

Assignment 3

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10/20/2022

```
(echo = TRUE)
```

```
## [1] TRUE
```

```
T1 <- read.table("C:/Users/dgmur/Downloads/T1-9.dat")
colnames(T1) <- c("Country", "100m", "200m", "400m", "800m", "1500m", "3000m", "Marathon")
#A
round(apply(T1[, -1], 2, mean), 2)
```

```
##      100m      200m      400m      800m      1500m      3000m Marathon
##      11.36      23.12      51.99       2.02       4.19       9.08      153.62
```

```
round(cov(T1[, -1]), 2)
```

```
##           100m  200m  400m  800m  1500m  3000m Marathon
## 100m      0.16  0.34  0.89  0.03  0.08  0.23      4.33
## 200m      0.34  0.86  2.19  0.07  0.20  0.55     10.38
## 400m      0.89  2.19  6.75  0.18  0.51  1.43     28.90
## 800m      0.03  0.07  0.18  0.01  0.02  0.06      1.22
## 1500m     0.08  0.20  0.51  0.02  0.07  0.22      3.54
## 3000m     0.23  0.55  1.43  0.06  0.22  0.66     10.71
## Marathon 4.33 10.38 28.90 1.22  3.54 10.71     270.27
```

```
round(cor(T1[, -1]), 2)
```

```
##           100m  200m  400m  800m  1500m  3000m Marathon
## 100m      1.00  0.94  0.87  0.81  0.78  0.73      0.67
## 200m      0.94  1.00  0.91  0.82  0.80  0.73      0.68
## 400m      0.87  0.91  1.00  0.81  0.72  0.67      0.68
## 800m      0.81  0.82  0.81  1.00  0.91  0.87      0.85
## 1500m     0.78  0.80  0.72  0.91  1.00  0.97      0.79
## 3000m     0.73  0.73  0.67  0.87  0.97  1.00      0.80
## Marathon 0.67 0.68 0.68 0.85  0.79 0.80      1.00
```

```
#B
track = as.matrix(T1[,-1])
colnames(track) <- c("100m", "200m", "400m", "800m", "1500m", "3000m", "Marathon")
V1 <- track[,1]/3+track[,2]/6+track[,3]/12
V2<- track[,4]/2.4+track[,5]/4.5+track[,6]/9
V3 <- track[,7]/60

V1[c(15,25,21,14)]
```

```
## [1] 12.28000 11.90250 12.69333 12.11000
```

```
V2[c(15,25,21,14)]
```

```
## [1] 2.978611 2.650833 2.970278 2.725000
```

```
V3[c(15,25,21,14)]
```

```
## [1] 2.774333 2.370500 2.855500 2.489000
```

```
#C

mean(V1)
```

```
## [1] 11.97144
```

```
mean(V2)
```

```
## [1] 2.782629
```

```
mean(V3)
```

```
## [1] 2.560321
```

```
var(cbind(V1,V2,V3))
```

```
##           V1           V2           V3
## V1 0.23678853 0.07004779 0.09307002
## V2 0.07004779 0.03350372 0.04140643
## V3 0.09307002 0.04140643 0.07507504
```

#D

```
xbar <- colMeans(track)
```

```
S <- cov(track)
```

```
x <- c(1/3,1/6,1/12,0,0,0,0)
```

```
y <- c(0,0,0,1/2.4,1/4.5,1/9,0)
```

```
z <- c(0,0,0,0,0,0,1/60)
```

```
meanV1<- x%*%xbar
```

```
meanV1
```

```
##           [,1]
```

```
## [1,] 11.97144
```

```
meanV2 <- y%*%xbar
```

```
meanV2
```

```
##           [,1]
```

```
## [1,] 2.782629
```

```
meanV3 <- z%*%xbar
```

```
meanV3
```

```
##           [,1]
```

```
## [1,] 2.560321
```

```
varV1 <- x%*%S%*%x
```

```
varV1
```

```
##           [,1]
```

```
## [1,] 0.2367885
```

```
varV2 <- y%*%S%*%y
```

```
varV2
```

```
##           [,1]
```

```
## [1,] 0.03350372
```

```
varV3 <- z%*%S%*%z
```

```
varV3
```

```
##           [,1]
```

```
## [1,] 0.07507504
```

```
covV1V2<- x%*%S%*%y
covV1V2
```

```
##           [,1]
## [1,] 0.07004779
```

```
covV1V3 <- x%*%S%*%z
covV1V3
```

```
##           [,1]
## [1,] 0.09307002
```

```
covV2V3 <- y%*%S%*%z
covV2V3
```

```
##           [,1]
## [1,] 0.04140643
```

```
X <- rbind(x,y,z)

MatMean<- X%*%xbar
MatMean
```

```
##           [,1]
## x 11.971435
## y  2.782629
## z  2.560321
```

```
MatCov <- X%*%S%*%t(X)
MatCov
```

```
##           x           y           z
## x 0.23678853 0.07004779 0.09307002
## y 0.07004779 0.03350372 0.04140643
## z 0.09307002 0.04140643 0.07507504
```

```
#E
#Showed the same results as part C.
```

```
bikes2 <- read.csv("C:/Users/dgmur/Downloads/SeoulBikes_F12022.csv")

dim(bikes2)
```

```
## [1] 730 16
```

```
View(bikes2)
bikes2 <- bikes2[,c(4:16)]
View(bikes2)
bikes2$Holiday<-factor(bikes2$Holiday)
bikes2$Seasons <- as.factor(bikes2$Seasons)
bikes2$Functioning.Day<- factor(bikes2$Functioning.Day)
bikes2$Time<- factor(bikes2$Time)
#A
lm1 <- lm(bikes2$RentedBikeCount~., data= bikes2)
summary(lm1)
```

```
##
## Call:
## lm(formula = bikes2$RentedBikeCount ~ ., data = bikes2)
##
## Residuals:
```

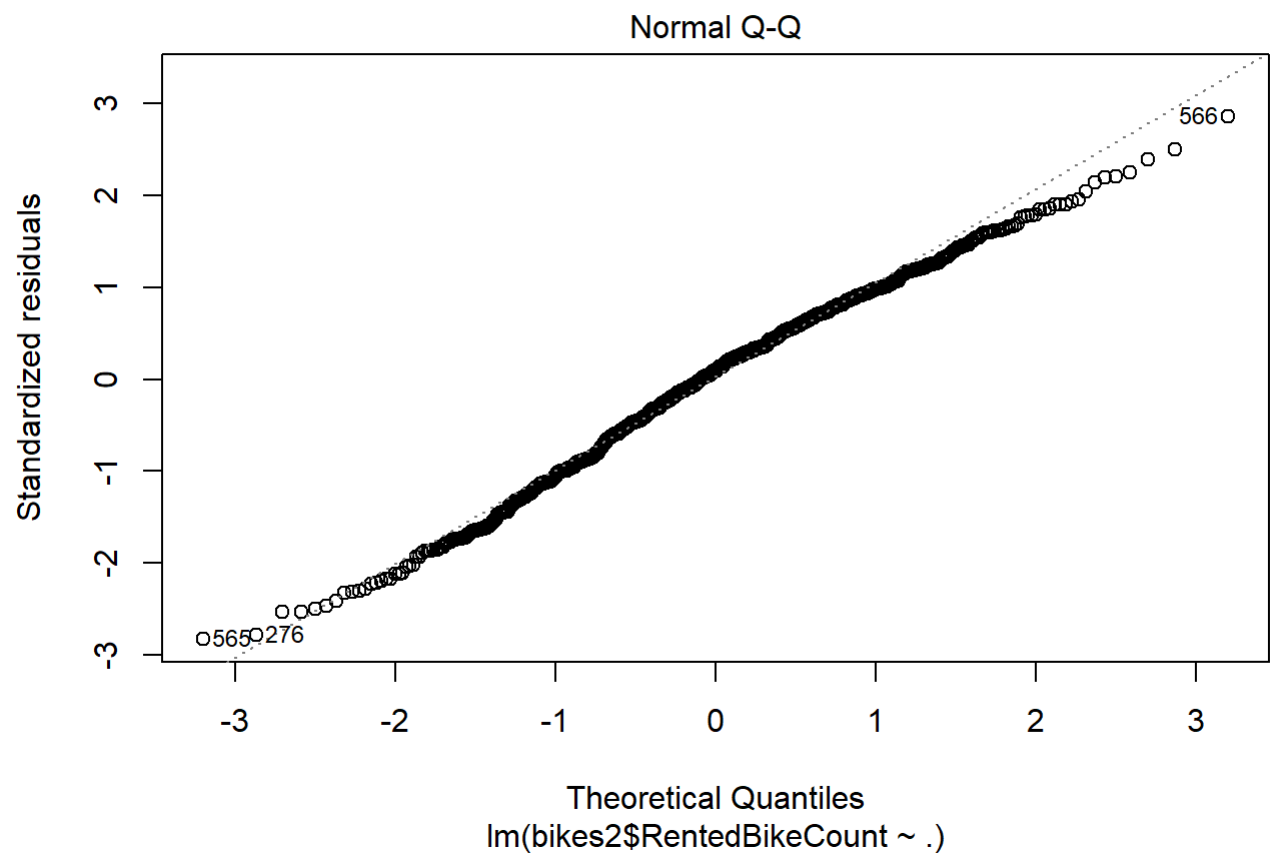
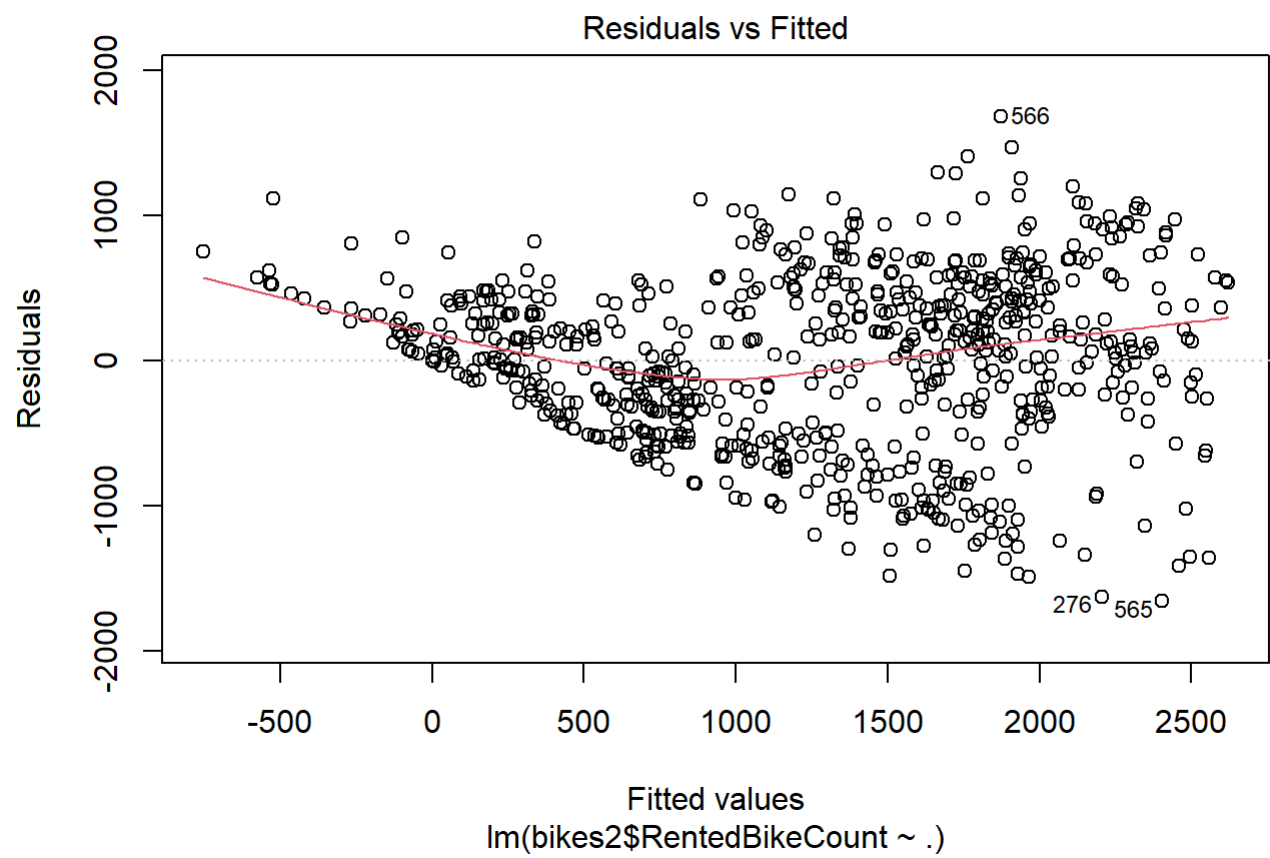
	Min	1Q	Median	3Q	Max
	-1657.94	-379.36	55.29	423.16	1682.70

