# Homework 6

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11/8/2021

# Question 1

# Part a

```
## Data from the website
x = c(110.5, 105.4, 118.1, 104.5, 93.6, 84.1, 77.8, 75.6)
y = c(5.755, 5.939, 6.010, 6.545, 6.730, 6.750, 6.899, 7.862)
## binds x and y into a dataframe
d = data.frame(c(x,y))
\#\# Creates variable fit_d that shows the slope and the y-intercept
fit_d = lm(y~x, data = d)
print(fit_d)
##
## Call:
## lm(formula = y \sim x, data = d)
## Coefficients:
## (Intercept)
                          Х
      10.13746
                   -0.03717
##
```

The least squares estimate is of  $\beta_1$  is -.037. This value represent the best fit of the trend of the data that reduces the distance from all of the points to the line itself

### Part B

```
## Runs an ANOVA test
anovad = anova(fit_d)
print(anovad)

## Analysis of Variance Table
##
```

```
## Response: y
##
           Df Sum Sq Mean Sq F value Pr(>F)
            1 2.42357 2.42357 18.455 0.005116 **
## Residuals 6 0.78794 0.13132
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## used to get the t-test p-value
summaryd = summary(fit_d)
print(summaryd)
##
## Call:
## lm(formula = y \sim x, data = d)
## Residuals:
##
       \mathtt{Min}
                 1Q Median
                                   3Q
## -0.34626 -0.27605 -0.09448 0.27023 0.53495
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 10.137455 0.842265 12.036
                                              2e-05 ***
                         0.008653 -4.296 0.00512 **
## x
              -0.037175
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3624 on 6 degrees of freedom
## Multiple R-squared: 0.7547, Adjusted R-squared: 0.7138
## F-statistic: 18.46 on 1 and 6 DF, p-value: 0.005116
```

Both the F-test and the T-test show a P-value <. 01 confirming the alternative hypothesis. This means that  $H_a \neq 0$  is true.

### Part C

#### Part D

```
## Calculates raw residuals
residd= resid(fit_d)

print(residd)

## 1 2 3 4 5 6 7

## -0.2746519 -0.2802428 0.2628757 0.2922999 0.0720958 -0.2610638 -0.3462643

## 8

## 0.5349514
```

Based off the values from part a, we know the y-intercept and the slope of the regression line. This give us the equation:

$$\hat{y} = 10.137 - 0.0372x$$

### Part E

```
##Print the error values
print(summaryd[6])
```

## \$sigma ## [1] 0.3623848

Based off the summary statistics generated in part b, we know that  $\hat{\sigma}^2 = .3624$ 

# Part F

```
## Predicts the 95% confidence interval based on the new rice of x = 100
mu0conf = predict(fit_d, newdata = data.frame(x = 100), interval = "confidence")
print(mu0conf)

## fit lwr upr
## 1 6.419986 6.096321 6.743651
```

# Part G

```
## Predicts the 95% prediction interval based on the new rice of x = 100
mu0pred = predict(fit_d, newdata = data.frame(x = 100), interval = "prediction")
print(mu0pred)
```

```
## fit lwr upr
## 1 6.419986 5.476038 7.363934
```

Compared to the confidence interval (range = 0.647), the prediction interval (range = 1.8879) is much wider

# Part H

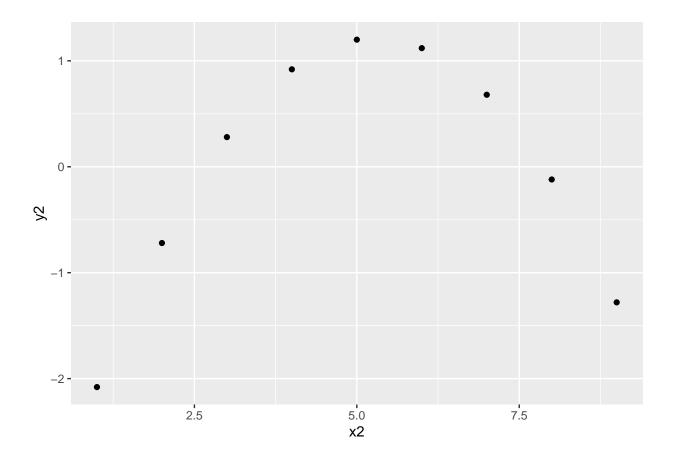
```
## Pulls the r^2 value from summaryd
d_r_squared = summaryd['r.squared']
```

The R-squared value represents how much variance that can be explained by the independent variable, in this case plant height.

# Question 2

### Part A

```
library(tidyverse, warn.conflicts = FALSE)
## -- Attaching packages ------ tidyverse 1.3.1 --
## v ggplot2 3.3.5
                  v purrr
                            0.3.4
## v tibble 3.1.5 v dplyr
                           1.0.7
## v tidyr 1.1.4 v stringr 1.4.0
## v readr
         2.0.2 v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                masks stats::lag()
## Data for question 2
x2 = c(1, 2, 3, 4, 5, 6, 7, 8, 9)
y2 = c(-2.08, -0.72, 0.28, 0.92, 1.20, 1.12, 0.68, -0.12, -1.28)
# Combines data into datafram
d2 = data.frame(x2,y2)
## Plots data
ggplot(d2, aes(x = x2, y = y2)) +
 geom_point()
```



Part B