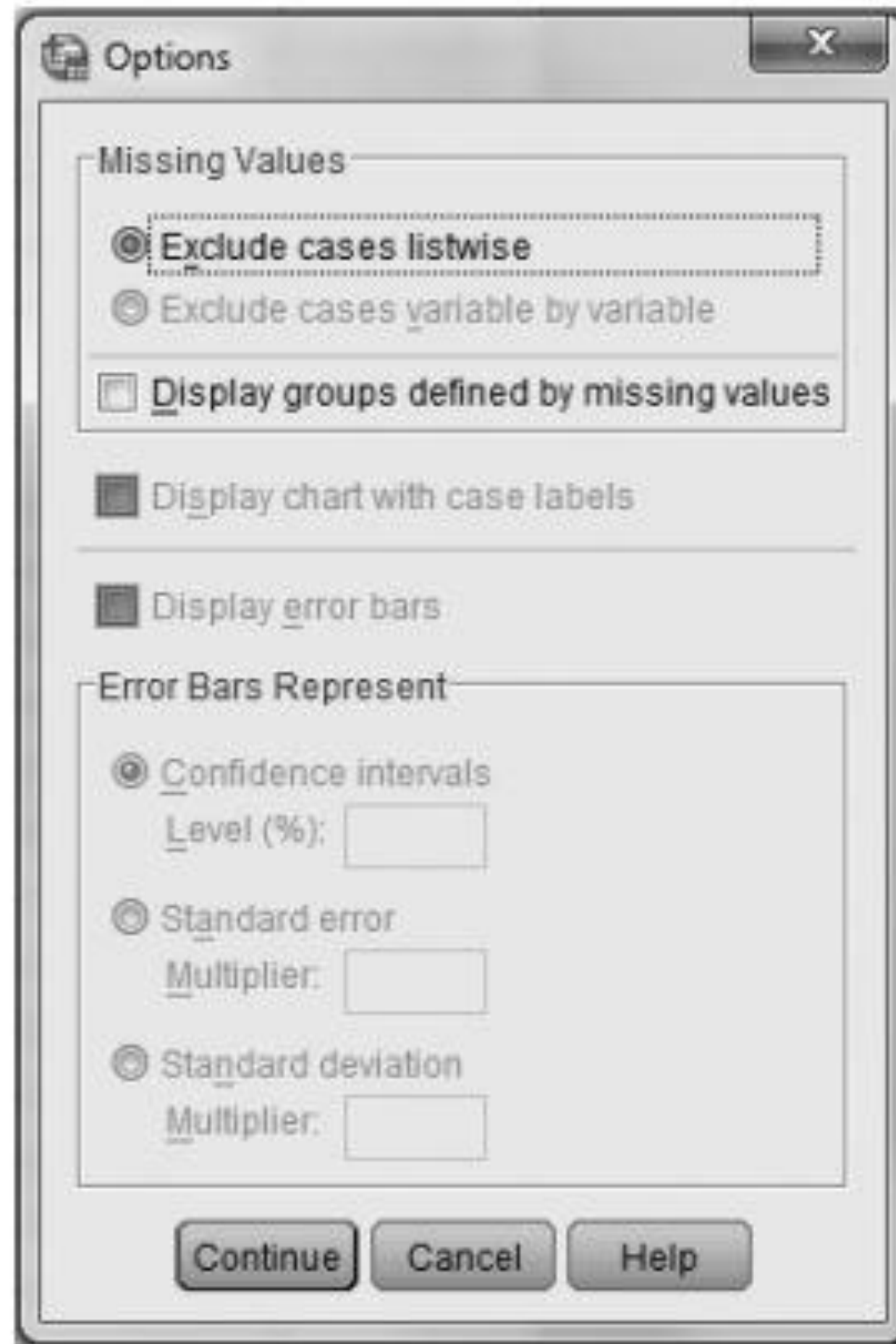
1. From the menu bar, click **Graphs**, then **Legacy Dialogs**, and then **Scatter/Dot....** The following **Scatter/Dot** window will open. Click (highlight) the  Simple Scatter icon.



2. Click **Define** to open the **Simple Scatterplot** window below.







The image shows the 'Options' dialog box in SPSS. It has a title bar with a close button (X). The dialog is divided into two main sections: 'Missing Values' and 'Error Bars Represent'. In the 'Missing Values' section, there are three radio buttons: 'Exclude cases listwise' (selected), 'Exclude cases variable by variable', and 'Display groups defined by missing values'. Below these are two checkboxes: 'Display chart with case labels' and 'Display error bars'. The 'Error Bars Represent' section contains three radio buttons: 'Confidence intervals' (selected), 'Standard error', and 'Standard deviation'. Each of the last three radio buttons has a corresponding 'Level (%)' or 'Multiplier' input field. At the bottom of the dialog are three buttons: 'Continue', 'Cancel', and 'Help'.

