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# Set Working directory
setwd('/home/noble mannu/Documents/PhD/First/
STAT 2131 Applied Statistical Methods I/HW6')
### Exercise 4 6.10 ###
## Part (a) ##
# Read the table with the information of the data
data <- read.delim('CH06PR09.txt', sep = '', dec = '.', header = FALSE)</pre>
names(data) <- c('Labor_Hours','Cases_Shipped','Indirect_Costs','Holiday')</pre>
# Make the multilinear regression model
linearMod = lm(formula = Labor_Hours~Cases_Shipped+Indirect_Costs+Holiday, data =
data)
# Display summary of our model
summary(linearMod)
# Obtain the regression coefficients
beta 0 = coef(linearMod)[1]
beta_1 = coef(linearMod)[2]
beta 2 = coef(linearMod)[3]
beta 3 = coef(linearMod)[4]
## Part (c) ##
# Obtain the residuals of the linear model
linearMod res <- resid(linearMod) # Or just linearMod$residuals</pre>
# Plot the residuals against the fitted values
plot(linearMod$fitted.values, resid(linearMod), xlab = "Fitted Values", ylab =
"Residuals",
     main = 'Residual plot against Y Hat', col = 'blue')
abline(a=0, b=0)
# Here I separated the 2 clusters and computed the variance of each one
cluster1 = linearMod$fitted.values[which(linearMod$fitted.values < 4600)]</pre>
var(cluster1)
cluster2 = linearMod$fitted.values[which(linearMod$fitted.values > 4600)]
var(cluster2)
# PLot the residuals against X1
plot(data$Cases_Shipped, resid(linearMod), xlab = "Number of cases shipped", ylab =
"Residuals",
     main = 'Residual plot against X1', col = 'blue')
abline(a=0, b=0)
# PLot the residuals against X2
plot(data$Indirect Costs, resid(linearMod), xlab = "Indirect cost", ylab =
"Residuals",
     main = 'Residual plot against X2', col = 'blue')
abline(a=0, b=0)
# PLot the residuals against X3
plot(data$Holiday, resid(linearMod), xlab = "Week with holiday", ylab =
"Residuals",
     main = 'Residual plot against X3', col = 'blue')
abline(a=0, b=0)
# PLot the residuals against X1X2
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ylab = "Estimated residual quantiles", col = 'red')
gqline(linearMod$residuals)
### Exercise 4 7.4 ###
## Part (b) ##
# Make the multilinear regression model (full and reduced models)
mod1 <- lm(formula = Labor_Hours ~ Cases_Shipped+Indirect_Costs+Holiday, data =</pre>
summary(mod1)
mod1 red <- lm(formula = Labor Hours ~ Cases Shipped+Holiday, data = data)
summary(mod1 red)
# Using an ANOVA table
anova(mod1,mod1 red)
# For the p-value
qf(.95, df1 = 1, df2 = length(data$Labor Hours)-4)
1 - pf(0.3251, df1 = 1, df2 = length(data$Labor_Hours)-4)
### Exercise 4 6.16 ###
## Part (a) ##
# Read the data
data2 <- read.delim('CH06PR15.txt', sep = '', dec = '.', header = FALSE)</pre>
names(data2) <- c('Satisfaction','Age','Severity Illness','Anxiety Level')</pre>
# Regressing Satisfaction (Y) onto Age (X1), Severity of Illness (X2) and Anxiety
Level (FULL MODEL)
m1 <- lm(Satisfaction ~ Age+Severity Illness+Anxiety Level, data = data2)
summary(m1)
# Making the reduced model with coefficients equal to 0, except beta 0 (REDUCED
m1 reduced <- lm(Satisfaction ~ 1, data = data2)</pre>
summary(m1 reduced)
# Conducting the F-test I did this in two ways
# First way, doing all the computations separately
x <- m1$fitted.values - mean(data2$Satisfaction)</pre>
SSR <- x%*%x
MSR < - SSR/(4-1)
y <- m1$fitted.values - data2$Satisfaction
SSE <- y%*%y
MSE <- SSE/(length(data2$Satisfaction)-4)</pre>
f <- MSR/MSE
# Second way, using an ANOVA table
anova(m1,m1_reduced)
# For the p-value
qf(.90, df1 = 4-1, df2 = length(data2$Satisfaction)-4)
1 - pf(f, df1 = 4-1, df2 = length(data2$Satisfaction)-4)
## Part (b) ##
b 1 <- coefficients(m1)[2]
```

```
sd b 2 <- 0.4920
sd_b_3 < -7.0997
n <- length(data2$Satisfaction)</pre>
p <- 4
g <- 3
alpha <- 0.1
t \leftarrow qt(1 - alpha/(2*g), df=n-p)
# For beta1
b_1 - t*sd_b_1
b^{-}1 + t*sd_{-}b_{-}^{-}1
# For beta2
b_2 - t*sd_b_2
b_2 + t*sd_b_2
# For beta3
b_3 - t*sd_b_3
b_3 + t*sd_b_3
## Part (c) ##
# We computed SSR and SSE in part (a) and we know SSTO = SSR+SSE
SSTO <- SSR + SSE
# We could also use the definition of SSTO
z <- data2$Satisfaction - mean(data2$Satisfaction)</pre>
SST01 <- z%*%z
# Computing the Coefficient of multiple determination
R squared <- SSR/SSTO
R squared1 <- 1 - SSE/SST0
```