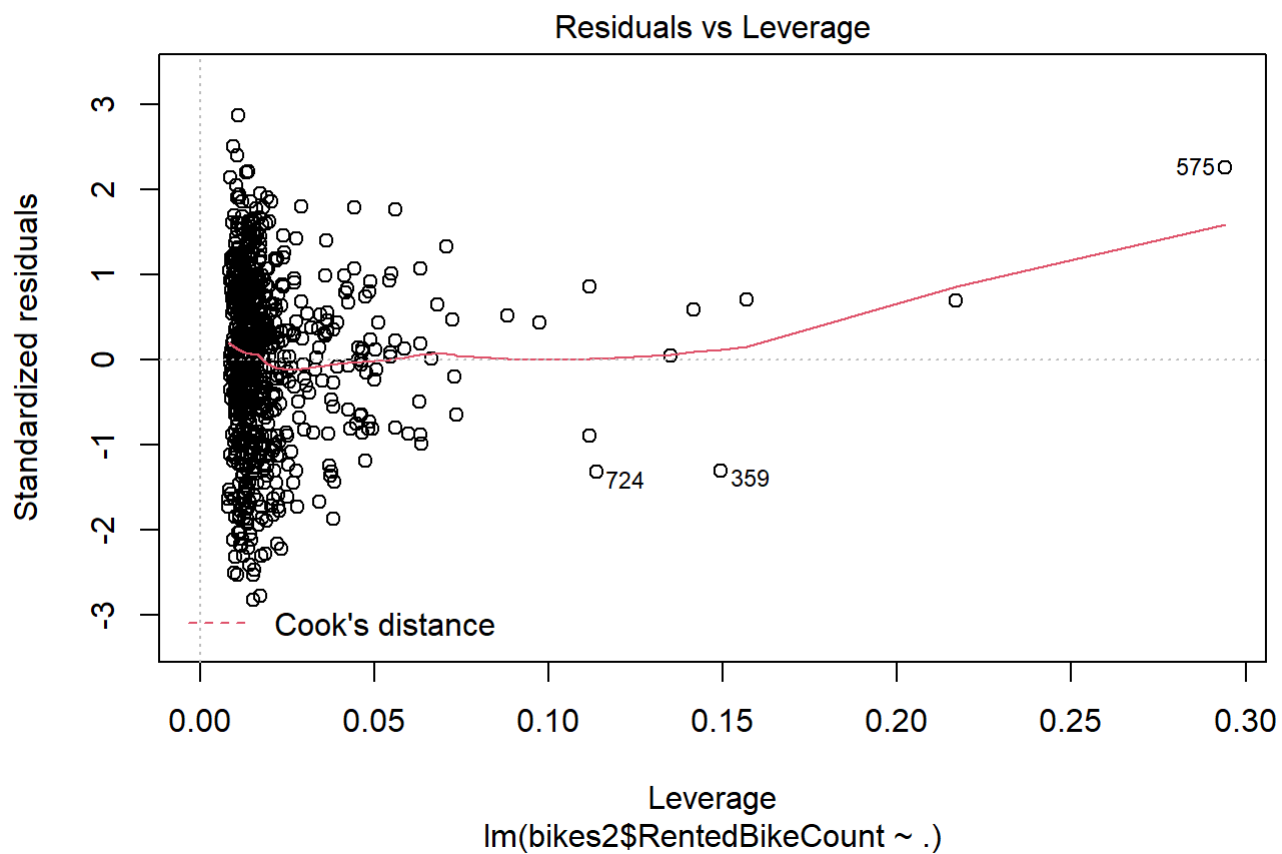
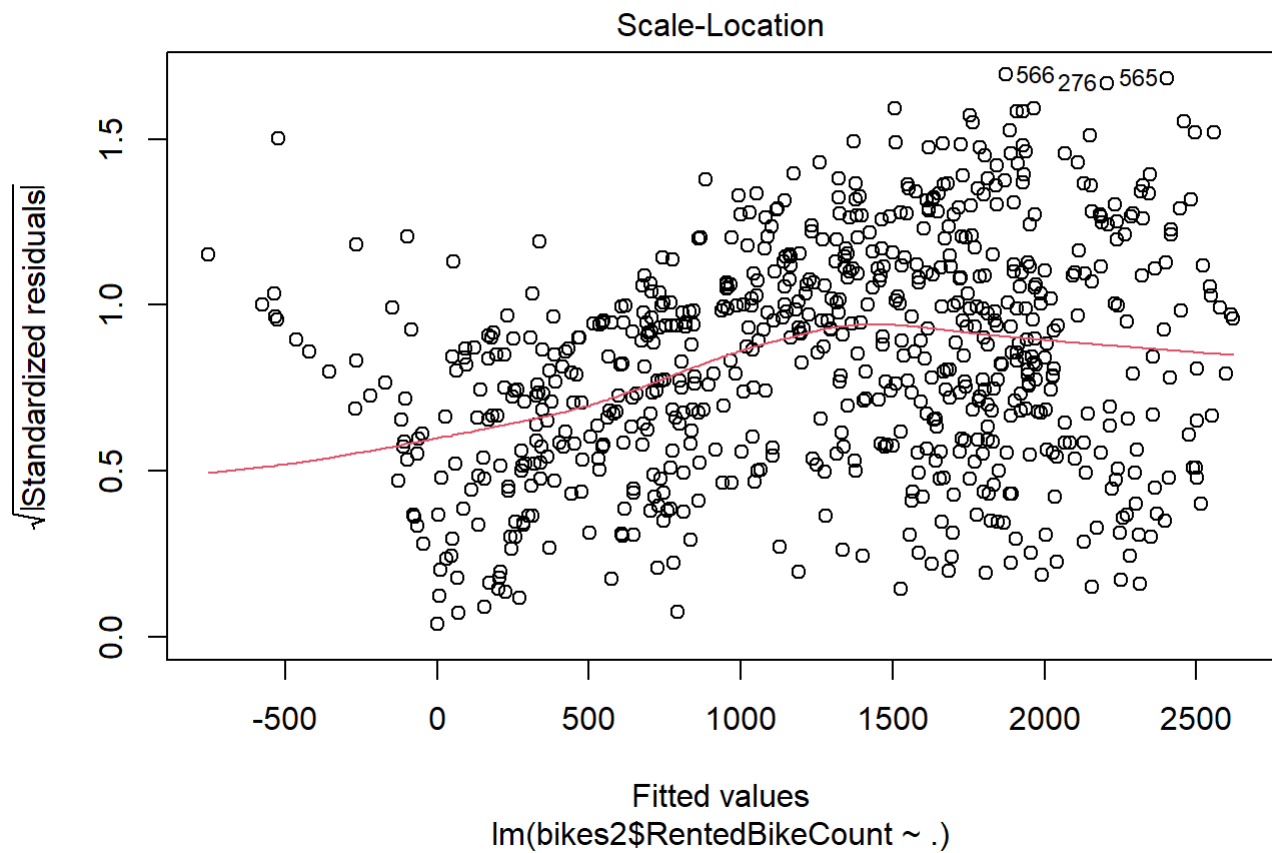
```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1266.52970	575.05965	2.202	0.027953	*
Temperature	-45.46624	21.00316	-2.165	0.030738	*
Humidity	-26.92340	6.22476	-4.325	1.74e-05	***
WindSpeed	13.02980	26.00716	0.501	0.616520	
Visibility	0.00403	0.04784	0.084	0.932890	
DewPointTemp	76.03379	22.57952	3.367	0.000799	***
SolarRadiation	550.55773	97.15453	5.667	2.11e-08	***
Rainfall	-74.51760	15.73064	-4.737	2.62e-06	***
Snowfall	4.51150	57.35246	0.079	0.937323	
SeasonsSpring	-470.07335	72.73163	-6.463	1.90e-10	***
SeasonsSummer	-519.76455	87.53538	-5.938	4.51e-09	***
SeasonsWinter	-781.32605	93.22146	-8.381	2.77e-16	***
HolidayNo Holiday	540.94438	102.04515	5.301	1.54e-07	***
Functioning.DayYes	1724.11080	127.14174	13.561	< 2e-16	***
TimeMorning	-237.60832	61.53045	-3.862	0.000123	***

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 590 on 715 degrees of freedom
## Multiple R-squared:  0.6112, Adjusted R-squared:  0.6035
## F-statistic: 80.27 on 14 and 715 DF,  p-value: < 2.2e-16
```