Options

Missing Values

☒ Exclude cases listwise

☐ Exclude cases variable by variable

☐ Display groups defined by missing values

☐ Display chart with case labels

☐ Display error bars

Error Bars Represent

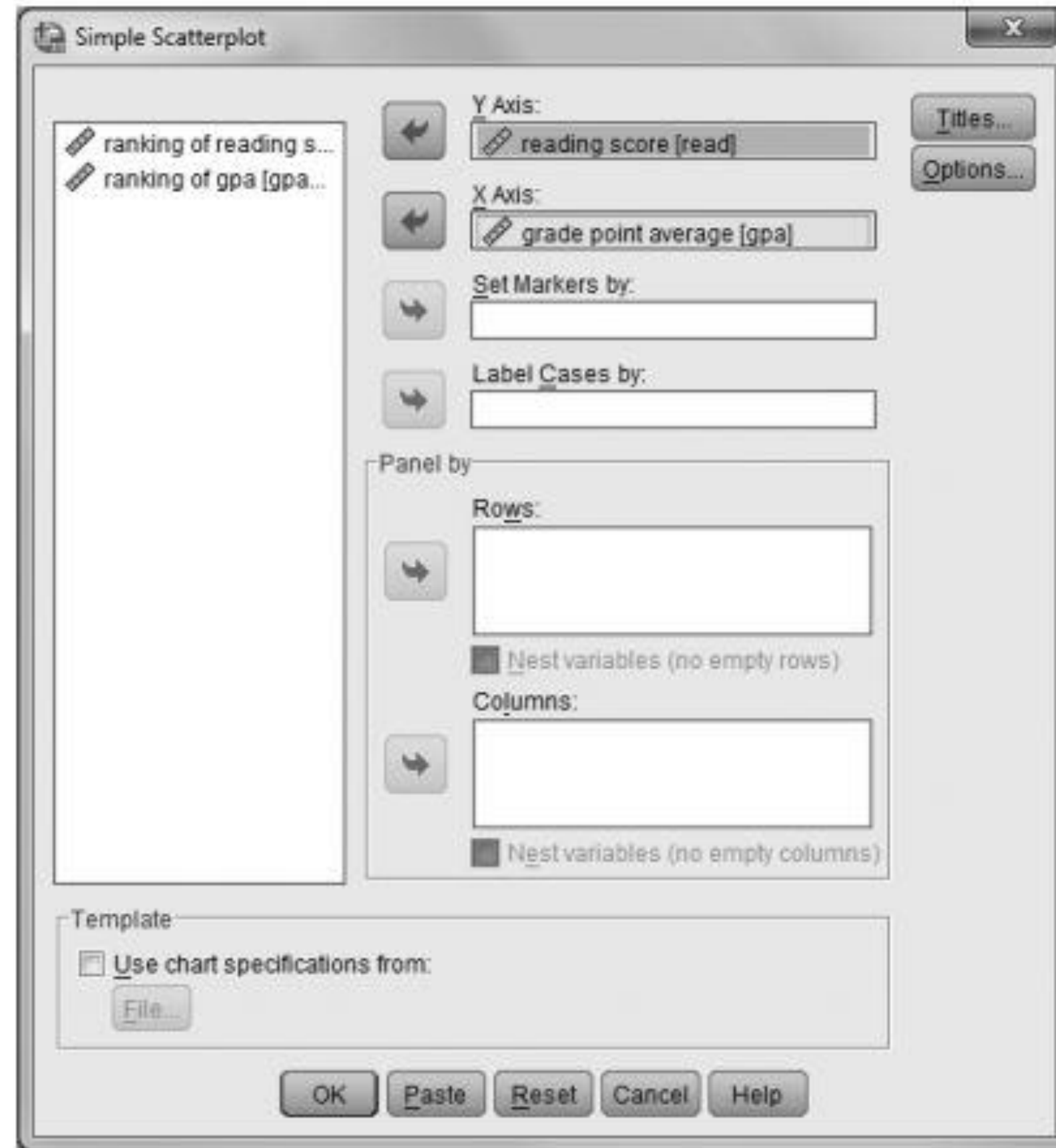
☒ Confidence intervals
Level (%):

☐ Standard error
Multiplier:

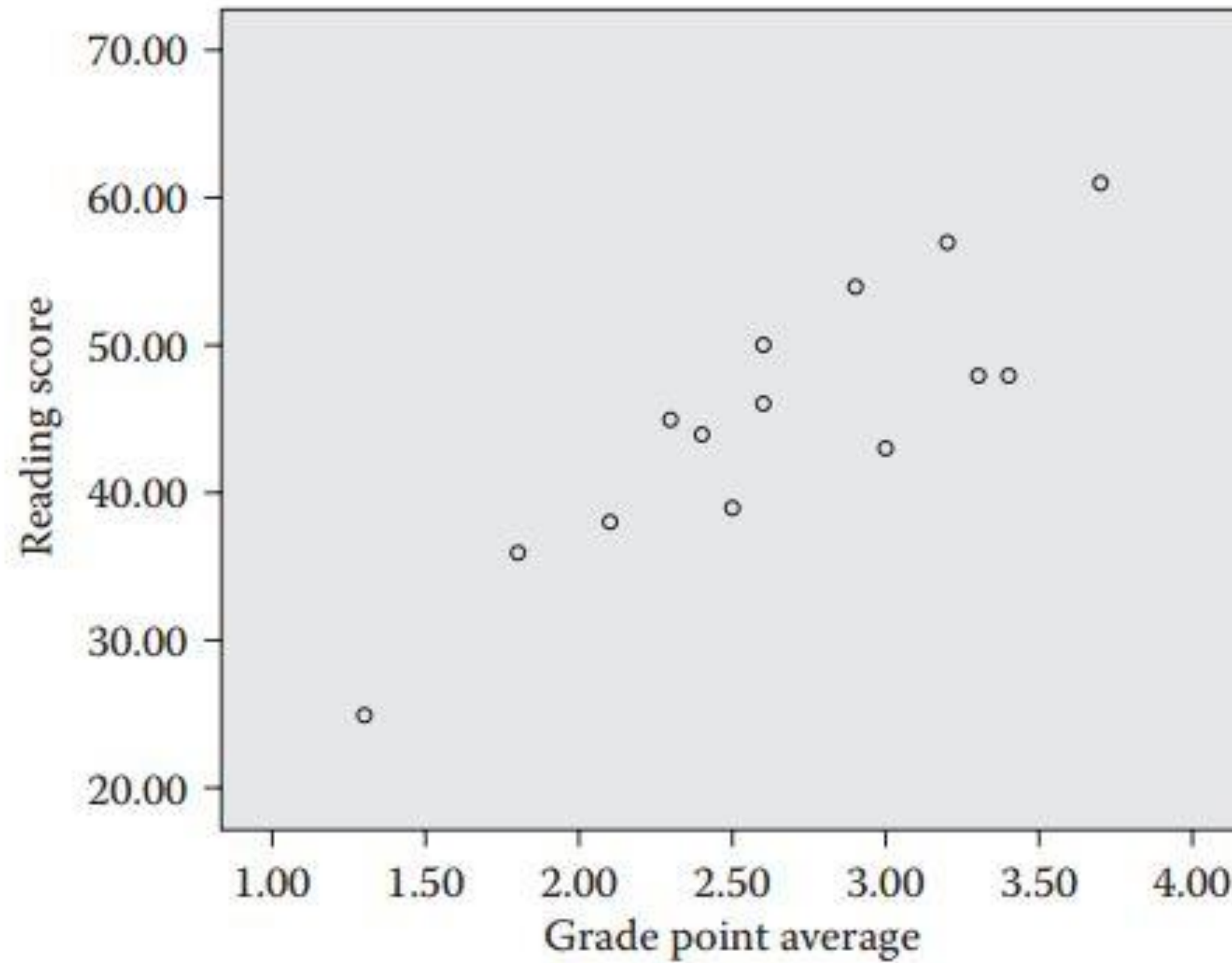
☐ Standard deviation
Multiplier:

