



Hypothesis Testing Fundamentals

Basic Concepts

<p><b>Hypothesis:</b> A statement about a population parameter.</p>
<p><b>Null Hypothesis (<math>H_0</math>):</b> A statement of no effect or no difference; it is assumed to be true until evidence indicates otherwise.</p>
<p><b>Alternative Hypothesis (<math>H_1</math> or <math>H_a</math>):</b> A statement that contradicts the null hypothesis; it represents what we are trying to find evidence for.</p>
<p><b>Types of Tests:</b></p> <ul style="list-style-type: none"><li>• <b>One-Tailed (Right-Tailed):</b> <math>H_1: \mu &gt; \text{value}</math></li><li>• <b>One-Tailed (Left-Tailed):</b> <math>H_1: \mu &lt; \text{value}</math></li><li>• <b>Two-Tailed:</b> <math>H_1: \mu \neq \text{value}</math></li></ul>
<p><b>Type I Error (<math>\alpha</math>):</b> Rejecting <math>H_0</math> when it is actually true (False Positive).</p>
<p><b>Type II Error (<math>\beta</math>):</b> Failing to reject <math>H_0</math> when it is actually false (False Negative).</p>
<p><b>Significance Level (<math>\alpha</math>):</b> The probability of making a Type I error.</p>
<p><b>Power (<math>1-\beta</math>):</b> The probability of correctly rejecting <math>H_0</math> when it is false.</p>

Steps in Hypothesis Testing

<p>1. <b>State the Hypotheses:</b> Define <math>H_0</math> and <math>H_1</math>.</p> <p>2. <b>Determine the Test Statistic:</b> Choose the appropriate test statistic (z, t, F, etc.).</p> <p>3. <b>Set the Significance Level:</b> Determine <math>\alpha</math> (e.g., 0.05, 0.01).</p> <p>4. <b>Calculate the Test Statistic:</b> Compute the value of the test statistic from the sample data.</p> <p>5. <b>Determine the p-value or Critical Value:</b></p> <ul style="list-style-type: none"><li>• <b>p-value:</b> The probability of observing a test statistic as extreme as, or more extreme than, the one computed if <math>H_0</math> is true.</li><li>• <b>Critical Value:</b> The value(s) that define the rejection region.</li></ul> <p>6. <b>Make a Decision:</b></p> <ul style="list-style-type: none"><li>• <b>p-value Method:</b> If <math>p\text{-value} \leq \alpha</math>, reject <math>H_0</math>.</li><li>• <b>Critical Value Method:</b> If the test statistic falls in the rejection region, reject <math>H_0</math>.</li></ul> <p>7. <b>State the Conclusion:</b> Interpret the decision in the context of the problem.</p>
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Z-Test vs T-Test

<p><b>Z-Test</b></p> <ul style="list-style-type: none"><li>• Population standard deviation (<math>\sigma</math>) is known.</li><li>• Sample size is large (<math>n \geq 30</math>).</li><li>• Used for testing hypotheses about a single mean or comparing two means.</li></ul> <p><b>Test Statistic:</b></p> <p><math display="block">z = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}</math></p>	<p><b>T-Test</b></p> <ul style="list-style-type: none"><li>• Population standard deviation (<math>\sigma</math>) is unknown.</li><li>• Sample size is small (<math>n &lt; 30</math>).</li><li>• Used for testing hypotheses about a single mean or comparing two means.</li></ul> <p><b>Test Statistic:</b></p> <p><math display="block">t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}}</math></p>
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## Two-Sample Tests

### Two-Sample Z-Test for Means

**Purpose:** To test if there is a significant difference between the means of two independent populations when the population standard deviations are known or sample sizes are large.

**Hypotheses:**

- $H_0: \mu_1 = \mu_2$
- $H_1: \mu_1 \neq \mu_2$  (two-tailed)
- $H_1: \mu_1 > \mu_2$  (right-tailed)
- $H_1: \mu_1 < \mu_2$  (left-tailed)

**Test Statistic:**

$$z = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Where:

- $\bar{x}_1, \bar{x}_2$  are the sample means.
- $\mu_1, \mu_2$  are the population means.
- $\sigma_1, \sigma_2$  are the population standard deviations.
- $n_1, n_2$  are the sample sizes.

**Example:** Testing the difference in weight loss between a low-carb diet ( $n_1 = 80, \bar{x}_1 = 13.5$  lbs,  $s_1 = 6.59$  lbs) and a low-fat diet ( $n_2 = 76, \bar{x}_2 = 15.1$  lbs,  $s_2 = 6.38$  lbs) at  $\alpha = 0.01$ .

- $H_0: \mu_1 = \mu_2$
- $H_1: \mu_1 \neq \mu_2$

$$z = \frac{(13.5 - 15.1) - 0}{\sqrt{\frac{6.59^2}{80} + \frac{6.38^2}{76}}} \approx -1.54$$

Since  $-2.58 < -1.54 < 2.58$ , fail to reject  $H_0$ . There is not sufficient evidence to conclude that the mean weight loss differs.

### Paired Sample T-Test

**Purpose:** To test if there is a significant difference between two related populations (e.g., before and after treatment).

**Hypotheses:**

- $H_0: \mu_d = 0$
- $H_1: \mu_d \neq 0$  (two-tailed)
- $H_1: \mu_d > 0$  (right-tailed)
- $H_1: \mu_d < 0$  (left-tailed)

**Test Statistic:**

$$t = \frac{\bar{d} - \mu_d}{\frac{s_d}{\sqrt{n}}}$$

Where:

- $\bar{d}$  is the mean of the differences.
- $\mu_d$  is the hypothesized mean difference (usually 0).
- $s_d$  is the standard deviation of the differences.
- $n$  is the number of pairs.

