

Regression and Time Series (MA60056) - Spring 2020

ASSIGNMENT 1

Suman Pal Roll no. 19BM6JP22 (PGDBA)

Collect the data as discussed in the class. Y= Height, X= length of palm of hand as shown in 2

```
In [1]: data = read.csv("D:/Suman/PGDBA/IIT Kgp/RTSM/RTSM_data.csv")
```

```
In [2]: myx = data$Palm.Length..cms.[data$Roll.Number == "19BM6JP22" ]
myy = data$Height..cms.[data$Roll.Number == "19BM6JP22" ]
mydata = as.data.frame(cbind(myx,myy))
colnames(mydata) = c("X", "Y")
Y = data$Height..cms.[!data$Roll.Number == "19BM6JP22" ]
X = data$Palm.Length..cms.[!data$Roll.Number == "19BM6JP22" ]
```

(1) Fit a simple liner regression model for Y on X [don't on your own observation]

```
In [3]: model = lm(Y~X)
plot(X,Y, pch = 16, col = "blue")
abline(model)
summary(model)
```

Call:

lm(formula = Y ~ X)

Residuals:

Min	1Q	Median	3Q	Max
-13.1257	-4.5391	-0.7559	4.3743	16.5576

Coefficients:

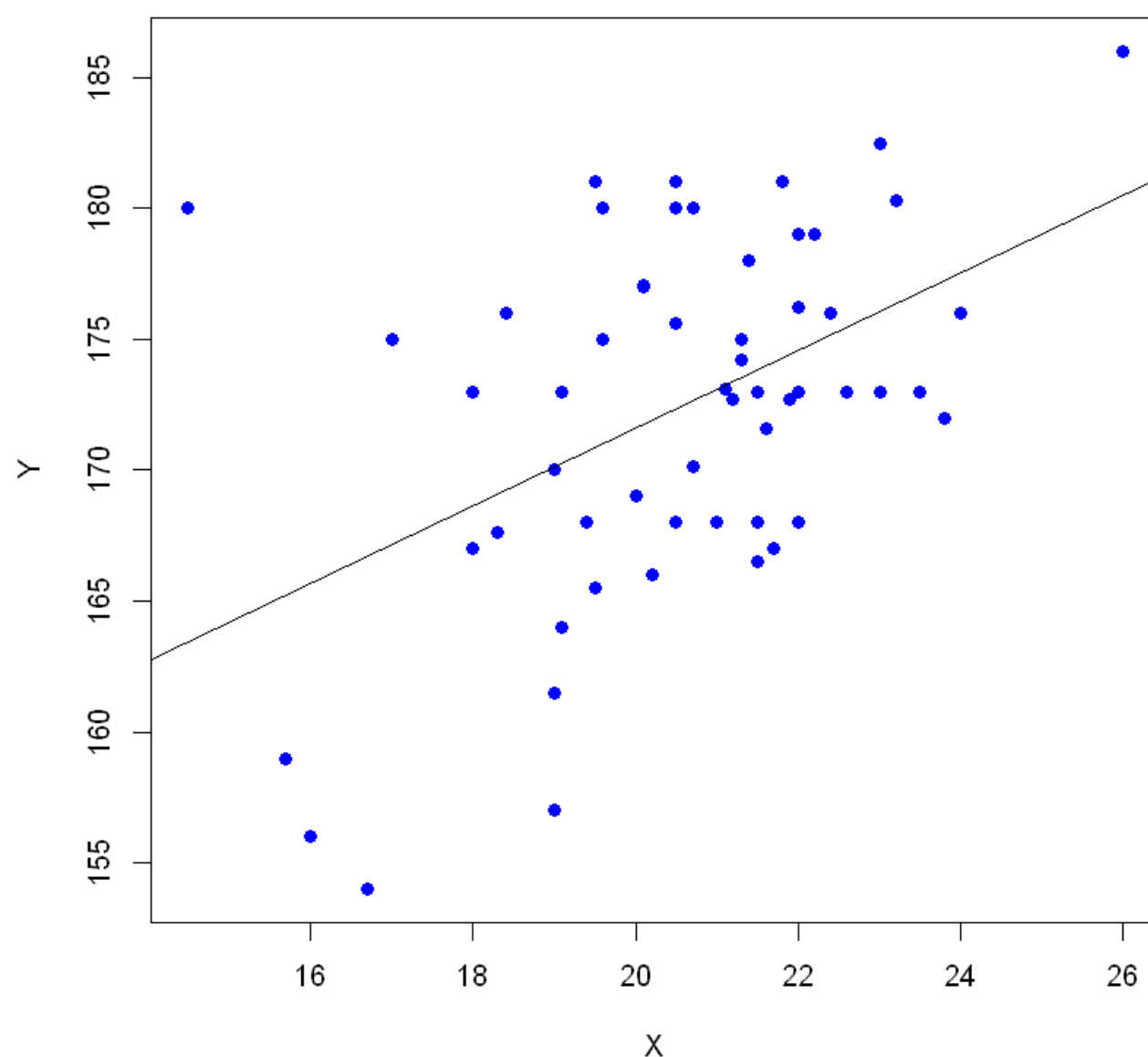
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	141.9073	7.8562	18.063	< 2e-16 ***
X	1.4852	0.3806	3.902	0.000266 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 6.119 on 54 degrees of freedom

