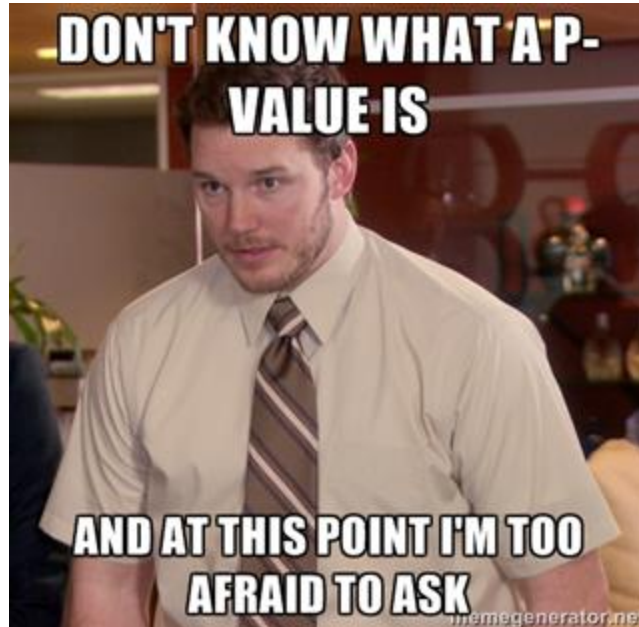
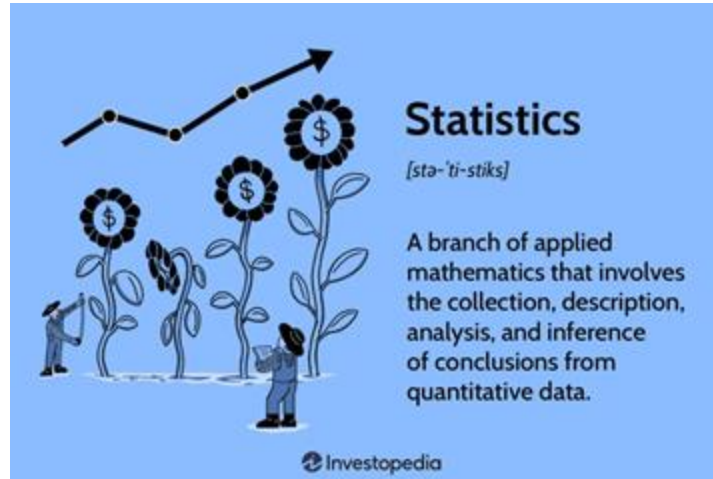


# t-statistics





William Sealy Gosset





If I add the barley to the beer, its taste is going to improve.





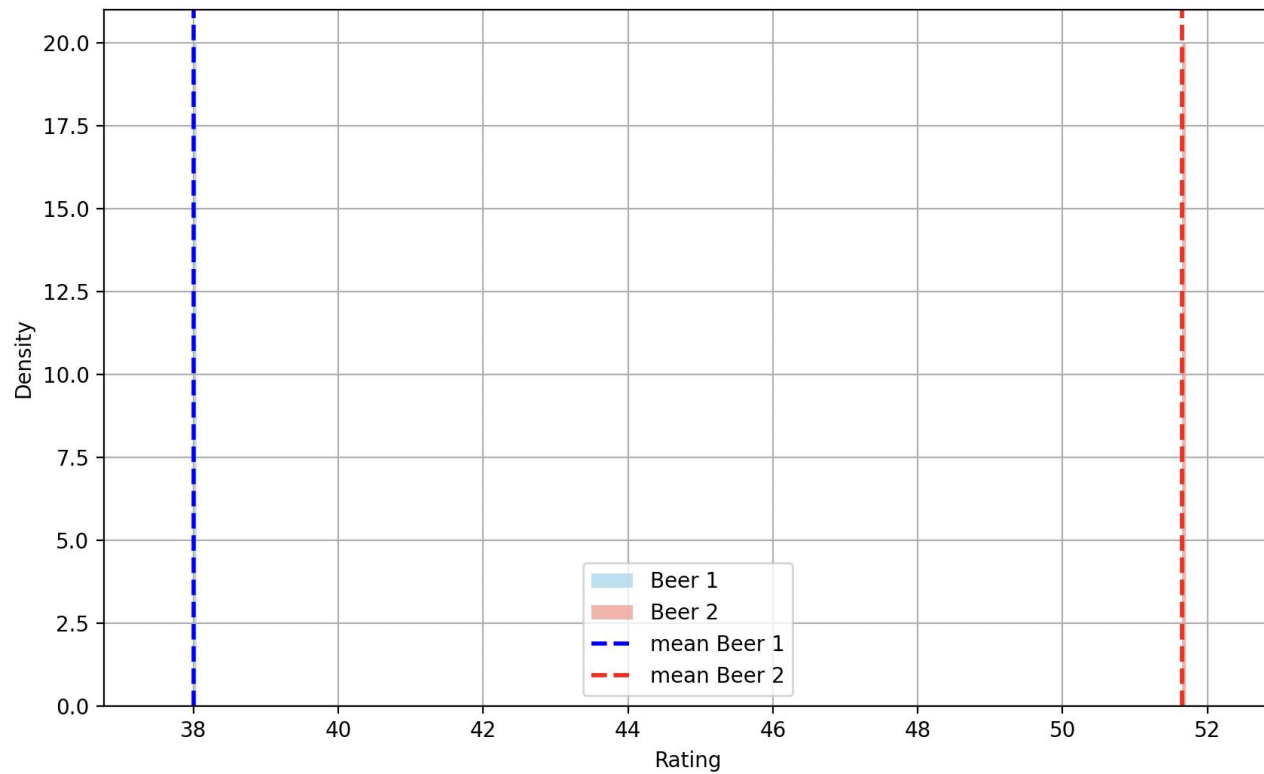
independent variable = CAUSE

If I add the barley to the beer, its taste is going to improve.

