# $HW4\_ISYE6414\_ashish\_dhiman$

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```
library(ggplot2)
setwd("~/data_projects/fall22_hw/isye6414_hw/hw4")
```

# Boxplot and ANOVA

# Read Data and Summary

##

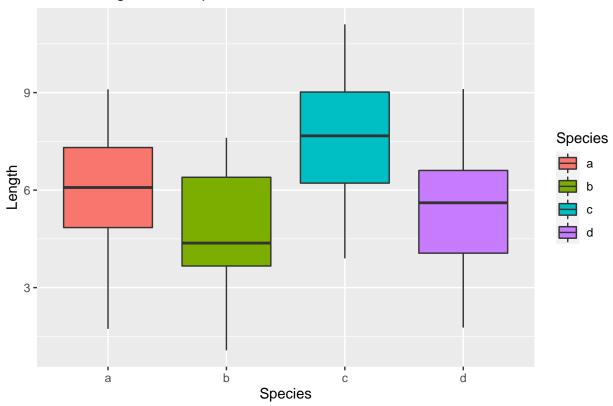
## a b c d ## 20 15 22 28

```
head -5 ./homework04data01.csv
## "Species", "Length"
## "a",7.72
## "a",7.26
## "a",9.1
## "a",3.69
data1 = read.csv("./homework04data01.csv",sep=",")
dim(data1)
## [1] 85 2
summary(data1)
##
     Species
                          Length
## Length:85
                            : 1.070
                     Min.
## Class:character 1st Qu.: 4.620
## Mode :character Median : 6.100
##
                      Mean : 6.015
##
                      3rd Qu.: 7.570
##
                      Max. :11.100
table(data1$Species)
```

#### **Question 1: Box Plots**

```
ggplot(data1, aes(x=Species, y=Length, fill=Species)) +
   geom_boxplot() + ggtitle("Antler Length basis Species")
```

# Antler Length basis Species



From the box plot above we see there is a significant overlap between the ranges of antler length between the different species. Hence, without further investigation, it is very difficult to conclude if there is a difference between the antler length among species.

The box plot however, does give us an indication of possible differences:

- 1. Mean of c is larger than every one else
- 2. Mean of a and d are very similar

## 6.014706

### Question 2: ANOVA to test difference in mean with $\alpha = 0.05$

```
anova_v0 = aov(Length ~ Species, data=data1)
model.tables(anova_v0, type = "means")

## Tables of means
## Grand mean
##
```

```
##
## Species
## a b c d
## 6.063 4.697 7.594 5.445
## rep 20.000 15.000 22.000 28.000
```

Hypothesis test for equal means:

```
H_0: \mu_a = \mu_b = \mu_c = \mu_d
```

 $H_a$ : Some means are different

# summary(anova\_v0)

```
## Df Sum Sq Mean Sq F value Pr(>F)
## Species   3 90.0 30.001 8.121 8.58e-05 ***
## Residuals 81 299.2 3.694
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

From above table we see that the p-value is 8.58e-5. This is fairly smaller than 5%, thus we have sufficient statistical evidence to reject Null Hypothesis, i.e. All species have similar means

# Question 3: Identify SSE, SSTr, MSE, MSTr

From above table:

```
• SSE = 299.2
```

• SSTr = 90

• MSE = SSE/(N-k) = 299.2/81 = 3.694 (from table)

```
299.2/81
```

```
## [1] 3.693827
```

• MSTr = SSTr/k-1 = 90.0/3 = 30.001 (from table)

```
90.0/3
```

## [1] 30

# Question 4: Pairwise comparison of means

```
TukeyHSD(anova_v0)
```

```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = Length ~ Species, data = data1)
##
## $Species
##
             diff
                          lwr
                                     upr
                                             p adj
## b-a -1.3656667 -3.08777369
                               0.3564404 0.1683081
  c-a 1.5306364 -0.02706805
                               3.0883408 0.0559481
## d-a -0.6176429 -2.09373459
                               0.8584489 0.6920003
## c-b 2.8963030 1.20807683
                               4.5845292 0.0001303
## d-b 0.7480238 -0.86520632
                               2.3612539 0.6183736
## d-c -2.1482792 -3.58469904 -0.7118594 0.0010291
```

The above table gives the 95% CI for the 6 pairwise difference of means above.

As expected from the box plot, c-b and d-c both have low p-values. Similarly, p-value for c-a is about 5%, hinting there is a slight difference in means, especially for alpha <6%.

While d-b has lowest absolute diff and the highest p-value.