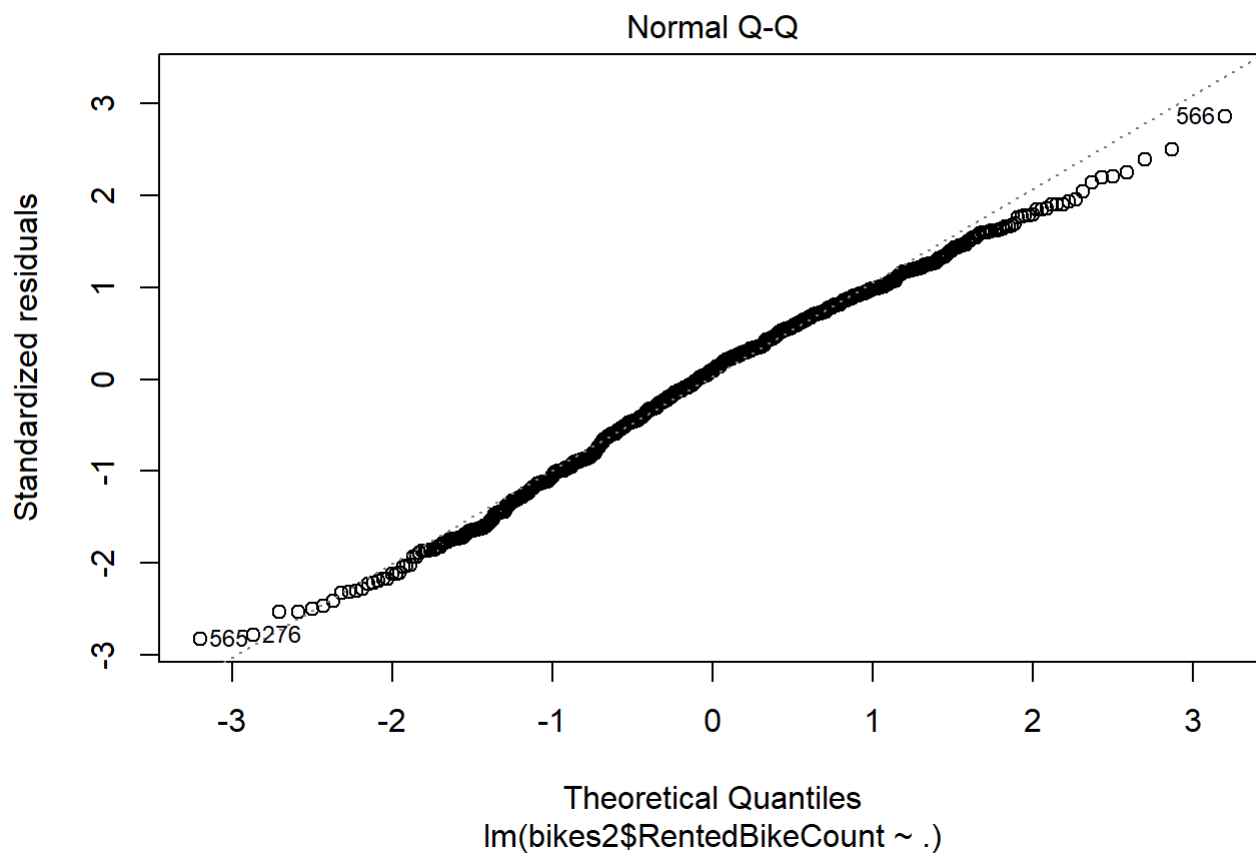
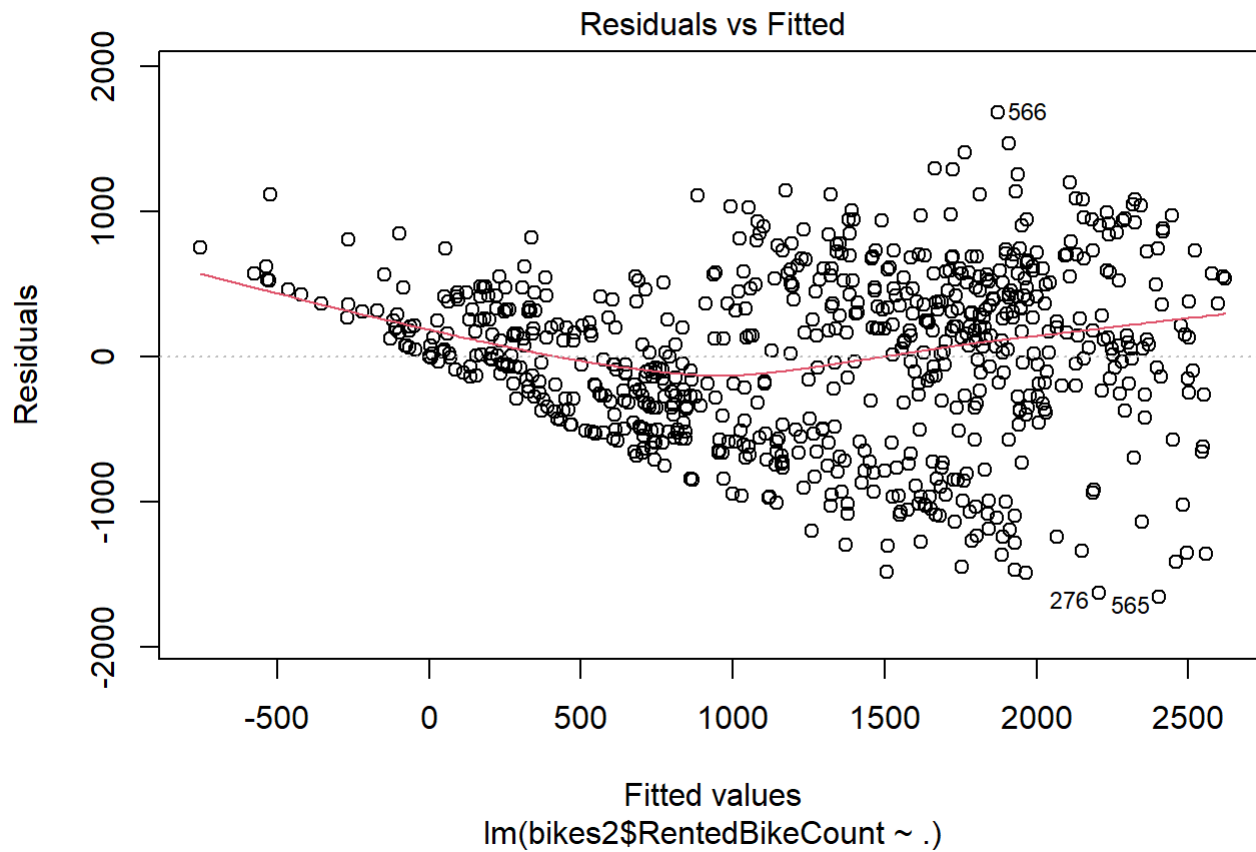
```
plot(lm1)
```

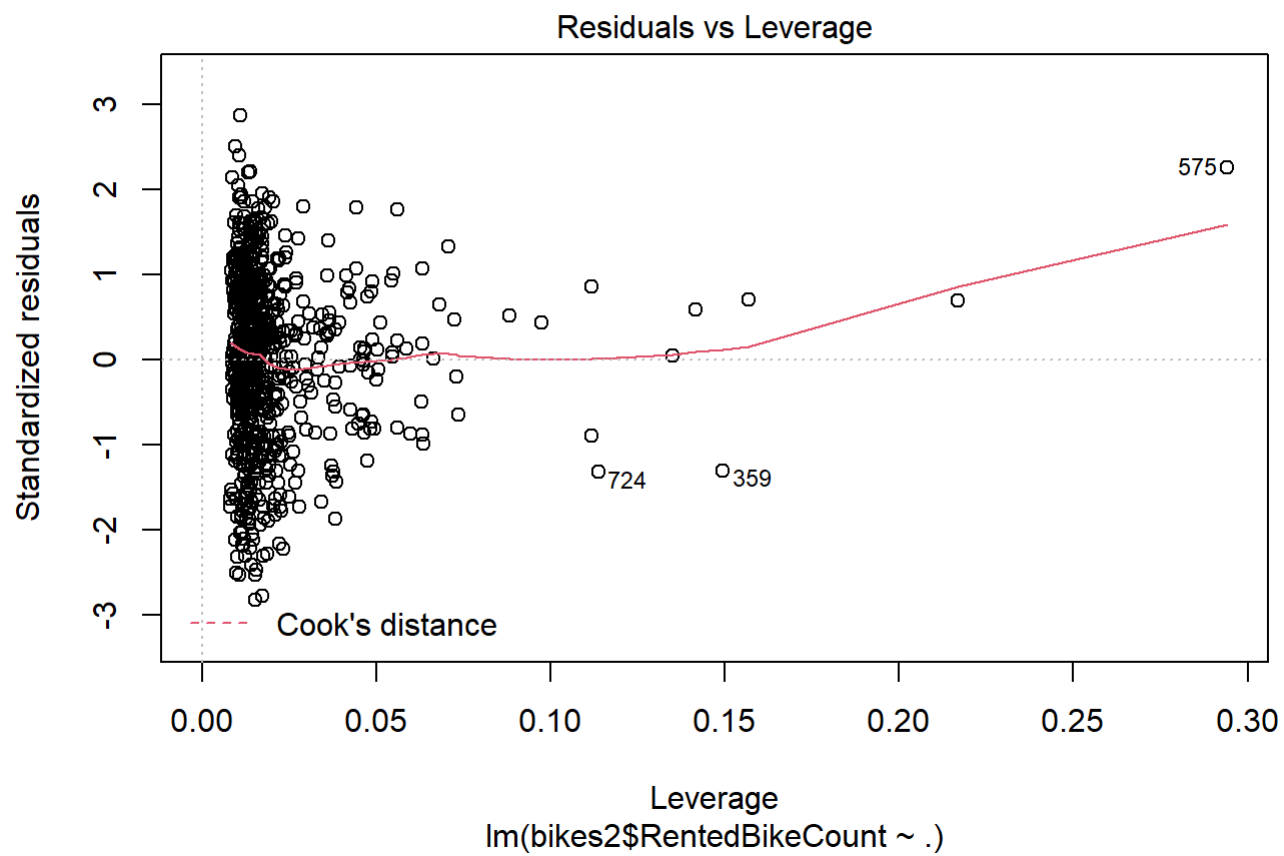
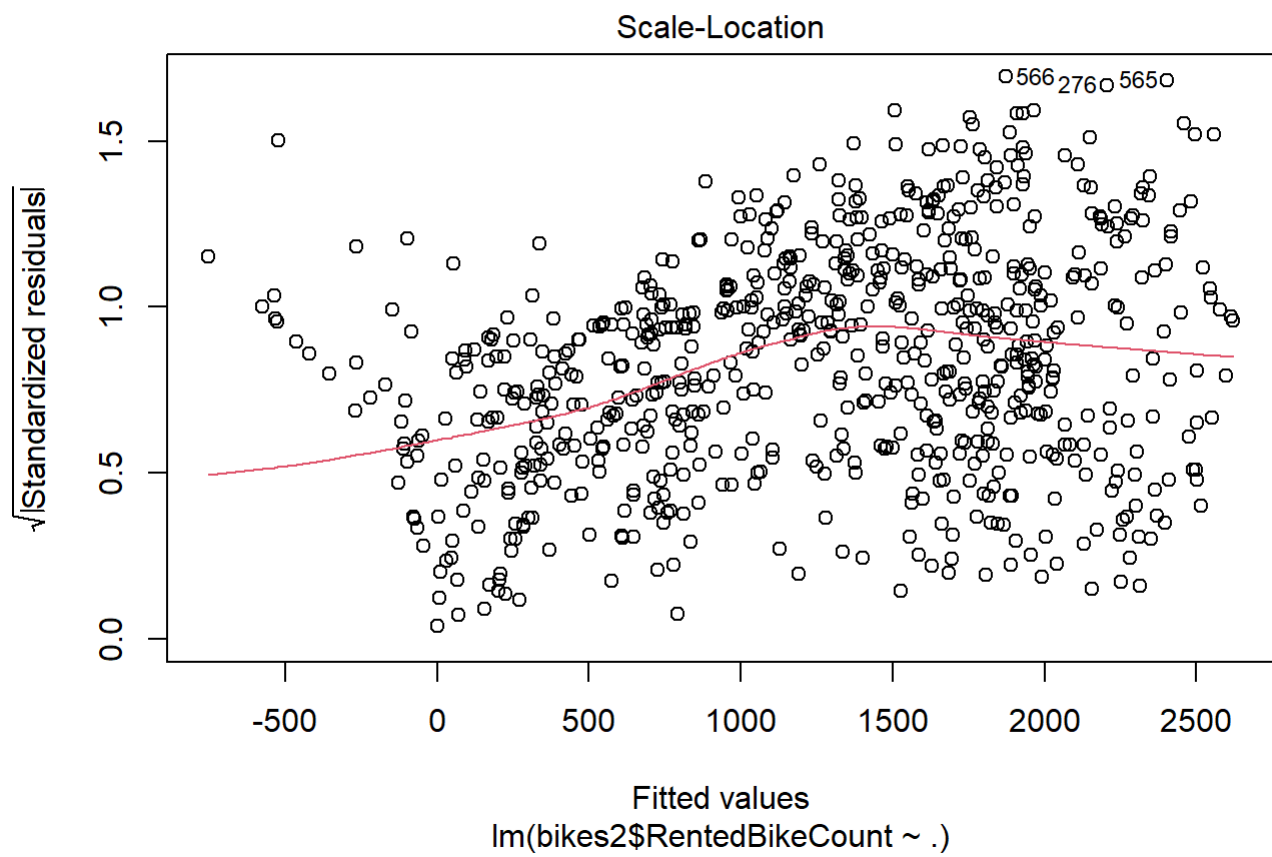





