#### Statistical Inference Overview

Young W. Lim

2024-07-22 Mon

#### Outline

Based on

- Overview
  - Statistical Inference
  - Types of Hypothesis Tests

#### Based on

#### "Understanding Statistics in the Behavioral Sciences" R. R. Pagano

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#### Populations and Samples (1)

- population: everything in the group that we want to learn about.
- sample: a part of the population.
- Examples of populations and a sample from those populations:

| Population                      | Sample                   |
|---------------------------------|--------------------------|
| All of the people in Germany    | 500 Germans              |
| All of the customers of Netflix | 300 Netflix customers    |
| Every car manufacturer          | Tesla, Toyota, BMW, Ford |

#### Populations and Samples (2)

- For good statistical analysis, the sample needs to be as <u>similar</u> as possible to the population.
- If they are <u>similar enough</u>, we say that the <u>sample</u> is representative of the population.
- The sample is used to make <u>conclusions</u> about the whole <u>population</u>.

#### Populations and Samples (3)

- If the sample is <u>not</u> <u>similar enough</u> to the whole <u>population</u>, the <u>conclusions</u> could be useless.
- Many words have specific meanings in statistics.
- The word population normally refers to a group of people.
- In statistics, it is any <u>specific group</u> that we are interested in learning about.

#### Statistical Inference

- Using <u>data analysis</u> and <u>statistics</u> to make <u>conclusions</u> about a <u>population</u> is called <u>statistical inference</u>.
- The main types of statistical inference are:
  - Estimation
  - Hypothesis testing

#### Estimation (1)

- <u>Statistics</u> from a <u>sample</u> are used to <u>estimate</u> population <u>parameters</u>.
- The most likely value is called a point estimate.
- There is always uncertainty when estimating.

## Estimation (2)

- The uncertainty is often expressed as confidence intervals defined by a likely lowest and highest value for the parameter.
- An example could be a confidence interval for the number of bicycles a Dutch person owns:
  - The average number of bikes a Dutch person owns is between 3.5 and 6.

## Hypothesis Testing (1)

- a method to check if a claim about a population is true.
- checks how <u>likely</u> it is that a <u>hypothesis</u> is <u>true</u> is based on the <u>sample</u> data.
- there are different types of hypothesis testing.

## Hypothesis Testing (2)

- the steps of the test depends on:
  - Type of data (categorical or numerical)
  - If you are looking at:
    - a single group
    - comparing one group to another
    - comparing the same group before and after a change

## Hypothesis Testing (3)

- a hypothesis is a claim about a population parameter
- a hypothesis test is a formal procedure to check if a hypothesis is true or not.
- examples of claims that can be checked:
  - the average height of people in Denmark is more than 170 cm.
  - the share of left handed people in Australia is not 10%.
  - The <u>average income</u> of dentists is <u>less</u> the <u>average income</u> of lawyers.

 $\verb|https://www.w3schools.com/statistics/statistics_hypothesis\_testing.php|$ 

#### Steps of Hypothesis Testing

- State your research hypothesis as a null hypothesis  $(H_0)$  and alternate hypothesis and  $(H_a \text{ or } H_1)$ .
- Collect data in a way designed to test the hypothesis.
- Operation Perform an appropriate statistical test
- Decide whether to reject or fail to reject your null hypothesis
- Present the findings in your results and discussion section.

https://www.scribbr.com/statistics/hypothesis-testing/

#### Statistical tests (1)

- Statistical tests are used in hypothesis testing.
   They can be used to:
  - determine whether a <u>predictor variable</u> has a statistically significant <u>relationship</u>
     with an outcome variable
  - estimate the difference between two or more groups.

#### Statistical tests (2)

 Statistical tests assume a null hypothesis of no relationship or no difference between groups.
 Then they determine whether the observed data fall outside of the range of values predicted by the null hypothesis.

## Statistical tests (3)

- Statistical tests work by calculating a test statistic
  - a number that describes how much the <u>relationship</u> between variables in your test differs from the <u>null hypothesis</u> of no relationship
- then calculates a p value (probability value).
   the p-value estimates how likely it is
   that you would see the <u>difference</u> described by the test statistic if the null hypothesis of no relationship were true.

#### Statistical tests (4)

- If the value of the test statistic is more extreme than the statistic calculated from the null hypothesis, then you can infer a statistically significant relationship between the predictor and outcome variables
- If the value of the test statistic is less extreme than the one calculated from the null hypothesis, then you can infer no statistically significant relationship between the predictor and outcome variables.

