

# Hypothesis Testing and The Null Hypothesis

Statistics and Big Data

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Academic Year 2025-2026

Course: Statistics and Big Data

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# What is Hypothesis Testing?

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Imagine we are testing two different drugs to see which one helps patients recover from a virus more effectively. How do we determine if one drug is truly better than the other? This leads us to the concept of hypothesis testing.

# Concrete Example of Drug Testing

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## Drug A Recovery Times

- Patient 1: 10 hours
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- Patient 3: 12 hours

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- Patient 5: 18 hours
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What do these recovery times suggest about the effectiveness of the drugs?

# Observing Differences

From our example, we observe that:

$$\bar{x}_A = \frac{10 + 15 + 12}{3} = 12.33 \text{ hours,}$$

$$\bar{x}_B = \frac{20 + 18 + 22}{3} = 20 \text{ hours.}$$

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This indicates that patients taking Drug A recover faster on average. But can we conclude that Drug A is definitively better than Drug B?

# Formulating a Hypothesis

Based on our preliminary data, we might hypothesize:

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# The Role of Random Variation

The differences in recovery times could be attributed to uncontrolled variables, such as:

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This variability raises an important question: How can we be sure of our hypothesis?

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# Rejecting the Hypothesis

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## Key Understanding

If data consistently contradicts our hypothesis, we must reject it.

# Introducing New Drugs

Now, let's consider two new drugs, C and D. We hypothesize that:

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Patients taking Drug C recover, on average, 13 hours faster than those taking Drug D.

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What happens when we test this hypothesis?

# Testing the New Hypothesis

Upon testing, we find:

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In this case, we cannot confidently reject the hypothesis that Drug C is better than Drug D. Instead, we "fail to reject" the hypothesis, indicating that while the evidence is not strong enough to confirm it, it is also not disproven.

# Understanding the Null Hypothesis

To clarify our testing process, we introduce the **Null Hypothesis ( $H_0$ )**:

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Why is this important?

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The null hypothesis allows us to focus on whether any observed differences are statistically significant. If we find a small difference, such as 0.5 hours, we can assess whether this difference is due to random variation or if it is significant enough to reject  $H_0$ .

# Summary of Key Concepts

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- ➋ We can reject a hypothesis if data consistently contradicts it.
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What are the implications of these concepts in real-world applications?

## Exercise 1

Define the null hypothesis in your own words and explain its significance in hypothesis testing.

# Exercises

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## Exercise 2

Given the following recovery times for two drugs, calculate the mean recovery time for each drug and determine if there is a significant difference:

- Drug A: 10, 12, 14 hours
- Drug B: 15, 18, 20 hours

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