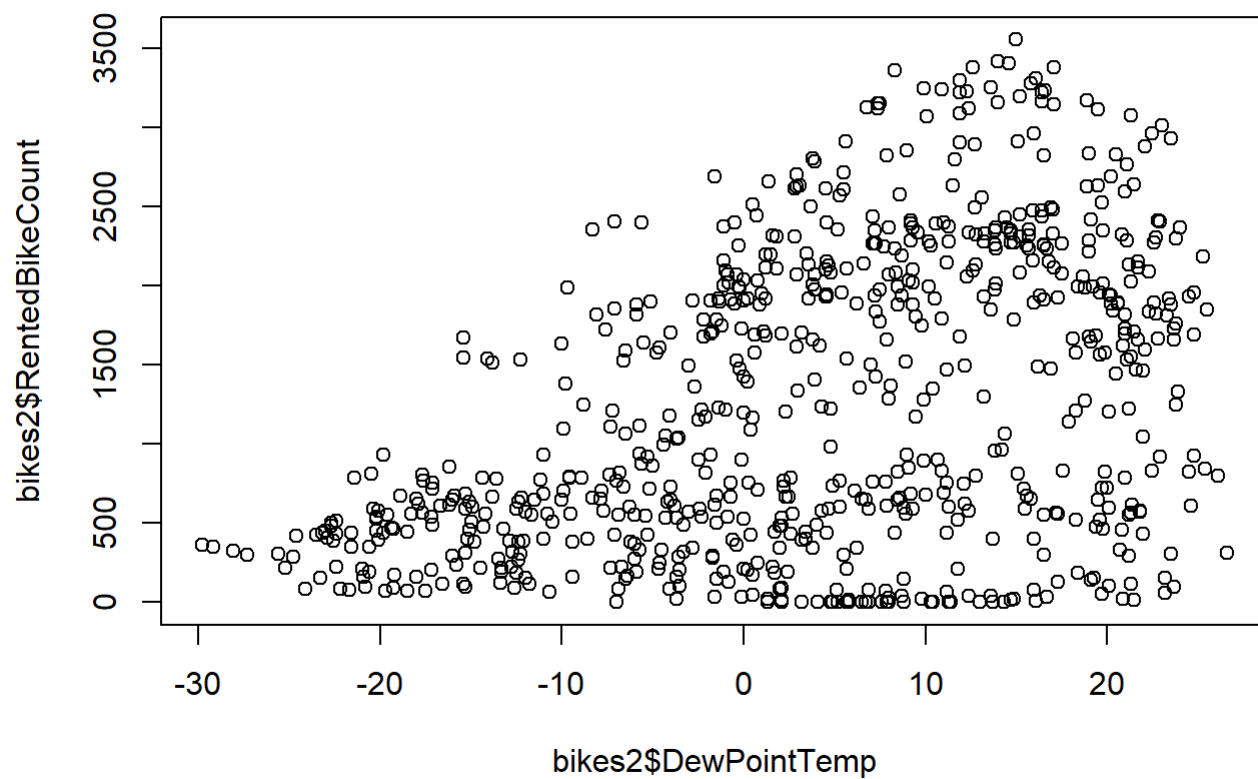
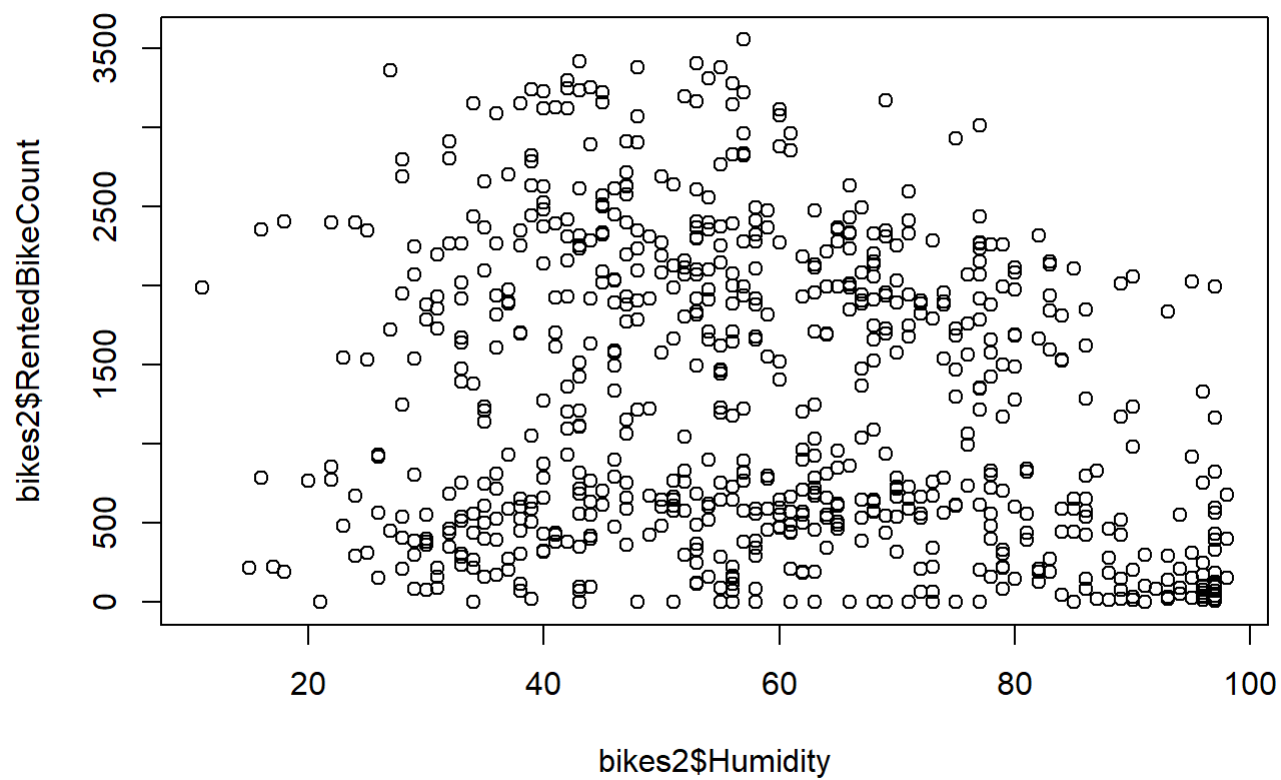
Residuals vs Fitted plot shows heteroscedasticity, qqplot steers away from the line once it hits between quartile 1 and 2. A log transformation on one of the predictor variables would fix this issue.

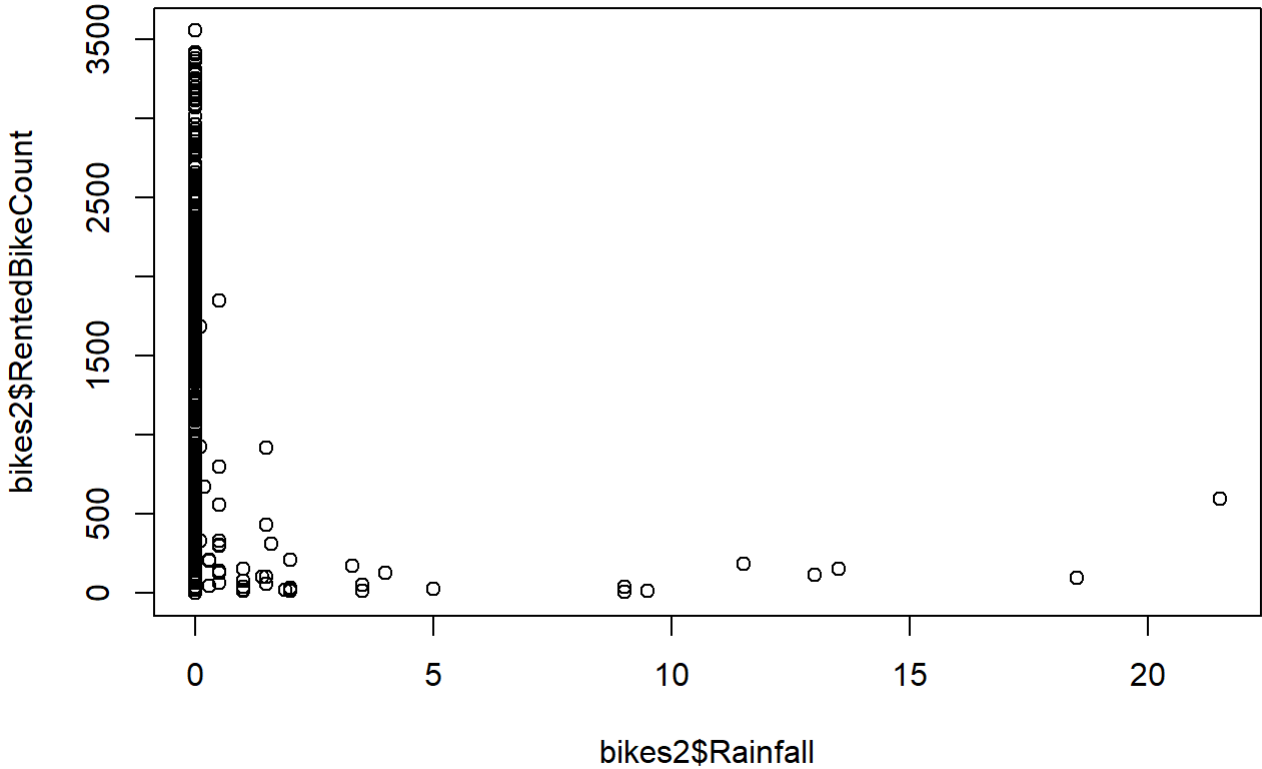
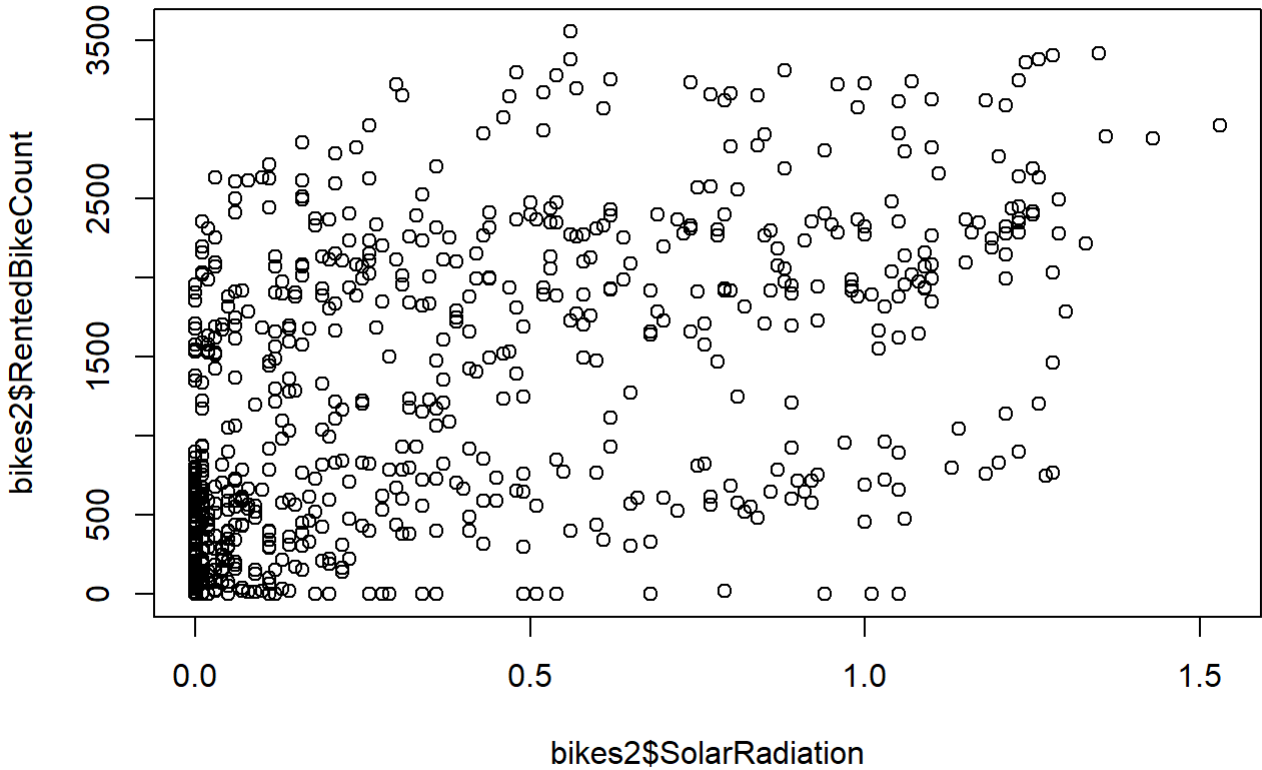
```
plot(lm1)
```