#### Statistical tests (5)

- You can <u>perform</u> <u>statistical</u> tests on data that have been <u>collected</u> in a <u>statistically valid</u> manner
  - either through an experiment,
  - or through observations made using probability sampling methods.
- For a statistical test to be <u>valid</u>, your sample size needs to be <u>large</u> enough to <u>approximate</u> the true distribution of the population being studied.
- To <u>determine</u> which <u>statistical</u> test to use, you need to know:
  - whether your data meets certain assumptions.
  - the types of variables that you're dealing with.

#### Statistical assumptions (1)

- Statistical tests make some common <u>assumptions</u> about the data they are testing:
  - Independence of observations (a.k.a. no autocorrelation):
  - Homogeneity of variance:
  - Normality of data:

#### Statistical assumptions (2)

- Independence of observations (a.k.a. no autocorrelation):
   The observations/variables you include in your test are not related (for example, multiple measurements of a single test subject are not independent, while measurements of multiple different test subjects are independent).
  - Homogeneity of variance: the variance within each group being compared is similar among all groups. If one group has much more variation than others, it will limit the test's effectiveness.
  - Normality of data: the data follows a normal distribution (a.k.a. a bell curve). This assumption applies only to quantitative data.

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## Types of data (1)

- Data is a specific measurement of a variable it is the value you record in your data sheet.
   Data is generally divided into two categories:
  - Quantitative data represents amounts
  - Categorical data represents groupings

## Types of data (2)

- a variable that contains quantitative data is a quantitative variable;
- a variable that contains categorical data is a categorical variable.
- Each of these types of variables can be broken down into further types.

#### Quantitative variables

 When you collect quantitative data, the numbers you record represent real amounts that can be added, subtracted, divided, etc. There are two types of quantitative variables: discrete and continuous.

#### Catagorical variables

- Categorical variables represent groupings of some kind. They are sometimes recorded as numbers, but the numbers represent categories rather than actual amounts of things.
- There are three types of categorical variables: binary, nominal, and ordinal variables.

#### Independant and dependant variables

- Experiments are usually designed to find out what effect one variable
  has on another in our example, the effect of salt addition on plant
  growth.
- You manipulate the independent variable (the one you think might be the cause) and then measure the dependent variable (the one you think might be the effect) to find out what this effect might be.
- You will probably also have variables that you hold constant (control variables) in order to focus on your experimental treatment.

#### The Null and Alternative Hypothesis

- Hypothesis testing is based on making two different claims about a population parameter.
- The null hypothesis  $(H_0)$  and the alternative hypothesis  $(H_1)$  are the claims.
- The two claims needs to be mutually exclusive, meaning only one of them can be true.
- The alternative hypothesis is typically what we are trying to prove.
- For example, we want to check the following claim:
  - "The average height of people in Denmark is more than 170 cm."

https://www.w3schools.com/statistics/statistics\_hypothesis\_testing.php

# Summary (1) comparing means

| tests                               |  |
|-------------------------------------|--|
| • one-sample test                   | comparing sample mean, population mean |
| <ul> <li>two-sample test</li> </ul> | comparing two independent sample means |
| • paired test                       | comparing two related sample means     |

| tests    | test conditions                            |
|----------|--|
| • t-test | 1. when the population variance is known   |
|          | 2. when the sample size is large           |
| • z-test | 1, when the population variance is unknown |
|          | 2. the sample size is small                |

# Summary (2) comparing means

| one sample z-test | sample mean, population mean               |  |
|-------------------|--|--|
|                   | known population var / large sample size   |  |
| one sample t-test | sample mean, population mean               |  |
|                   | unknown population var / small sample size |  |
| two sample z-test | two independent sample means               |  |
|                   | known population var / large sample size   |  |
| two sample t-test | two independent sample means               |  |
|                   | unknown population var / small sample size |  |
| paired t-test     | two <i>related</i> sample means            |  |
|                   | unknown population var / small sample size |  |
|                   | -  |  |

## Summary (3) comparing proportions

| one sample propotion  | sample proportion, population proportion   |
|-----------------------|--|
| test                  | when $np \geq 10$ and $n(1-p) \geq 10$     |
| two sample proportion | two independent sample proportions         |
| test                  | when $n ho \geq 10$ and $n(1- ho) \geq 10$ |

#### test conditions

```
the normal approximation is used when both np \geq 10 and n(1-p) \geq 10 (data should have at least 10 "successes" and at least 10 "failures" )
```

# Summary (4)

| compare variances between                  |                             |
|--|-----------------------------|
| sample variance, known population variance | Chi-square test             |
| two independent sample variances           | F-test                      |
| observed frequencies, expected frequencies | goodness of fit test        |
| observed frequencies, expected frequencies | contingency tables          |
| means of three or more independent samples | ANOVA (Analysis of Variance |
|  |                             |

## Tests for Comparing Means (1)

- One-sample z-test:
  - used to <u>compare</u> the <u>mean</u> of a <u>sample</u> to a known population <u>mean</u>
  - used when the population variance is known, or the sample size is large (n > 30).
- Two-sample z-test:
  - used to compare the means of two independent samples.
  - used when the population variances are known, or the sample sizes are large (n > 30).

