

Homework7

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10/27/2022

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load data

```
mmdata<-read.table("../homework7/CH01PR27.txt",header = F)
colnames(mmdata)<-c("Y", "X")
head(mmdata,3)

##      Y  X
## 1 106 43
## 2 106 41
## 3  97 47

# tail check
tail(mmdata,2)

##      Y  X
## 59 70 72
## 60 74 76
```

a) Fit regression model (8.2). Plot the fitted regression function and the data. Does the quadratic

#regression function appear to be a good fit here? Find R^2 .

```
#center the variables
t<-(mmdata$X-mean(mmdata$X))
t_power<- t^2
```

```

mmdata<-cbind(mmdata,t,t_power)

mmdata.reg=lm(Y~t + t_power,data = mmdata)
mmdata.reg

##
## Call:
## lm(formula = Y ~ t + t_power, data = mmdata)
##
## Coefficients:
## (Intercept)          t          t_power
##   82.93575     -1.18396      0.01484

```

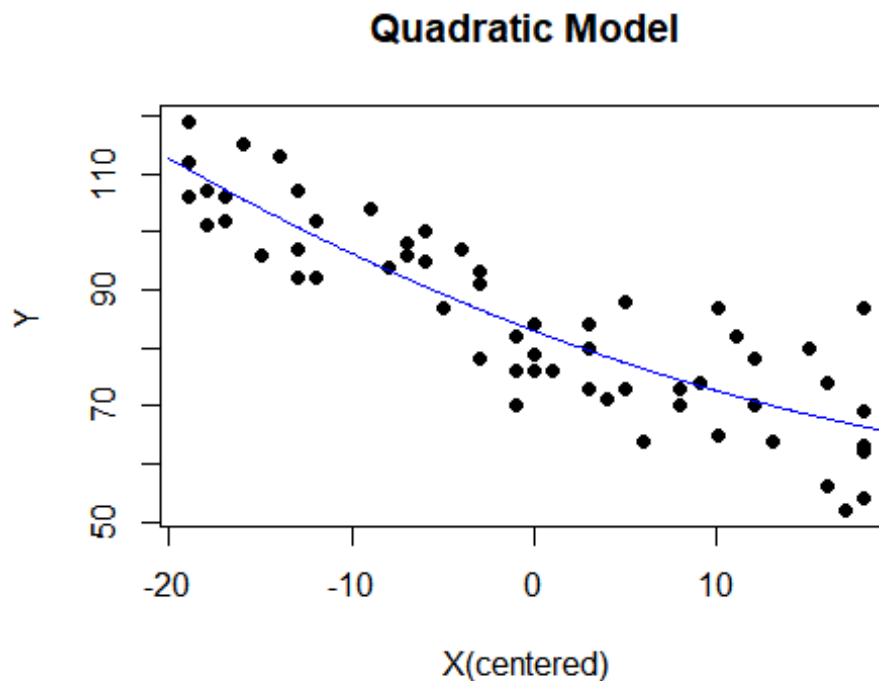
Fit the regression model 8.2 $\hat{Y} = 82.93575 - 1.18396x + 0.01484x^2$

plot the fitted regression function and data

```

plot(mmdata$t,mmdata$Y,main="Quadratic
Model",xlab="X(centered)",ylab="Y",pch=19)
x=seq(-20,20,by=.1)
y=82.93575-1.18396*x+0.01484*x^2
lines(x,y,col="blue")

```



appears to be a good fit?

Yes it is a good fit

Find R^2 $R^2 = 0.7632$

Does the quadratic

summary

```
summary(mdata.reg)
```

```
##
## Call:
## lm(formula = Y ~ t + t_power, data = mdata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -15.086  -6.154  -1.088   6.220  20.578
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 82.935749   1.543146  53.745  <2e-16 ***
## t           -1.183958   0.088633 -13.358  <2e-16 ***
## t_power      0.014840   0.008357   1.776   0.0811 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.026 on 57 degrees of freedom
## Multiple R-squared:  0.7632, Adjusted R-squared:  0.7549
## F-statistic: 91.84 on 2 and 57 DF,  p-value: < 2.2e-16
```

Test whether or not there is a regression relation; use $\alpha = .05$. State the alternatives, decision rule, and conclusion.

Hypothesis $H_0 : \beta_1 = \beta_{11} = 0$, H_a : not both β_1 and $\beta_{11} = 0$. Decision rule, and conclusion: - we reject H_0 If p_value is $< \alpha$

conclusion , we reject the H_0 in this case because p_value ($2.2e-16$) is less than 0.05. there is sufficient evidence to indicate that a regression relation exist between muscle mass and the centered age variable and its squared value .