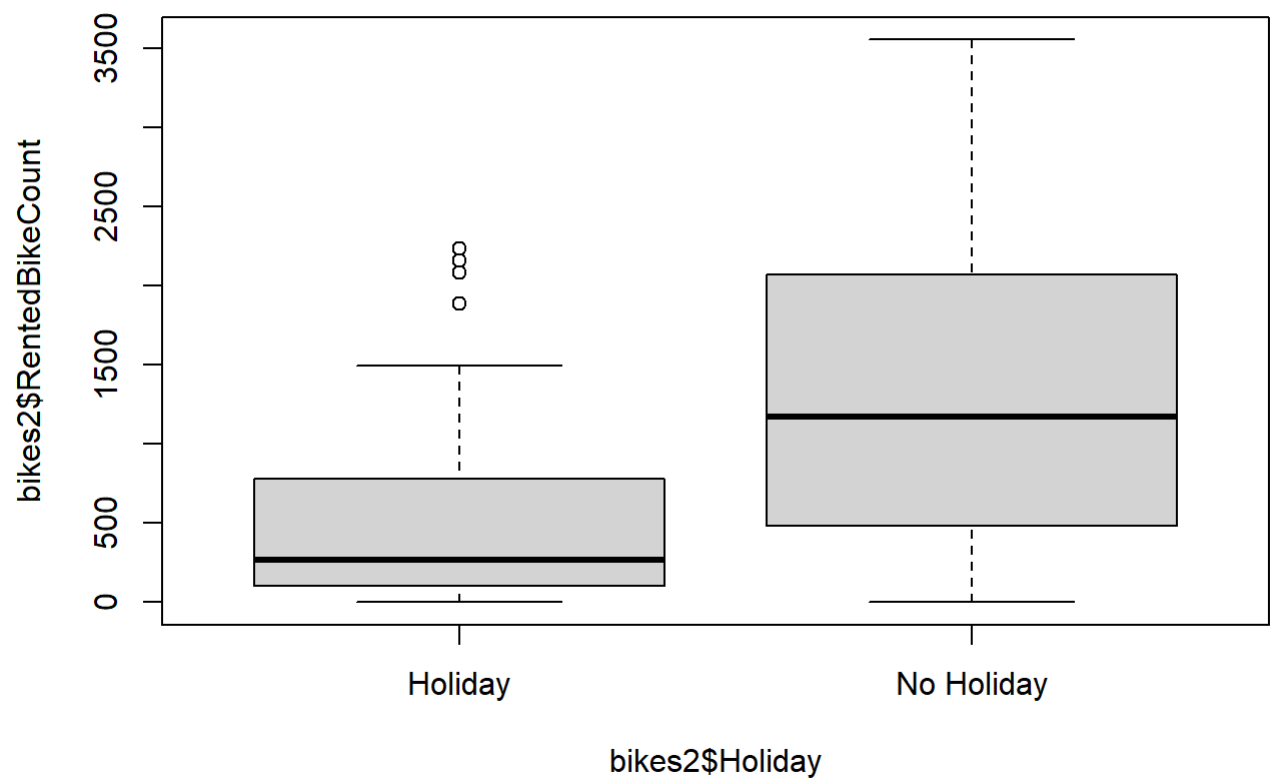
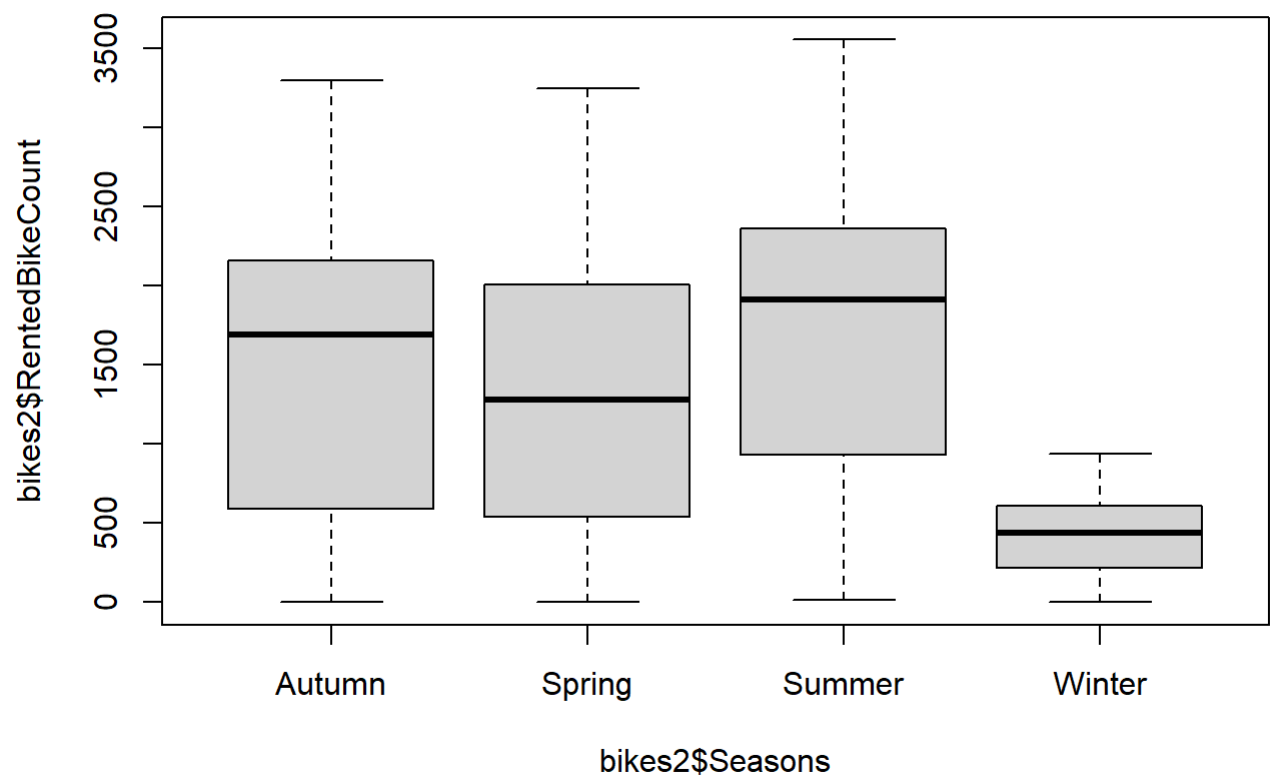


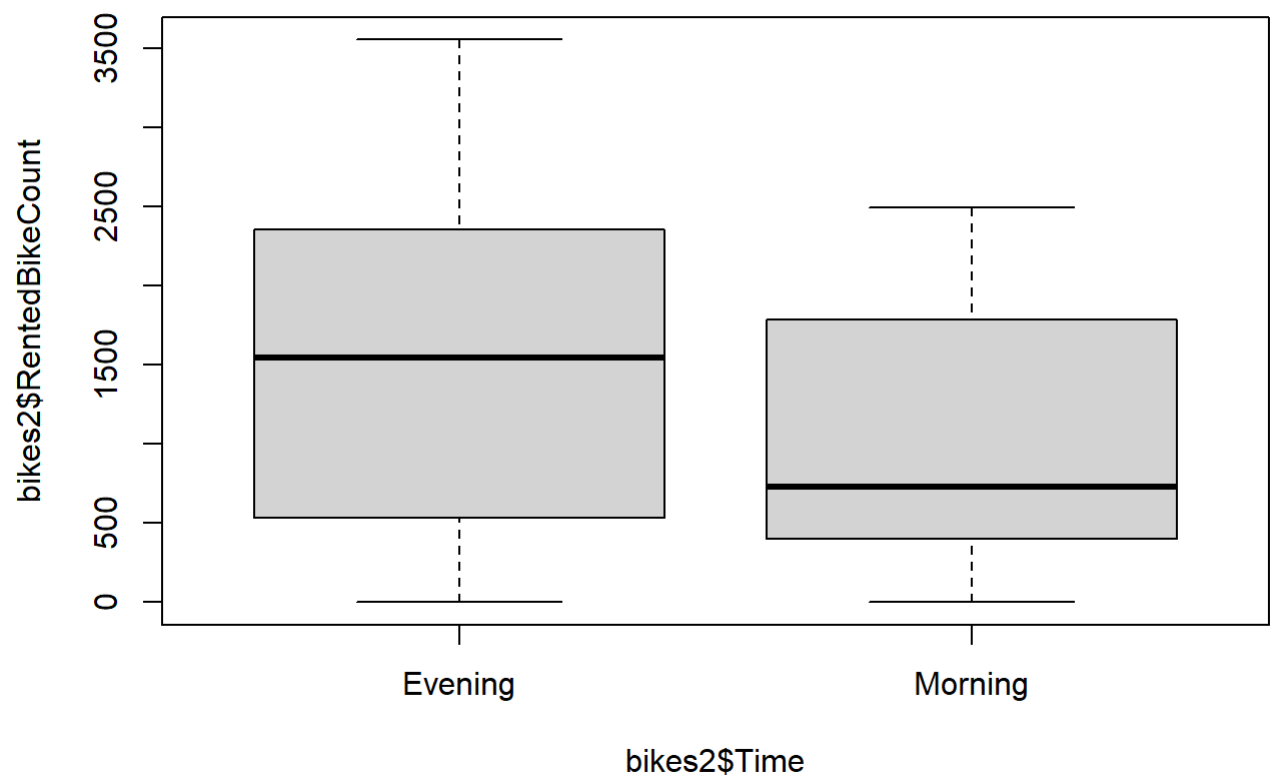
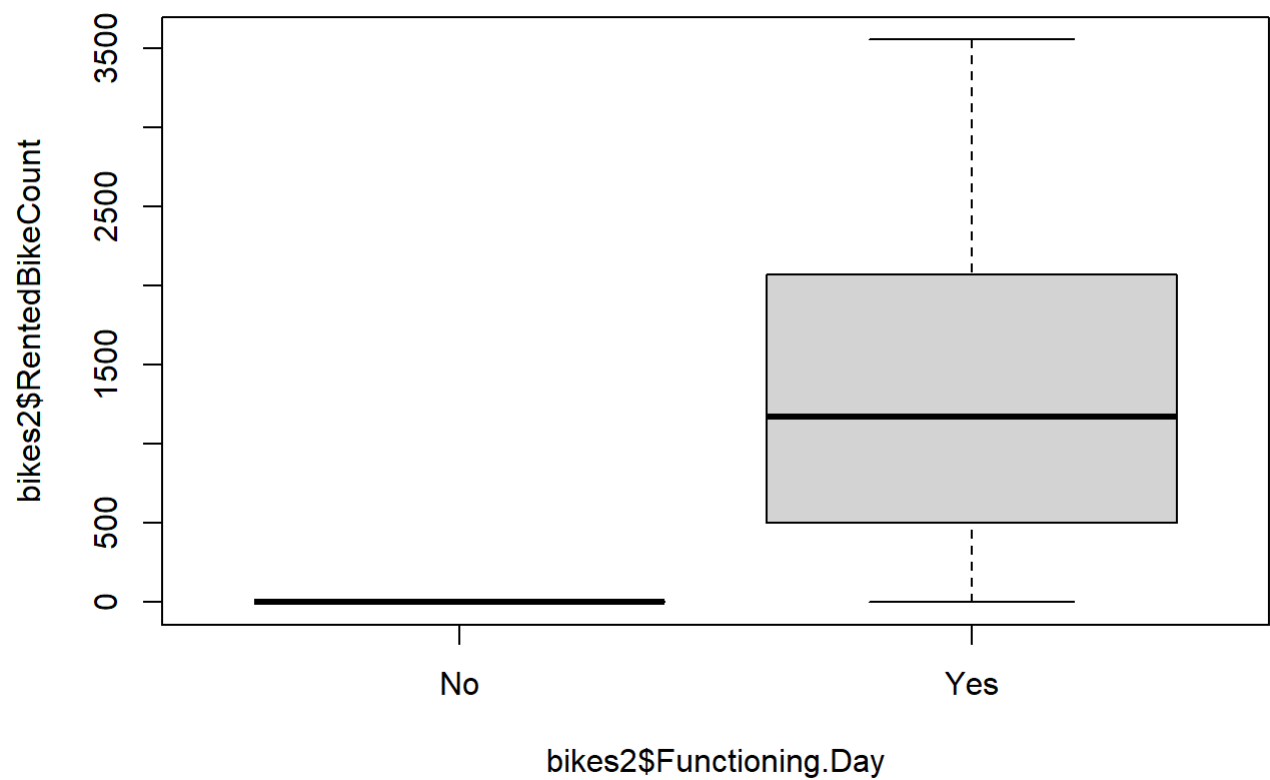


```
plot(bikes2$RentedBikeCount~bikes2$Humidity+bikes2$DewPointTemp+bikes2$SolarRadiation+bikes2$Rainfall+bikes2$Seasons  
      +bikes2$Holiday+bikes2$Functioning.Day+bikes2$Time, data = bikes2)
```

