

**P-VALUES, NULL HYPOTHESIS, TYPE I & TYPE II ERRORS**

**GROUP 4**

# INTRODUCTION

- Hypothesis is an assumption about a population that can be tested with a statistical method.
- It serves as the foundation for scientific research and statistical testing
- Formulated based on prior observations, or theories and are evaluated through experiments.
- **Key Characteristics of a Hypothesis**
  - **Testability:** Allows it to be tested using data or experiments.
  - **Falsifiability:** It must be possible to prove the hypothesis false.
  - **Specificity:** Defines the variables and the expected relationship or difference.

# Why formulate a hypothesis?

- Data-driven decision making:
  - Provides a systematic approach to analyze data and make informed decisions based on statistical evidence rather than personal opinions or biases.
- Testing theories and assumptions:
  - Researchers can test their hypotheses against empirical data to determine whether their initial ideas are supported or need to be revised.
- Identifying significant relationships
  - Hypothesis testing reveal whether observed relationships between variables are statistically significant, allowing for meaningful interpretations of the results.

# The Null and Alternative Hypotheses

- **Null Hypothesis ( $H_0$ )**

- A statement of "no effect" : There is no relationship between A and B

- **Alternative Hypothesis ( $H_1$ ):**

- A statement that challenges  $H_0$ . Posits that there is some difference between A and B

- **Key Points**

- $H_0$  and  $H_1$  are **mutually exclusive**: meaning that only one of them can be true at any given time.
- Hypothesis testing assumes  $H_0$  is true until evidence suggests rejecting it.
- Failing to reject  $H_0$  does not prove  $H_0$  is true; it means there's insufficient evidence against it.

# P-VALUE

- The p-value is the probability of obtaining the observed results or extreme ones if the null hypothesis is true.
- **Explanation:**
  - A small p-value (e.g.):  **$p < 0.05$** : Evidence against  $H_0$  : Reject  $H_0$ .
  - A large p-value : Insufficient evidence : Fail to reject  $H_0$ .
- **Misconceptions:**
  - The p-value **does not** measure the probability that  $H_0$  is True or False.
    - Explain the likelihood of observing the data
  - It also **does not** indicate the strength or size of an effect.
    - Indicates a statistical difference but not the magnitude of this effect. Use Cohen's d test

# Setting The Significance Level (Alpha, $\alpha$ )

- **Significance Level ( $\alpha$ )**
  - The threshold used to determine whether to reject the null hypothesis ( $H_0$ ).
  - It is set before conducting the analysis.
  - Represents the probability of making a **Type I error** (rejecting  $H_0$  when it is actually true).
- **Relationship to the P-Value:**
  - Compare the p-value to  $\alpha$ :
    - **If p-value <  $\alpha$ :** Reject  $H_0$  (result is statistically significant).
    - **If p-value  $\geq \alpha$ :** Fail to reject  $H_0$  (result is not statistically significant).

# Confidence Level and Confidence Interval

## Confidence level (C. L)

- The probability that the true population parameter lies within the confidence interval (e.g., 95%) in repeated studies
- Commonly expressed as a percentage, such as 90%, 95%, or 99%.
  - A 95% confidence level means we expect the true population parameter to fall within the range 95% of the time across repeated samples.
- Relationship with significant level
  - Confidence level =  $1 - \alpha$
  - If you use an alpha value of  $p < 0.05$  for statistical significance, then the confidence level would be  $1 - 0.05 = 0.95$ , or 95%.

## Confidence interval (C.I)

- Provides an estimated range, based on sample data, where the true population parameter is likely to fall.
- Typically written as a range: [lower bound, upper bound]
  - A 95% CI of [10,20] means the true parameter is likely to fall between 10 and 20, with 95% confidence.

# Confidence Level and Significance Level: Inversely Proportional

- Inversely Proportional:
  - As you increase the **confidence level** (e.g., from 95% to 99%), you decrease the **significance level** ( $\alpha$ ), which reduces the likelihood of making a **Type I error** (rejecting  $H_0$  when it is true).
  - However, increasing the confidence level can increase the likelihood of making a **Type II error** (failing to reject  $H_0$  when it is false).



# What Are Errors in Hypothesis Testing?

- Hypothesis testing involves making decisions based on sample data to draw conclusions about a population.
- These decisions are not always correct, and errors can occur when conclusions about the null hypothesis ( $H_0$ ) are drawn incorrectly.
- There are two types of errors in hypothesis testing: Type I and Type II.

## Type I Error (False Positive)

- Rejecting the null hypothesis when it is actually true.
- We conclude that there is an effect or difference when, in reality, there isn't.
- **Example:** A drug test shows effectiveness when the drug has no effect.
- **Significance Level ( $\alpha$ ):** Type I error is directly related to the significance level;  $\alpha$  represents the probability of making a Type I error.

## Type II Error (False Negative)

- Failing to reject the null hypothesis when it is actually false.
- We conclude that there is no effect or difference when there actually is.
- **Example:** A drug test shows no effect when the drug actually works.
- **Power ( $1-\beta$ ):** The probability of correctly rejecting the null hypothesis (avoiding a Type II error).

# Type I and Type II Error Table

## Hypothesis testing:

		Decision	
		$H_0$ true (Fail to reject)	$H_0$ false (Rejecting $H_0$ )
Actual	$H_0$ true	<b>TRUE NEGATIVE</b>  Correct decision: <b>Confidence level</b> (prob $1 - \alpha$ )	<b>FALSE POSITIVE</b>  <b>Type I Error:</b> <b>Significance level/Size (<math>\alpha</math>)</b> (prob $\alpha$ )
	$H_0$ false	<b>FALSE NEGATIVE</b>  <b>Type II Error:</b> fail to reject (prob $\beta$ )	<b>TRUE POSITIVE</b>  Correct decision: <b>Power</b> (prob $1 - \beta$ )

# Balancing Type I and Type II Errors

- Understanding the Tradeoff

## Disease prediction:

- **Type I Error:** Incorrectly diagnosing a healthy patient with a disease (false positive).
- **Type II Error:** Failing to diagnose a patient who actually has the disease (false negative).
- **Balance:** In situations where the cost of incorrectly diagnosing a healthy patient (Type I error) is high, a lower  $\alpha$  might be used. However, if the consequences of missing a disease diagnosis (Type II error) are severe, a higher confidence level (95%) might be chosen.

## Court Case:

- **Type I Error:** Convicting an innocent person (wrongful punishment).
- **Type II Error:** Letting a guilty person go free (failure to protect society).
- **Balance:** In a criminal trial, a higher confidence level may be required (e.g., 99% or more) to avoid convicting innocent people, though this may increase the risk of letting some guilty individuals go free.