

# PLS 120: Applied Statistics in Agricultural Sciences

ANOVA and Linear Regression Analysis



## Week 10 Tutorial Guide

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**Important Links****Course Website**

All course materials available at: Course Website Link

**Interactive Binder Environment**

Access Week 10 lab materials: Week 10 Binder Link

**Welcome to Week 10: ANOVA and Linear Regression**

This week, we explore **Analysis of Variance (ANOVA)** and **Linear Regression** - powerful statistical methods for comparing multiple groups and modeling relationships between variables in agricultural research.

**Analysis of Variance (ANOVA)****When to Use ANOVA**

ANOVA compares means across multiple groups simultaneously, such as testing different diet treatments on pig growth or comparing crop yields across varieties.

**ANOVA Components and Formulas****ANOVA Table Structure:**

Source	SS	df	MS	F
Between Groups	SSB	$k - 1$	MSB	$MSB/MSE$
Within Groups	SSE	$N - k$	MSE	
Total	SST	$N - 1$		

**Sum of Squares Formulas:**

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

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

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

**Mean Squares:**

$$MSB = \frac{SSB}{k-1} \text{ (Between groups mean square)}$$

$$MSE = \frac{SSE}{N-k} \text{ (Error mean square)}$$

**F-Statistic:**

$$F = \frac{MSB}{MSE} \text{ (Ratio of between to within group variation)}$$

## ANOVA Hypotheses

### ANOVA Hypotheses:

**Null Hypothesis ( $H_0$ ):** All group means are equal

$$H_0 : \mu_1 = \mu_2 = \mu_3 = \dots = \mu_k$$

**Alternative Hypothesis ( $H_1$ ):** At least one group mean differs

$$H_1 : \text{Not all } \mu_i \text{ are equal}$$

### Decision Rule:

If  $F > F_{critical}$  or  $p < \alpha \rightarrow$  Reject  $H_0$

If  $F \leq F_{critical}$  or  $p \geq \alpha \rightarrow$  Fail to reject  $H_0$

### R Implementation:

```
model <- lm(response ~ factor, data)
anova(model)
summary(model)
```

## Manual ANOVA Calculations

### Step-by-Step ANOVA Construction

Building ANOVA tables from scratch helps understand the underlying statistical concepts.

### Manual ANOVA Steps:

#### 1. Calculate Group Means:

$$\bar{x}_i = \frac{\sum_{j=1}^{n_i} x_{ij}}{n_i} \text{ for each group } i$$

#### 2. Calculate Overall Mean:

$$\bar{x} = \frac{\sum_{i=1}^k \sum_{j=1}^{n_i} x_{ij}}{N} \text{ where } N = \sum n_i$$

#### 3. Calculate Sum of Squares:

SSB: Weighted squared deviations of group means from overall mean

SST: Squared deviations of all observations from overall mean

SSE: SST - SSB (or sum of within-group squared deviations)

#### 4. Calculate Degrees of Freedom:

Between:  $df_1 = k - 1$  (number of groups - 1)

Within:  $df_2 = N - k$  (total observations - number of groups)

Total:  $df_3 = N - 1$  (total observations - 1)

#### 5. Calculate Mean Squares and F-statistic:

$$MSB = SSB/df_1, MSE = SSE/df_2$$

$$F = MSB/MSE$$

## Linear Regression Analysis

### Simple Linear Regression

Linear regression models the relationship between a continuous predictor and response variable.

**Simple Linear Regression Model:**

$$y = \beta_0 + \beta_1 x + \epsilon$$

Where:

$y$  = response variable (dependent)

$x$  = predictor variable (independent)

$\beta_0$  = intercept ( $y$ -value when  $x = 0$ )

$\beta_1$  = slope (change in  $y$  per unit change in  $x$ )

$\epsilon$  = random error term

**Least Squares Estimates:**

$$\hat{\beta}_1 = \frac{\sum(x_i - \bar{x})(y_i - \bar{y})}{\sum(x_i - \bar{x})^2}$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

**Correlation Coefficient:**

$$r = \frac{\sum(x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum(x_i - \bar{x})^2 \sum(y_i - \bar{y})^2}}$$

**R-squared (Coefficient of Determination):**

$R^2 = r^2$  = Proportion of variance explained by the model

**R Implementation:**

```
model <- lm(y ~ x, data)
summary(model)
cor(x, y)
```

## Multiple Linear Regression

**Multiple Linear Regression Model:**

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_p x_p + \epsilon$$

**Categorical Predictors:**

When using factors (like Species), R creates dummy variables:

Reference category (baseline)

Coefficients represent differences from reference

**Model Interpretation:**

Each  $\beta_i$  represents change in  $y$  per unit change in  $x_i$

Holding all other variables constant

Intercept is  $y$ -value when all predictors = 0

**R Implementation:**

```
model <- lm(y ~ x1 + x2 + factor, data)
model <- lm(Petal.Width ~ Species, data)
model <- lm(Petal.Width ~ Petal.Length, data)
```

## Assignment 10 Overview

### Assignment Structure (20 points total)

**1. Part 1: Data Loading and Study Design (6 points)**

- Identify treatment, response variable, and experimental unit (3 points)
- Formulate null and alternative hypotheses (2 points)
- Load and examine pig weight data (1 point)

**2. Part 2: ANOVA Table Calculations (7 points)**

- Calculate sum of squares (SSB, SSE, SST) (2 points)
  - Calculate degrees of freedom (1 point)
  - Calculate mean squares and F-statistic (2 points)
  - Complete ANOVA table (2 points)
3. **Part 3: Statistical Interpretation and Conclusions (7 points)**
- Interpret F-statistic and p-value (3 points)
  - Draw conclusions about diet effects (2 points)
  - Discuss practical implications (2 points)

## Agricultural Applications

### Real-World ANOVA and Regression Applications:

- **Variety Trials** - Compare crop performance across multiple varieties
- **Treatment Comparisons** - Test fertilizer, pesticide, or management effects
- **Feed Efficiency Studies** - Analyze animal growth under different diets
- **Soil Management** - Compare tillage or amendment effects on soil health
- **Environmental Factors** - Model relationships between climate and yield
- **Quality Assessment** - Analyze factors affecting product quality
- **Breeding Programs** - Compare genetic lines or breeding methods
- **Economic Analysis** - Model cost-benefit relationships in agriculture

## Getting Started

1. Launch Week 10 Binder environment
2. Navigate to `class_activity` folder
3. Open `Week10_ANOVA_Regression.ipynb`
4. Work through ANOVA and regression examples
5. Complete Assignment 10 in `assignment` folder

## Learning Objectives

By the end of this week, you will be able to:

- Build ANOVA tables from scratch with manual calculations
- Understand sum of squares, degrees of freedom, and F-statistics
- Perform one-way ANOVA using R functions
- Fit and interpret simple and multiple linear regression models

- Check assumptions for ANOVA and regression analyses
- Choose appropriate statistical methods for different research questions
- Interpret statistical output in agricultural research context
- Understand the relationship between ANOVA and regression

## Need Help?

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Zoom Link: [Join Office Hours](#)