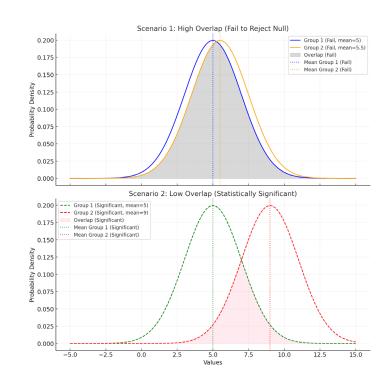


# **T-Test**

#### **Key Steps**

- 1. Calculate the means
- 2. Calculate the standard error
- 3. Compute the degrees of freedom
- 4. Determine the p-value





## **T-Test: Mathematical Formulation**

T-statistic Calculation

$$t = \frac{\overline{x_1} - \overline{x_2}}{SE}$$

#### Where:

- $\overline{x_i}$ : Mean of sample i
- SE: Standard Error of the Difference

- SE Calculation depends on observed variances
- If samples have equal variances:
  Pooled T-Test
- If not, Welch's T-Test

## **T-Test: Mathematical Formulation**

T-statistic Calculation

$$t = \frac{\overline{x_1} - \overline{x_2}}{SE}$$

#### Where:

- $\bar{x_i}$ : Mean of sample i
- SE: Standard Error of the Difference

Pooled T-Test (Equal Variances)

$$SE = \sqrt{s_p^2 \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}$$

Welch's T-Test (Unequal Variances)

$$SE = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

#### Where:

 $s_p$ : Standard deviation of pooled data

 $s_i$ : Standard deviation of sample i

 $n_i$ : Size of sample i



# T-Test: Mathematical Formulation, Degrees of Freedom (df)

Pooled T-Test

$$df = n_1 + n_2 - 2$$

Welch's T-Test

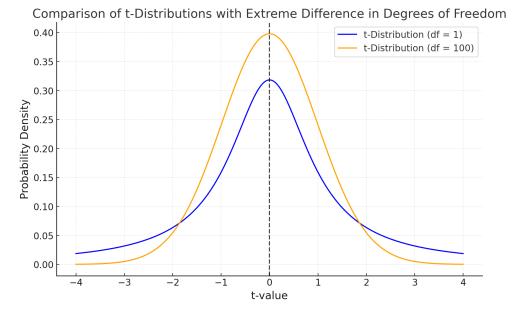
$$df = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\left(\frac{s_1^2}{n_1}\right)^2 + \left(\frac{s_2^2}{n_2}\right)^2}$$

Where:

 $s_n$ : Standard deviation of pooled data

 $s_i$ : Standard deviation of sample i

 $n_i$ : Size of sample i



# T-Test: Mathematical Formulation, p-value

#### 1-Tailed Test

$$p = 1 - F_t(t, df) \text{ if } t>0$$
$$p = F_t(t, df) \text{ if } t<0$$

#### 2-Tailed Test

$$p = 2 \cdot (1 - F_t(|t|, df))$$

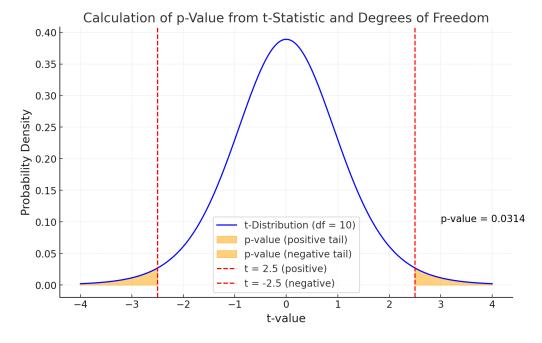
#### Where:

t: t-statistic

*df*: Degrees of freedom

 $F_t(t, df)$ : Cumulative distribution function (CDF)

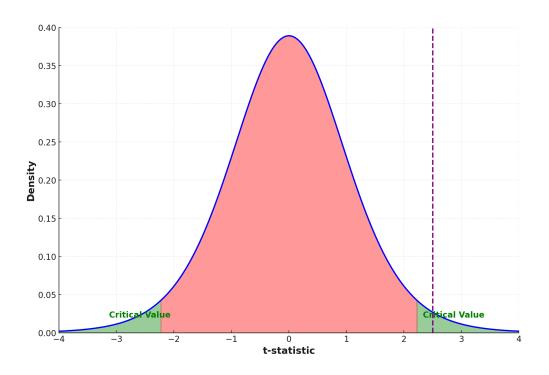
of the t-distribution given t and df



### **T-Test: Make a Decision**

**Reject Ho** if the p-value is less than a; otherwise, **fail to reject Ho**.

In the example on the right, we would **reject Ho** as our t-statistic (dotted line) is to the right of the critical value





### **T-Test**

#### When to use it

- Comparing two groups
- Single dependent variable
- Small sample size

#### When not to use it

- Comparing more than 2 groups (use ANOVA)
- Non-normal Data
- Multiple Dependent Variables



# **Analysis of Variance (ANOVA)**

#### What is it?

**ANOVA** is used to compare the means of three or more groups to determine if at least one group mean is significantly different.

It partitions the total variability into between-group and within-group variability. The test statistic follows an F-distribution, which evaluates the ratio of variances.

#### **Types of ANOVA**

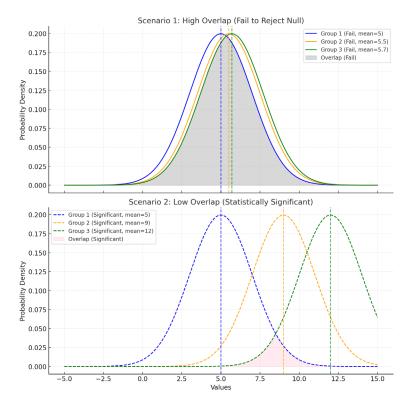
- One-Way ANOVA: Compares means across a single factor with multiple groups.
- Two-Way ANOVA: Analyzes the effect of two factors and their interaction.
- Repeated Measures ANOVA:
  Compares means when the same subjects are measured under different conditions.



# **Analysis of Variance (ANOVA)**

#### **Key Calculations**

- 1. Between-Group Variance
- 2. Within-Group Variance
- 3. Mean-Squared Between
- 4. Mean-Squared Within





# ANOVA: Mathematical Formulation

- Between-group variance (SSB)
- Mean Square Between (MSB)

$$SSB = \sum_{i=1}^{k} n_i (\bar{x}_i - \bar{x})^2$$

$$MSB = \frac{SSB}{k-1}$$

- Within-group variance (SSW)

$$SSW = \sum_{i=1}^{k} \sum_{j=1}^{n_i} (x_{ij} - \bar{x}_i)^2$$

Mean Square Within (MSW)

$$MSW = \frac{SSW}{N - k}$$

F-Statistic

#### Where:

- k: Number of groups  $\bar{x_i}$ : Mean of group I
- N: Overall number of observations
- $n_i$ : Size of group i
- $\bar{x}$ : Overall mean of the data
- $x_{ii}$ : Observation j in group i

$$F = \frac{MSB}{MSW}$$

# **ANOVA**

#### When to use it

- Comparing three or more groups
- Single dependent variable
- Exploring factor effects

#### When not to use it

- Fewer than 3 groups
- Non-normal Data
- Multiple Dependant Variables



