

Calculate Spearman Rank Correlation in Google Sheets

Authored by
Mohammed looti

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In the realm of [statistics](#) and data analysis, the concept of [correlation](#) is foundational. It provides a quantitative measure of the strength and directional relationship between two distinct variables within a dataset. Analysts rely on this measure to quickly understand if, and how, changes in one variable correspond to changes in the other. The result of this calculation is the [correlation coefficient](#), a standardized index that always ranges precisely between -1 and +1.

Interpreting the correlation coefficient offers immediate and powerful insight into the relationship's nature. Understanding these standardized values is the first step toward advanced inferential analysis.

-1 (Perfect Negative Correlation): Indicates a perfect inverse relationship. As one variable consistently increases, the other decreases.

0 (No Linear Relationship): Suggests that there is no discernible linear association between the movement of the two variables.

+1 (Perfect Positive Correlation): Signifies a perfect direct relationship, where both variables increase or decrease together consistently.

Understanding Spearman Rank Correlation

While the widely known Pearson correlation coefficient is used to assess linear relationships in raw, often normally distributed data, the **Spearman Rank Correlation** (often symbolized as r_s or ρ_s) serves a different, yet equally critical, purpose. Spearman's method is a crucial [non-parametric measure](#) specifically designed to evaluate the monotonic relationship between two sets of data when the original scores are converted into ranks.

The key advantage of utilizing the Spearman method lies in its resilience against extreme values, or **outliers**. Because it operates solely on the ordinal ranking of the data points--assigning ranks from 1 (highest) to N (lowest)--it is highly applicable to ordinal data where the magnitude of the difference between scores is less important than the relative order. This makes it ideal for fields like social science and psychology where data often comes in the form of ratings or rankings.

This comprehensive guide will demonstrate precisely how to compute the **Spearman Rank Correlation** coefficient accurately and efficiently using the powerful spreadsheet application, [Google Sheets](#). We will break down the process into clear, manageable steps, starting with data preparation and concluding with the interpretation of the final result.

Practical Example: Calculating Rank Correlation in Google Sheets

To illustrate the methodology, we will calculate the Spearman rank correlation coefficient (r_s) between two variables: Math exam scores and Science exam scores, collected from a sample group of 10 students (N=10). Precision is paramount; follow these detailed instructions carefully to

ensure your analysis is reproducible in your own spreadsheet environment.

Step 1: Input the Raw Data. The procedure begins by accurately entering the raw, numerical scores for both variables into two distinct, adjacent columns within your Google Sheet. It is strongly recommended to use descriptive labels for your columns, such as "Math Score" (Column B) and "Science Score" (Column C), to maintain clarity throughout the analysis.

The initial data setup should strictly adhere to the arrangement visualized in the image below. This structure establishes the foundation for all subsequent ranking calculations.

<i>fx</i>					
	A	B	C	D	
1	Student	Math	Science		
2	A	70	90		
3	B	78	94		
4	C	90	79		
5	D	87	86		
6	E	84	84		
7	F	86	83		
8	G	91	88		
9	H	74	92		
10	I	83	76		
11	J	85	75		
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Step 2: Deriving the Rank Values

Since **Spearman Rank Correlation** is based entirely on the relative position of values, the next critical phase involves converting the raw data into its corresponding ranks. This is essential for converting interval or ratio data into **ordinal data** that the Spearman formula requires. For this conversion, we must use the dedicated Google Sheets function **RANK.AVG()**. This function is vital because it correctly handles instances of tied scores by assigning the average rank to all tied values, ensuring mathematical validity.

To calculate the ranks for the first entry (Student 1, Row 2), you will need to input the formula twice--once for the Math Rank (Cell D2) and once for the Science Rank (Cell E2). Pay close attention to the use of **absolute references** (e.g., **\$B\$2:\$B\$11**) within the range argument; these ensure that when the formula is copied down, the data range remains fixed, preventing errors.

The required formulas for the first row are as follows:

Cell D2 (Math Rank Calculation):

=RANK.AVG(B2, \$B\$2:\$B\$11, 0)

Cell E2 (Science Rank Calculation):

=RANK.AVG(C2, \$C\$2:\$C\$11, 0)

Upon successfully entering these formulas into the first row, your sheet will temporarily reflect the ranks for the initial data point, as demonstrated below:

<i>fx</i>						
	A	B	C	D	E	
1	Student	Math	Science	Math Rank	Science Rank	
2	A	70	90	10	3	
3	B	78	94			
4	C	90	79			
5	D	87	86			
6	E	84	84			
7	F	86	83			
8	G	91	88			
9	H	74	92			
10	I	83	76			
11	J	85	75			
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Next, efficiently apply this ranking calculation to the rest of the dataset. Select the entire range where the ranks need to be generated (D2:E11).

<i>fx</i>	=RANK.AVG(B2, \$B\$2:\$B\$11, 0)				
	A	B	C	D	E
1	Student	Math	Science	Math Rank	Science Rank
2	A	70	90	10	3
3	B	78	94		
4	C	90	79		
5	D	87	86		
6	E	84	84		
7	F	86	83		
8	G	91	88		
9	H	74	92		
10	I	83	76		
11	J	85	75		
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To finalize the ranking, use the fill handle (dragging down from the corner) or the shortcut **Ctrl+D** (Command+D on Mac) to automatically compute the ranks for the remaining students. Once complete, your dataset is prepared for the final correlation calculation.

