

# P-values, confidence intervals and point estimates. Inference in Statistics

Anne Lyngholm Sørensen

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# Outline

Introduction

Hypothesis-testing

P-values

Point estimates

Confidence intervals

Summary

# Plan for this talk

- We are going to talk about statistical inference.
  - Especially statistical inference in publications.
- This includes a discussion about whether p-values are the most important statistical metric in publications. And why it has so much attention.
- We will be using simulations for understanding p-values, point estimates and confidence intervals in more depth.

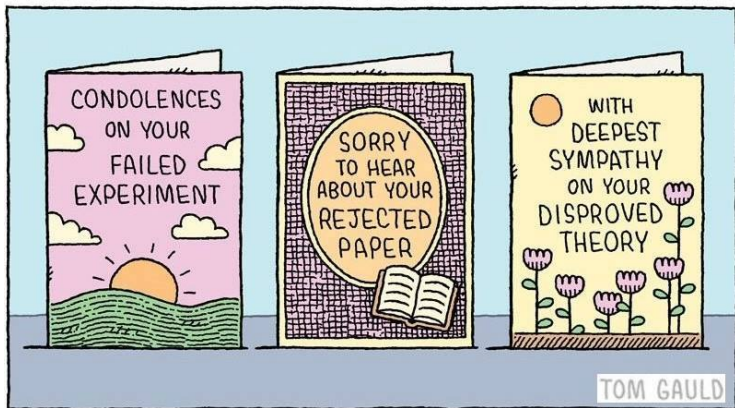
What is statistical inference? Is it this?

$p < 0.05$



What is statistical inference? Is it this?

## WHEN THE P-VALUE IS JUST ABOVE .05



## Different kinds of hypotheses

- "The hazard of death is different in the treatment group compared to the control group."
- "The level of testosterone is higher in males with XXX gene-mutation than other males"
- "The rate of infection per year is lower in people treated with medicine y compared with people treated with medicine x"

# Different kinds of hypotheses

## Two-sided

- $H_0$  : Hazard Ratio = 1 vs.  $H_A$  : Hazard Ratio  $\neq$  1

## One-sided

- $H_0$  :  $Level_A > Level_B$  vs.  $H_A$  :  $Level_A \leq Level_B$ 
  - Rewrite  $H_0$  :  $Level_A - Level_B > 0$ .
- $H_0$  : Rate Ratio  $< 1$  vs.  $H_A$  : Rate Ratio  $\geq 1$

# What difference really matters?

We can test these hypotheses, but maybe we should think more in depth about what we are testing.

Say we do a hypothesis test of:

$$H_0 : Level_A - Level_B > 0 \text{ vs. } H_A : Level_A - Level_B \leq 0$$

If the true difference is 0.0001, we will with enough participants be able to reject this hypothesis. Is that true difference a relevant difference?



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A p-value below 0.05 (a rejected null hypothesis) does not equal relevance

# Hypothesis for this talk

Are Australians the same height as New Zealanders?

$$H_0 : \text{Average Height}_{AUS} - \text{Average Height}_{NZ} = 0$$

vs.

$$H_A : \text{Average Height}_{AUS} - \text{Average Height}_{NZ} \neq 0$$

We are going to assume that the heights for our two populations are approximately normal distributed.

# Simulation

Let us simulate some data!

- Everything that exists is an object.
- Everything that happens is a function call.

# P-values and hypothesis testing

To be able to reject a hypothesis, we need the following three things:

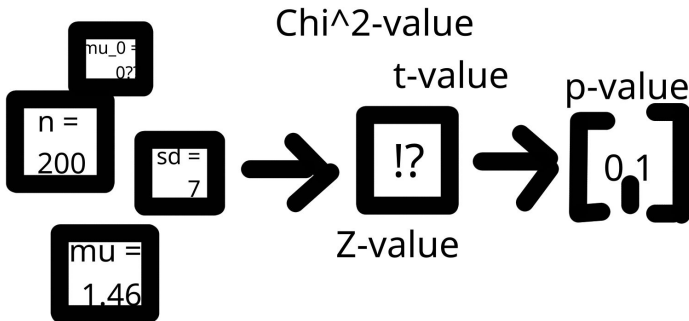
1. A hypothesis to test.
2. A metric to evaluate our data in relation to the hypothesis.
3. Some cut-off value related to when we reject or can not reject the hypothesis.

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3. Some cut-off value related to when we reject or can not reject the hypothesis.

The  $p$  value is the metric that we most often use. And the most common cut-off value is 0.05.



## Z-value and P-value

The  $p$  value is between 0 and 1. And it tells the probability of observing something at least as extreme as a specific value in relation to the null hypothesis.

