

## Chapter 14

### Nonparametric Statistics

In **all of the inference procedures** we have seen **IN THE PREVIOUS CHAPTERS** there are some assumptions that must be satisfied and we must check them before proceeding with the procedure.

Response	Inferences for	Assumptions
Categorical	<ul style="list-style-type: none"> <li>▪ Proportions (<math>p</math> or <math>p_1 - p_2</math>)</li> <li>▪ Contingency tables</li> <li>▪ Logistic Regression</li> </ul>	SRS Large Sample(s)
Quantitative	Means ( $\mu$ , $\mu_d$ , or $\mu_1 - \mu_2$ ) ANOVA SRL, MLR	SRS Normality Constant $\sigma^2$

In general we prefer to use tests based on normality assumptions since they perform “better” even when there is a slight deviation from a normal distribution.

However, we should not use them when

- There is some evidence in the sample data to indicate that the population is highly skewed
- The sample sizes are too small

In such cases we prefer to use non-parametric methods where there are fewer and less restrictive assumptions. In fact, the methods of nonparametric statistics are also called “distribution free” methods.

### Nonparametric Procedures

- The response variable is quantitative but does not have to satisfy the assumptions of the methods based on normal distribution of response.
- Use these methods with
  - Small samples and/or
  - Skewed distributions and/or
  - Groups have very different variances
- Assumptions
  - Random samples
  - Quantitative Data
- Inferences are about the **population medians**, **NOT** about mean(s)

We will see three of these tests:

Type of inference	Rank Tests	Normal Tests
Two independent samples	Wilcoxon Rank Sum Test (Mann-Whitney Test)	CIs and Significance tests for $\mu_1 - \mu_2$ (t-test)
Matched pairs	Wilcoxon signed rank test	CIs and Significance tests for $\mu_d$ (t-test)
Several independent groups	Kruskal Wallis Test	ANOVA