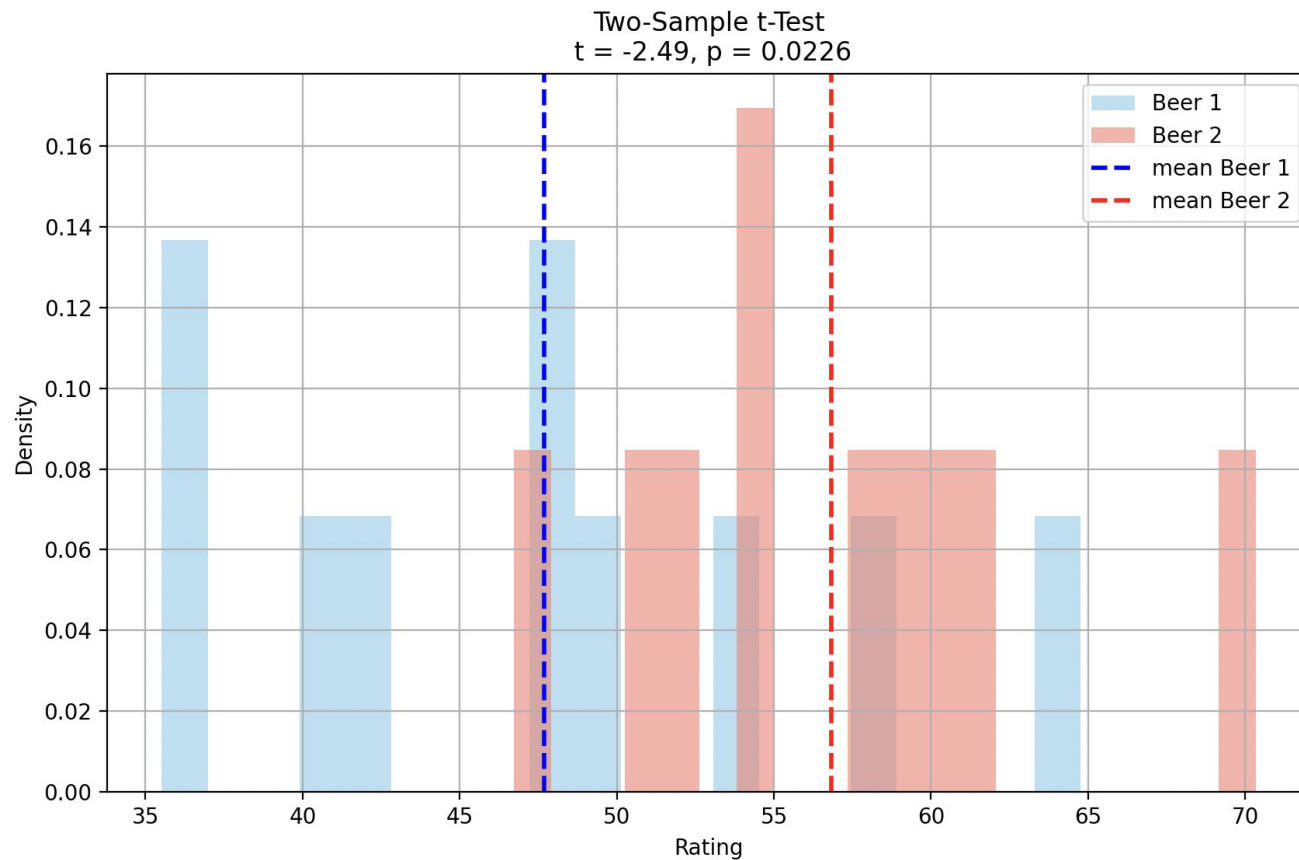


dependent variable = EFFECT

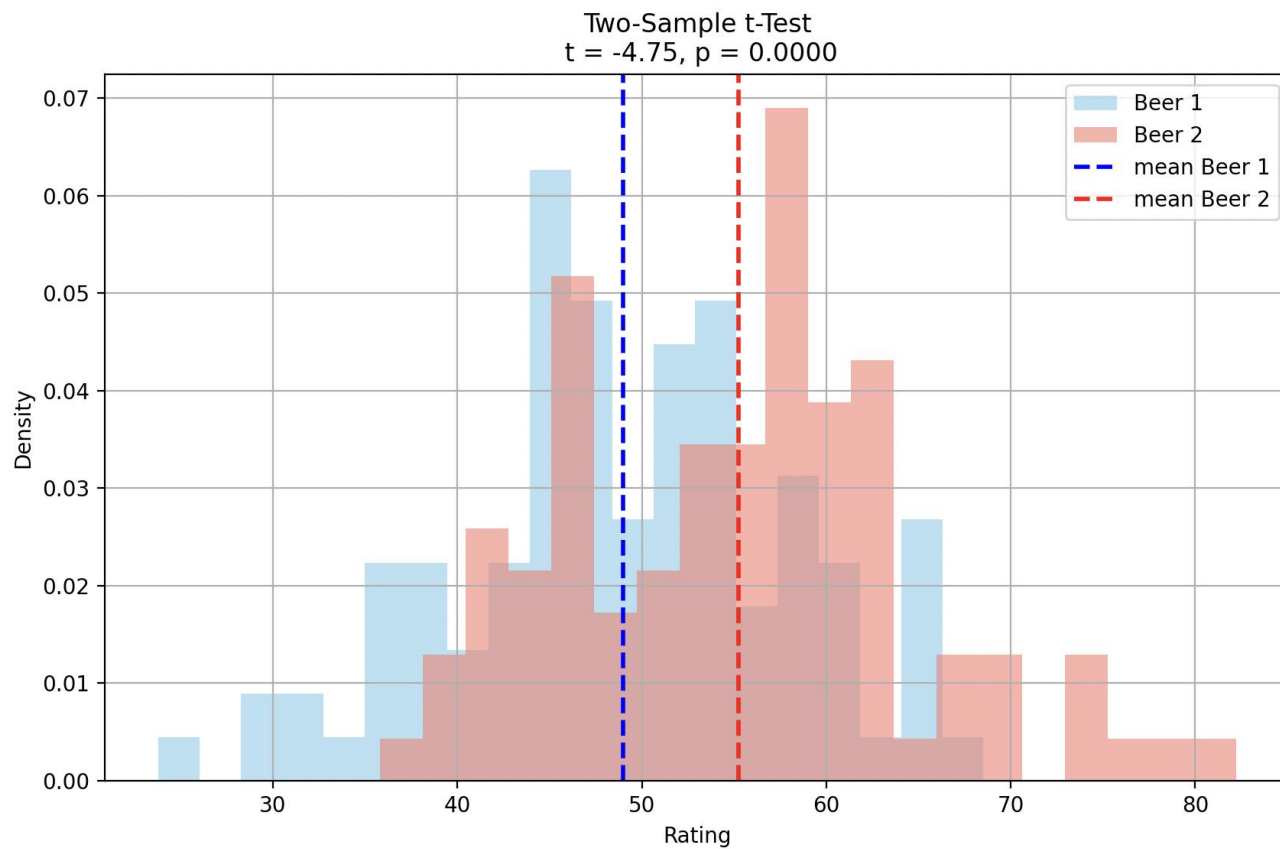
**n = 1**



**n = 10**

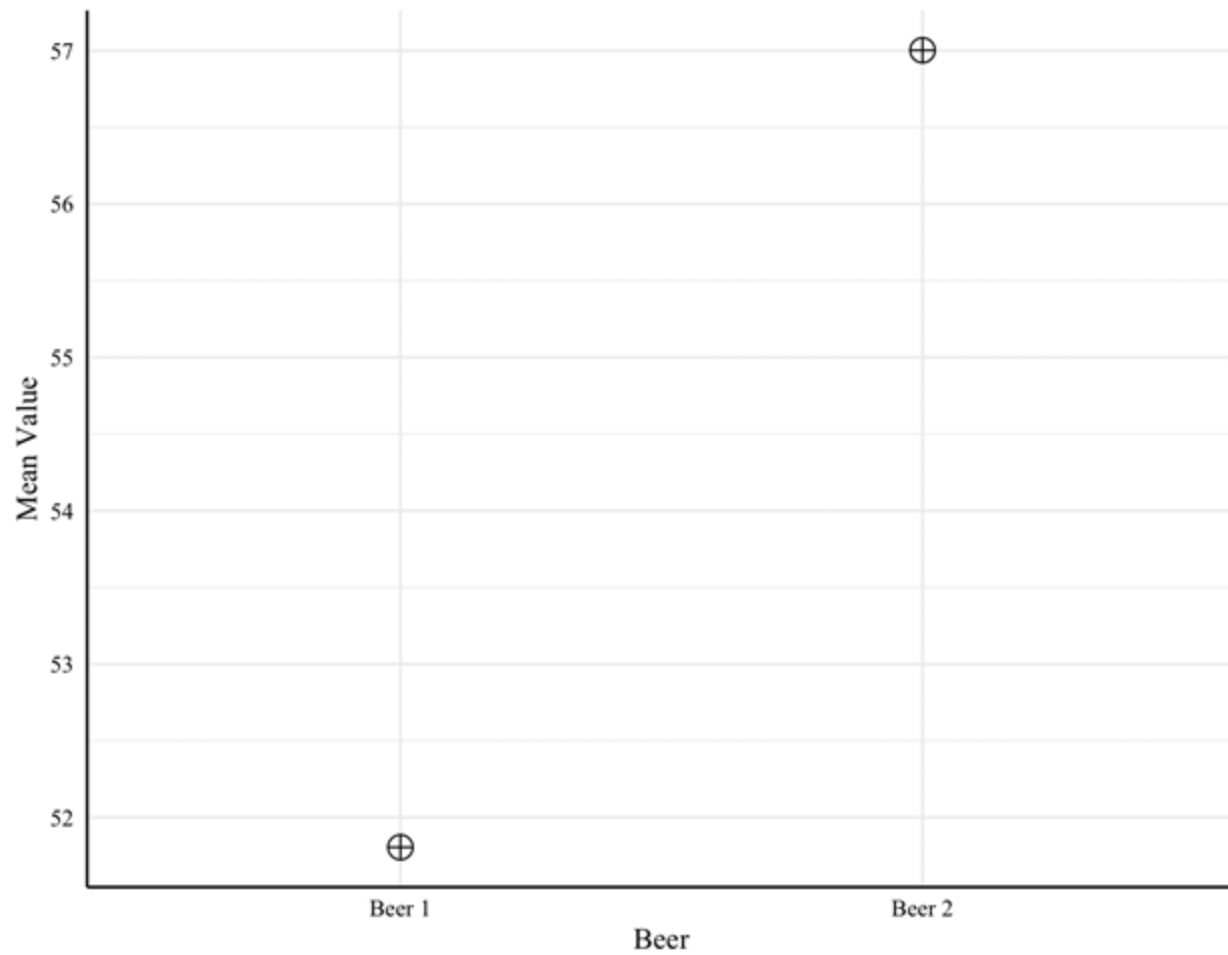


**n = 100**



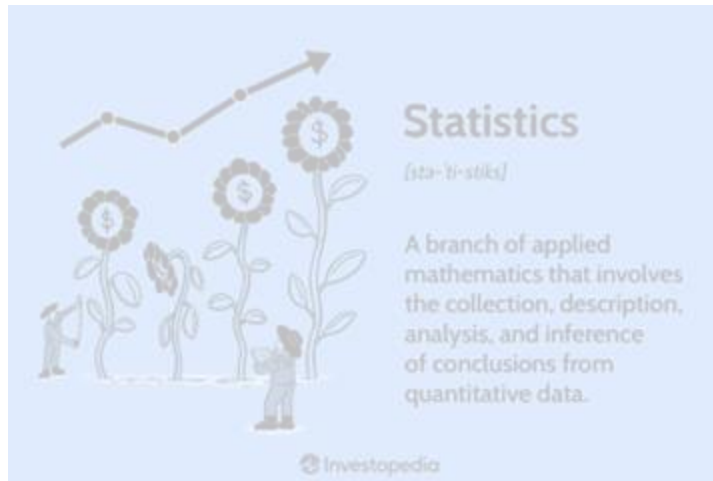


Mean Values of Beer 1 and Beer 2





# Student's t-test



William Sealy Gosset

# What is t-test?

T-test is a statistical test that is used to compare the means of two groups.

