



Day 7: Spearman's Rank Correlation Coefficient ★

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Spearman's Rank Correlation Coefficient

We have two random variables, X and Y :

- $X = \{x_1, x_2, x_3, \dots, x_n\}$
- $Y = \{y_1, y_2, y_3, \dots, y_n\}$

If Rank_X and Rank_Y denote the respective ranks of each data point, then the Spearman's rank correlation coefficient, r_s , is the Pearson correlation coefficient of Rank_X and Rank_Y .

Example

- $X = \{0.2, 1.3, 0.2, 1.1, 1.4, 1.5\}$
- $Y = \{1.9, 2.2, 3.1, 1.2, 2.2, 2.2\}$

Rank_X :

X :	0.2	1.3	0.2	1.1	1.4	1.5
Rank :	1	3	1	2	4	5

So, $\text{Rank}_X = \{1, 3, 1, 2, 4, 5\}$

Similarly, $\text{Rank}_Y = \{2, 3, 4, 1, 3, 3\}$

r_s equals the Pearson correlation coefficient of Rank_X and Rank_Y , meaning that $r_s = 0.158114$.

Special Case: X and Y Don't Contain Duplicates

$$r_s = 1 - \frac{6 \cdot \sum d_i^2}{n \cdot (n^2 - 1)}$$

Here, d_i is the difference between the respective values of Rank_X and Rank_Y .

Proof

Let's define P be the rank of X and Q be the rank of Y . Both P and Q are permutations of set $\{1, 2, 3, \dots, n\}$, because data sets X and Y contain no duplicates in this special case.

Mean of P and Q :

$$\begin{aligned} \sum_i p_i &= \sum_i q_i = \frac{n \cdot (n+1)}{2} \\ \Rightarrow \mu_P &= \mu_Q = \mu = \frac{(n+1)}{2} \end{aligned}$$

Standard Deviation of P and Q :

$$n \cdot (n^2 - 1)$$



$$\sum_i (p_i - \mu_p)^2 = \sum_i (p_i - \mu)^2 = \sum_i p_i^2 - 2\mu \sum_i p_i + \mu^2 \sum_i 1 = \frac{n^2 - 1}{12}$$

So,

$$\sigma_P = \sigma_Q = \sigma = \sqrt{\frac{\sum_i (p_i - \mu_p)^2}{n}} = \sqrt{\frac{n^2 - 1}{12}}$$

Calculating $\sum_i d_i^2$:

$$\sum_i d_i^2 = \sum_i (p_i - q_i)^2 = \sum_i p_i^2 - 2 \sum_i (p_i q_i) + \sum_i q_i^2$$

We know that:

$$\sum_i p_i^2 = \sum_i q_i^2 = \frac{n \cdot (n+1) \cdot (2n+1)}{6}$$

So,

$$\sum_i (p_i q_i) = \frac{n \cdot (n+1)(n^2+1)}{6} - \frac{1}{2} \sum_i d_i^2$$

Covariance of P and Q :

$$\begin{aligned} \text{cov}(P, Q) &= \frac{\sum_i (p_i - \mu_p)(q_i - \mu_q)}{n} = \frac{\sum_i (p_i - \mu)(q_i - \mu)}{n} \\ \Rightarrow \text{cov}(P, Q) &= \frac{\sum_i (p_i q_i) - \mu (\sum_i p_i + \sum_i q_i) + \mu^2 \sum_i 1}{n} \\ \Rightarrow \text{cov}(P, Q) &= \frac{\frac{n \cdot (n+1) \cdot (n^2+1)}{6} - \frac{1}{2} \sum_i d_i^2 - \mu (\sum_i p_i + \sum_i q_i) + \mu^2 \sum_i 1}{n} \\ \Rightarrow \text{cov}(P, Q) &= \frac{\frac{n \cdot (n^2-1)}{12} - \frac{1}{2} \sum_i d_i^2}{n} \end{aligned}$$

Spearman's Rank Correlation Coefficient:

We know that the Spearman's rank correlation coefficient (r_s) of X and Y is equal to the Pearson correlation coefficient of P and Q . So,

$$\begin{aligned} r_s &= \frac{\text{cov}(P, Q)}{\sigma_P \sigma_Q} = \frac{\text{cov}(P, Q)}{\sigma^2} \\ \Rightarrow r_s &= \frac{\frac{\frac{n \cdot (n^2-1)}{12} - \frac{1}{2} \sum_i d_i^2}{n}}{\frac{(n^2-1)}{12}} \\ \Rightarrow r_s &= \frac{\frac{n \cdot (n^2-1)}{12} - \frac{1}{2} \sum_i d_i^2}{\frac{n \cdot (n^2-1)}{12}} \\ \Rightarrow r_s &= 1 - \frac{6 \sum_i d_i^2}{n \cdot (n^2 - 1)} \end{aligned}$$



