Statistical Significance and Univariate and Bivariate Tests

BUS 230: Business and Economics Research and Communication

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1.1 Goals

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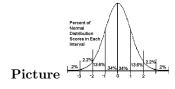
- Specific goals:
 - Re-familiarize ourselves with basic statistics ideas: sampling distributions, hypothesis tests, p-values.
 - Be able to distinguish different types of data and prescribe appropriate statistical methods.
 - Conduct a number of hypothesis tests using methods appropriate for questions involving only one or two variables.
- Learning objectives:
 - LO2: Interpret data using statistical analysis.
 - LO2.3: Formulate conclusions and recommendations based upon statistical results.

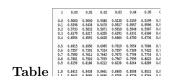
2 Statistical Significance

2.1 Sampling Distribution

Probability Distribution

- **Probability distribution:** summary of all possible values a variable can take along with the probabilities in which they occur.
- Usually displayed as:





Formula

$$f(x|\mu,\sigma) =$$

$$\frac{1}{\sigma\sqrt{2\pi}}e^{-\left(\frac{(x-\mu)^2}{2\sigma^2}\right)}$$

• Normal distribution: often used "bell shaped curve", reveals probabilities based on how many standard deviations away an event is from the mean.

Sampling distribution

- Imagine taking a sample of size 100 from a population and computing some kind of statistic.
- The statistic you compute can be anything, such as: mean, median, proportion, difference between two sample means, standard deviation, variance, or anything else you might imagine.
- Suppose you repeated this experiment over and over: take a sample of 100 and compute and record the statistic.
- A sampling distribution is the probability distribution of the statistic
- Is this the same thing as the probability distribution of the population? NO! They may coincidentally have the same shape though.

Example

- Sampling Distribution Simulator
- In reality, you only do an experiment once, so the sampling distribution is a hypothetical distribution.
- Why are we interested in this?

Desirable qualities

What are some qualities you would like to see in a sampling distribution?

- The average of the sample statistics is equal to the true population parameter.
- Want the variance of the sampling distribution to be as small as possible. Why?
- Want the *sampling distribution* to be normal, regardless of the distribution of the population.

2.2 Central Limit Theorem

Central Limit Theorem

- Given:
 - Suppose a RV x has a distribution (it need not be normal) with mean μ and standard deviation σ .
 - Suppose a sample mean (\bar{x}) is computed from a sample of size n.
- Then, if n is sufficiently large, the sampling distribution of \bar{x} will have the following properties:
 - The sampling distribution of \bar{x} will be normal.
 - The mean of the sampling distribution will equal the mean of the population (consistent):

$$\mu_{\bar{x}} = \mu$$

- The standard deviation of the sampling distribution will decrease with larger sample sizes, and is given by:

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

Central Limit Theorem: Small samples

If n is small (rule of thumb for a single variable: n < 30)

- The sample mean is still consistent.
- Sampling distribution will be normal if the distribution of the population is normal.

Example 1

Suppose average birth weight is $\mu = 7lbs$, and the standard deviation is $\sigma = 1.5lbs$.

What is the probability that a sample of size n = 30 will have a mean of 7.5lbs or greater?

$$z = \frac{\bar{x} - \mu_{\bar{x}}}{\sigma_{\bar{x}}} = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$$

$$z = \frac{7.5 - 7}{1.5/\sqrt{30}} = 1.826$$

In Excel, we can compute P(z > 1.826) using the formula

=1 - normsdist(1.826). The probability the sample mean is greater than 7.5lbs is:

$$P(\bar{x} > 7.5) = P(z > 1.826) = 0.0339$$

Example 2

Suppose average birth weight is $\mu = 7lbs$, and the standard deviation is $\sigma = 1.5lbs$.

What is the probability that a randomly selected baby will have a weight of 7.5lbs or more? What do you need to assume to answer this question?

Must assume the population is normally distributed. Why?

$$z = \frac{\bar{x} - \mu}{\sigma} = \frac{7.5 - 7}{1.5} = 0.33$$

In Excel, we can compute P(z > 0.33) using the formula

=1 - normsdist(0.33). The probability that a baby is greater than 7.5lbs is:

$$P(x > 7.5) = P(z > 0.33) = 0.3707$$

Example 3

- Suppose average birth weight of all babies is $\mu = 7lbs$, and the standard deviation is $\sigma = 1.5lbs$.
- Suppose you collect a sample of 30 newborn babies whose mothers smoked during pregnancy.
- Suppose you obtained a sample mean $\bar{x} = 5$ lbs. If you assume the mean birth weight of babies whose mothers used illegal drugs has the same sampling distribution as the rest of the population, what is the probability of getting a sample mean this low or lower?

Example 3 continued

$$z = \frac{\bar{x} - \mu_{\bar{x}}}{\sigma_{\bar{x}}} = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$$
$$z = \frac{5 - 7}{1.5 / \sqrt{30}} = -7.30$$

In Excel, we can compute P(z < -7.30) using the formula

=normsdist(-7.30). The probability the sample mean is less than or equal to 5lbs is:

$$P(\bar{x} < 5) = P(z < -7.30) = 0.000000000000014$$

That is, if using drugs during pregnancy actually does still lead to an average birth weight of 7 pounds, there was only a 0.000131 (or 0.0131%) chance of getting a sample mean as low as six or lower. This is an extremely unlikely event if the assumption is true. Therefore it is likely the assumption is not true.

2.3 Hypotheses Tests

Statistical Hypotheses

- A hypothesis is a claim or statement about a property of a population.
 - Example: The population mean for systolic blood pressure is 120.
- A hypothesis test (or test of significance) is a standard procedure for testing a claim about a property of a population.
- Recall the example about birth weights with mothers who use drugs.
 - Hypothesis: Using drugs during pregnancy leads to an average birth weight of 7 pounds (the same as with mothers who do not use drugs).