## Two-Sample T-Test

$$t = \frac{(\bar{X}_1 - \bar{X}_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$\bar{X}_1$  = observed mean of 1<sup>st</sup> sample

$\bar{X}_2$  = observed mean of 2<sup>nd</sup> sample

$s_1$  = standard deviation of 1<sup>st</sup> sample

$s_2$  = standard deviation of 2<sup>nd</sup> sample

$n_1$  = sample size of 1<sup>st</sup> sample

$n_2$  = sample size of 2<sup>nd</sup> sample

# T-test assumptions

The data is continuous

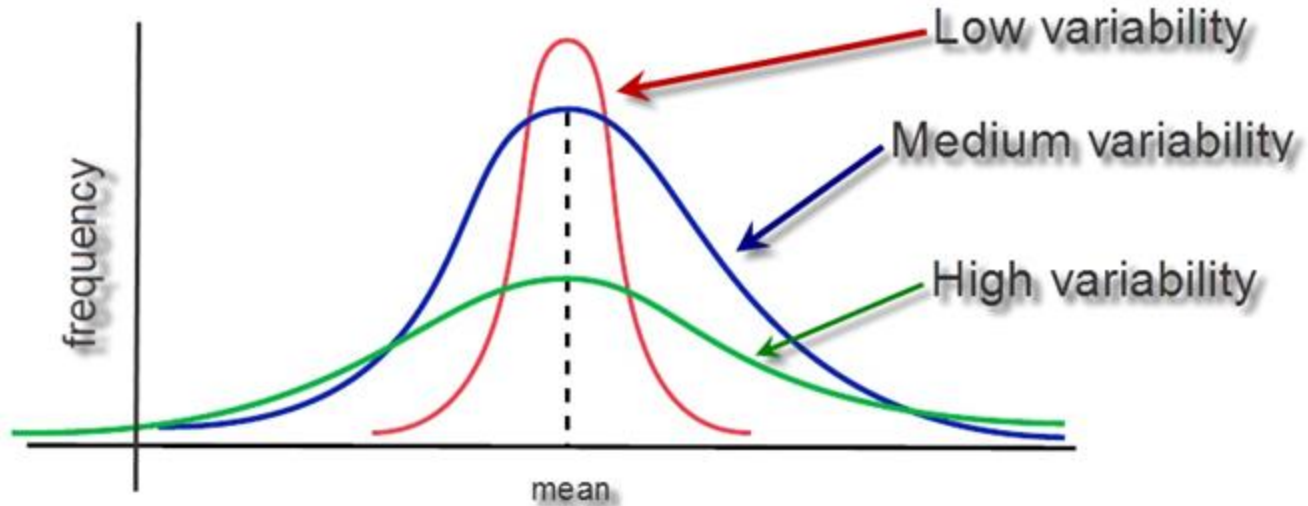
The data is randomly sampled from the population