Multiple R-squared: 0.22, Adjusted R-squared: 0.2055

F-statistic: 15.23 on 1 and 54 DF, p-value: 0.0002662



Linear Regression line Y on X : $141.9073 + 1.4852 \cdot X$

(2) Predict your height when your palm length is known.

```
In [4]: predict(model, newdata = mydata ,interval = "confidence" ) #95% CI
```

fit	lwr	upr
172.3535	170.714	173.993

```
In [6]: predict(model, newdata = mydata, interval = 'prediction') #95% Prediction Interval
```

fit	lwr	upr
172.3535	159.9766	184.7304

3) Test for Ho : Bo =0 vs Bo != 0 at level 0.05

```
In [7]: print(paste("Intercept value = " , model$coefficients[1]))
print(paste("Standard Error of intercept = " , summary(model)$coefficients[1,2]))
print(paste("t-value = " ,summary(model)$coefficients[1,3]))
print(paste("p-value =",summary(model)$coefficients[1,4] ))
```

[1] "Intercept value = 141.90731264563"
[1] "Standard Error of intercept = 7.85617522367301"
[1] "t-value = 18.0631552384449"
[1] "p-value = 1.50531361840329e-24"

Ans: : From the model,
Bo (intercept) = 141.9073.
SE of intercept = 7.8561
Under Ho, t statistic = 141.9073/7.8561 = 18.0631 ~ t distribution with df = n-2 = 54
p-value = 1.5e-24.
Since p-value < 0.05, we reject the Null Hypothesis Bo = 0.

4) Test for Ho : B1 =8 vs B1 != 8 at level 0.05

```
In [10]: print(paste("Slope value = " , model$coefficients[2]))
print(paste("Standard Error of slope = " , summary(model)$coefficients[2,2]))
print(paste("t-value = " ,(model$coefficients[2]-8)/(summary(model)$coefficients[2,2])))
print(paste("p-value =",pt((model$coefficients[2]-8)/(summary(model)$coefficients[2,2] ) , 54) ))
```

[1] "Slope value = 1.48517917008326"
[1] "Standard Error of slope = 0.380583082357847"
[1] "t-value = -17.1179990175998"
[1] "p-value = 8.97167196799356e-24"

Ans: : From the model,
B1 (slope) = 1.4851
SE of slope = 0.3805
Under Ho, t statistic = (1.4851 - 8)/0.3805 = -17.1179 ~ t distribution with df = n-2 = 54
p-value = 8.97e-24
Since p-value < 0.05, we reject the Null Hypothesis slope = 8.

5) Consider (X,Y) follows bivariate normal distribution. Find the regression of Y on X

Ans: When (X,Y) follows bivariate normal distribution, then
regression of Y on X = $E(Y|X = x) = (\mu_y) + (\text{correlation}(S_y/S_x))(x - \mu_x)$

```
In [11]: print(paste("μy = " , mean(Y)))
print(paste("μx = " , mean(X)))
print(paste("Sy = " , sqrt(var(Y))))
print(paste("Sx = " , sqrt(var(X))))
print(paste("correlation = " , cor(X,Y)))
```

[1] "μy = 172.398571428571"
[1] "μx = 20.5303571428571"
[1] "Sy = 6.86498663934909"
[1] "Sx = 2.16794159958722"
[1] "correlation = 0.469014999564399"

Thus Regression of Y on X = $E(Y|X=x) = 172.3985 + 0.4690(6.8649/2.1679)(x - 20.5303)$
or **172.3985 + 1.4851*(x-20.5303)**

Prediction on my data for x= 22.4 :

```
In [12]: 172.3985 + 1.4851*( mydata$X - 20.5303)
```

172.35350147

6) Consider L1 norm and fit a simple linear regression model for Y on X.

```
In [13]: library(L1pack)
l1model = l1fit(X, Y, intercept = TRUE, tolerance = 1e-07, print.it = TRUE)
l1model$coefficients
```

