

MVE055 2017 Lecture 15

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Parametric vs. Nonparametric

- The data have to be numeric:
- We need/can make assumptions on the underlying distribution of the data (e.g. normality).
- They often have higher statistical power than corresponding nonparametric test. That is, When a significant effect exists, parametric test are more likely to detect this effects.
- If the sample size is large "enough", they can still perform well even in the case where assumptions are not met (e.g. T test with equal variances, nonnormal data)

Nonparametric test

- The data can be numeric, ordinal, or nominal.
- We do not specify the underlying distribution of the data ("distribution-free").
- Usable even with small samples;
- T-test (parametric) is testing equality of means, where sometimes the mean might not be the "best" measure of location.

Wilcoxon Rank-Sum test

- Let X_1, \dots, X_m and Y_1, \dots, Y_n be two independent random samples of, respectively, X and Y .
- We want to test if X and Y have the same distribution, i.e.

$$H_0 : X \text{ and } Y \text{ have the same distribution}$$

$$H_1 : X \text{ and } Y \text{ do not have the same distribution}$$

- Under some assumptions, Wilcoxon test can be interpreted as a test on medians, i.e. denote by M_X, M_Y the medians

$$H_0 : M_X = M_Y$$

$$H_1 : M_X \neq M_Y$$

- Similarly, one can consider one-sided tests.

An example

- Suppose we have the following data regarding the height of individuals (in centimeters):

Male	Female
180	175
193	168
178	163
188	165
185	173
170	
183	

- We would like to investigate if the distribution of the height is the same in this two groups, but (for one of the reason specified above) we cannot use a T-test.

An example

- Sort the data in ascending order: (Men, Women)

163 165 168 170 173 175 178 180 183 185 188 193

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- Rank the values starting from 1 for the lowest value:

Height	163	165	168	170	173	175	178	180	183	185	188	193
Rank	1	2	3	4	5	6	7	8	9	10	11	12

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- Define W_5 to be the sum of the ranks for women, i.e.

$$W_5 = 1 + 2 + 3 + 5 + 6 = 17$$

- We look in the Statistical table for the critical values for Wilcoxon Rank Sum test with $m = 5$ and $n = 7 = m + 2$

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- We look in the Statistical table for the critical values for Wilcoxon Rank Sum test with $m = 5$ and $n = 7 = m + 2$
- We reject H_0 at a 5% significance level if $W_5 < 20$ or $W_m > 45$.
We conclude: men are taller than women.

Wilcoxon Rank-Sum test

- Let X_1, \dots, X_m and Y_1, \dots, Y_n be two independent random samples of, respectively, X and Y . Assume $m \leq n$.
- Sort the values of $X_1, \dots, X_m, Y_1, \dots, Y_n$ in ascending order and rank them starting from 1 up to $m + n$.
- If there are ties, all the tied values get the average rank of the tied value, for example if we have three tied observation with height 180 and ranks $i, i + 1, i + 2$ then each of those observation get the rank

$$\frac{i + (i + 1) + (i + 2)}{3}$$

- Compute W_m , the sum of the ranks of the X 's;
- Look at the statistical tables (Table X in (MA)) for the critical values;

Wilcoxon Signed Rank test

- Is a nonparametric test for testing equality of distribution of two **DEPENDENT** samples (null and alternative hypothesis as before);
- Example: we would like to test the effect of a new drug on blood pressure. We measure the (systolic) blood pressure of 7 patients before and after they take the drug.

Before	After
120	115
113	108
118	113
128	125
115	123
110	112
123	122

- We can see that the measurements of the blood pressure are naturally paired as "before" (X s) and "after" (Y s) $(X_1, Y_1), \dots, (X_7, Y_7)$;

Wilcoxon Signed Rank test

- We compute the pair differences $X_i - Y_i$, $i = 1, \dots, 7$.

X	Y	$ X-Y $	$\text{sign}(X-Y)$	rank	signed rank
120	115	5	+	5	+5
113	108	5	+	5	+5
118	113	5	+	5	+5
128	125	3	+	3	+3
115	123	8	-	7	-7
110	112	2	-	2	-2
123	122	1	+	1	+1

- Define

$$W_+ = \sum_{\text{all positive ranks } R_i} R_i, \quad |W_-| = \sum_{\text{all negative ranks } R_i} |R_i|$$

- zero differences are assigned the sign that is "least leading" to the rejection of the null hypothesis;

Wilcoxon Signed Rank test

- In our example

$$W_+ = 19, \quad |W_-| = 9$$

- For a two sided test, the statistic to be used is $\min\{W_+, |W_-|\}$; for right tailed test ("X>Y") we use $|W_-|$, for left tailed test ("X<Y") W_+ ;
- In any case we reject the null hypothesis if the test statistic is smaller than the critical value found in Table VIII of (MA);
- We perform the two tailed test at the level $\alpha = 0.05$. The critical point is 2 and $\min\{W_+, |W_-|\} = 9 > 2$. Thus we do not reject the null hypothesis.

Wilcoxon Signed Rank-Sum test

- Let X_1, \dots, X_n and Y_1, \dots, Y_n be two dependent (i.e. paired) random samples of, respectively, X and Y .
- Sort the absolute values of $X_1 - Y_1, \dots, X_n - Y_n$ in ascending order and rank them starting from 1 up to n .
- If there are ties, all the tied values get the average rank of the tied value;
- Compute the signed ranks by multiplying the sign of the difference $X_i - Y_i$ by its rank;
- Compute $\{W_+, |W_-|\}$
- Look at the statistical tables (Table VIII in (MA)) for the critical value;
- Reject H_0 if the test statistic is smaller than the critical value