The variance is homogenous

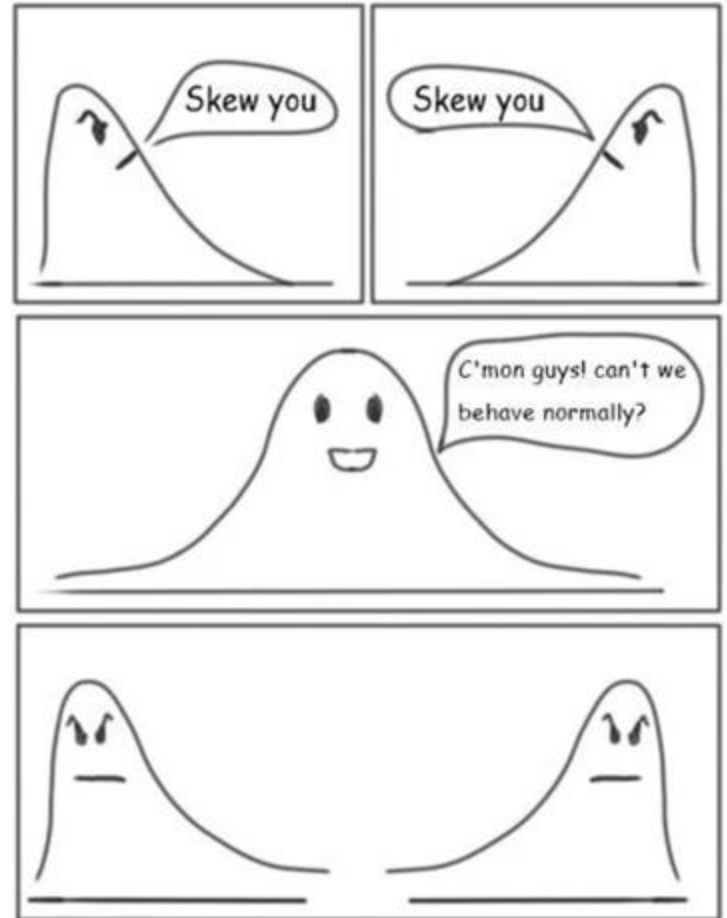
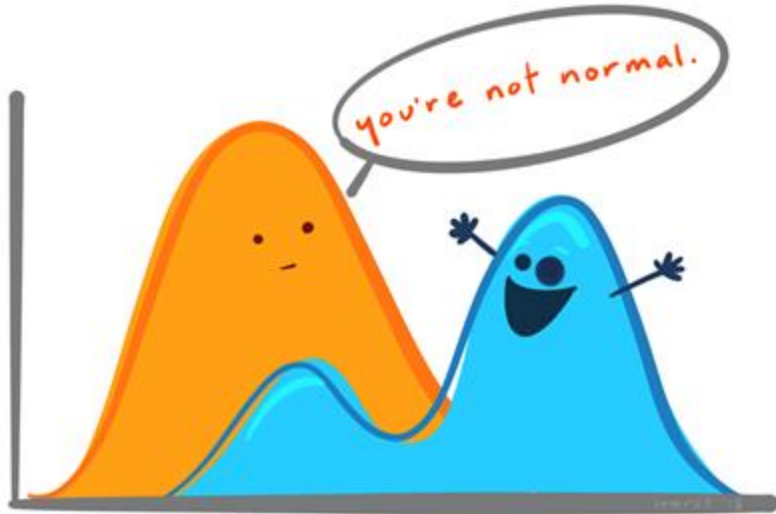
The data distribution is approx. normal

# What is variance?

How spreaded are our data points.



# Normal distribution



# Hypothesis testing

The null hypothesis ( $H_0$ ) is that the true difference between these group means is zero.

The alternative hypothesis ( $H_a$ ) is that the true difference is different from zero.

# Hypothesis testing

The null hypothesis ( $H_0$ ):

$$\mu_1 = \mu_2$$

The alternate hypothesis ( $H_a$ ):

$$\mu_1 \neq \mu_2 \text{ (two-tailed test)}$$

$$\mu_1 > \mu_2 \text{ or } \mu_1 < \mu_2 \text{ (one-tailed test)}$$



# Hypothesis testing

The null hypothesis ( $H_0$ ):

Both beers are rated similarly.

The alternate hypothesis ( $H_a$ ):

There is a difference in rating the taste of beer 1 and beer 2. (two-tailed test)

Beer 1 is better than beer 2. or Beer 2 is better than beer 1. (one-tailed test)

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## The results from our data

$t = -1.7067$ ,  $df = 98$ ,  $p\text{-value} = 0.09106$

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**t = -1.7067**, df = 98, p-value = 0.09106

The t-value measures the difference between the means of the two groups relative to the variance of the data.

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The t-value (-1.7067) indicates the direction of the difference between the means. Since it is negative, it suggests that the mean of group 1 is lower than the mean of group 2.