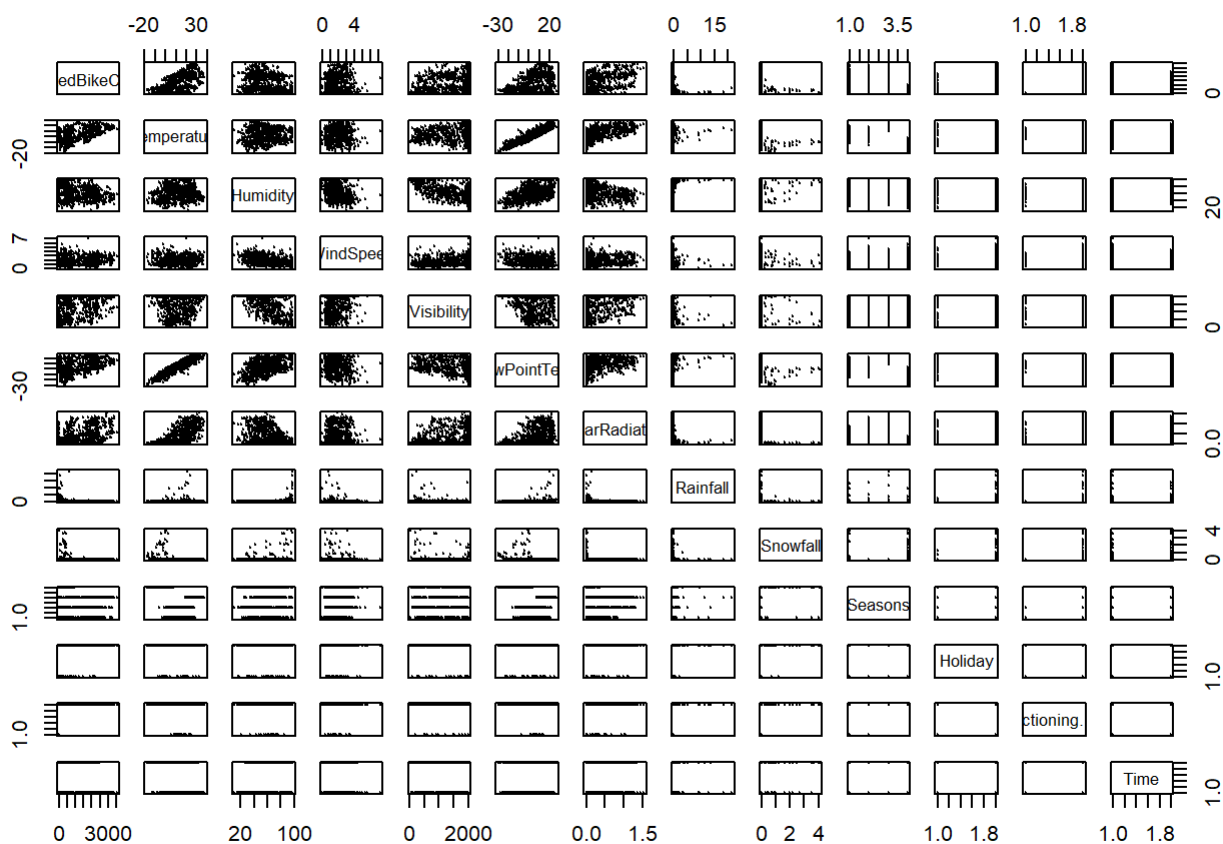








```
pairs(bikes2, labels = names(bikes2), cex = 0.1)
```



```
library(leaps)
```

```
## Warning: package 'leaps' was built under R version 4.1.3
```

```
regfit_full = regsubsets(bikes2$RentedBikeCount~., data = bikes2)
summary(regfit_full)
```

```
## Subset selection object
## Call: regsubsets.formula(bikes2$RentedBikeCount ~ ., data = bikes2)
## 14 Variables (and intercept)
##
```

		Forced in	Forced out
## Temperature		FALSE	FALSE
## Humidity		FALSE	FALSE
## WindSpeed		FALSE	FALSE
## Visibility		FALSE	FALSE
## DewPointTemp		FALSE	FALSE
## SolarRadiation		FALSE	FALSE
## Rainfall		FALSE	FALSE
## Snowfall		FALSE	FALSE
## SeasonsSpring		FALSE	FALSE
## SeasonsSummer		FALSE	FALSE
## SeasonsWinter		FALSE	FALSE
## HolidayNo Holiday		FALSE	FALSE
## Functioning.DayYes		FALSE	FALSE
## TimeMorning		FALSE	FALSE

```
## 1 subsets of each size up to 8
## Selection Algorithm: exhaustive
##
```

		Temperature	Humidity	WindSpeed	Visibility	DewPointTemp	SolarRadiation
## 1 (1)	"*"	" "	" "	" "	" "	" "	" "
## 2 (1)	" "	"*"	" "	" "	" "	"*"	" "
## 3 (1)	" "	"*"	" "	" "	" "	"*"	" "
## 4 (1)	" "	"*"	" "	" "	" "	"*"	" "
## 5 (1)	" "	"*"	" "	" "	" "	"*"	" "
## 6 (1)	" "	"*"	" "	" "	" "	"*"	" "
## 7 (1)	" "	"*"	" "	" "	" "	"*"	" "
## 8 (1)	" "	"*"	" "	" "	" "	"*"	" "

```
##
```

		Rainfall	Snowfall	SeasonsSpring	SeasonsSummer	SeasonsWinter
## 1 (1)	" "	" "	" "	" "	" "	" "
## 2 (1)	" "	" "	" "	" "	" "	" "
## 3 (1)	" "	" "	" "	" "	" "	" "
## 4 (1)	" "	" "	" "	" "	" "	"*"
## 5 (1)	"*"	" "	" "	" "	" "	"*"
## 6 (1)	"*"	" "	" "	" "	" "	"*"
## 7 (1)	"*"	" "	" "	" "	"*"	"*"
## 8 (1)	"*"	" "	"*"	" "	"*"	"*"

```
##
```

		HolidayNo	Holiday	Functioning.DayYes	TimeMorning
## 1 (1)	" "	" "	" "	" "	" "
## 2 (1)	" "	" "	" "	" "	" "
## 3 (1)	" "	" "	"*"	" "	" "
## 4 (1)	" "	" "	"*"	" "	" "
## 5 (1)	" "	" "	"*"	" "	" "
## 6 (1)	"*"	" "	"*"	" "	" "
## 7 (1)	"*"	" "	"*"	" "	" "
## 8 (1)	"*"	" "	"*"	" "	" "

```
reg <- summary(regfit_full)
names(reg)
```

