

## Chapter 25

### Nonparametric Tests



## Inference Methods So Far

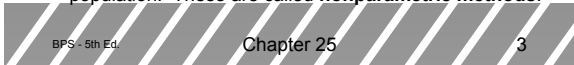
- ◆ Variables have had **Normal distributions**.
- ◆ In practice, few variables have true Normal distributions, but our methods have been **robust** (not sensitive to moderate lack of Normality).
- ◆ If the data is clearly not Normal, then using the methods of the previous chapters will yield inaccurate results. Other approaches must be investigated.



## Options for Non-Normal Data

Suppose plots suggest the data is **non-Normal**.

- ◆ If there are **extreme outliers**, investigate their cause before deciding how to proceed; do not remove outliers indiscriminately.
- ◆ In some instances, data can be **transformed** to be nearly Normal.
- ◆ Perhaps the data follow **another standard distribution**. If so, use methods based on that distribution.
- ◆ Finally, there are inference procedures that do not assume any specific form of the distribution of the population. These are called **nonparametric methods**.



## Ranking Procedures

- ◆ The procedures we will study will replace *t* tests and ANOVA when the Normality condition is not met.
- ◆ Analyses will be based on the **ranks** (place in order) of the data and not their actual values.
- ◆ Rank tests concern the **center** of the population(s).
  - for Normal distributions, center is described by the mean.
  - for skewed distributions, center is described by the median.
- ◆ Ranking the data is essential to finding the median as you need to order the values before you can determine which one is in the center.



## Comparison of Testing Procedures

Setting	Normal test	Rank test
One sample	One-sample <i>t</i> test <a href="#">Chapter 17</a>	Wilcoxon signed rank test
Matched pairs	Apply one-sample test to differences within pairs	
Two independent samples	Two-sample <i>t</i> test <a href="#">Chapter 18</a>	Wilcoxon rank sum test
Several independent samples	One-way ANOVA <i>F</i> test <a href="#">Chapter 24</a>	Kruskal-Wallis test



## Ranks

- ◆ To rank observations, first arrange them in **order** from smallest to largest.
- ◆ The **rank** of each observation is its position in this ordered list, starting with rank 1 for the smallest value.
- ◆ The **sum of the ranks** for *N* total observations will be:

$$1 + 2 + \dots + N = \frac{N(N+1)}{2}$$



### Case Study Test Scores



The test scores shown below were recorded by two different professors for two sections of the same course. Test to determine if the distributions are identical or not.

Professor A	74	78	68	72	76	69	71	74
Professor B	75	80	87	81	72	73	80	76

### Case Study: Test Scores Ranks



Professor A	Overall Rank	Professor B	Overall Rank
74	7.5	75	9
78	12	80	13.5
68	1	87	16
72	4.5	81	15
76	10.5	72	4.5
69	2	73	6
71	3	80	13.5
74	7.5	76	10.5

The sum of Professor A's ranks is  $W = 48$ .

### Ties in Ranks

- ◆ If two or more values are tied, assign each the **average** of the corresponding ranks. Then continue the ranking with the next value.
  - if the 3<sup>rd</sup> and 4<sup>th</sup> values are tied, assign each a rank of 3.5; the 5<sup>th</sup> value will get a rank of 5. (Ranks: 1, 2, 3.5, 3.5, 5).
  - if the 4<sup>th</sup>, 5<sup>th</sup>, and 6<sup>th</sup> values are tied, assign each a rank of 5; the 7<sup>th</sup> value will get a rank of 7. (Ranks: 1, 2, 3, 5, 5, 5, 7).
- ◆ The **sum of the ranks** for  $N$  total observations remains

$$\frac{N(N+1)}{2}$$

### Comparing Two Samples Wilcoxon Rank Sum Test

- ◆ The **Wilcoxon Rank Sum** procedure tests the null hypothesis that the two populations have identical distributions. ( $H_0$ : two distributions are the same)
- ◆ If the two distributions are identical, then samples of the same size should have roughly the same number of small values, the same number of medium values, and the same number of large values. Thus, the sums of ranks for each sample should be roughly the same.
- ◆ If instead, the sample sizes differ, then they should have proportionally similar sums of ranks.