# The results from our data

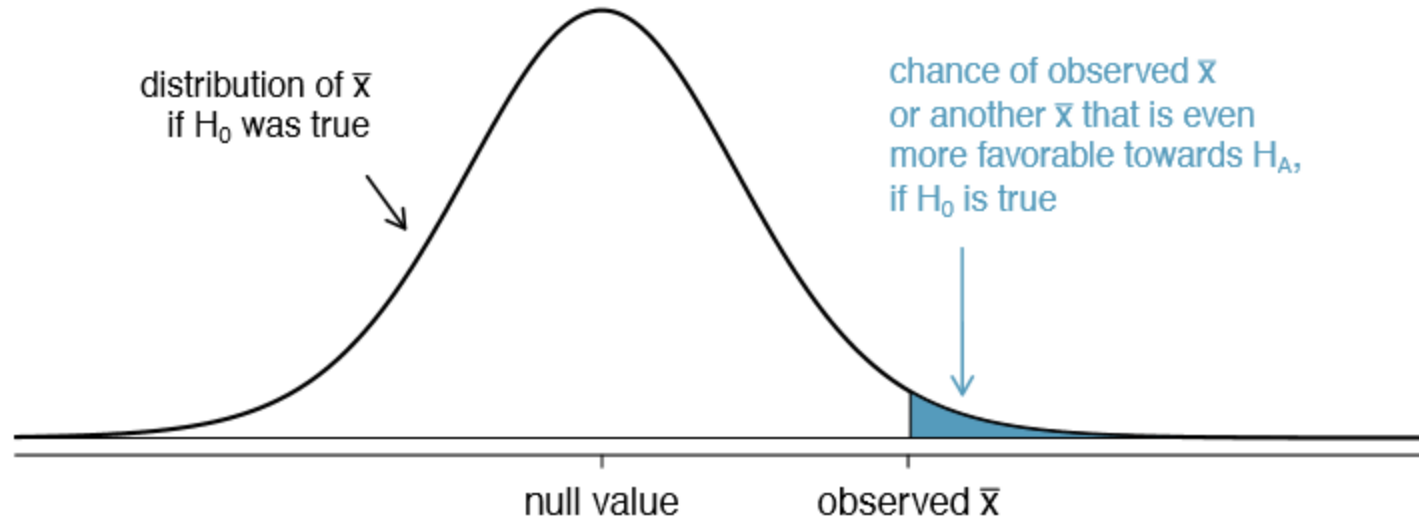
$t = -1.7067$ , **df = 98**,  $p\text{-value} = 0.09106$

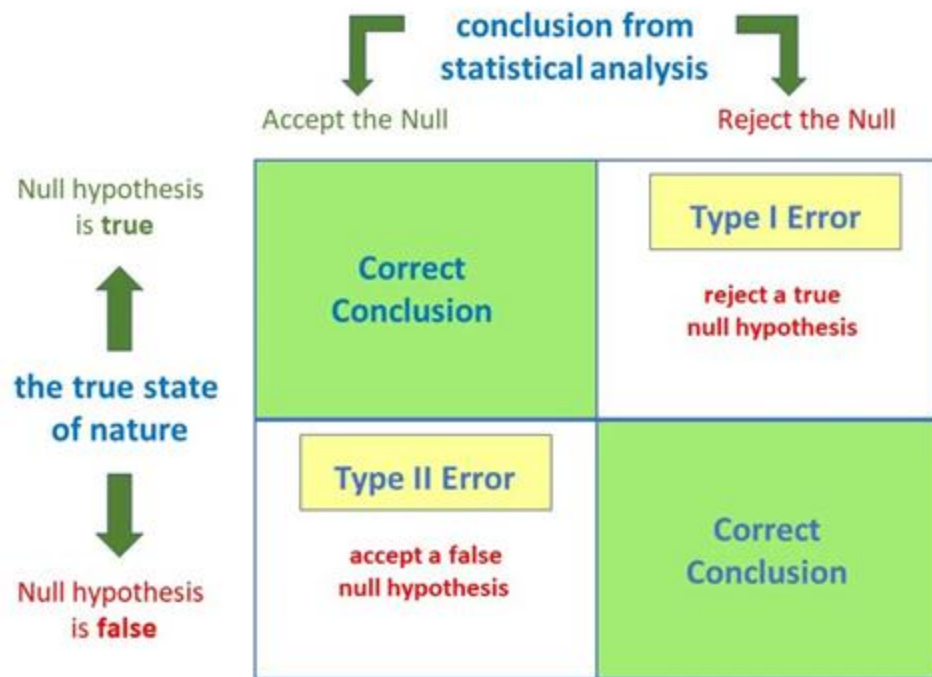
Degrees of freedom are the maximum number of logically independent values, which may vary in a data sample. Degrees of freedom are calculated by subtracting one from the number of items within the data sample.

$$df = n_1 + n_2 - 2$$

# The results from our data

$t = -1.7067$ ,  $df = 98$ , **p-value = 0.09106**





**Type I Error (false-positive)**



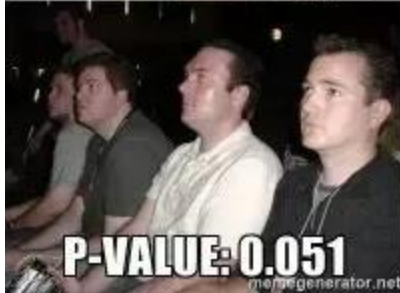
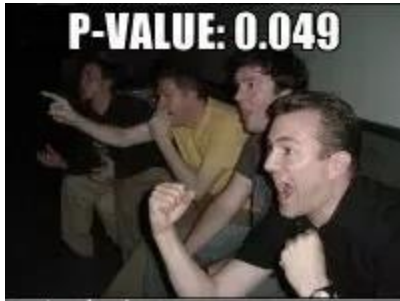
**Type II Error (false-negative)**





# The results from our data

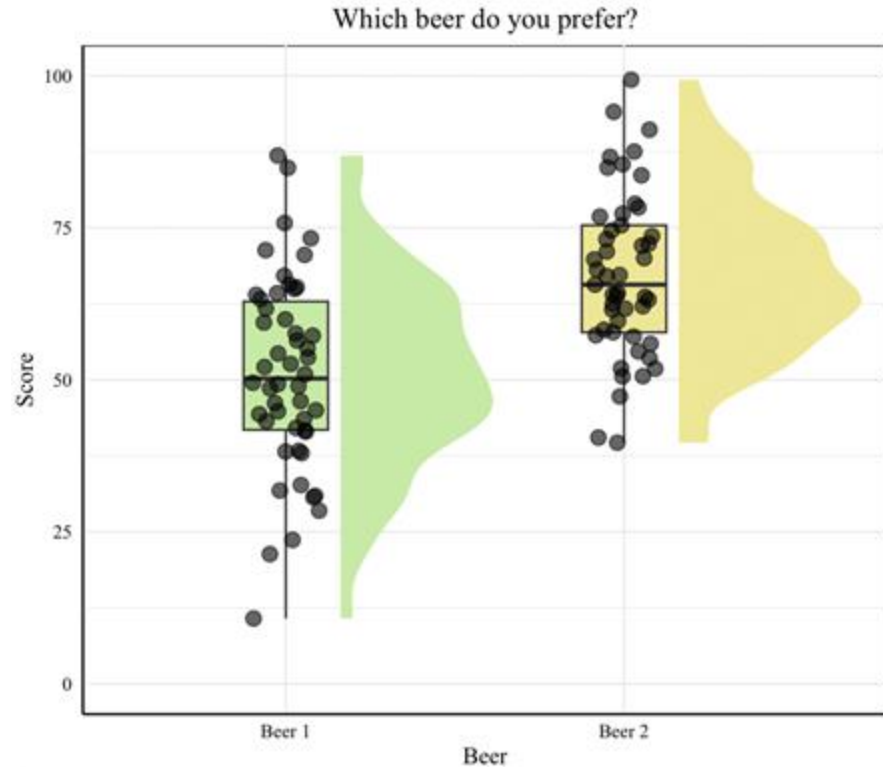
$t = -1.7067$ ,  $df = 98$ , **p-value = 0.09106**



There is no difference between preferences for any of the beer types - we confirm  $H_0$  and reject  $H_a$ .