**Example:** Testing if runners ran faster after eating spaghetti ( $n = 6, \bar{d} = 3.0$ ,  $sd = 3.742$ ) at  $\alpha = 0.05$ .

- $H_0: \mu_d = 0$
- $H_1: \mu_d > 0$

$$t = \frac{3.0 - 0}{\frac{3.742}{\sqrt{6}}} \approx 1.963$$

Since  $1.963 < 2.015$ , fail to reject  $H_0$ . There is not sufficient evidence to support the claim that runners run faster after eating spaghetti.

# Correlation and Regression

## Correlation

**Purpose:** To measure the strength and direction of the linear relationship between two variables.

**Pearson Correlation Coefficient (r):**

$$r = \frac{n(\sum xy) - (\sum x)(\sum y)}{\sqrt{[n(\sum x^2) - (\sum x)^2][n(\sum y^2) - (\sum y)^2]}}$$

**Interpretation:**

- 1 ≤ r ≤ 1
- r > 0: Positive correlation
- r < 0: Negative correlation
- r ≈ 0: No correlation
- |r| close to 1: Strong correlation

**Example:** Calculating correlation between non-member price (x) and member price (y) for n = 8 pairs.

r ≈ 0.857 indicates a strong positive correlation. As non-member price increases, member price tends to increase as well.

## Linear Regression

**Purpose:** To model the relationship between two variables and predict the value of one variable based on the other.

**Regression Line Equation:**

$$\hat{y} = a + bx$$

Where:

- $\hat{y}$  is the predicted value of the dependent variable.
- x is the independent variable.
- a is the y-intercept.
- b is the slope.

**Formulas for Slope (b) and Y-Intercept (a):**

$$b = \frac{n(\sum xy) - (\sum x)(\sum y)}{n(\sum x^2) - (\sum x)^2}$$
$$a = \bar{y} - b\bar{x}$$

**Example:** Predicting number of mistakes (y) based on hours without sleep (x).

$$\hat{y} = -2.769 + 0.288x$$

For x = 40 hours:

$$\hat{y} = -2.769 + 0.288(40) \approx 8.751 \approx 9 \text{ mistakes.}$$

## Scatter Plots

**Visual Representation:** A scatter plot is a graph that displays the relationship between two variables.

**Types of Relationships:**

- Positive Linear:** As x increases, y increases.
- Negative Linear:** As x increases, y decreases.
- No Relationship:** Points are scattered randomly.
- Nonlinear:** Points follow a curved pattern.

**Example:** Plotting age (x) vs. income (y) data points on a scatter plot can reveal a positive linear relationship, indicating that income tends to increase with age.

# Analysis of Variance (ANOVA)

## ANOVA Fundamentals

**Purpose:** To test if there is a significant difference between the means of three or more independent groups.

**Hypotheses:**

- $H_0: \mu_1 = \mu_2 = \mu_3 = \dots = \mu_k$  (all means are equal)
- $H_1$ : At least one mean is different

**Test Statistic:**  $F = MS_{\text{between}} / MS_{\text{within}}$

Where:

- $MS_{\text{between}}$  is the mean square between groups.
- $MS_{\text{within}}$  is the mean square within groups.

**Degrees of Freedom:**

- $df_{\text{Numerator}} = k - 1$  (k is the number of groups)
- $df_{\text{Denominator}} = N - k$  (N is the total number of observations)

**F-Distribution:**

- Always right-tailed.
- Critical value is found using  $\alpha$ ,  $df_{\text{Numerator}}$ , and  $df_{\text{Denominator}}$ .

**Sums of Squares:**

- $SSTotal = \sum x^2 - (\sum x)^2 / N$
- $SS_{\text{Between}} = \sum n_i (\bar{x}_i - \bar{XGM})^2$
- $SS_{\text{Within}} = SSTotal - SS_{\text{Between}}$

## ANOVA Calculation Steps

1. **Calculate the Group Sums, Means, and Total Mean.**
2. **Calculate the Sum of Squares Total ( $SSTotal$ ).**
3. **Calculate the Sum of Squares Between ( $SS_{\text{Between}}$ ).**
4. **Calculate the Sum of Squares Within ( $SS_{\text{Within}}$ ).**
5. **Calculate the Mean Squares:**
  - $MS_{\text{Between}} = SS_{\text{Between}} / df_{\text{Numerator}}$
  - $MS_{\text{Within}} = SS_{\text{Within}} / df_{\text{Denominator}}$
6. **Calculate the F statistic:**  $F = MS_{\text{Between}} / MS_{\text{Within}}$
7. **Find the Critical Value:** Using  $\alpha$ ,  $df_{\text{Numerator}}$ , and  $df_{\text{Denominator}}$ .
8. **Make a Decision:** If  $F > F_{\text{crit}}$ , reject  $H_0$ .
9. **State the Conclusion.**

## Example

**Example:** Testing if average number of students in an English course differs by time of day (Morning, Afternoon, Evening) at  $\alpha = 0.05$ .

$F \approx 4.776$ ,  $F_{\text{crit}} = 3.89$ . Since  $4.776 > 3.89$ , reject  $H_0$ . There is sufficient evidence to conclude that mean attendance differs by time of day.