## PS: Indpendent and Dependent t-test for TAs

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```
R practice.
Install packages.
```

```
library(car)
## Loading required package: carData
library(ggplot2)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following object is masked from 'package:car':
##
##
       recode
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
```

Load the water pollution data into R.

```
##
     Birds
              Species Gender Treatment value
## 1
         1 Hylocichla
                          M Pre_weight
## 2
         2 Hylocichla
                           M Pre_weight
                                           44
## 3
        3 Hylocichla
                                           48
                           M Pre_weight
## 4
        4 Hylocichla
                           M Pre_weight
                                           42
## 5
        5 Hylocichla
                           M Pre_weight
                                           51
## 6
         6 Hylocichla
                           M Pre_weight
                                           51
```

Make a subgroup of Weight before migration (Pre\_weight)

```
df_bird_Pre_weight <- df_bird_all %>%
filter(Treatment == 'Pre_weight')
```

```
summary(df_bird_Pre_weight)
```

```
{\tt Treatment}
##
        Birds
                      Species
                                          Gender
## Min. : 1.00
                   Length:28
                                       Length:28
                                                          Length:28
  1st Qu.: 7.75
                    Class :character
                                       Class : character
                                                          Class : character
## Median :14.50
                   Mode :character
                                       Mode :character
                                                          Mode :character
```

```
## Mean :14.50
##
  3rd Qu.:21.25
  Max.
         :28.00
##
##
       value
## Min.
          :41.00
##
  1st Qu.:46.75
## Median :49.00
         :48.68
## Mean
## 3rd Qu.:52.00
## Max. :55.00
```

## Part I.

2. Analyze data distribution. Is your pre-weight data normally distributed?

```
shapiro.test(df_bird_Pre_weight$value)
```

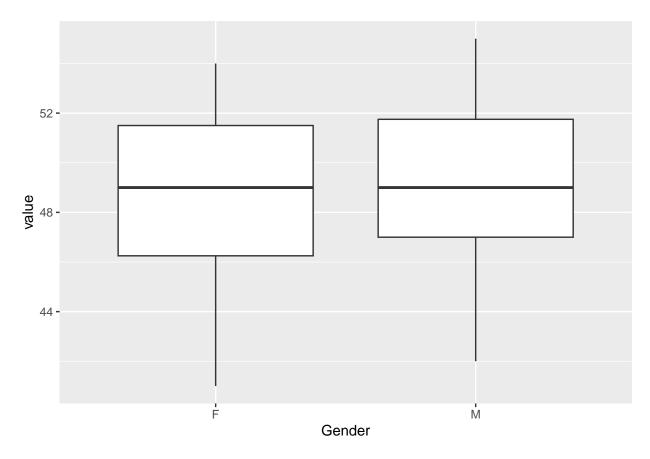
```
##
## Shapiro-Wilk normality test
##
## data: df_bird_Pre_weight$value
## W = 0.95557, p-value = 0.2725
```

3. Test to see if you meet the assumption of equal variance for this test. Based on the results of this test for unequal variance, which test should you use?

```
# Perform Levene's test
result <- leveneTest(value~Gender, data=df_bird_Pre_weight)</pre>
```

```
## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.
print(result)
## Levene's Test for Homogeneity of Variance (center = median)
         Df F value Pr(>F)
## group 1 0.0061 0.9381
##
         26
p value <- result$Pr[1]</pre>
# Significance level
alpha <- 0.05
# Interpret the result
if (p_value < alpha) {</pre>
 cat("Reject null hypothesis: Variances are significantly different.\n")
} else {
  cat("Fail to reject null hypothesis: Variances are not significantly different.\n")
}
## Fail to reject null hypothesis: Variances are not significantly different.
4. Run this analysis. What is the test statistic (obtained) value for the test you selected?
Use alternative to determine your hypothesis. Default= "two.sided".
alternative = c("two.sided", "less", "greater"),
# Perform t-test
t_test_result <- t.test(value~Gender,alternative = "two.sided",</pre>
                         var.equal=TRUE, data=df_bird_Pre_weight)
print(t_test_result)
##
##
   Two Sample t-test
##
## data: value by Gender
## t = -0.78654, df = 26, p-value = 0.4387
## alternative hypothesis: true difference in means between group F and group M is not equal to 0
## 95 percent confidence interval:
## -4.387692 1.959120
## sample estimates:
## mean in group F mean in group M
          48.07143
                          49.28571
##
# Create a boxplot plot using ggplot2
ggplot(df_bird_Pre_weight, aes(x = Gender, y = value)) +
```

geom\_boxplot() # Add jitter for better visualization



Part II. Analyze data distribution. Is your pre-weight data normally distributed?

```
##
## Shapiro-Wilk normality test
##
## data: df_bird_Pre_weight$value
## W = 0.95557, p-value = 0.2725
```

Test to see if you meet the assumption of equal variance for this test. Based on the results of this test for unequal variance, which test should you use?

