

Project 4 Part II

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Part 2

Using “USArrests” Dataset

A) Create an “crime” dataset containing all the variables of USArrests

```
crime <- USArrests
```

B) Create correlation matrix plot of the crime data and interpret each scatterplot carefully

```
cor(crime)
```

```
##           Murder  Assault  UrbanPop   Rape
## Murder    1.00000000 0.8018733 0.06957262 0.5635788
## Assault    0.80187331 1.0000000 0.25887170 0.6652412
## UrbanPop    0.06957262 0.2588717 1.00000000 0.4113412
## Rape       0.56357883 0.6652412 0.41134124 1.0000000
```

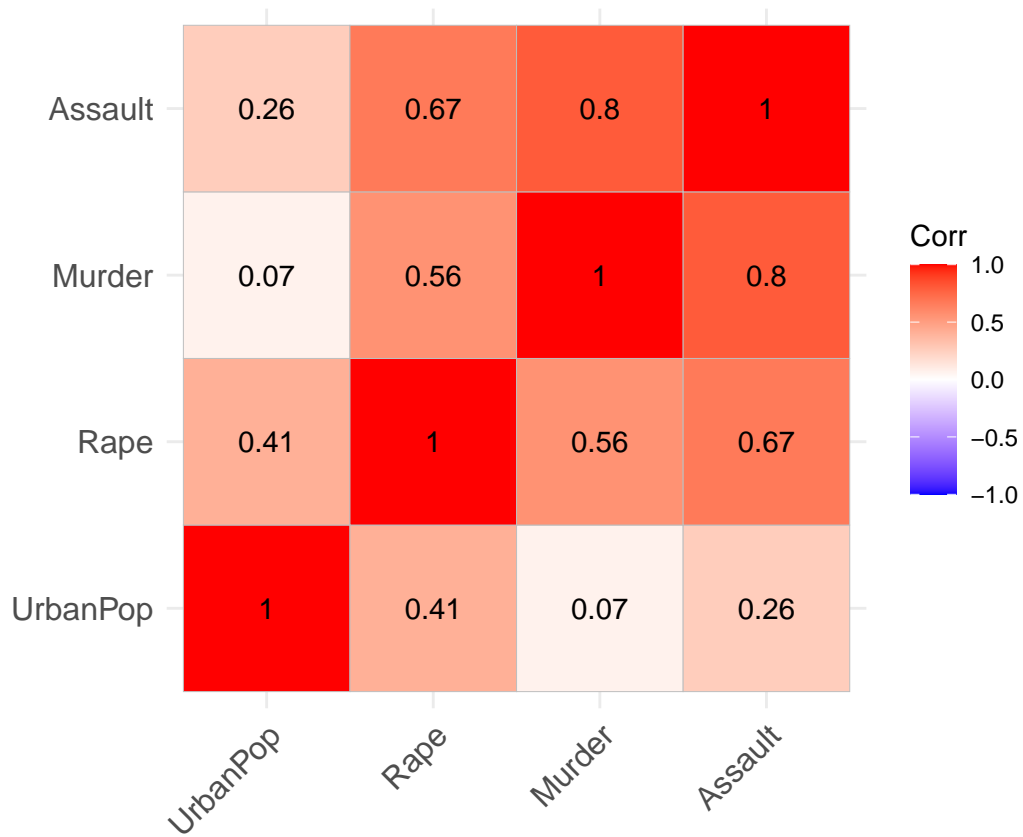
visualize correlation matrix

```
library(ggcorrplot)
```

```
## Loading required package: ggplot2
```