To evaluate if the value we have measured is close to the null hypothesis, we need to understand which values are likely under the null hypothesis.

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To evaluate if the value we have measured is close to the null hypothesis, we need to understand which values are likely under the null hypothesis.

$$Z = \frac{\hat{\mu} - \mu_0}{s^2 / \sqrt{n}}$$

- $\hat{\mu}$  is the estimate
- $\mu_0$  is the null hypothesis value
- $s^2$  is the variance estimate
- $n$  is the number of observations



# When is the $Z$ value large?

What can give you a large  $Z$  value?

- The larger the distance between  $\hat{\mu} - \mu_0$ .
- The larger the number of observations  $n$ .

So do not get fooled by very small p-values. There could be testing of a too loose hypothesis. And p-values can be really small by having a really large number of observations.

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So do not get fooled by large p-values. They could test too strict a hypothesis. And they can become significant by getting a larger number of number of observations.

# Height example

We are going to use the height example to illustrate some of the points.

$$t = \frac{\hat{\mu}_{AUS} - \hat{\mu}_{NZ}}{\sqrt{s^2 \cdot (1/n_1 + 1/n_2)}}$$

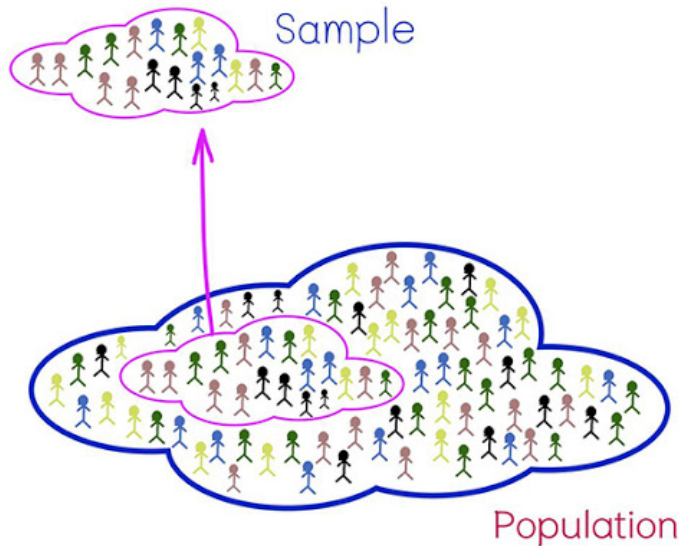
# Simulation

Let us do some simulations to get smarter!

# Point Estimates

- A crucial information is the point estimate. It is the best guess in data about the metric of interest.
- There is not that many ways to go wrong with a point estimate. The most important thing to remember is that it is just an estimate.
- The more observations, the better and more precise estimate.

## Point Estimates



# Simulation

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# The good old formula

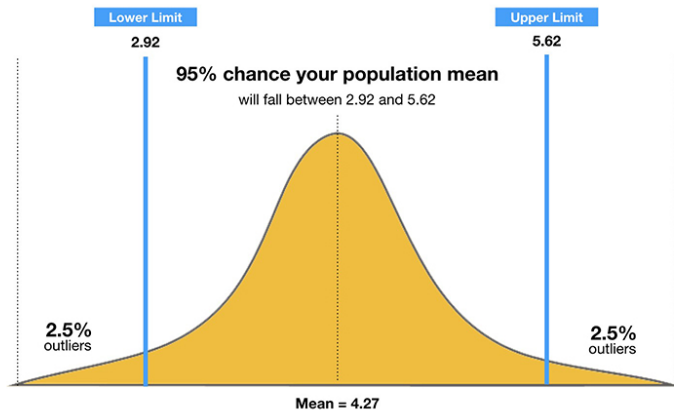
One sample:

$$t = \hat{\mu} \pm 1.96 \frac{s}{\sqrt{n}}$$

Our example:

$$95 \text{ CI} = \hat{\mu}_{AUS} - \hat{\mu}_{NZ} \pm 1.96 \frac{s}{\sqrt{1/n_{AUS} + 1/n_{NZ}}}$$





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The choice of the word "confidence" is actually quite crucial in understanding what a confidence interval is. And most importantly, confidence does not equal probability!

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- This means that before the interval is computed, there is a 95% chance that the true value will be contained in the interval. But when it is computed, it will either contain the true value or not. And we will never know the true value.
- A 95% confidence level does not mean that 95% of the sample data lie within the confidence interval.

# The non-fun tombola example





# Simulation

Let us do the last round of simulation for understanding confidence better!

# Summary

- P-values can be deceiving in relation to the relevance of ones findings.
- Always think about whether the point estimate and belonging confidence interval are showing something that can be relevant.