Continue Cancel Help

Click **Continue** to return to the **Simple Scatterplot** window.



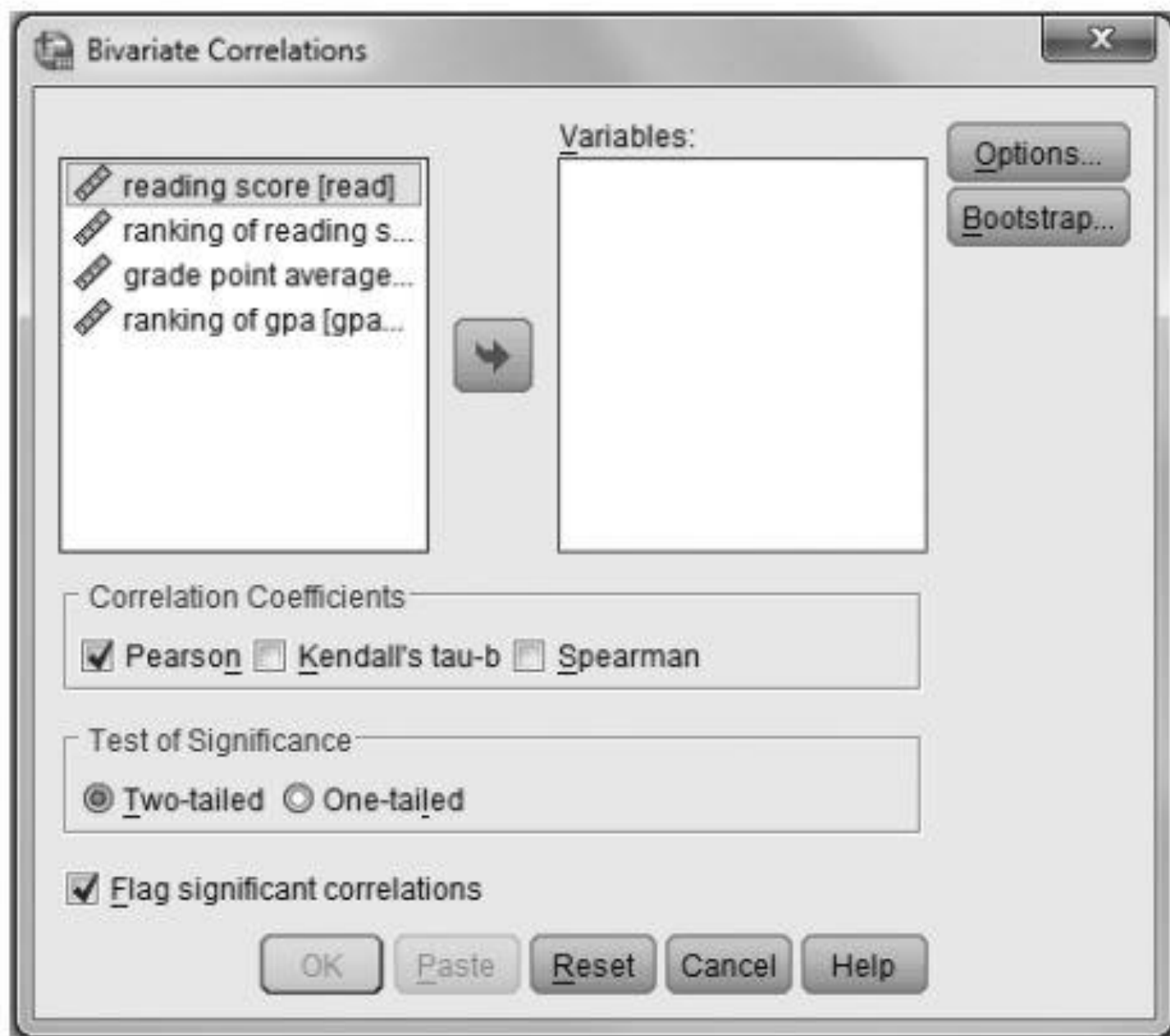
Scatterplot

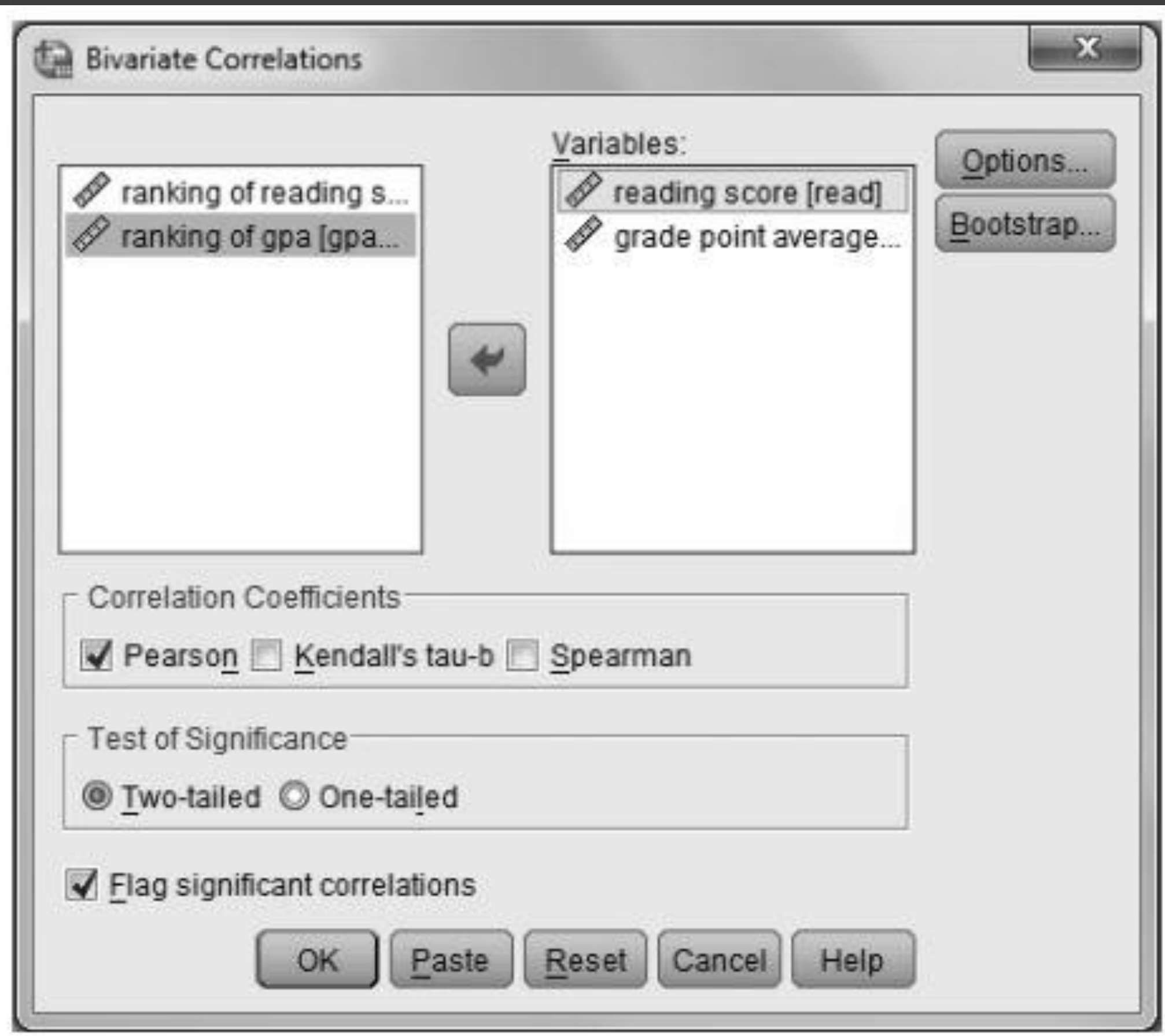


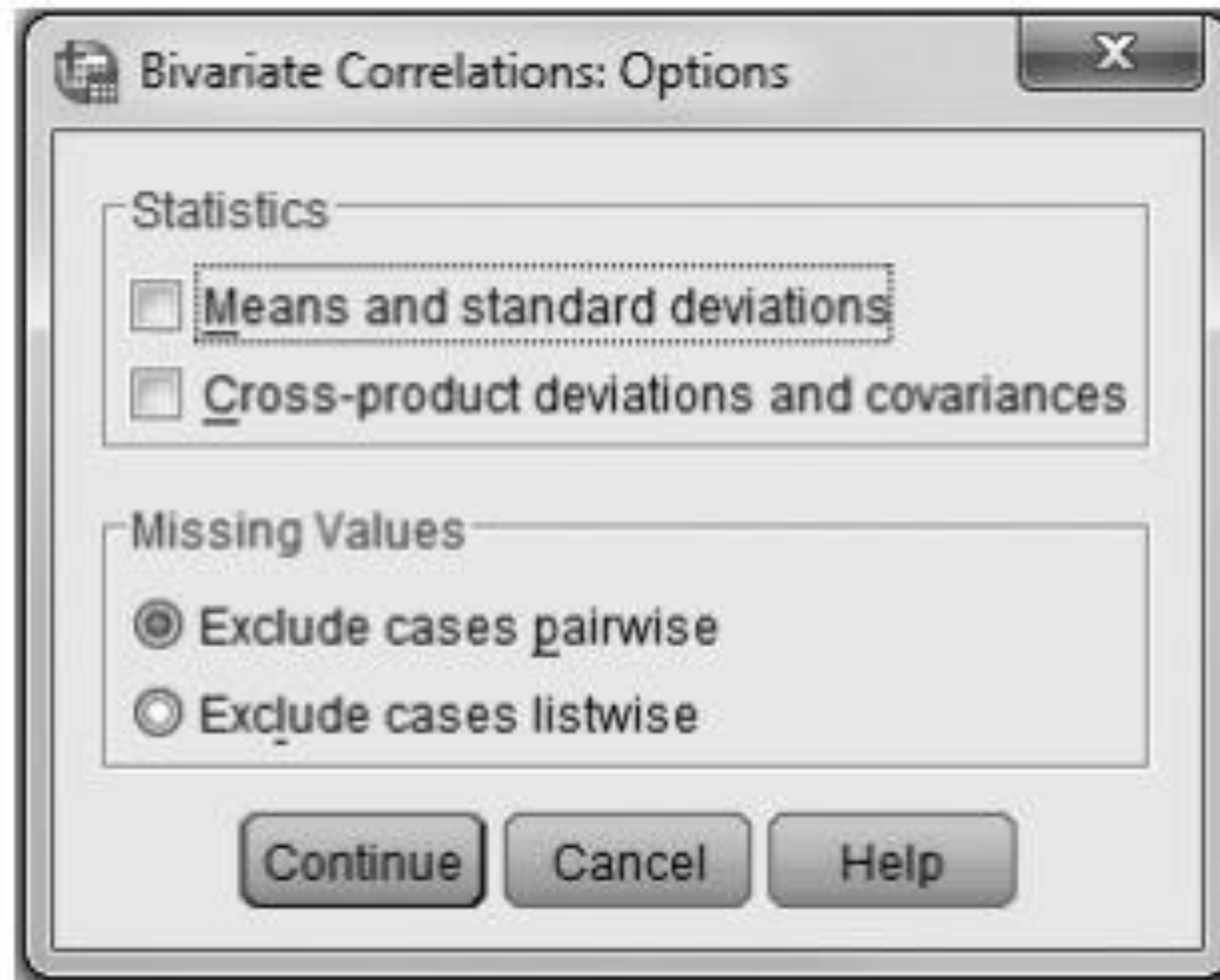
As can be seen from Figure 10.1, there is a linear relationship between the variables of reading score and grade point average, such that as reading score increases, so does grade point average.