#### Question 5: Multiple Linear regression

```
y: monthly labor hours required
```

x1: monthly X-ray exposures

x2: monthly occupied bed days

x3: average length of patients' stays (in days).

```
data2 = read.csv("./homework04Hospital.csv",sep=",")
dim(data2)
```

## [1] 16 4

```
summary(data2)
```

```
##
         Xray
                        BedDays
                                            Length
                                                              Hours
##
                            : 472.9
                                               : 3.900
                                                                 : 566.5
           : 2048
    1st Qu.: 5765
                     1st Qu.: 1359.3
                                        1st Qu.: 4.810
                                                          1st Qu.: 1609.4
                     Median : 2279.7
##
    Median: 9616
                                       Median : 5.560
                                                          Median: 2233.1
##
    Mean
           :17036
                            : 4280.4
                                       Mean
                                               : 5.824
                                                          Mean
                                                                 : 4643.1
                     Mean
##
    3rd Qu.:16684
                     3rd Qu.: 3879.5
                                        3rd Qu.: 6.223
                                                          3rd Qu.: 3812.7
           :86533
                            :15524.0
                                               :10.780
                                                                 :18854.5
##
    Max.
                     Max.
                                       Max.
                                                          Max.
```

```
lr_model = lm(Hours ~ ., data = data2)
summary(lr_model)
```

```
##
## Call:
  lm(formula = Hours ~ ., data = data2)
##
##
  Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
   -677.23 -270.19
                     60.93
                            228.32
                                   517.70
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
  (Intercept) 1946.80204
                           504.18193
                                       3.861
                                              0.00226 **
                             0.01304
                                       2.958
                                              0.01197 *
## Xray
                  0.03858
## BedDays
                  1.03939
                             0.06756
                                      15.386 2.91e-09 ***
                            98.59828
                                      -4.196 0.00124 **
## Length
               -413.75780
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 387.2 on 12 degrees of freedom
## Multiple R-squared: 0.9961, Adjusted R-squared: 0.9952
## F-statistic: 1028 on 3 and 12 DF, p-value: 9.919e-15
```

From the above table we can have beta of the different predictors

```
• \beta_{xray} = 0.03858
```

- $\beta_{bed\ days} = 1.03939$
- $\beta_{length} = -413.75780$

#### Question 6: Interpretation of each coefficient

- $\beta_{xray} = 0.03858$ : This implies that with each unit monthly exposure of xray the expected monthly labor hours required increase by 0.03858, given rest of the predictors remain constant.
- $\beta_{bed\_days} = 1.03939$ . Similarly with each unit increase in monthly occupied bed days the expected monthly labor hours required increase by 1.03939, given rest of the predictors remain constant.
- $\beta_{length} = -413.75780$  With each unit increase in average length of patients' stay (in days), the expected monthly labor hours required decrease by 413.7580, given rest of the predictors remain constant. The negative effect seems a little counter intuitive here, but might because with greater length of stay in hospital, the labor requirement is more spread out, but we need more information on data collection to identify this.

```
test = lm(Hours ~ Length, data = data2)
summary(test)
```

```
##
## Call:
## lm(formula = Hours ~ Length, data = data2)
##
## Residuals:
## Min 1Q Median 3Q Max
## -6068.1 -2395.8 -1177.8 180.5 13193.4
```

```
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6636.4
                           4613.4 -1.439
                                             0.172
## Length
                1936.6
                            765.2
                                    2.531
                                             0.024 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4769 on 14 degrees of freedom
## Multiple R-squared: 0.3139, Adjusted R-squared: 0.2649
## F-statistic: 6.406 on 1 and 14 DF, p-value: 0.02399
```

Thus in absence of other predictors, Length is postiively effecting y. Also note that the intercept value is negative here, and p-value has also changed.

#### Question 7: Hypothesis Test on BedDays

```
H_0: \beta_{bed\ days} = 0 H_a: \beta_{bed\ days} \neq 0
```

test statistic = 15.386

Critical t value, from qt function

```
print ("Upper tale t")

## [1] "Upper tale t"

qt(0.975, df = 12)

## [1] 2.178813

print ("Lower tale t")

## [1] "Lower tale t"

qt(0.025, df = 12)
```

## [1] -2.178813

Thus are  $|{\rm critical}\ {\rm t}\ {\rm value}|$  is 2.178813

Now given out test statistic is significantly larger than critical upper tail t value, we have sufficient evidence to reject the null hypothesis and conclude that BedDays variable is a significant predictor, given all other predictor variables in the model.

### Question 8: Hypothesis Test on Xray

$$H_0: \beta_{xray} = 0 \quad H_a: \beta_{xray} \neq 0$$

For the above we have p-value as 0.01197 = 1.197% (this is two tailed probability here)

For  $\alpha = 5\%$ ,  $p < \alpha$  Hence we can reject the null hypothesis in this case, and have sufficient statistical evidence to conclude that **Xray variable** is a **significant predictor**, given all other predictor variables in the model.

However if our confidence level changes, i.e. for  $\alpha = 1\%$ ,  $p > \alpha$  Hence we can not reject the null hypothesis in this case, and conclude that **Xray variable is not a significant predictor**, given all other predictor variables in the model.

### Question 8: Hypothesis Test on Length

$$H_0: \beta_{length} = 0 \quad H_a: \beta_{length} \neq 0$$

For the above we have p-value as 0.00124 = 0.124% (this is two tailed probability here)

For  $\alpha = 5\%$ ,  $p < \alpha$  Hence we can reject the null hypothesis in this case, and have sufficient statistical evidence to conclude that **length variable** is a significant **predictor**, given all other predictor variables in the model.

Even for more stricter confidence level, i.e. for  $\alpha = 1\%, p < \alpha$  Hence we can reject the null hypothesis again in this case, and conclude that **length variable is a significant predictor**, given all other predictor variables in the model.