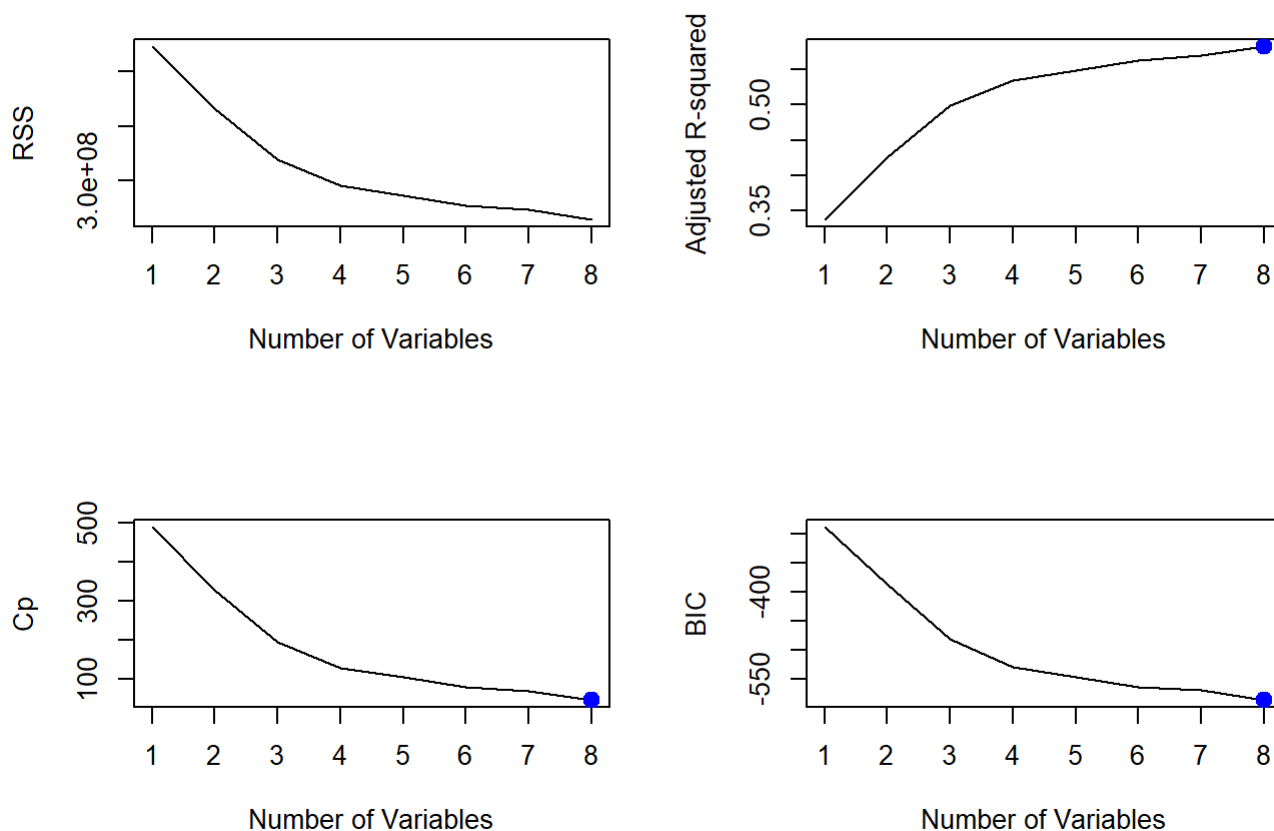
```
## [1] "which" "rsq" "rss" "adjr2" "cp" "bic" "outmat" "obj"
```

```
par(mfrow=c(2,2))
plot(reg$rss, xlab = "Number of Variables", ylab = "RSS", type = "l")

plot(reg$adjr2, xlab = "Number of Variables", ylab= "Adjusted R-squared", type = "l")
adjr2max = which.max(reg$adjr2)
points(adjr2max, reg$adjr2[adjr2max], col="blue", cex = 2, pch=20)

plot(reg$cp, xlab = "Number of Variables", ylab= "Cp", type = "l")
cpmin = which.min(reg$cp)
points(cpmin, reg$cp[cpmin], col="blue", cex=2, pch=20)

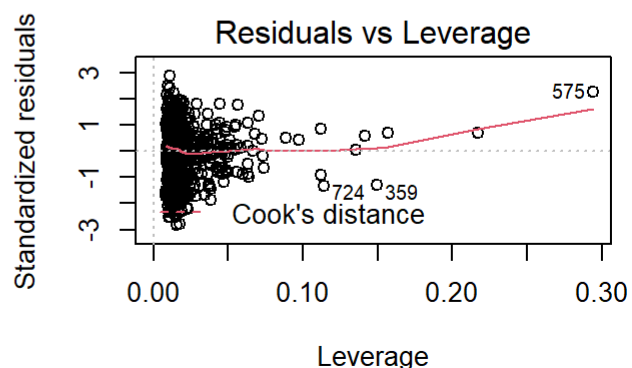
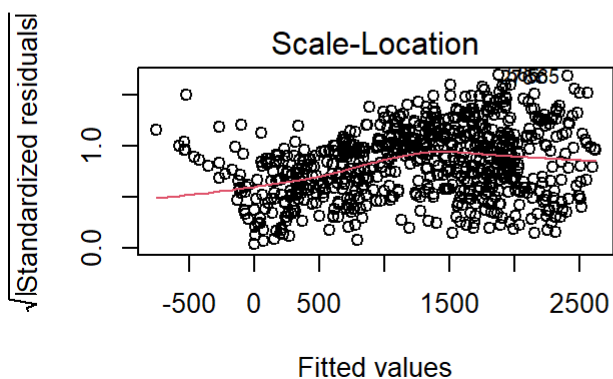
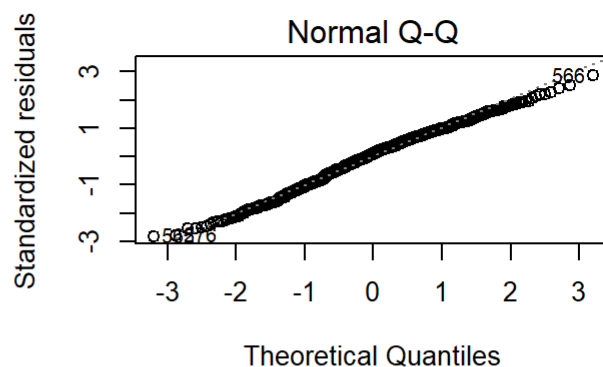
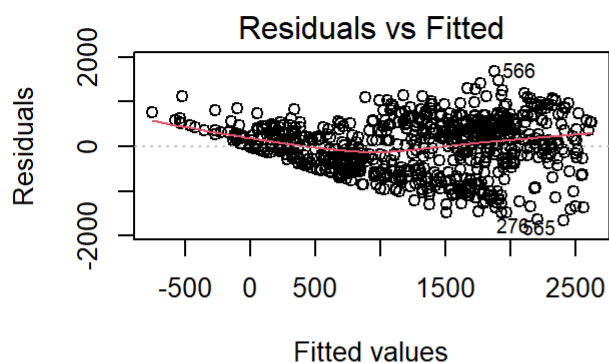
plot(reg$bic, xlab = "Number of Variables", ylab = "BIC", type="l")
bicmin = which.min(reg$bic)
points(bicmin, reg$bic[bicmin], col="blue", cex=2, pch=20)
```



```
coef(regfit_full,8)
```

```
##      (Intercept)      Humidity      DewPointTemp      Rainfall
##      706.52873      -24.58723      43.85470      -78.75892
##      SeasonsSpring  SeasonsSummer  SeasonsWinter  HolidayNo Holiday
##      -316.16994      -420.48498      -667.49369      559.13015
## Functioning.DayYes
##      1706.17969
```

```
#B
lm2 <- lm(RentedBikeCount~ Humidity+DewPointTemp+Rainfall+Seasons+
          Functioning.Day+Holiday, data = bikes2)
plot(lm1)
```



```
#c
x.new <- data.frame(Temperature=12.1, Humidity=29, WindSpeed=2.3,
                    Visibility=1734, DewPointTemp=-5.4, SolarRadiation=2.26,
                    Rainfall=0, Snowfall=0, Seasons="Spring", Functioning.Day="Yes",
                    Holiday="Holiday")

predict(lm2, x.new, interval = "predict")
```

```
##      fit      lwr      upr
## 1 1146.693 -63.59837 2356.985
```

#3

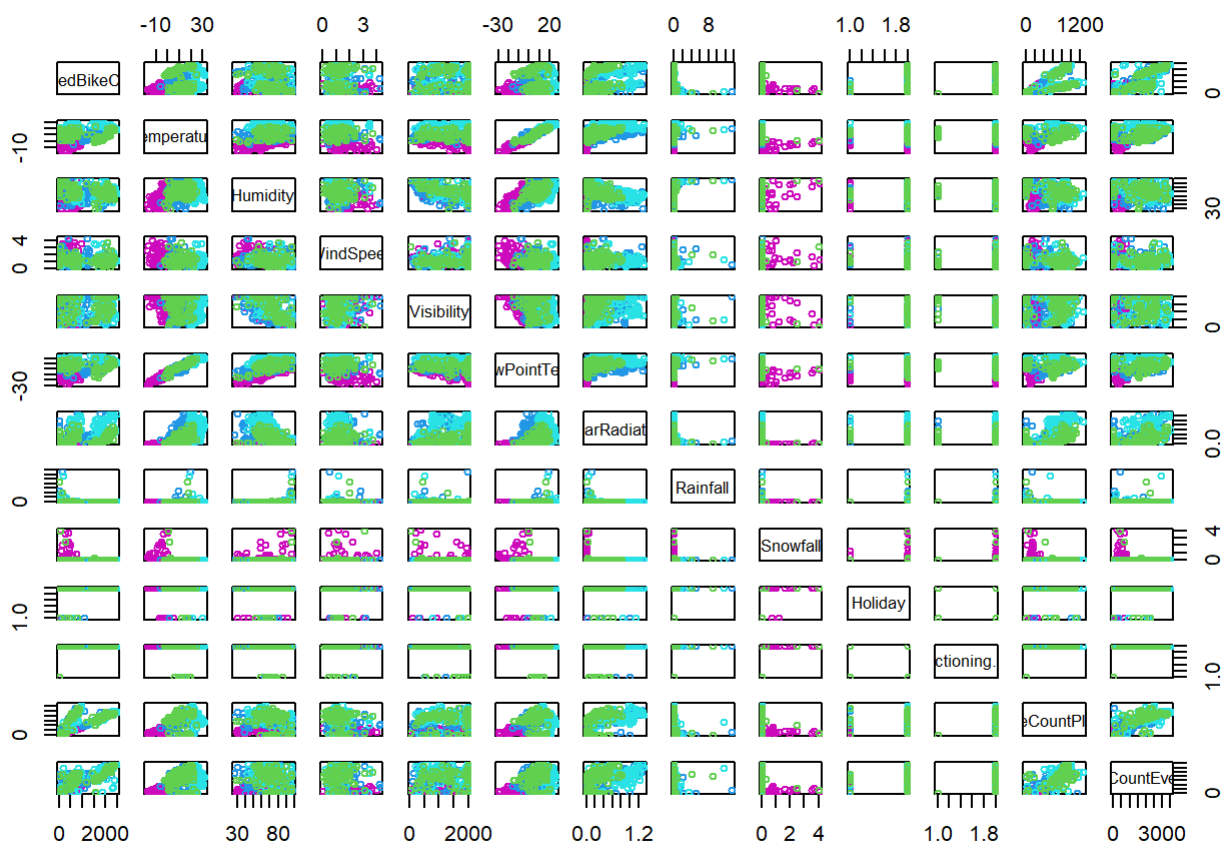
```
bikes3 <- read.csv("C:/Users/dgmur/Downloads/SeoulBikesVer2_F12022.csv")
dim(bikes3)
```

[1] 365 17

```
bikes3$Seasons<- as.factor(bikes3$Seasons)
bikes3$Holiday <- as.factor(bikes3$Holiday)
bikes3$Functioning.Day <- as.factor(bikes3$Functioning.Day)
```

```
bikes3 <- bikes3[,c(4:17)]
```

```
pairs(bikes3[,c(1:9,11:14)], label=names(bikes3[,c(1:9,11:14)]),cex=0.75, col=2+as.numeric(bikes3$Seasons))
```



```
names(bikes3[10])
```