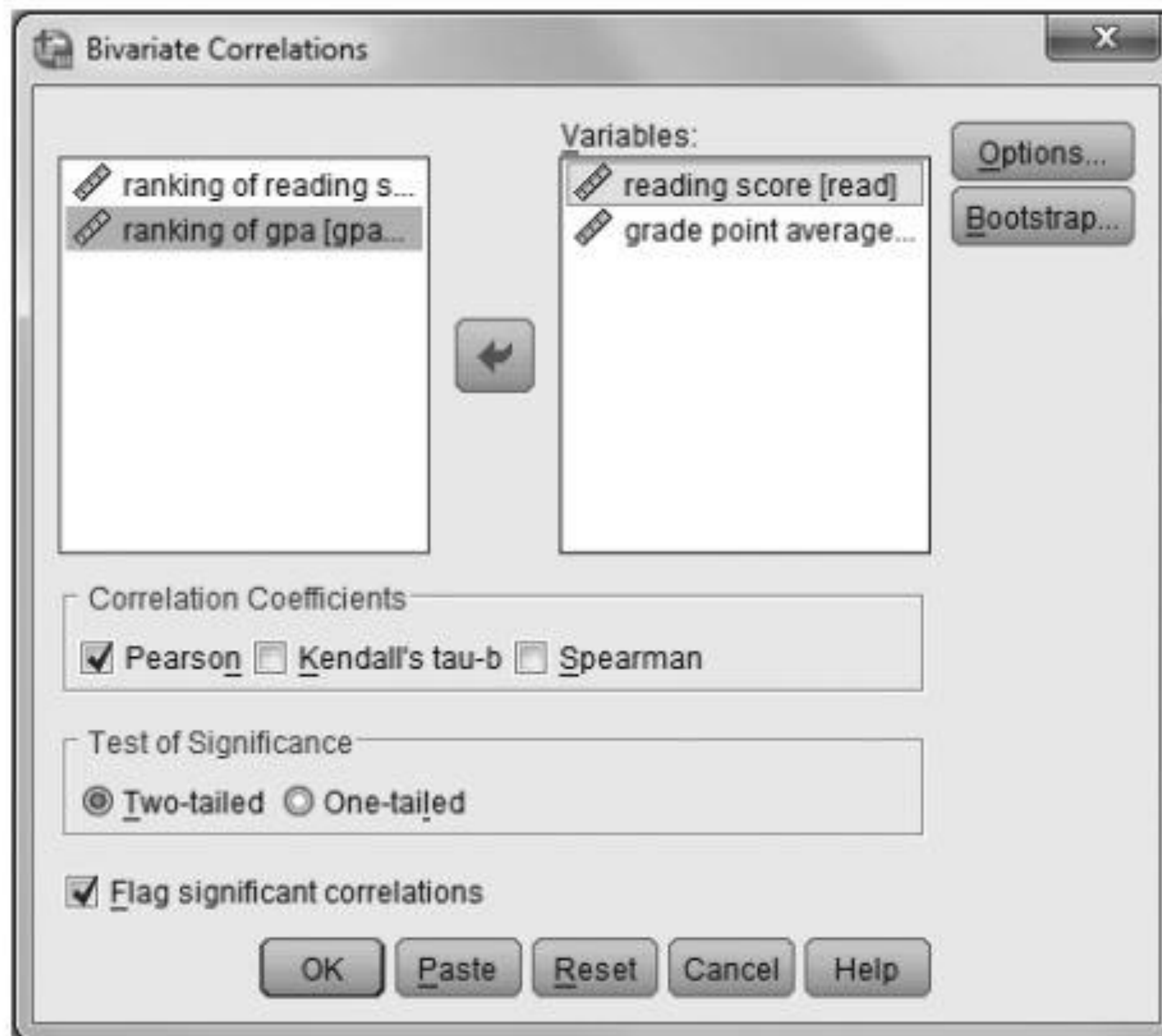
The figure also shows that the homoscedasticity assumption is met, because the variability of the READ score remains relatively constant from one GPA score to the next. Heteroscedasticity is usually shown by a cluster of points that is wider as the values for the Y variable (READ) get larger.

From the menu bar, click **Analyze**, then **Correlate**, and then **Bivariate...** The following **Bivariate Correlations** window will open.