## Tests for Comparing Means (2)

- One-sample t-test:
  - used to <u>compare</u> the <u>mean</u> of a <u>sample</u> to a known population <u>mean</u>.
  - used when the population variance is unknown, and the sample size is small (n < 30).
- Two-sample t-test:
  - used to compare the means of two independent samples.
  - used when the population variances are unknown, and the sample sizes are small (n < 30).

## Tests for Comparing Means (3)

#### Paired t-test:

- used to <u>compare</u> the <u>means</u> of two <u>related</u> <u>samples</u>, such as the <u>before</u> and <u>after measurements</u> of the same group of subjects.
- used when the <u>population</u> variances are <u>unknown</u>, and the sample size is *small* (n < 30).

## Tests for Comparing Proportions (1)

- Let us consider the parameter p of the population proportion
  - eg) we might want to know the proportion of males within a total population of adults when we conduct a survey.
- A test of proportion will assess
   whether or not a sample from a population represents
   the true proportion of the entire population

https://online.stat.psu.edu/statprogram/reviews/statistical-concepts/proportions

#### Examples of proportions

- an example
  - newborn babies are more likely to be boys than girls.
  - a random sample found 13,173 boys were born among 25,468 newborn children
  - the sample proportion of boys was 0.5172 (=  $\frac{13173}{25468}$ )
  - is this sample evidence that the birth of boys is more common than the birth of girls in the entire population? (0.5172 > 0.4828)

https://online.stat.psu.edu/statprogram/reviews/statistical-concepts/proportions

#### Tests for Comparing Proportions (2-1)

- examples involved testing whether a single population proportion p equals some value.
- different examples of testing whether one population proportion equals a second population proportion

# Tests for Comparing Proportions (2-2)

- Additionally, most of our examples thus far have involved
  - left-tailed tests in which the alternative hypothesis involved
  - right-tailed tests in which the alternative hypothesis involved
- Here, let's consider an example that tests the equality of two proportions against the alternative that they are not equal

# Tests for Comparing Proportions (2-3)

- Time magazine reported the result of a telephone poll of 800 adult Americans.
- the question posed of the Americans who were surveyed was:
   "Should the federal tax on cigarettes be raised to pay for health care reform?"
- the results of the survey were:

| Smokers                             |
|-------------------------------------|
| $n_2 = 195$                         |
| $y_2=41$ said yes                   |
| $\hat{p}_2 = \frac{41}{195} = 0.21$ |
|                                     |

### Tests for Comparing Proportions (2-4)

- If  $p_1$  = the proportion of the <u>non-smoker</u> population who reply "yes"
- and  $p_2 =$  the proportion of the <u>smoker</u> population who reply "yes,"
- then we are interested in testing the <u>null hypothesis</u>:

 $H_0: p_1 = p_2$  against the <u>alternative hypothesis</u>:

 $H_A: p_1 \neq p_2$ 

 Before we can actually conduct the hypothesis test, we'll have to derive the appropriate test statistic.

# Tests for Comparing Proportions (2-5)

• The overall sample proportion is:

$$\hat{p} = \frac{41+351}{195+605} = \frac{392}{800} = 0.49$$

• that implies then that the test statistic for testing:

$$H_0: p_1 = p_2 \text{ versus } H_A: p_1 \neq p_2$$
 is:  

$$Z = \frac{(0.58 - 0.21) - 0}{\sqrt{0.49(0.51)(\frac{1}{195} + \frac{1}{605})}} = 8.9$$

# Tests for Comparing Proportions (3)

- One-sample proportion test :
  - used to <u>compare</u> the <u>proportion</u> of a <u>sample</u> to a known population <u>proportion</u>.
  - the normal approximation is used when both  $np \geq 10$  and  $n(1-p) \geq 10$  (data should have at least 10 "successes" and at least 10 "failures" ) (in some books, it is 5)

# Tests for Comparing Proportions (4)

- Two-sample proportion test :
  - used to <u>compare</u> the <u>proportions</u> of two independent <u>samples</u>.
  - the normal approximation is used when both  $np \geq 10$  and  $n(1-p) \geq 10$  (data should have at least 10 "successes" and at least 10 "failures" ) (in some books, it is 5)

```
https://www.qualitygurus.com/common-types-of-hypothesis-tests/
```

# Tests for Comparing Proportions (5)

- Time magazine reported the result of a telephone poll of 800 adult Americans.
- The question posed of the Americans who were surveyed was:
   "Should the federal tax on cigarettes be raised to pay for health care reform?"