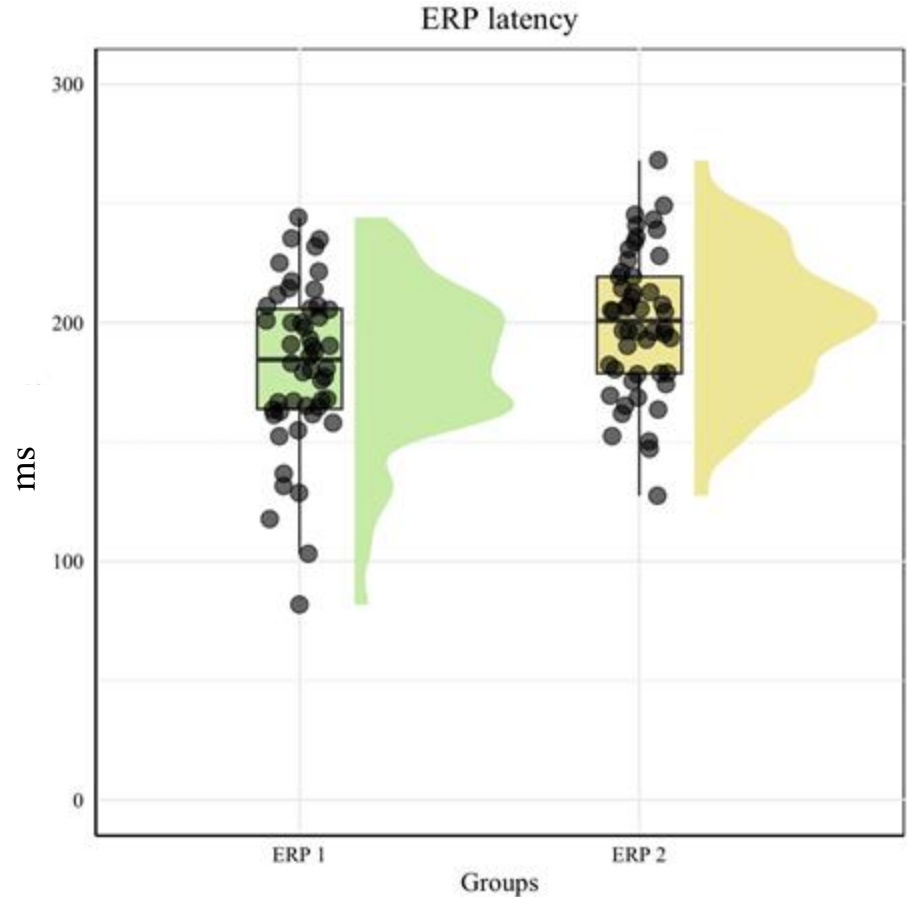
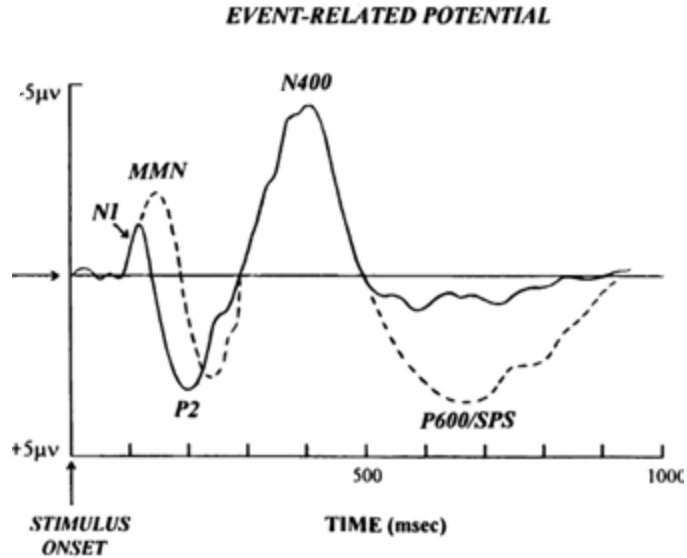
# Exercise - interpret the following results

$t = -5.63$ ,  $df = 98$ ,  $p\text{-value} = 0.0000001752$



# T-test in the EEG world

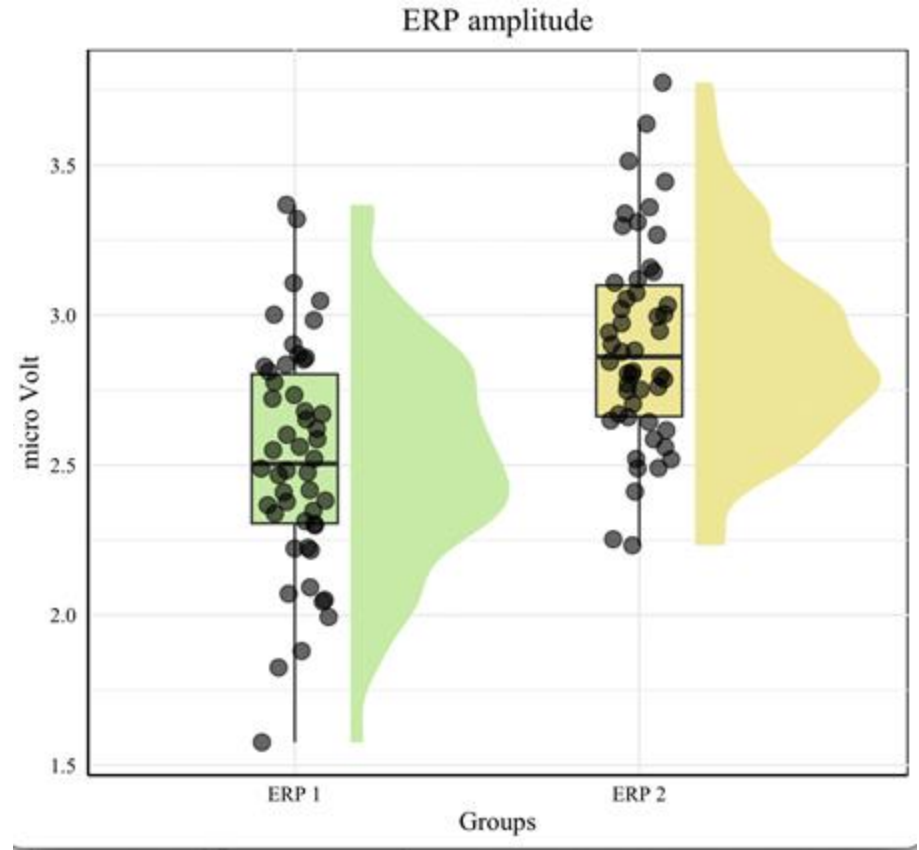
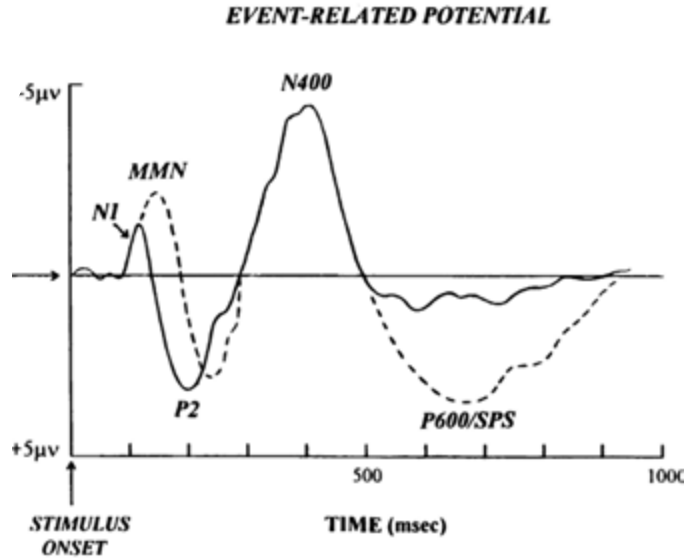
- Latency of ERP peak