Pearson Product Moment Correlation

Correlations

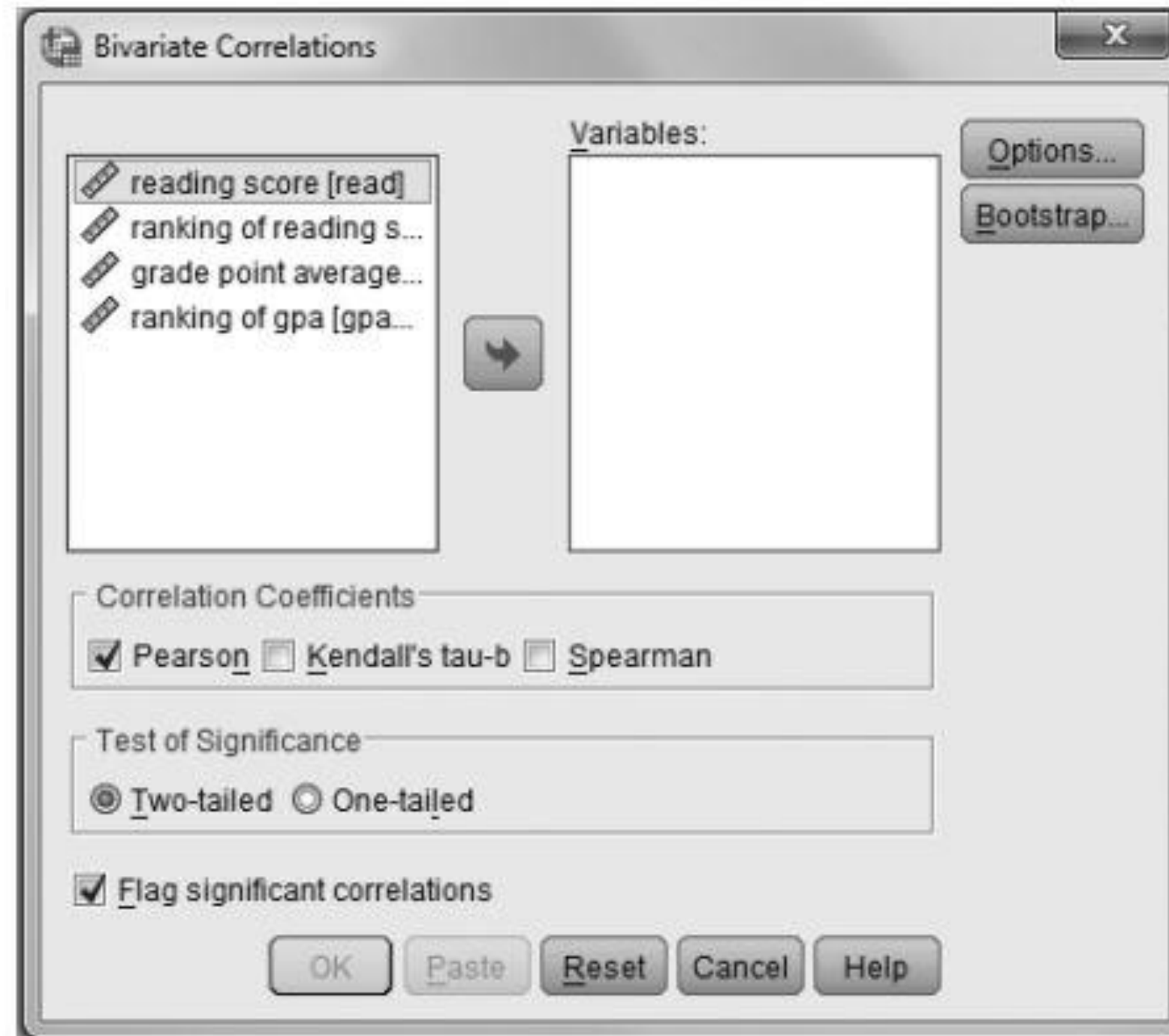
		Reading Score	Grade Point Average
reading score	Pearson Correlation	1	.867**
	Sig. (2-tailed)		.000
	N	15	15
grade point average	Pearson Correlation	.867**	1
	Sig. (2-tailed)	.000	
	N	15	15