The results of the survey were:

### Tests for Comparing Proportions (6)

| Non-Smokers                          | Smokers                             |
|--------------------------------------|-------------------------------------|
| $n_1 = 605$                          | $n_2 = 195$                         |
| $y_1 = 351$ said "yes"               | $y_2 = 41 \text{ said "yes"}$       |
| $\hat{p}_1 = \frac{351}{605} = 0.58$ | $\hat{p}_2 = \frac{41}{195} = 0.21$ |

 Is there sufficient evidence at the , say, to conclude that the two populations - smokers and non-smokers - differ significantly with respect to their opinions?

# Tests for Comparing Proportions (7)

- Errr.... that Z-value is off the charts, so to speak. Let's go through the formalities anyway making the decision first using the rejection region approach, and then using the P-value approach. Putting half of the rejection region in each tail, we have:
- That is, we reject the null hypothesis if or if . We clearly reject , since 8.99 falls in the "red zone," that is, 8.99 is (much) greater than 1.96. There is sufficient evidence at the 0.05 level to conclude that the two populations differ with respect to their opinions concerning imposing a federal tax to help pay for health care reform.

# Tests for Comparing Proportions (8)

- That is, the P-value is less than 0.0001. Because, we reject the null hypothesis. Again, there is sufficient evidence at the 0.05 level to conclude that the two populations differ with respect to their opinions concerning imposing a federal tax to help pay for health care reform.
- Thankfully, as should always be the case, the two approaches.... the critical value approach and the P-value approach... lead to the same conclusion

# Tests for Comparing Variance

- Chi-square test for variance :
  - used to <u>compare</u> the <u>variance</u> of a <u>sample</u> to a <u>known population variance</u>
- F-test for variance :
  - used to <u>compare</u> the <u>variances</u> of two <u>independent samples</u>

#### Other Common Tests (1)

- Goodness of fit test :
- used to determine whether a sample fits a specific distribution.
- used to compare the observed frequencies of a categorical variable to the expected frequencies under a particular distribution.

### Other Common Tests (2)

- Testing for independence of two attributes (Contingency Tables) :
- used to determine whether there is a <u>relationship</u> between two *categorical variables*.
- often used in the form of a chi-square test,
   which compares the observed frequencies in a contingency table
   to the expected frequencies under the assumption of independence.

### Other Common Tests (3)

- ANOVA (Analysis of Variance) :
- used to compare the means of three or more independent samples.
- used to <u>determine</u> whether there is a significant <u>difference</u> between the <u>means</u> of the groups.

#### One-sample z-test

- used to test a hypothesis about the population mean
- based on the assumption that the <u>sample</u> is drawn from a <u>normally distributed population</u>.
  - the null hypothesis
     the population mean is equal to a specific value
  - the alternative hypothesis
     the population mean is not equal to that value

#### Two-sample z-test

- based on the assumption that both <u>samples</u> are drawn from <u>normally distributed populations</u> with <u>equal variances</u>.
- the two-sample z-test requires
   that the population standard deviations be known or
   that the sample sizes be large (30 or more),
  - the null hypothesis
     the means of the two samples are equal
  - the alternative hypothesis the means are not equal

#### One-sample t-test

- used to test a hypothesis about the population mean
- based on the assumption that the <u>sample</u> is drawn from a <u>normally distributed population</u>
  - the null hypothesis
     the population mean is equal to a specific value
  - the alternative hypothesis
     the population mean is not equal to that value

#### Two-sample t-test

- based on the assumption that the samples are drawn from populations with normal distributions.
- the two-sample t-test
  that the <u>population</u> standard deviations <u>need not</u> be <u>known</u> or
  that the sample sizes need not be <u>large</u> (30 or more),
  - the null hypothesis
     the means of the two samples are equal
  - the alternative hypothesis the means are not equal

#### Paired t-test

- used to test a hypothesis about the <u>difference</u> between the means of the two samples
- based on the assumption that the <u>differences</u> between the pairs are <u>normally distributed</u>
- In a <u>dependent</u> <u>two-sample</u> <u>t-test</u> (a <u>paired</u> <u>t-test</u>), the <u>samples</u> in the two <u>groups</u> being compared are <u>related</u> in some way.
  - the null hypothesis there is no difference between the means of the two samples
  - the alternative hypothesis there is a difference between the means

#### Two proportions z-test

- used to test a hypothesis about the <u>difference</u> between the proportions of the two samples and
- based on the assumption that the <u>samples</u> are drawn from populations with a <u>normal distribution</u>
  - the null hypothesis:
     there is no difference between the proportions of the two samples
  - the alternative hypothesis:
     there is a <u>difference</u> between the <u>proportion</u>

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