

# Summative Assessment 2

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## Introduction

This performs the necessary assumption checks for conducting a 2-Factor ANOVA on the Alzheimer's Mice Data. Assumptions include normality, homogeneity of variance, and independence of observations.

```
# Load required libraries
library(tidyverse)
```

```
## — Attaching core tidyverse packages — tidyverse 2.0.0 —
## ✓ dplyr      1.1.4      ✓ readr      2.1.5
## ✓ forcats    1.0.0      ✓ stringr    1.5.1
## ✓ ggplot2    3.5.1      ✓ tibble     3.2.1
## ✓ lubridate  1.9.3      ✓ tidyr      1.3.1
## ✓ purrr      1.0.2
## — Conflicts — tidyverse_conflicts() —
## ✖ dplyr::filter() masks stats::filter()
## ✖ dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(car)
```

```
## Loading required package: carData
##
## Attaching package: 'car'
##
## The following object is masked from 'package:dplyr':
##
##     recode
##
## The following object is masked from 'package:purrr':
##
##     some
```

```
# Load the data
data <- read.csv("Alzheimers Mice Data.csv")

# View the first few rows
head(data)
```

```
##   AD_Status Treatment Training Memory
## 1         1         1      12      10
## 2         1         1      15      12
## 3         1         1      13      13
## 4         1         1      12      10
## 5         1         1      14      13
## 6         1         2      15      13
```

### ##Check Assumptions

####Assumption 1: Normality The Shapiro-Wilk test is used to check normality for the Training and Memory columns. Histograms with density curves and QQ plots are also generated for visual inspection.

```
# Normality check
training_normality <- shapiro.test(data$Training)
memory_normality <- shapiro.test(data$Memory)

# Display results
training_normality
```

```
##
##   Shapiro-Wilk normality test
##
## data:  data$Training
## W = 0.93583, p-value = 0.02507
```

```
memory_normality
```

```
##
##   Shapiro-Wilk normality test
##
## data:  data$Memory
## W = 0.97184, p-value = 0.4107
```

```
# Visual inspection: histograms
par(mfrow = c(1, 2)) # Set layout
hist(data$Training, breaks = 10, col = "blue", main = "Training Errors", xlab = "Errors")
lines(density(data$Training), col = "red")

hist(data$Memory, breaks = 10, col = "green", main = "Memory Errors", xlab = "Errors")
lines(density(data$Memory), col = "red")
```



## Assumption 2: Homogeneity of Variance

Levene's test is performed to assess homogeneity of variances.

```
# Levene's Test for Training Errors
levene_training <- leveneTest(Training ~ as.factor(AD_Status) * as.factor(Treatment), data = data)
levene_training
```

```
## Levene's Test for Homogeneity of Variance (center = median)
##      Df F value Pr(>F)
## group  7  0.4346 0.8731
##      32
```

```
# Levene's Test for Memory Errors
levene_memory <- leveneTest(Memory ~ as.factor(AD_Status) * as.factor(Treatment), data = data)
levene_memory
```

```
## Levene's Test for Homogeneity of Variance (center = median)
##      Df F value Pr(>F)
## group  7  0.8275 0.5722
##      32
```

### ##Results