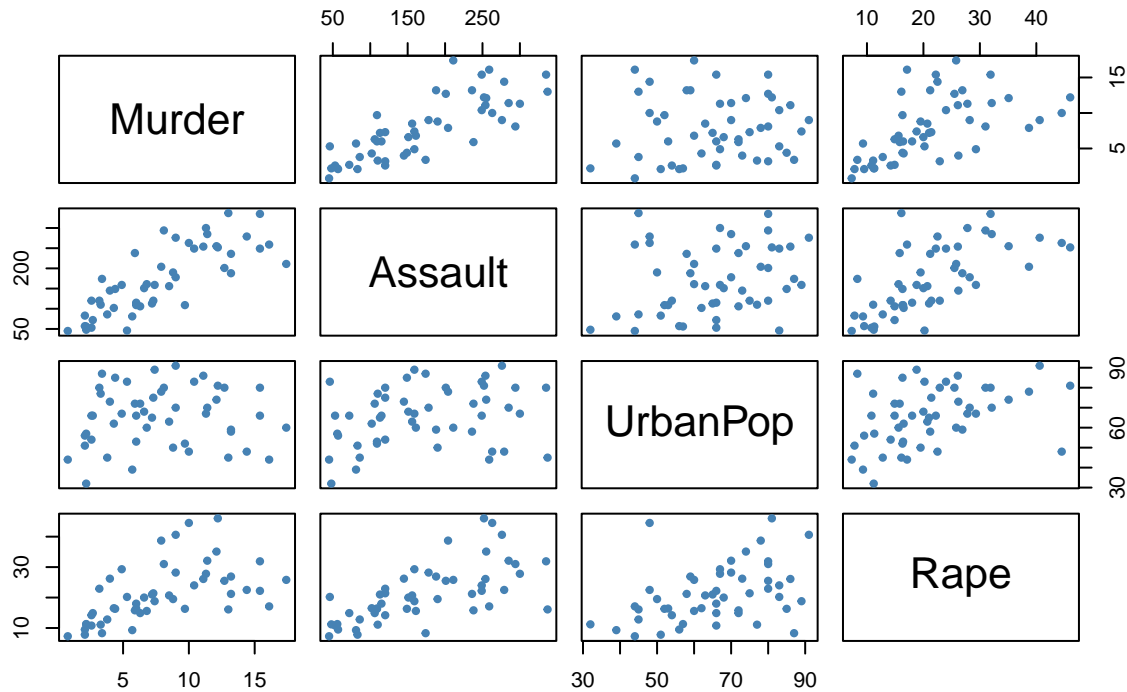
```
library(ggplot2)
```

```
ggcorrplot(cor(crime),hc.order = T,lab = T)
```



```
plot(crime, pch=20,cex=1,col='steelblue',main="Scatter Plot")
```

Scatter Plot



```
#Relationship between two variable
# -1 inverse linear relationship,
# 0 no linear correlation,
# 1 linear relationship]
```

C) Split the crime dataset into training and testing data with 70% and 30% cases

```
set.seed(13)
index <- sample(2,nrow(crime),replace = T,prob = c(0.7,0.3)) #Random sampling into two independent vari
train.crime <- crime[index==1,] #Training set
test.crime <- crime[index==2,] #Test set
```

D) Fit a multiple linear regression on training data with Murder as dependent variable and all other variables as independent variables and interpret the results carefully using R-squared, RMSE, Regression ANOVA and Regression Coefficients (BLUE?)

```
mlr <- lm(Murder ~ .,data=train.crime)
summary(mlr)
```

```
##
```

```
## Call:
## lm(formula = Murder ~ ., data = train.crime)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.8132 -1.7164 -0.5668  1.2990  7.1819
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.16946    2.05877   2.025   0.051 .
## Assault      0.04653    0.00782   5.950 1.12e-06 ***
## UrbanPop    -0.05151    0.03217  -1.601   0.119
## Rape        -0.02633    0.06981  -0.377   0.709
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.518 on 33 degrees of freedom
## Multiple R-squared:  0.6648, Adjusted R-squared:  0.6343
## F-statistic: 21.81 on 3 and 33 DF,  p-value: 5.711e-08
```