$$t = -2.8006, df = 98, p\text{-value} = 0.006166$$

# T-test in the EEG world

- Amplitude of ERP peak



$$t = -4.7139, df = 98, p\text{-value} = 0.000008064$$

# What about EEG?

EEG data often do not fulfill t-test assumptions.

# T-test assumptions

The data is continuous

The data is randomly sampled from the population

The variance is homogenous

**The data distribution is approx. normal**

# What about EEG?

EEG data often do not fulfill t-test assumptions.

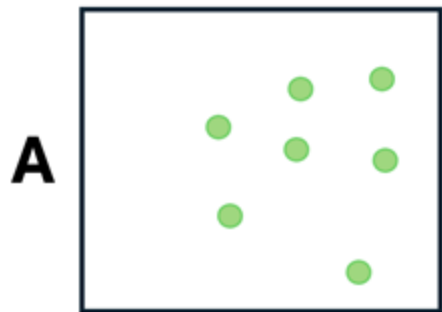
**Surrogate tests** → no assumptions about data distribution.

- Permutation
- Bootstrapping

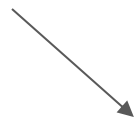
# Bootstrapping

- We assume that the best estimation of the population is our sample
- Recipe:
  - Collect data - there are at least two conditions
  - Check the difference between means
  - Randomly assign data to two groups and check the difference
  - Repeat 1000 times
  - Create a distribution of means
  - Check the probability of the original difference - if it is in the 5% tail - you can assume that the difference is not due to the chance

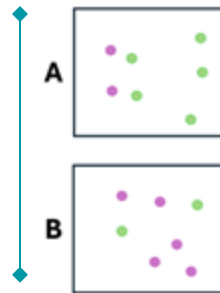
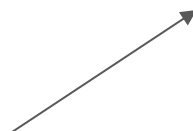




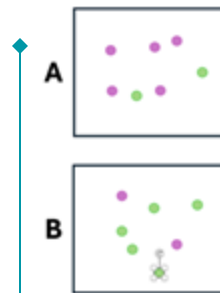
Observed mean



All data points



Difference 1



Difference 2



Difference 3

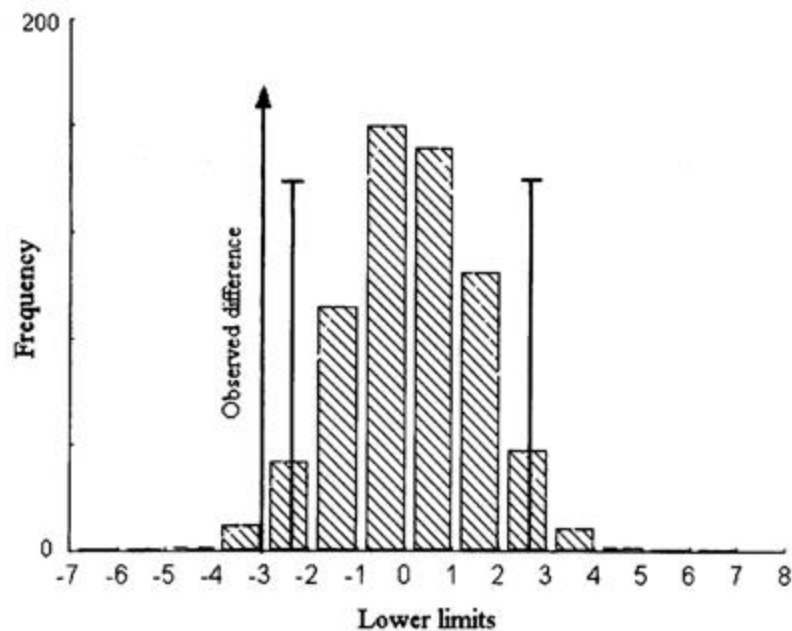


Figure 3. Histogram from 1,000 bootstrap resamplings on the difference between event-related potentials (Cz) to old and new low-imagery words for Subject A. Relative to this distribution, the observed difference (from the experiment) has a probability of less than .05.

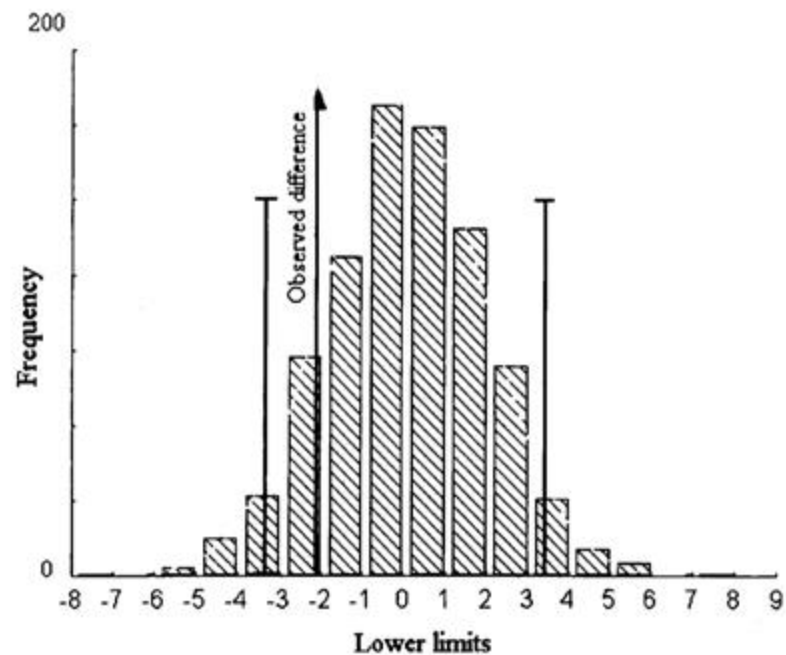


Figure 4. Histogram from 1,000 bootstrap resamplings on the difference between event-related potentials (Pz) to old and new low-imagery words. Relative to this distribution, the observed difference (from the experiment) has a probability of greater than .05.