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

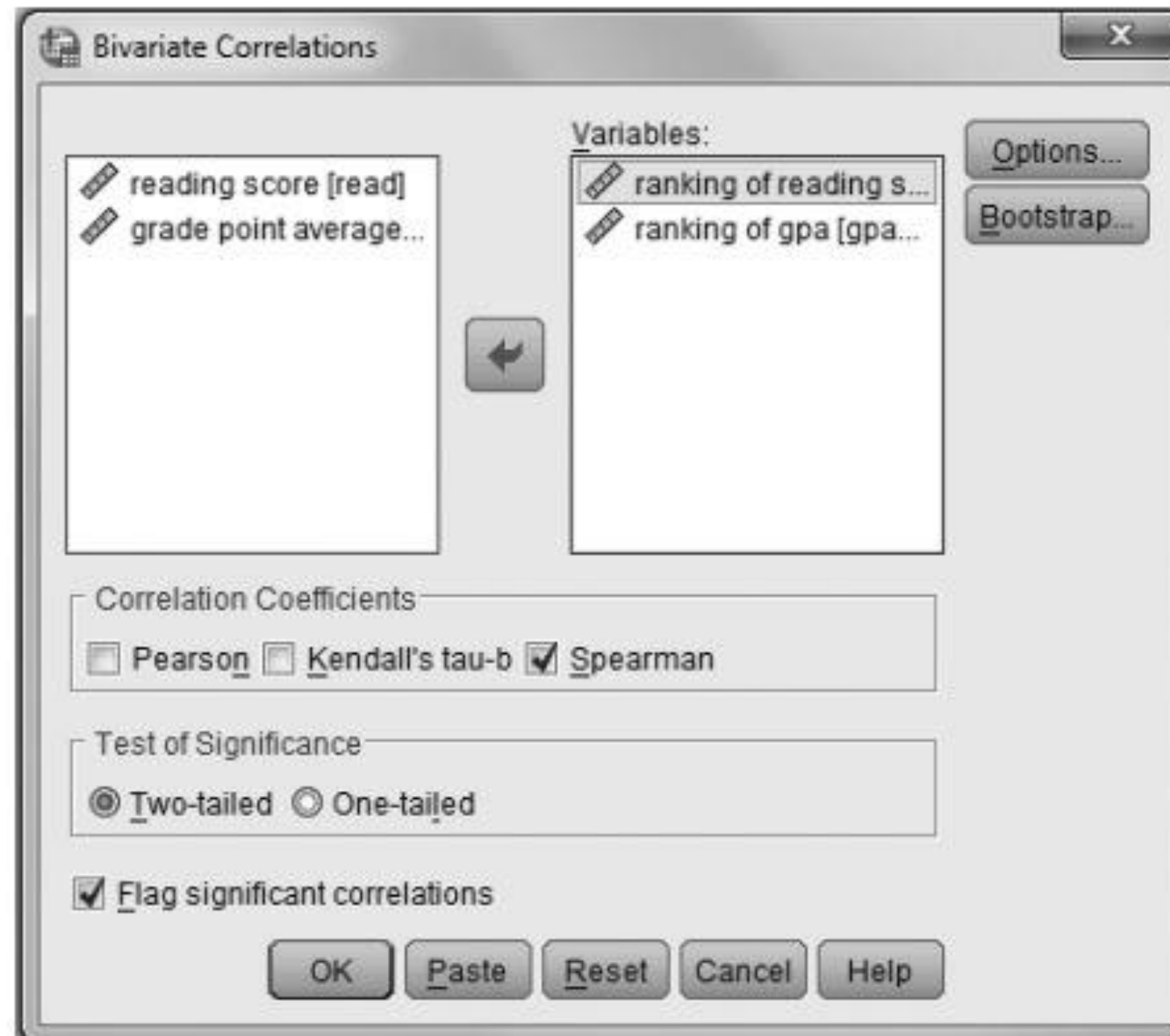
The correlation between reading scores and grade point average is positive and statistically significant ($r = 0.867, p < .001$). This means that as the students' reading scores increase, so do their grade point averages. Please note that this interpretation in no way implies *causality*—that increases in reading scores caused increases in GPA scores. The significant relationship merely indicates that the two variables *covary*.

For this example, the same data set will be used. However, the rank order of the two variables (READ_RANK, GPA_RANK) will be used instead of their actual values as recorded. Thus, the computation for this coefficient is not sensitive to asymmetrical distributions or to the presence of outliers.

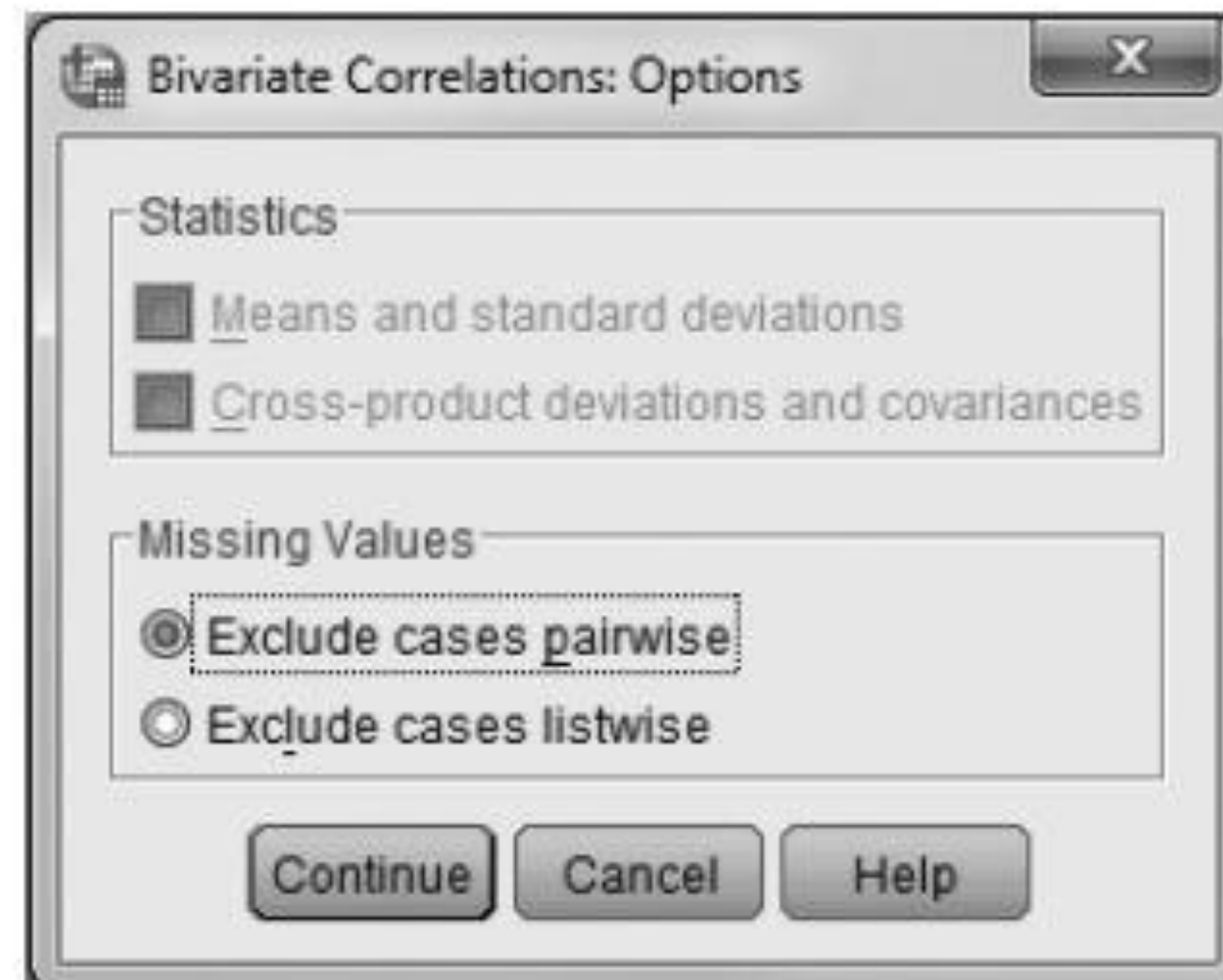
1. From the menu bar, click **Analyze**, then **Correlate**, and then **Bivariate...**
The following **Bivariate Correlations** window will open.



