

DMPM Assignment 2 part 2

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Question: Lab Assign-02- Linear Regression Model for Toyota Used-Car Price

Code

```
library(Metrics)
```

```
library(caret)
```

```
library(dplyr)
```

```
library(scales)
```

```
# 1. Read the dataset
```

```
cars<-read.csv("ToyotaCorolla.csv")
```

```
# 2. Create a model
```

```
model<-lm(Price ~ Age + KM + FuelType + HP + MetColor + Automatic + CC +  
Doors + Weight,
```

```
data = cars)
```

```
print(model)
```

```
print(summary(model))
```

3. Filter out the parameters with less significance

```
model<-lm(Price ~ Age + KM + FuelType + HP + CC + Weight, data = cars)
```

```
print(model)
```

```
print(summary(model))
```

```
pred1<-predict(model)
```

```
resd1<-residuals(model)
```

```
predict(model, data.frame(Age=5, KM=2000, FuelType="Diesel", HP=90,  
CC=2000, Weight=1200))
```

4. Scatter and Residual Plots

```
par(mfrow = c(2,1))
```

```
plot(cars$Age,resd1,main = "Residual Plot(Age and Price)",abline(0,0,col =  
"red"),ylab =
```

```
  "Residuals",xlab = "Price in $")
```

```
plot(cars$KM,resd1,main = "Residual Plot(KM and Price)",abline(0,0,col =  
"red"),ylab =
```

```
  "Residuals",xlab = "Price in $")
```

```
plot(cars$HP,resd1,main = "Residual Plot(HP and Price)",abline(0,0,col =  
"red"),ylab =
```

```
  "Residuals",xlab = "Price in $")
```

```
plot(cars$CC,resd1,main = "Residual Plot(CC and Price)",abline(0,0,col =  
"red"),ylab =
```

```
"Residuals",xlab = "Price in $")
```

5. Metrics and Evaluation

```
x<-cbind(cars$Price,pred1)
```

```
x<-data.matrix(x)
```

```
x<-rescale(x)
```

```
x<-as.data.frame(x)
```

```
mae<-MAE(x$V1,x$pred1)
```

```
mse<-mse(x$V1,x$pred1)
```

```
rmse<-RMSE(x$V1,x$pred1)
```

```
r2<-R2(x$V1,x$pred1)
```

```
cat("\nMean Absolute Error:",mae,"\n\nMean Squared Error:",mse)
```

```
cat("\n\nRoot Mean Squared Error:",rmse,"\n\nR-squared:",r2,"\n\n")
```

6. Predictions

```
x=1:length(pred1)
```

```
plot(x, cars$Price,
```

```
  pch=19, col = "yellow",main = "Model Evaluation",
```

```
  xlab = "Count", ylab = "Price")
```

```
lines(x, pred1,col="red")
```

```
legend("topright", legend = c("y-original", "y-predicted"),
      col = c("yellow", "red"),
      pch = c(19,NA), lty = c(NA,1))
```

Output

Creating the model

```
> print(model)

Call:
lm(formula = Price ~ Age + KM + FuelType + HP + MetColor + Automatic +
    CC + Doors + Weight, data = cars)

Coefficients:
(Intercept)          Age           KM  FuelTypeDiesel  FuelTypePetrol
-3.801e+03    -1.220e+02   -1.621e-02    3.390e+03    1.121e+03
          HP      MetColor    Automatic           CC           Doors
 6.081e+01    5.716e+01    3.303e+02   -4.174e+00   -7.776e+00
      Weight
 2.001e+01
```