#-Estimate the mean muscle mass for women aged 48 years; use a 95 percent confidence interval. Interpret your interval.

```
mdata.reg2 <- lm(Y~X, data = mdata)
predict(mdata.reg2, list(X=48), se.fit=T, interval="confidence", level=0.95)

## $fit
##      fit      lwr      upr
## 1 99.22678 96.20318 102.2504
##
## $se.fit
## [1] 1.510501
##
## $df
## [1] 58
```

```
##
## $residual.scale
## [1] 8.173177
```

From the out above , the std.err. for \hat{Y} (mean) is 1.510501. The 95% CI for \bar{Y} is $\hat{Y} \pm t(df=45-2, 1-\alpha/2) \times se(\hat{y}_{mean})$. Putting the values to the formula the 90% CI is:

```
tcrit<-qt(0.975,df=57)
c(99.22678-tcrit*1.510501,99.22678+tcrit*1.510501)
## [1] 96.20205 102.25151
```

At confidence level of 95%, the estimate mean muscle mass for women aged 48 years is $96.20205 \leq E(\hat{Y}_h) \leq 102.25151$

#d Predict the muscle mass for a woman whose age is 48 years; use a 95 percent prediction interval. Interpret your interval.

```
mmdata.reg2<- lm(Y~X,data = mmdata)
predict(mmdata.reg2,list(X=48),se.fit =T,interval ="prediction",level=0.95)
## $fit
##      fit      lwr      upr
## 1 99.22678 82.58934 115.8642
##
## $se.fit
## [1] 1.510501
##
## $df
## [1] 58
##
## $residual.scale
## [1] 8.173177
```

From the output above $\hat{Y}_{mean}=99.22678$ and the std.err for \hat{y}_{mean} is 1.510501. to compute the standard error of the individual prediction

$se(\hat{y}_{mean}) = \sqrt{(se(\hat{y}_{mean}))^2 + MSE}$

```
se_EY_new <- (sqrt((1.510501)^2+(8.173177)^2))
se_EY_new
## [1] 8.311584
```

so, by inserting the values to the formula at 95% PI is;

```
tcrit<-qt(0.975,df=57)
c(99.22678-tcrit*8.311584,99.22678+tcrit*8.311584)
## [1] 82.58312 115.87044
```

At confidence level of 95%, the predic mass for a woman aged 48 years is

$$82.58312 \leq E(Y_h) \leq 115.87044$$

e) Test whether the quadratic term can be dropped from the regression model; use $\alpha = .05$. State the alternatives, decision rule, and conclusion.

Full regression model: $\hat{Y} = 82.93575 - 1.18396x + 0.01484x^2$

Hypothesis

$$H_0: \beta_1 = 0 \quad H_a: \beta_1 \neq 0$$

Reduced model

$$\hat{y} = 82.93575 - 1.18396x + 0x^2$$

$$\hat{y} = 82.93575 - 1.18396x$$

Partial F-test

```
mmdata.full.reg=lm(Y~t+t_power,data =mmdata)
mmdata.reduced.reg<- lm(Y~t,data = mmdata)
anova(mmdata.reduced.reg,mmdata.full.reg)

## Analysis of Variance Table
##
## Model 1: Y ~ t
## Model 2: Y ~ t + t_power
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1      58 3874.4
## 2      57 3671.3  1    203.13 3.1538 0.08109 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

We obtain our SSE from the anova table output

$$\text{THE F-PARTIAL} = (\text{SSE(R)} - \text{SSE(F)} / (\text{dfe(R)} - \text{dfe(f)})) / (\text{SSE(F)} / \text{dfe(F)})$$

$$= (203.13) / ((58 - 57) / 3671.3 / 57)$$

$$= 3.153763$$

$$F_{\text{partial}} <- ((203.13) / (1) / (3671.3 / 57))$$

F_{partial}

```
## [1] 3.153763
```

The P-value calculation

```
p_value<-1-pf(3.153763,1,57)
p_value
```

```
## [1] 0.08108997
```

```
p-value =P(F(df1=dfe(R)-dfe(F),df2=dfe(F))) >Fpartial)
=P(F(df1=1,df2=57)>3.153763)=0.08108997
```

Decision rule

Reject H_0 if $p\text{-value} < \alpha(0.025)$

Statistical conclusion

since $p\text{-value}$ (0.08108997) is greater than α (0.025), we do not reject H_0 . Therefore, we do not have significant evidence to support that x^2 is needed in the model, so x^2 can be dropped from the model when x is in the model.