ST 516: Foundations of Data Analytics Wilcoxon Signed-Rank Test

Wilcoxon Signed-Rank Test: Introduction

The Wilcoxon signed-rank test is used to assess whether a particular value is the 'center' of a distribution of values (one-sample setting) or differences (paired two-sample setting). We will try to clarify what we mean by 'center' later, but it turns out to be rather challenging to precisely define what the signed-rank test actually tests.

Be careful not to confuse the *signed-rank* test (today) with the *rank-sum* test (different lecture). The signed-rank test is used on single-sample or paired-sample data; the rank-sum test is used on two independent samples.

- Setting: One sample of independent observations
 - Sample of size n from population of interest: $X_1, X_2, ..., X_n$

OR a sample of independent differences between paired observations

• n differences between paired observations in population 1 and population 2: $D_1, D_2, ..., D_n$

Wilcoxon Signed-Rank Test: Procedure

To test that C_0 is the population 'center', we do the following:

- 1. Calculate the distance of each observation/difference from C_0 .
- 2. Rank the observations by their distance (absolute value) from C_0 . The observation closest to C_0 gets rank 1, the next closest gets rank 2, and so on.
- 3. Let the test statistic S be the sum of the ranks that correspond to observations *larger* than C_0 .

Suppose we have observed the following data, and we wish to test the hypothesis that the 'center' of the differences is $C_0 = 3$ vs. the alternative that the 'center' is not $C_0 = 3$.

Sample A	8.6	12.3	14.0	18.1	6.3	7.5	10.7	6.2	7.0
Sample B	7.4	5.4	4.9	4.0	3.8	4.4	6.9	11.8	11.6
Differences	1.2	6.9	9.1	14.1	2.5	3.1	3.8	-5.6	-4.6

Suppose we have observed the following data, and we wish to test the hypothesis that the 'center' of the differences is $C_0 = 3$ vs. the alternative that the 'center' is not $C_0 = 3$.

1. Compute the distance of each *difference* from C_0 .

Sample A	8.6	12.3	14.0	18.1	6.3	7.5	10.7	6.2	7.0
Sample B	7.4	5.4	4.9	4.0	3.8	4.4	6.9	11.8	11.6
Differences	1.2	6.9	9.1	14.1	2.5	3.1	3.8	-5.6	-4.6
Distance from $C_0 = 3$	1.8	3.9	6.1	11.1	0.5	0.1	0.8	8.6	7.6

Suppose we have observed the following data, and we wish to test the hypothesis that the 'center' of the differences is $C_0 = 3$ vs. the alternative that the 'center' is not $C_0 = 3$.

2. Rank the distances from smallest to largest.

Sample A	8.6	12.3	14.0	18.1	6.3	7.5	10.7	6.2	7.0
Sample B	7.4	5.4	4.9	4.0	3.8	4.4	6.9	11.8	11.6
Differences	1.2	6.9	9.1	14.1	2.5	3.1	3.8	-5.6	-4.6
Distance from $C_0 = 3$	1.8	3.9	6.1	11.1	0.5	0.1	0.8	8.6	7.6
Distance Ranks	4	5	6	9	2	1	3	8	7

Suppose we have observed the following data, and we wish to test the hypothesis that the 'center' of the differences is $C_0 = 3$ vs. the alternative that the 'center' is not $C_0 = 3$.

3. Add up ranks corresponding to differences greater than C_0 .

Sample A	8.6	12.3	14.0	18.1	6.3	7.5	10.7	6.2	7.0
Sample B	7.4	5.4	4.9	4.0	3.8	4.4	6.9	11.8	11.6
Differences	1.2	6.9	9.1	14.1	2.5	3.1	3.8	-5.6	-4.6
Distance from $C_0 = 3$	1.8	3.9	6.1	11.1	0.5	0.1	0.8	8.6	7.6
Distance Ranks	4	5	6	9	2	1	3	8	7

$$S = 5 + 6 + 9 + 1 + 3 = 24$$

Reference Null Distribution

How do we decide whether a resulting signed-rank test statistic is 'significant'? How do we compute p-values?

We use a *permutation-type* approach to decide how unusual an observed test statistic is if each rank is equally likely to correspond to an observation larger than C_0 or smaller than C_0 .

- If the population distribution were symmetric about C_0 , each rank 1, ..., n has probability $\frac{1}{2}$ of being assigned to an observation above C_0 .
- We can consider all possible ways of assigning the ranks 1,..., n above and below C₀ to work out the exact reference distribution for the test statistic S.

Reference Null Distribution: Normal Approximation

For large sample sizes, it is time-consuming to work out the exact reference distribution. Instead, we can use a Normal approximation to the reference null distribution of *S*:

$$S \sim \text{Normal}\left(\frac{n(n+1)}{4}, \frac{n(n+1)(2n+1)}{24}\right)$$

when the population distribution is symmetric around C_0 .

We can therefore construct a z-statistic and compare it to a standard normal reference distribution:

$$Z = \frac{S - \frac{n(n+1)}{4}}{\sqrt{\frac{n(n+1)(2n+1)}{24}}} \stackrel{\cdot}{\sim} \text{Normal}(0,1)$$

Reference Null Distribution: Example

Sample A	8.6	12.3	14.0	18.1	6.3	7.5	10.7	6.2	7.0
Sample B	7.4	5.4	4.9	4.0	3.8	4.4	6.9	11.8	11.6
Differences	1.2	6.9	9.1	14.1	2.5	3.1	3.8	-5.6	-4.6
Distance from $C_0 = 3$	1.8	3.9	6.1	11.1	0.5	0.1	0.8	8.6	7.6
Distance Ranks	4	5	6	9	2	1	3	8	7

$$S = 5 + 6 + 9 + 1 + 3 = 24$$

Using our example from before, we compute the z-statistic:

$$Z = \frac{S - \frac{n(n+1)}{4}}{\sqrt{\frac{n(n+1)(2n+1)}{24}}} = \frac{24 - \frac{9(10)}{4}}{\sqrt{\frac{9(10)(19)}{24}}} = 0.177$$

and find the p-value:

$$2*(1 - pnorm(abs(0.1777)))$$

0.859

Signed-Rank Test Assumptions and Summary

Assumptions required to perform the signed-rank test:

 Individual observations in each sample are independent of each other (as usual)... and that's about it!

What have we learned after performing the Wilcoxon Signed-Rank test?

- It is very difficult to say: we compute the null distribution assuming the population is symmetric around its center—but the signed-rank test does not answer the question 'Is the population symmetric around C_0 ?'.
- The signed-rank test also does not answer questions about means or medians (...unless we assume that the population distribution is symmetric—a pretty big and unverifiable assumptions!).

It is important that you know how the signed-rank test works and what it does so that you can understand what people are doing if they use it—but keep in mind the above cautions: it is difficult to precisely say what you learn from a signed-rank test.