[1] "Seasons"

```
lm.multi <- lm(cbind(RentedBikeCount, BikeCountPlus2,BikeCountEvening)~
  Temperature+Humidity+WindSpeed+Visibility+
  SolarRadiation+Seasons, data = bikes3)

summary.aov(lm.multi)
```

```
## Response RentedBikeCount :
##              Df      Sum Sq Mean Sq F value    Pr(>F)
## Temperature    1  40358746 40358746  98.6585 < 2.2e-16 ***
## Humidity        1   8129664  8129664  19.8733 1.110e-05 ***
## WindSpeed       1     2818    2818    0.0069 0.9339030
## Visibility      1     1777    1777    0.0043 0.9474863
## SolarRadiation  1   5370405  5370405  13.1282 0.0003332 ***
## Seasons         3  11635416  3878472   9.4811 4.848e-06 ***
## Residuals      356 145630770   409075
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Response BikeCountPlus2 :
##              Df      Sum Sq Mean Sq F value    Pr(>F)
## Temperature    1 12374911 12374911 259.6806 < 2.2e-16 ***
## Humidity        1  1902007  1902007  39.9126 7.932e-10 ***
## WindSpeed       1   147770   147770   3.1009  0.07911 .
## Visibility      1    40585    40585   0.8517  0.35671
## SolarRadiation  1  2776087  2776087  58.2546 2.134e-13 ***
## Seasons         3   3772427 1257476  26.3874 1.959e-15 ***
## Residuals      356 16964953   47654
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Response BikeCountEvening :
##              Df      Sum Sq Mean Sq F value    Pr(>F)
## Temperature    1 129164247 129164247 226.3397 < 2.2e-16 ***
## Humidity        1   6669541   6669541  11.6873 0.0007022 ***
## WindSpeed       1     5104     5104   0.0089 0.9247088
## Visibility      1  1100014   1100014   1.9276 0.1658895
## SolarRadiation  1  19931795  19931795  34.9273 8.013e-09 ***
## Seasons         3   25616518   8538839  14.9630 3.406e-09 ***
## Residuals      356 203156902   570665
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(manova(lm.multi), test = c("Wilks"))
```

```
##
##      Df    Wilks approx F num Df den Df    Pr(>F)
## Temperature      1 0.53589  102.195      3 354.00 < 2.2e-16 ***
## Humidity          1 0.88994   14.594      3 354.00 5.551e-09 ***
## WindSpeed         1 0.98603    1.672      3 354.00  0.1728
## Visibility         1 0.98989    1.205      3 354.00  0.3076
## SolarRadiation     1 0.84872   21.033      3 354.00 1.463e-12 ***
## Seasons           3 0.80230    9.070      9 861.69 3.509e-13 ***
## Residuals        356
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
library(car)
```

```
## Warning: package 'car' was built under R version 4.1.3
```

```
## Loading required package: carData
```

```
## Warning: package 'carData' was built under R version 4.1.3
```

```
Manova(lm.multi, type="II", test=c("Wilks"))
```

```
##
## Type II MANOVA Tests: Wilks test statistic
##      Df test stat approx F num Df den Df    Pr(>F)
## Temperature      1  0.99461  0.6400      3 354.00  0.58968
## Humidity          1  0.97994  2.4149      3 354.00  0.06632 .
## WindSpeed         1  0.98942  1.2617      3 354.00  0.28737
## Visibility         1  0.98952  1.2493      3 354.00  0.29173
## SolarRadiation     1  0.84764 21.2108      3 354.00 1.171e-12 ***
## Seasons           3  0.80230  9.0697      9 861.69 3.509e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```



```

rstandard.mlm <- function(model){
  Q <- with(model, qr.qy(qr, diag(1, nrow = nrow(qr$qr), ncol = qr$rank)))
  hii <- rowSums(Q^2)
  RSS <- colSums(model$residuals^2)
  sigma <- sqrt(RSS/model$df.residual)
  pointwise_sd <- outer(sqrt(1-hii), sigma)
  model$residuals/pointwise_sd
}

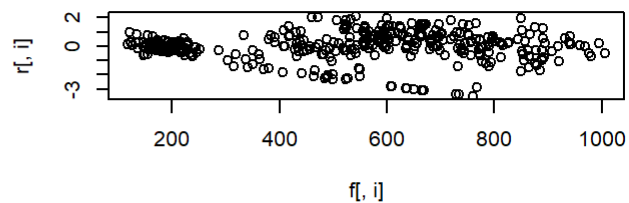
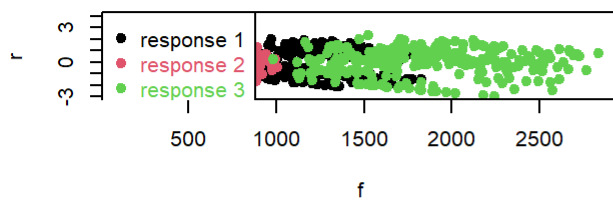
f<- fitted(lm.multi);
r<- rstandard(lm.multi);

par(mfcol=c(3,2))
a <- plot(f, r, col = as.numeric(col(f)), pch = 19, ylim = c(-3, 4))
legend("topleft", legend = paste0("response ", 1:ncol(f)), pch = 19,
      col = 1:ncol(f), text.col = 1:ncol(f))

for(i in 1:ncol(f)){
  plot(f[,i],r[,i], main = paste("Reponse",i))
  qqnorm(r[,i], main = paste("Reponse", i))
  qqline(r[,i])
}

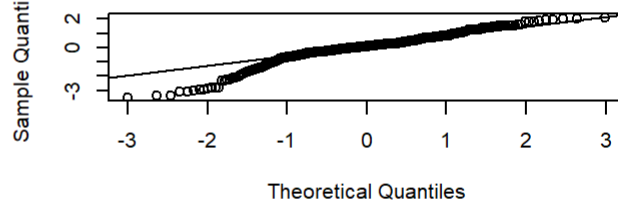
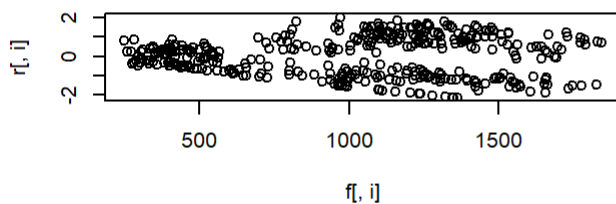
```

Reponse 2



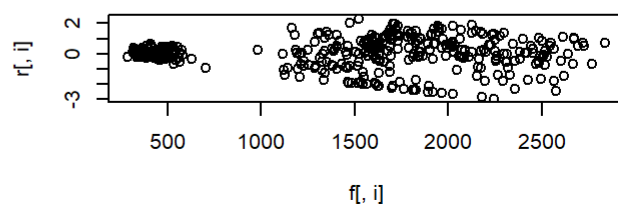
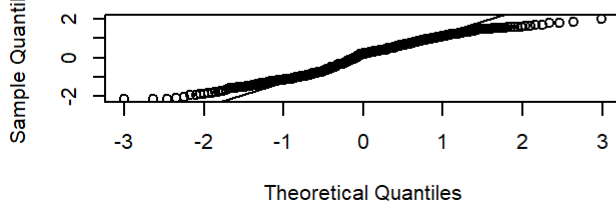
Reponse 1

Reponse 2



Reponse 1

Reponse 3



```
#fitted model
Nlm.multi <- lm(cbind(RentedBikeCount, BikeCountPlus2,BikeCountEvening)~
                Temperature+Humidity+
                SolarRadiation+Seasons, data = bikes3)
summary.aov(Nlm.multi)
```

```
## Response RentedBikeCount :
##              Df      Sum Sq  Mean Sq F value    Pr(>F)
## Temperature    1  40358746 40358746 99.0865 < 2.2e-16 ***
## Humidity        1   8129664  8129664 19.9595 1.062e-05 ***
## SolarRadiation  1   5324622  5324622 13.0727 0.0003426 ***
## Seasons         3  11500164  3833388  9.4115 5.309e-06 ***
## Residuals      358 145816400    407308
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Response BikeCountPlus2 :
##              Df      Sum Sq  Mean Sq F value    Pr(>F)
## Temperature    1 12374911 12374911 257.380 < 2.2e-16 ***
## Humidity        1  1902007  1902007  39.559 9.282e-10 ***
## SolarRadiation  1  2787828  2787828  57.983 2.377e-13 ***
## Seasons         3  3701248  1233749  25.660 4.658e-15 ***
## Residuals      358 17212745    48080
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Response BikeCountEvening :
##              Df      Sum Sq  Mean Sq F value    Pr(>F)
## Temperature    1 129164247 129164247 225.914 < 2.2e-16 ***
## Humidity        1   6669541   6669541  11.665 0.0007098 ***
## SolarRadiation  1  20728763  20728763  36.255 4.291e-09 ***
## Seasons         3  24398197   8132732  14.225 8.888e-09 ***
## Residuals      358 204683373    571741
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(manova(Nlm.multi), test = c("Wilks"))
```

```
##              Df    Wilks approx F num Df den Df    Pr(>F)
## Temperature    1 0.53923   101.400      3 356.00 < 2.2e-16 ***
## Humidity        1 0.89092    14.529      3 356.00 6.004e-09 ***
## SolarRadiation  1 0.84847    21.194      3 356.00 1.181e-12 ***
## Seasons         3 0.80564     8.941      9 866.56 5.654e-13 ***
## Residuals      358
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

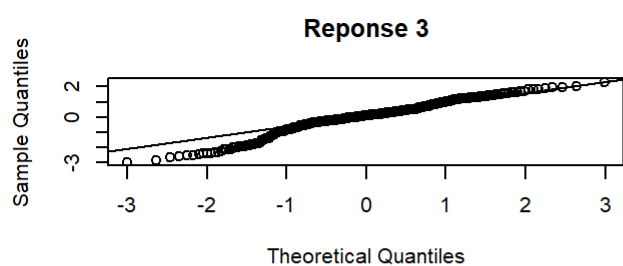
```
library(car)
```

```
Manova(Nlm.multi, type="II", test=c("Wilks"))
```

```
##
## Type II MANOVA Tests: Wilks test statistic
##
```

	Df test	stat	approx F	num Df	den Df	Pr(>F)
Temperature	1	0.99564	0.5191	3	356.00	0.66936
Humidity	1	0.97467	3.0840	3	356.00	0.02741 *
SolarRadiation	1	0.84650	21.5178	3	356.00	7.873e-13 ***
Seasons	3	0.80564	8.9411	9	866.56	5.654e-13 ***

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



shows lack of normality in each of the responses for the qqplots. A log transformation on one or more of the predictor variables may fix this issue.

```
#C
x1<- c(1, Temperature=12.1, Humidity=29, SolarRadiation=2.26, SeasonsSpring=1, SeasonsSummer=0, SeasonsWinter=0)
x1
```

```
##           Temperature      Humidity SolarRadiation  SeasonsSpring
##           1.00           12.10      29.00           2.26           1.00
## SeasonsSummer SeasonsWinter
##           0.00           0.00
```

```
h.Beta <- Nlm.multi$coeff
h.Beta
```

```
##           RentedBikeCount BikeCountPlus2 BikeCountEvening
## (Intercept)      1393.6422623      635.4826330      1455.3299530
## Temperature        0.9199711        0.1228168        7.2503531
## Humidity           -4.8995808       -1.3913855       -0.2964755
## SolarRadiation     556.1318554     393.5483466     1017.0309208
## SeasonsSpring     -319.9633232    -167.0434712    -326.1688311
## SeasonsSummer     -40.5792335     -90.0289049    -180.2825661
## SeasonsWinter     -686.0779732    -371.5310146    -972.9115205
```

```
x
```

```
## [1] 0.33333333 0.16666667 0.08333333 0.00000000 0.00000000 0.00000000 0.00000000
```

```
p<- x1%*%h.Beta
```

```
X<-model.matrix(~Temperature+Humidity+SolarRadiation+Seasons, data = bikes3)
```

```
m=3
```

```
r=nrow(h.Beta)-1
```

```
Resid=lm.multi$residuals
```

```
n=nrow(Resid)
```

```
h.Beta<- as.vector(h.Beta)
```

```
hat.sigma = t(Resid)%*%Resid/n
```

```
hat.sigma
```

```
##           RentedBikeCount BikeCountPlus2 BikeCountEvening
## RentedBikeCount      398988.41      72593.13      329508.54
## BikeCountPlus2       72593.13      46479.32      98767.32
## BikeCountEvening     329508.54      98767.32      556594.25
```

```
multiplier <- sqrt(qf(0.95,m,n-r-m)*(m*(n-r-1))/(n-r-m))
for(i in 1:m){
  pred.se <- sqrt((1 + x1**solve(t(X)**X)**x1)*hat.sigma[i,i]*(n/(n-r-1)))
  cat("Response", i, (p)[i] - multiplier*pred.se ,
      (p)[i] + multiplier*pred.se, "\n")
}
```

```
## Response 1 287.2362 4111.925
## Response 2 666.2909 1971.698
## Response 3 1248.099 5765.466
```