<i>fx</i>						
	A	B	C	D	E	
1	Student	Math	Science	Math Rank	Science Rank	
2	A	70	90	10	3	
3	B	78	94	8	1	
4	C	90	79	2	8	
5	D	87	86	3	5	
6	E	84	84	6	6	
7	F	86	83	4	7	
8	G	91	88	1	4	
9	H	74	92	9	2	
10	I	83	76	7	9	
11	J	85	75	5	10	
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Step 3: Calculating the Spearman Rank Correlation Coefficient

Once the raw scores have been accurately converted into rank values, determining the Spearman coefficient (r_s) becomes remarkably simple, despite the lack of a dedicated Spearman function in Google Sheets. Crucially, the standard **CORREL()** function--which calculates the Pearson correlation--when applied to ranked data, mathematically produces the exact **Spearman Rank Correlation** coefficient.

To execute this final step, enter the following formula into any unoccupied cell: **=CORREL(D2:D11, E2:E11)**. This powerful function compares the entire array of Math ranks (D2:D11) against the array of Science ranks (E2:E11), treating them as simple numerical values to determine their linear association (which, on ranks, is the monotonic association).

fx =CORREL(D2:D11, E2:E11)						
	A	B	C	D	E	
1	Student	Math	Science	Math Rank	Science Rank	
2	A	70	90	10	3	
3	B	78	94	8	1	
4	C	90	79	2	8	
5	D	87	86	3	5	
6	E	84	84	6	6	
7	F	86	83	4	7	
8	G	91	88	1	4	
9	H	74	92	9	2	
10	I	83	76	7	9	
11	J	85	75	5	10	
12						
13	Spearman Correlation: -0.41818182					
14	? =CORREL(D2:D11, E2:E11)					
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After the calculation is performed, the resulting **Spearman Rank Correlation** coefficient for our sample dataset is **-0.41818**. This negative value indicates a moderate, inverse monotonic relationship: students who ranked higher in Math tended to rank slightly lower in Science, and vice versa.

Step 4: Interpreting the Result and Assessing Significance

The calculated correlation coefficient of **-0.41818** confirms a negative trend in the sample data, but this finding is incomplete without assessing its **statistical significance**. Statistical significance determines whether the observed relationship is likely a genuine phenomenon within the wider population or merely a random fluctuation due to sample variation.

Determining significance for the Spearman coefficient typically involves consulting a specific table of critical values. These external tables are necessary because Google Sheets does not provide a built-in P-value calculation for the correlation function. These tables require two inputs: the sample size (n) and the chosen alpha level (significance level, usually $\alpha = 0.05$ or $\alpha = 0.01$).

The fundamental rule for hypothesis testing the Spearman correlation is straightforward: if the absolute value of the calculated coefficient (denoted as $|r_s|$) is greater than the critical value

retrieved from the table, we conclude that the correlation is **statistically significant** at the selected alpha level, meaning we reject the null hypothesis of no relationship.

For reference, a standard excerpt from a Spearman critical values table is provided below:

n	α				
	0.10	0.05	0.025	0.01	0.005
5	0.800	0.900	1.000	1.000	-
6	0.657	0.829	0.886	0.943	1.000
7	0.571	0.714	0.786	0.893	0.929
8	0.524	0.643	0.738	0.833	0.881
9	0.483	0.600	0.700	0.783	0.833
10	0.455	0.564	0.648	0.745	0.794
11	0.427	0.536	0.618	0.709	0.755
12	0.406	0.503	0.587	0.678	0.727
13	0.385	0.484	0.560	0.648	0.703
14	0.367	0.464	0.538	0.626	0.679
15	0.354	0.446	0.521	0.604	0.654
16	0.341	0.429	0.503	0.582	0.635
17	0.328	0.414	0.488	0.566	0.618
18	0.317	0.401	0.472	0.550	0.600
19	0.309	0.391	0.460	0.535	0.584
20	0.299	0.380	0.447	0.522	0.570
21	0.292	0.370	0.436	0.509	0.556
22	0.284	0.361	0.425	0.497	0.544
23	0.278	0.353	0.416	0.486	0.532
24	0.271	0.344	0.407	0.476	0.521
25	0.265	0.337	0.398	0.466	0.511
26	0.259	0.331	0.390	0.457	0.501
27	0.255	0.324	0.383	0.449	0.492
28	0.250	0.318	0.375	0.441	0.483
29	0.245	0.321	0.368	0.433	0.475
30	0.240	0.306	0.362	0.425	0.467

In our running example, the sample size (n) is 10 students. If we adopt the common scientific standard of a 0.05 significance level ($\alpha = 0.05$), we locate the corresponding critical value for $n=10$, which is **0.564**.

We must compare our calculated absolute coefficient ($|-0.41818| = 0.41818$) against the critical threshold (0.564). Since 0.41818 is clearly not larger than 0.564, we must retain the null hypothesis. Therefore, we conclude that the observed negative correlation between Math and Science rankings is not **statistically significant** within this limited sample.

Further Reading: [How to Calculate Spearman Rank Correlation in Excel](#)