Intercept	135.583450317383
X	1.75172352790833

Regression model for Y on X using L1 norm : 135.5834 + 1.7517*X

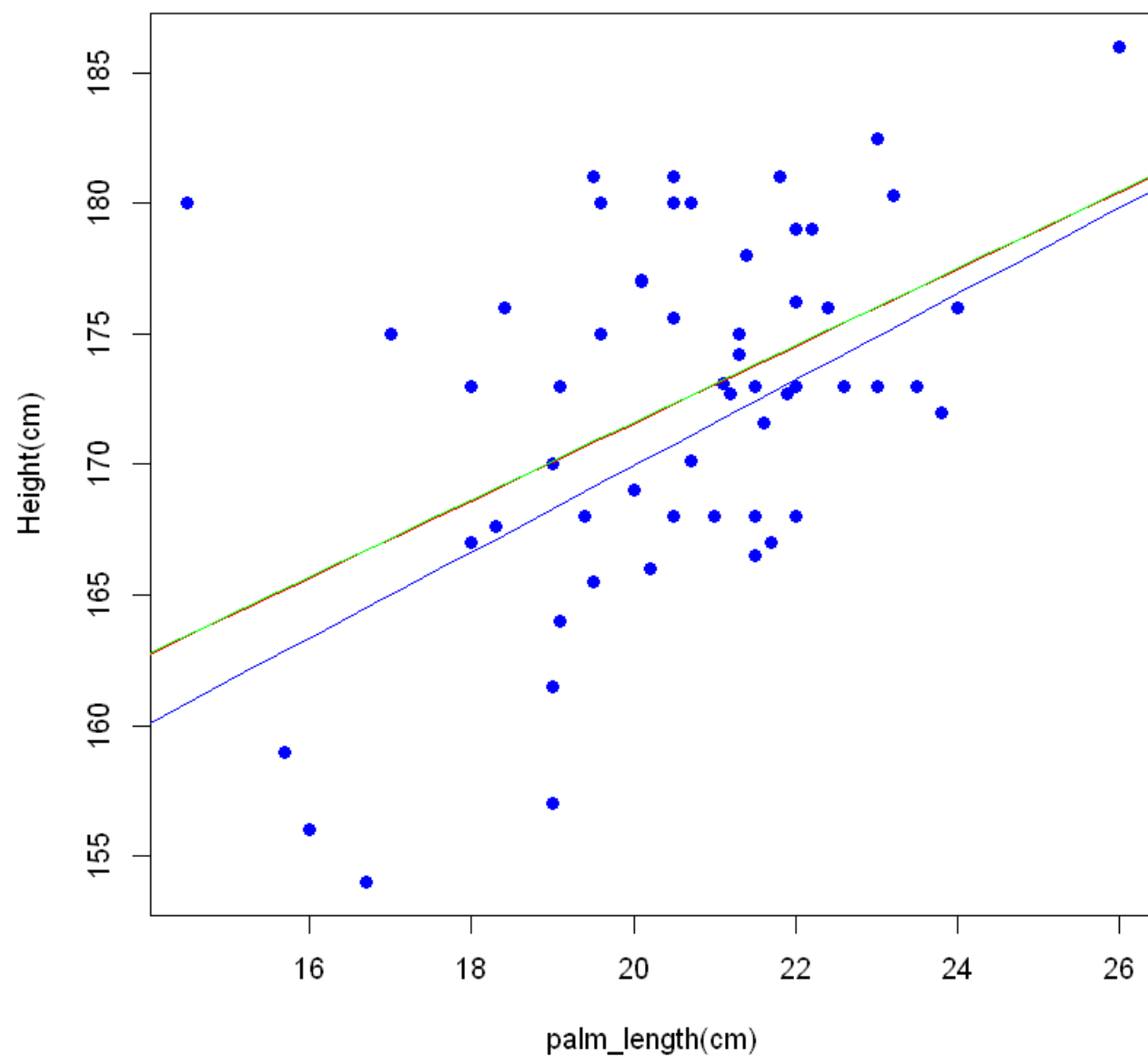
Prediction on my data for x= 22.4 :

```
In [14]: 135.5834 + 1.7517*22.4
```

174.82148

(7) Plot the there regression lines obtained in 1,5 and 6 in a diagram.

```
In [15]: plot(X,Y, pch = 16, col = "blue",xlab = "palm_length(cm)" , ylab = "Height(cm)")
lines(seq(0,30,0.4) , 141.9744 + 1.4790*(seq(0,30,0.4)),type = "l", col = "red" ) #Q1
lines(seq(0,30,0.4) , 142.04+ 1.478*(seq(0,30,0.4)),type = "l", col = "green") #Q5
lines(seq(0,30,0.4) , 136.9875 + 1.6479*(seq(0,30,0.4)),type = "l", col = "blue") #Q6
```



Blue Line is line obtained by fitting regression line using L1 norm.

Red Line is least squares line.

Green Line is obtained by bivariate normal distribution.