Null and Alternative Hypotheses

• The **null hypothesis** is a statement that the value of a population parameter (such as the population mean) *is equal to* some claimed value.

$$- H_0$$
: $\mu = 7$.

- The alternative hypothesis is an alternative to the null hypothesis; a statement that says a parameter differs from the value given in the null hypothesis.
- Pick *only one* of the following for your alternative hypothesis. Which one depends on your research question.
 - H_a : $\mu < 7$.
 - H_a : $\mu > 7$.
 - $-H_a$: $\mu \neq 7$.
- In hypothesis testing, assume the null hypothesis is true until there is strong statistical evidence to suggest the alternative hypothesis.
- Similar to an "innocent until proven guilty" policy.

Hypothesis tests

• (Many) hypothesis tests are all the same:

 $z \text{ or } t = \frac{\text{sample statistic} - \text{null hypothesis value}}{\text{standard deviation of the sampling distribution}}$

- Example: hypothesis testing about μ :
 - Sample statistic = \bar{x} .
 - Standard deviation of the sampling distribution of \bar{x} :

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

P-values

- The P-value of the test statistic, is the area of the sampling distribution from the sample result in the direction of the alternative hypothesis.
- Interpretation: If the null hypothesis is correct, than the p-value is the probability of obtaining a sample that yielded your statistic, or a statistic that provides even stronger evidence of the null hypothesis.
- The p-value is therefore a measure of statistical significance.
 - If p-values are very small, there is strong statistical evidence in favor of the alternative hypothesis.
 - If p-values are large, there is insignificant statistical evidence. When large, you fail to reject the null hypothesis.
- Best practice is writing research: report the p-value. Different readers may have different opinions about how small a p-value should be before saying your results are statistically significant.

3 Univariate Tests

3.1 Types of Data/Tests

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.

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 data's ranks instead of actual data.

3.2 Hypothesis Testing about Mean

Single Mean T-Test

- Test whether the population mean is equal or different to some value.
- Uses the sample mean its statistic.
- T-test is used instead of Z-test for reasons you may have learned in a statistics class. Interpretation is the same.
- 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.

Example Questions

- Suppose we have a dataset of individual schools and the average pay for teachers in each school, and the level of spending as a ratio of the number of students (spending per pupil).
 - Show some descriptive statistics for teacher pay and expenditure per pupil.
 - Is there statistical evidence that teachers make less than \$50,000 per year?
 - Is there statistical evidence that expenditure per pupil is more than \$7,500?

3.3 Hypothesis Testing about Proportion

Single Proportion T-Test

- Proportion: Percentage of times some characteristic occurs.
- Example: percentage of consumers of soda who prefer Pepsi over Coke.

Sample proportion =
$$\frac{\text{Number of items that has characteristic}}{\text{sample size}}$$

- Example questions:
 - Are more than 50% of potential voters most likely to vote for Barack Obama in the next presidential election?
 - Suppose typical brand-loyalty turn-over in the mobile phone industry is 10%. Is there statistical evidence that AT&T has brand-loyalty turnover more than 10%?

3.4 Nonparametric Testing about Median

Single Median Nonparametric Test

- Why?
 - 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.
- Sign test: can use tests for proportions for testing the median.
 - For a null hypothesized population median...
 - Count how many observations are above the median.
 - Test whether that proportion is greater, less than, or not equal to 0.5.
 - For small sample sizes, use binomial distribution instead of normal distribution.

Example Questions

- Instructor evaluations have an ordinal scale: Excellent, Very Good, Good, Poor, Very Poor.
 - Is there statistical evidence that the median rating for a professor is below 'Very Good'?
- Suppose you have a frequency determination question on your survey. Is this an ordinal scale? Is the median an appropriate measure of center?

4 Bivariate Tests

4.1 Difference in Populations (Independent Samples)

Difference in Means (Independent Samples)

- Suppose you want to know whether the mean from one population is larger than the mean for another.
- Independent samples means you have different individuals in your two sample groups.
- Examples:
 - Compare sales volume for stores that advertise versus those that do not.
 - Compare production volume for employees that have completed some type of training versus those who have not.
- Statistic: Difference in the sample means $(\bar{x}_1 \bar{x}_2)$.
- Hypotheses:
 - Null hypothesis: the difference between the two means is zero.
 - Alternative hypothesis: the difference is [above/below/not equal] to zero.

Nonparametric Tests for Differences in Medians

- Mann-Whitney U test: nonparametric test to determine difference in medians.
- Can you suggest some examples?
- Assumptions:
 - Samples are independent of one another.

- The underlying distributions have the same shape (i.e. only the location of the distribution is different).
- It has been argued that violating the second assumption does not severely change the sampling distribution of the Mann-Whitney U test.
- Null hypothesis: medians for the two populations are the same.
- Alternative hypotheses: medians for the two populations are different.

4.2 Paired Samples

Dependent Samples - Paired Samples

- Use a **paired sampled test** if instead the two samples have the same individuals before and after some treatment.
- Examples:
 - 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.
- Really simple: for each individual subtract the before treatment measure from the after treatment measure (or vice-versa).
- Treat your new series as a single series.
- Conduct one-sample tests.

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Conclusions

- Ideas to keep in mind:
 - What is a sampling distribution? What does it imply about p-values and statistical significance?
 - When it is appropriate to use parametric versus non-parametric methods.
 - Most univariate and bivariate questions have a parametric and nonparametric approach.
- Next class: closed-book, closed note quiz (so study!); and an in-class exercise practicing this stuff in SPSS.
- Next: Statistics to understand the relationship between two variables.