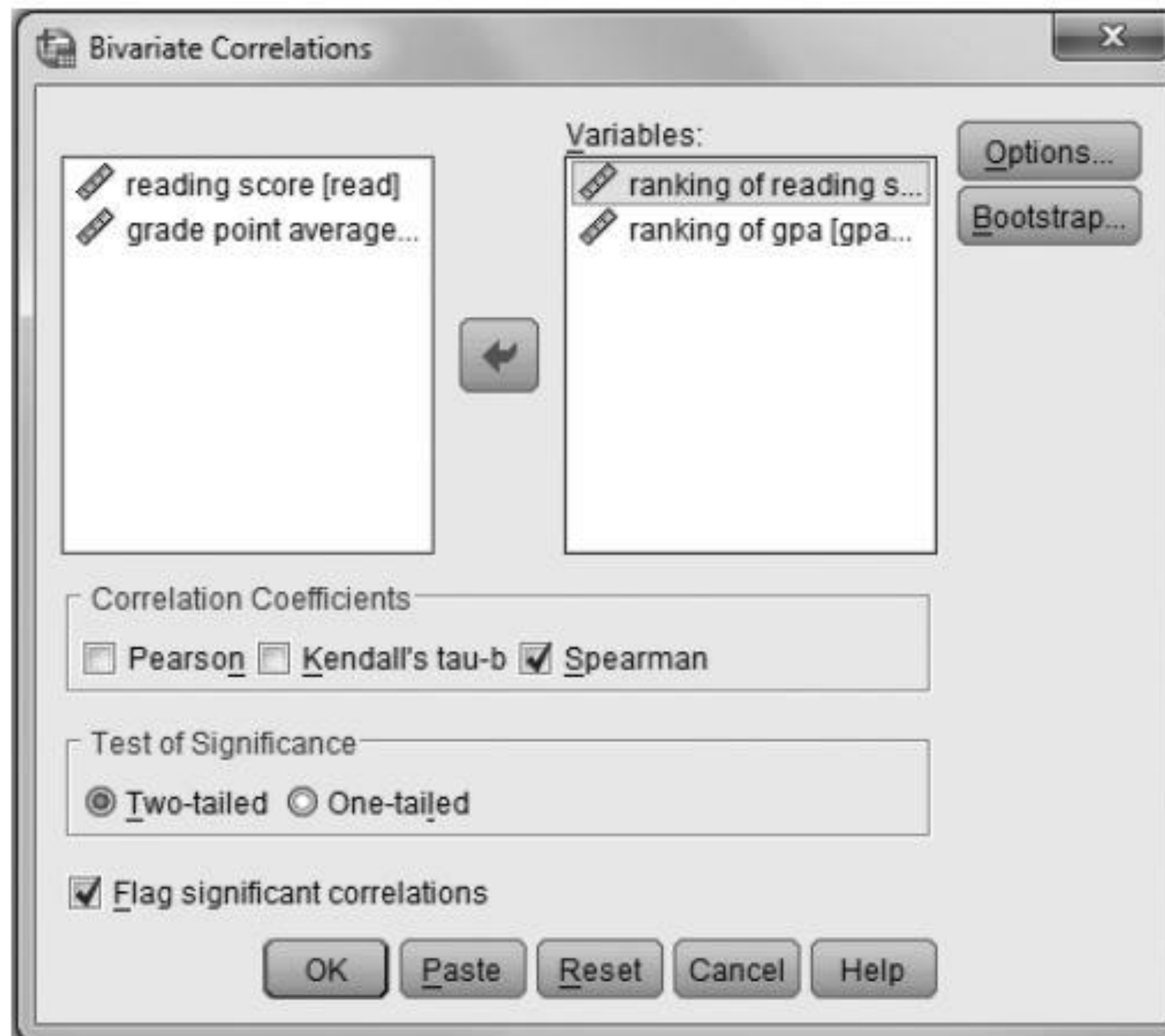
2. Transfer the **READ_RANK** and **GPA_RANK** variables to the **Variables:** field by clicking (highlight) them and then clicking . By default, SPSS will employ the **Pearson correlation analysis** (this field is checked). Uncheck the **Pearson** field and check the **Spearman** field.



3. Click **Options...** to open the **Bivariate Correlation: Options** window. Ensure that the **Exclude cases pairwise** field is checked.



4. Click **Continue** to return to the **Bivariate Correlations** window.



Spearman Rank Order Correlation

Correlations			Ranking of Reading Scores	Ranking of gpa
Spearman's rho	ranking of reading scores	Correlation	1.000	.826**
		Coefficient	.	.000
		Sig. (2-tailed) N	15	15
	ranking of gpa	Correlation	.826**	1.000
		Coefficient	.000	.
		Sig. (2-tailed) N	15	15

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

The obtained Spearman rank-order coefficient ($\rho = 0.826$, $p < .001$) is highly similar in magnitude and direction to that in the Pearson correlation table . Thus, similar to the Pearson coefficient, the Spearman coefficient indicates that as the students' ranked reading scores increase, so do their ranked grade point average scores.

Thank You!

Any Questions?