#### THE WILCOXON RANK SUM TEST

Draw an SRS of size  $n_1$  from one population and draw an independent SRS of size  $n_2$  from a second population. There are  $N$  observations in all, where  $N = n_1 + n_2$ . Rank all  $N$  observations. The sum  $W$  of the ranks for the first sample is the Wilcoxon rank sum statistic. If the two populations have the same continuous distribution, then  $W$  has mean

$$\mu_W = \frac{n_1(N+1)}{2}$$

and standard deviation

$$\sigma_W = \sqrt{\frac{n_1 n_2 (N+1)}{12}}$$

The Wilcoxon rank sum test rejects the hypothesis that the two populations have identical distributions when the rank sum  $W$  is far from its mean.

### Case Study: Test Scores Computations



$$\mu_W = \frac{n_1(N+1)}{2} = \frac{8(16+1)}{2} = 68$$

$$\sigma_W = \sqrt{\frac{n_1 n_2 (N+1)}{12}} = \sqrt{\frac{(8)(8)(16+1)}{12}} = 9.52$$

Standardizing gives the **Wilcoxon test statistic** as:

$$z = \frac{W - \mu_W}{\sigma_W} = \frac{48 - 68}{9.52} = -2.10$$

### Case Study: Test Scores



- Hypotheses:**  
 $H_0$ : Professors A & B have same grade distributions  
 $H_a$ : Professors A & B have different grade distributions
- Test Statistic:**  

$$z = \frac{W - \mu_W}{\sigma_W} = \frac{48 - 68}{9.52} = -2.10$$
- P-value:**  
 $P\text{-value} = 2 P(Z < -2.10) = 0.0358$  (using a computer)  
 $P\text{-value} = 2 P(Z < -2.10) = 0.0358$  (Table A)
- Conclusion:**  
 Since the P-value is smaller than  $\alpha = 0.05$ , there is strong evidence that the two professors have different grade distributions.

### Matched Pairs Wilcoxon Signed Rank Test

- The **Wilcoxon Signed Rank** procedure tests the null hypothesis that the two populations have identical distributions ( $H_0$ : two distributions are the same) against the alternate hypothesis that one is systematically higher ( $H_a$ : scores are systematically higher in distribution 2)
- If the two distributions are identical, then differences should be centered around 0. Thus, the **sums of ranks of the positive differences** should be close to that of the negative scores.

### Case Study Tell me a Story



A study of early childhood education asked kindergarten students to tell fairy tales that had been read to them earlier in the week. Each child told two stories. The first had been read to them and the second had been read but also illustrated with pictures. An expert listened to a recording of the children and assigned a score for certain uses of language.

### Case Study Tell me a Story



Story 2	0.77	0.49	0.66	0.28	0.38
Story 1	0.40	0.72	0.00	0.36	0.55
Difference	0.37	-0.23	0.66	-0.08	-0.17

- $H_0$ : scores have the same distributions  
 $H_a$ : scores are systematically higher in Story 2

### Case Study: Tell me a Story Ranks



Absolute difference	0.08	0.17	0.23	0.37	0.66
Rank	1	2	3	4	5

The sum of ranks of positive differences is  $W^+ = 9$ .

$$\mu_{W^+} = \frac{(5)(6)}{4} = 7.5 \quad \sigma_{W^+} = \sqrt{\frac{(5)(6)(11)}{24}} = 3.708$$

$$z = \frac{W^+ - \mu_{W^+}}{\sigma_{W^+}} = \frac{9 - 7.5}{3.708} = 0.27$$

This gives a one-sided p-value of 0.394. Not enough significant evidence. Maybe sample size is too small.

#### THE WILCOXON SIGNED RANK TEST FOR MATCHED PAIRS

Draw an SRS of size  $n$  from a population for a matched pairs study and take the differences in responses within pairs. Rank the absolute values of these differences. The sum  $W^+$  of the ranks for the **positive differences** is the **Wilcoxon signed rank statistic**. If the distribution of the responses is not affected by the different treatments within pairs, then  $W^+$  has mean

$$\mu_{W^+} = \frac{n(n+1)}{4}$$

and standard deviation

$$\sigma_{W^+} = \sqrt{\frac{n(n+1)(2n+1)}{24}}$$

The **Wilcoxon signed rank test** rejects the hypothesis that there are no systematic differences within pairs when the rank sum  $W^+$  is far from its mean.