```
# Perform Levene's test
result <- leveneTest(value~Species, data=df_bird_Pre_weight)

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

print(result)

## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 1 1.2128 0.2809
## 26
```

```
p_value <- result$Pr[1]
# Significance level
alpha <- 0.05

# Interpret the result
if (p_value < alpha) {
   cat("Reject null hypothesis: Variances are significantly different.\n")
} else {
   cat("Fail to reject null hypothesis: Variances are not significantly different.\n")
}</pre>
```

## Fail to reject null hypothesis: Variances are not significantly different.

8. Run this analysis. What is the test statistic (obtained) value for the test you selected? Use alternative to determine your hypothesis. Default= "two.sided". alternative = c("two.sided", "less", "greater"),

```
##
## Two Sample t-test
##
## data: value by Species
## t = -2.891, df = 26, p-value = 0.007655
## alternative hypothesis: true difference in means between group Hylocichla and group Progne is not eq
## 95 percent confidence interval:
## -6.721783 -1.135360
## sample estimates:
## mean in group Hylocichla mean in group Progne
## 46.71429 50.64286

## Create a boxplot plot using ggplot2
```

# Create a boxplot plot using ggplot2
ggplot(df\_bird\_Pre\_weight, aes(x = Species, y = value)) +
 geom\_boxplot() # Add jitter for better visualization



Part III. 12. Analyze data distribution. Is your pre-weight data normally distributed?

```
##
## Shapiro-Wilk normality test
##
## data: df_bird_all$value
## W = 0.97528, p-value = 0.3028
```

Test to see if you meet the assumption of equal variance for this test. Based on the results of this test for unequal variance, which test should you use?

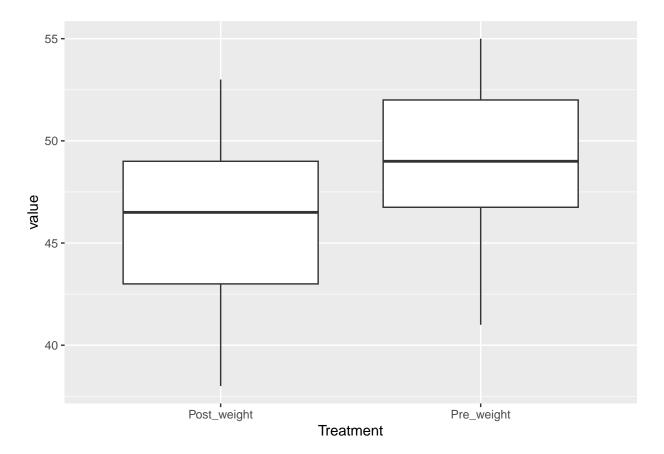
```
# Perform Levene's test
result_paired_t_test <- leveneTest(value~Treatment, data=df_bird_all)

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

print(result_paired_t_test)

## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 1 0.0035 0.9529
## 54
```

```
p_value <- result_paired_t_test$Pr[1]</pre>
# Significance level
alpha <- 0.05
# Interpret the result
if (p_value < alpha) {</pre>
 cat("Reject null hypothesis: Variances are significantly different.\n")
  cat("Fail to reject null hypothesis: Variances are not significantly different.\n")
## Fail to reject null hypothesis: Variances are not significantly different.
t_paired_result <- t.test(value ~ Treatment, paired = TRUE,</pre>
                          data = df_bird_all)
print(t_paired_result)
##
## Paired t-test
##
## data: value by Treatment
## t = -3.4184, df = 27, p-value = 0.002013
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -4.400616 -1.099384
## sample estimates:
## mean of the differences
##
                     -2.75
# Create a boxplot plot using ggplot2
ggplot(df_bird_all, aes(x = Treatment, y = value)) +
 geom_boxplot() # Add jitter for better visualization
```



Part IV. From the data frame make two groups, make two groups (females and males)

```
df_bird_female <- df_bird_all %>%
  filter(Gender == 'F')

df_bird_male <- df_bird_all %>%
  filter(Gender == 'M')
```

## **Assumptions for Females**

```
## ## Shapiro-Wilk normality test
## data: df_bird_female$value
## W = 0.95617, p-value = 0.282

leveneTest(value~Treatment, data=df_bird_female)

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

## Levene's Test for Homogeneity of Variance (center = median)
```

```
Df F value Pr(>F)
## group 1 0.0053 0.9424
##
         26
t_paired_female <- t.test(value ~ Treatment, paired = TRUE,</pre>
                          alternative= 'less',
                          data = df_bird_female)
print(t_paired_female)
##
## Paired t-test
##
## data: value by Treatment
## t = -1.9868, df = 13, p-value = 0.03422
\#\# alternative hypothesis: true difference in means is less than 0
## 95 percent confidence interval:
##
          -Inf -0.1164104
## sample estimates:
## mean of the differences
                 -1.071429
Assumptions for Males
shapiro.test(df_bird_male$value)
##
## Shapiro-Wilk normality test
##
## data: df_bird_male$value
## W = 0.97278, p-value = 0.6567
leveneTest(value~Treatment, data=df_bird_male)
## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.334 0.5683
         26
##
t_paired_male <- t.test(value ~ Treatment, paired = TRUE,</pre>
                        alternative= 'less',
                        data = df_bird_male)
print(t_paired_male)
## Paired t-test
## data: value by Treatment
```

```
## t = -3.1601, df = 13, p-value = 0.003762
## alternative hypothesis: true difference in means is less than 0
## 95 percent confidence interval:
## -Inf -1.946804
## sample estimates:
## mean of the differences
## -4.428571
```