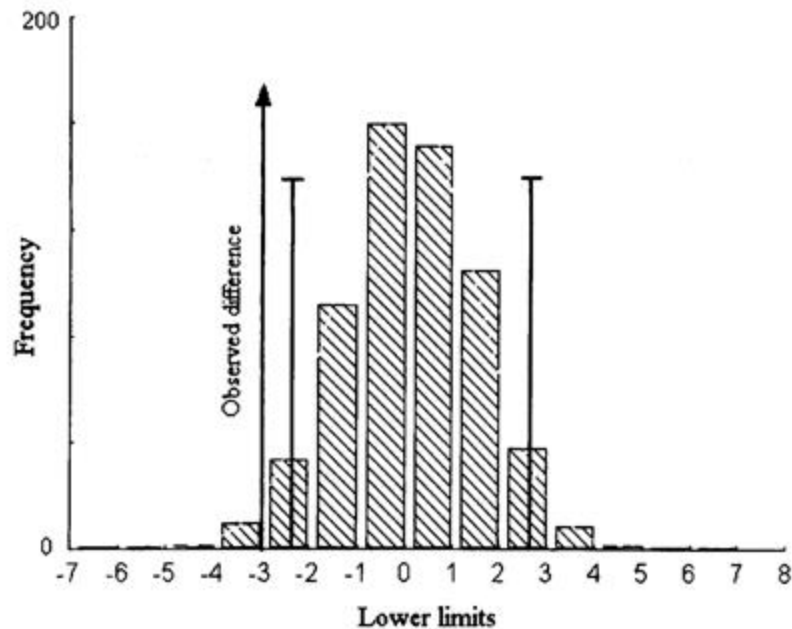


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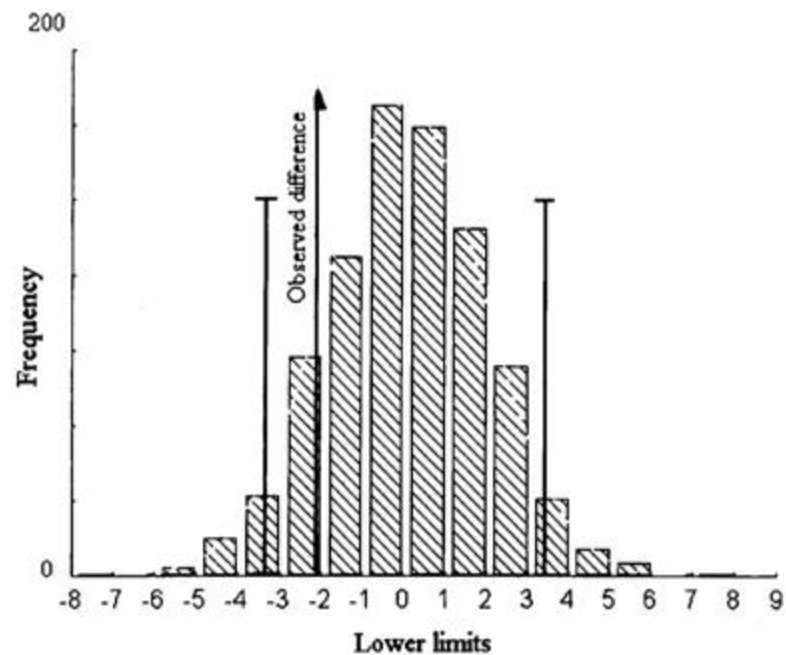


Figure 4. Histogram from 1,000 bootstrap resamplings on the difference between event-related potentials (Pz) to old and new low-imagery words. Relative to this distribution, the observed difference (from the experiment) has a probability of greater than .05.

# What should I use for my EEG data analysis?

Steve Luck says:

*“Use simple  $t$ -test for the exploratory analysis and surrogate tests for publications.”*

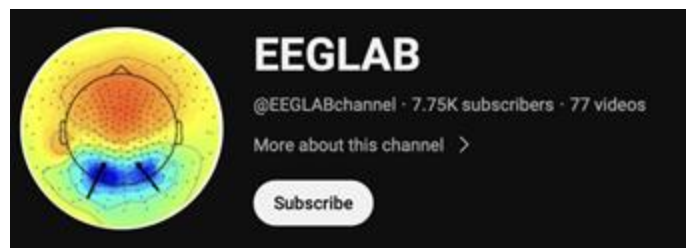
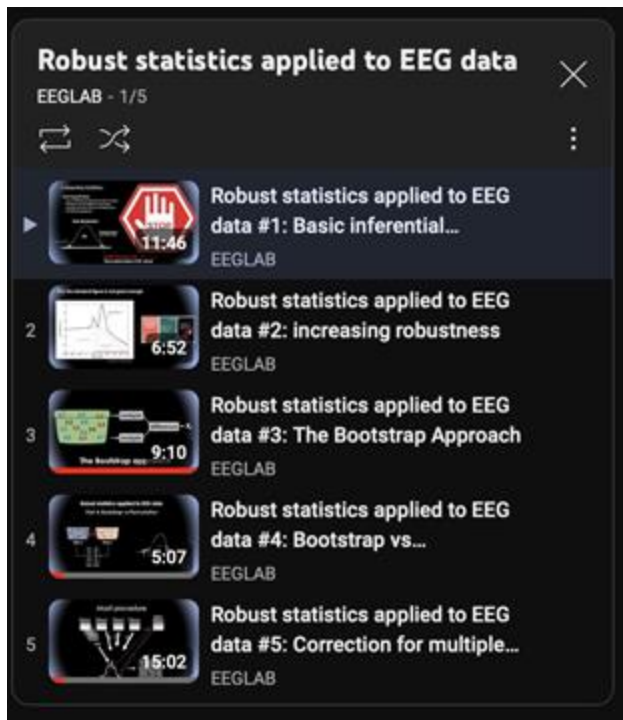


# Cluster-based Permutation Testing

Many data points → high chance that we will find something significant

1. **Identify clusters**
2. Compute the size of the cluster
3. Randomly assign data to clusters
4. Check what is the most probable size of a cluster

If you want to know more...



<https://www.youtube.com/@EEGLABchannel>