Summary of model

```
> print(summary(model))

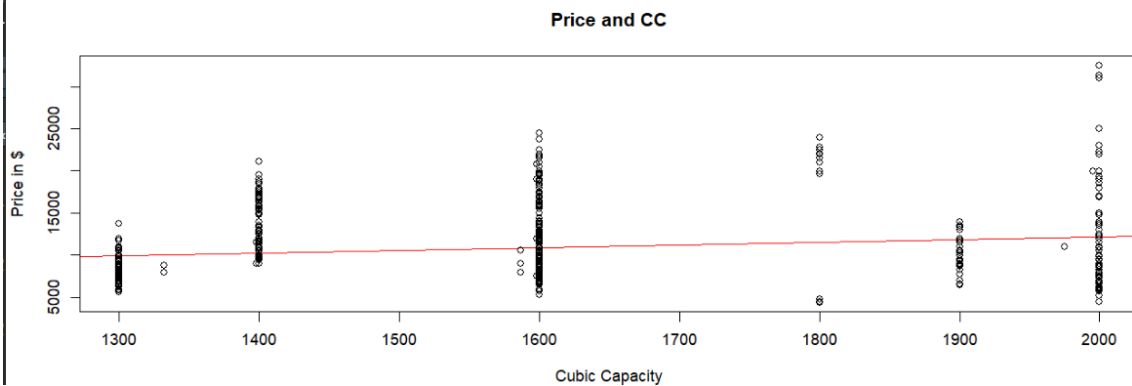
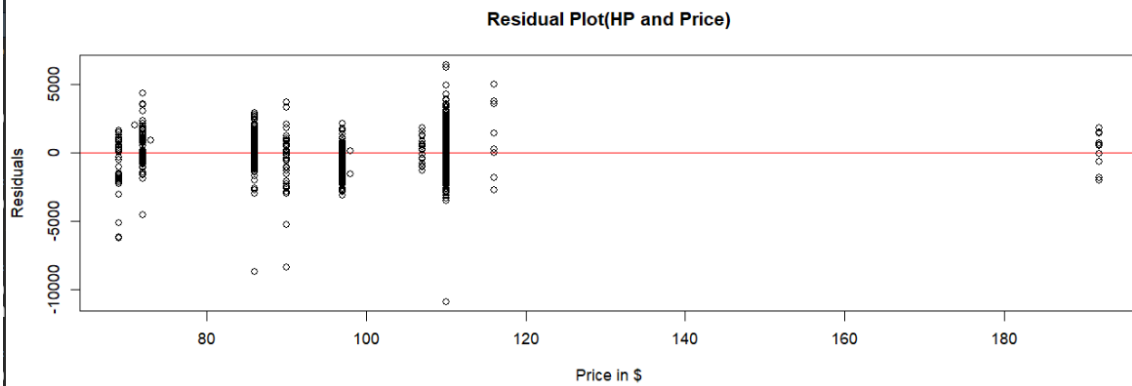
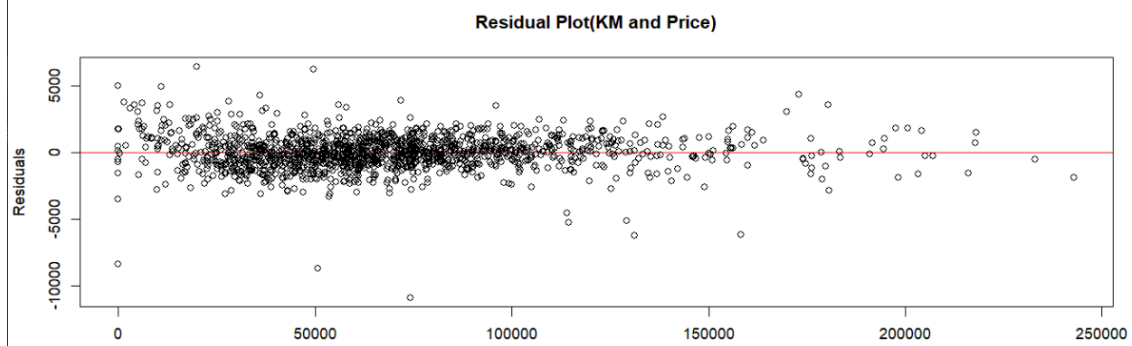
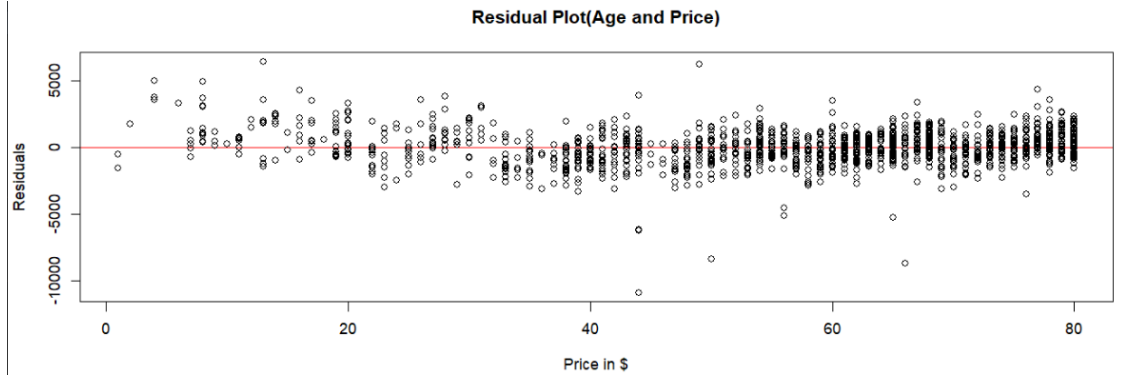
Call:
lm(formula = Price ~ Age + KM + FuelType + HP + MetColor + Automatic +
    CC + Doors + Weight, data = cars)

Residuals:
    Min       1Q   Median       3Q      Max
-10642.3  -737.7     3.1    731.3   6451.5

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) -3.801e+03  1.304e+03  -2.915  0.003613 **
Age          -1.220e+02  2.602e+00 -46.889 < 2e-16 ***
KM           -1.621e-02  1.313e-03 -12.347 < 2e-16 ***
FuelTypeDiesel 3.390e+03  5.188e+02  6.535 8.86e-11 ***
FuelTypePetrol 1.121e+03  3.324e+02  3.372 0.000767 ***
HP           6.081e+01  5.756e+00 10.565 < 2e-16 ***
MetColor      5.716e+01  7.494e+01  0.763 0.445738
Automatic     3.303e+02  1.571e+02  2.102 0.035708 *
CC           -4.174e+00  5.453e-01 -7.656 3.53e-14 ***
Doors        -7.776e+00  4.006e+01 -0.194 0.846129
Weight       2.001e+01  1.203e+00 16.629 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1316 on 1425 degrees of freedom
Multiple R-squared:  0.8693,    Adjusted R-squared:  0.8684
F-statistic: 948 on 10 and 1425 DF,  p-value: < 2.2e-16
```

Plotting the residual values



Metrics

```
> cat("\nMean Absolute Error:",mae,"\n\nMean Squared Error:",mse)

Mean Absolute Error: 0.03382402

Mean Squared Error: 0.00217556
> cat("\n\nRoot Mean Squared Error:",rmse,"\n\nR-squared:",r2,"\n\n")

Root Mean Squared Error: 0.0466429

R-squared: 0.8688574
```

The observed errors are very small, so the accuracy of our model is good

Prediction

```
> predict(model, data.frame(Age=5, KM=2000, FuelType="Diesel", HP=90, CC=2000, Weight=1200))
1
20091.74
```

Prediction of a 5 year old car is 20 thousand bucks