

# Univariate and Bivariate Tests

BUS 735: Business Decision Making and Research

# Goals

- 1 Be able to distinguish different types of data and prescribe appropriate statistical methods.
- 2 Conduct a number of hypothesis tests using methods appropriate for questions involving only one or two variables.



# Learning Outcomes

- LO1: Be able to construct and test hypotheses using a variety of bivariate statistical methods to compare characteristics between two populations.
- LO6: Be able to use standard computer packages such as R to conduct the quantitative analyses described in the learning objectives above.



# Agenda

<b>Learning Objective</b>	<b>Active Learning Activity</b>
Be able to distinguish different types of data.	Lecture / Discussion
Learn and conduct hypothesis tests on single variables.	Learn by doing: work together on examples using R.
Learn and conduct hypothesis tests for differences between two variables.	Learn by doing: work together on examples using R.

# Types of Data

- Nominal data: consists of categories that cannot be ordered in a meaningful way.
- Ordinal data: order is meaningful, but not the distances between data values.
  - Excellent, Very good, Good, Poor, Very poor.
- Interval data: order is meaningful, *and* distances are meaningful. However, there is *no natural zero*.
  - Examples: temperature, time.
- Ratio data: order, differences, and zero are all meaningful.
  - Examples: weight, prices, speed.

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# Types of Tests

- Different types of data require different statistical methods.
- Why? With interval data and below, operations like addition, subtraction, multiplication, and division are *meaningless!*
- Parametric statistics:
  - Typically take advantage of central limit theorem (imposes requirements on probability distributions)
  - Appropriate only for interval and ratio data.
  - More **powerful** than nonparametric methods.
- Nonparametric statistics:
  - Do not require assumptions concerning the probability distribution for the population.
  - There are many methods appropriate for ordinal data, some methods appropriate for nominal data.
  - Computations typically make use of *ranks* instead of actual data.

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# Single Mean T-Test

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- Test whether the population mean is equal or different to some value.
- Uses the sample mean its statistic.
- Parametric test that depends on results from Central Limit Theorem.
- Hypotheses
  - Null: The population mean is equal to some specified value.
  - Alternative: The population mean is [greater/less/different] than the value in the null.

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Dataset: Current Population Survey from 2004 that includes data on average hourly earnings, marital status, gender, and age for thousands of people.

<http://murraylax.org/datasets/cps04.csv>

Answer the following questions:

- 1 Report the mean average hourly earnings in the sample.
- 2 Construct a 95% confidence interval estimate for the average hourly earnings.
- 3 Test the hypothesis that average hourly earnings is greater than \$16.50.
- 4 Test the hypothesis that average hourly earnings *is different than* \$16.50.

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Why not perform a t-test on the mean?

- Ordinal data: Cannot compute sample means (they are meaningless), only median is meaningful.
- Small sample size and you are not sure the population is not normal.

Hypothesis test appropriate for medians:

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# Wilcoxon Signed Rank Test

## Hypotheses:

- Null: The population is centered around the null specified value.
- Alternative: The population is centered around a value different from the null specified value.

## Sample estimates:

- Sample median (middle number)
- Interpolated median: for ordinal data with limited number of outcomes, this takes into account the percentage of the data that is *strictly below* versus *strictly above* the median.

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- Dataset: 438 students in grades 4 through 6 were sampled from three school districts in Michigan. Students ranked from 1 (most important) to 4 (least important) how important grades, sports, being good looking, and having lots of money were to each of them.
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- Answer some of these questions:
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# Difference in Means (Independent Samples)

- Suppose you want to know whether the mean from one population is larger than the mean for another.
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Answer the following questions:

- 1 What is the average hourly earnings for males versus females?
- 2 Estimate a 95% confidence interval for the difference in average hourly earnings between males and females.
- 3 Test the hypothesis that men and women earn have *different average hourly earnings*
- 4 Test the hypothesis that men earn on average *more than \$2.00* per hour above women.

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# Nonparametric Tests for Differences in Medians

- Mann-Whitney U test: nonparametric test to determine difference in *medians*.
- Assumptions:
  - Samples are independent
  - The underlying distributions have the same shape (i.e. only the location of the distribution is different).  
(violating this assumption does not severely change the sampling distribution of the Mann-Whitney U test)
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- Use a **paired sampled test** if the two samples have the same individuals or sampling units.
- Many examples include before/after tests for differences:
  - The Biggest Loser: Compare the weight of people on the show before the season begins and one year after the show concludes.
  - Training session: Are workers more productive 6 months after they attended some training session versus before the training session.
- Examples besides before/after tests for differences:
  - Do students spend more time studying than watching TV?
  - Does the unemployment rate for White/Caucasian differ from the unemployment rate for African Americans (sampling unit = U.S. state).
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  - Do students spend more time studying than watching TV?
  - Does the unemployment rate for White/Caucasian differ from the unemployment rate for African Americans (sampling unit = U.S. state).
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# Parametric and Nonparametric Paired Samples Tests 16/ 18

## Paired-samples difference in means t-test

- Appropriate for interval or ratio data
- Appropriate when assumptions of CLT are met

## Wilcoxon-Signed Rank Test for Paired-samples

- Tests for a difference in medians (center of distribution)
- Appropriate for ordinal, interval, or ratio data
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# Paired Samples Means: Motor Vehicle Fatalities

17 / 18

- Centers for Disease Control and Prevention (CDC) state-level data (50 obs) on motor vehicle fatalities by state, age, and sex
- Variables: Motor vehicle occupant fatality rate per 100,000 members of the population
  - Over all age groups
  - Individual age groups: 0-20, 21-34, 35-54, and 55+.
  - Male versus Female
- Dataset: <http://murraylax.org/datasets/vehiclefatalities.csv>
- Answer the following questions:
  - Report the sample average mortality rate for age groups 21-34 and 35-54.
  - Report the difference in the average mortality rate for age groups 21-34 and 35-54.
  - Report a confidence interval for the difference above.
  - Do age groups 21-34 and 35-54 have different average mortality rates?

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# Paired-Samples Medians: Grade School Attitudes

- Dataset: 438 students in grades 4 through 6 were sampled from three school districts in Michigan. Students ranked from 1 (most important) to 4 (least important) how important grades, sports, being good looking, and having lots of money were to each of them.
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