R-squared = 0.6648

```
mean(mlr$residuals^2)    #RMSE= 5.656679
```

```
## [1] 5.656679
```

```
mlr$coefficients        #Regression Coefficient (Intercept)
```

```
## (Intercept)    Assault    UrbanPop      Rape
##  4.16946411  0.04653252 -0.05150877 -0.02632538
```

```
anova(mlr)              #Regression ANOVA
```

```
## Analysis of Variance Table
##
## Response: Murder
##              Df Sum Sq Mean Sq F value    Pr(>F)
## Assault      1 393.06  393.06 61.9733 4.475e-09 ***
## UrbanPop     1  21.07   21.07  3.3216  0.07745 .
## Rape         1   0.90    0.90  0.1422  0.70852
## Residuals   33 209.30    6.34
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

“Assault” is highly significance, p-value<0.05

E) Check multicollinearity and finalize this model with the appropriate VIF cut-off value

```
library(car)
```

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

```
vif(mlr)
```

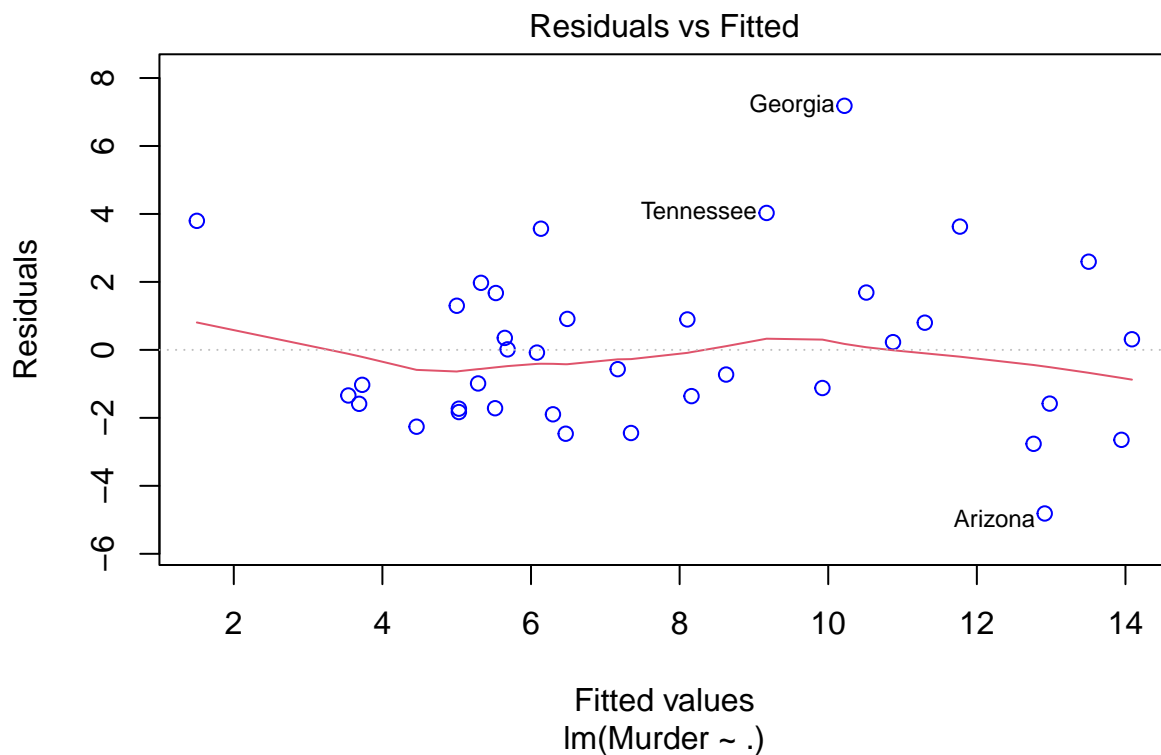
```
## Assault UrbanPop Rape  
## 2.087282 1.156333 2.305631
```

No Multicollinearity because VIF of all variable are <10

F) Perform residual analysis of this model i.e. LINE tests using suggestive graphs and confirmatory tests and interpret the results carefully

```
##LINE ## L = Linearity of residuals
```

```
### Suggestive  
plot(mlr,which = 1,col=c("blue")) # LOESS line lies in the zero line so residuals are linear
```

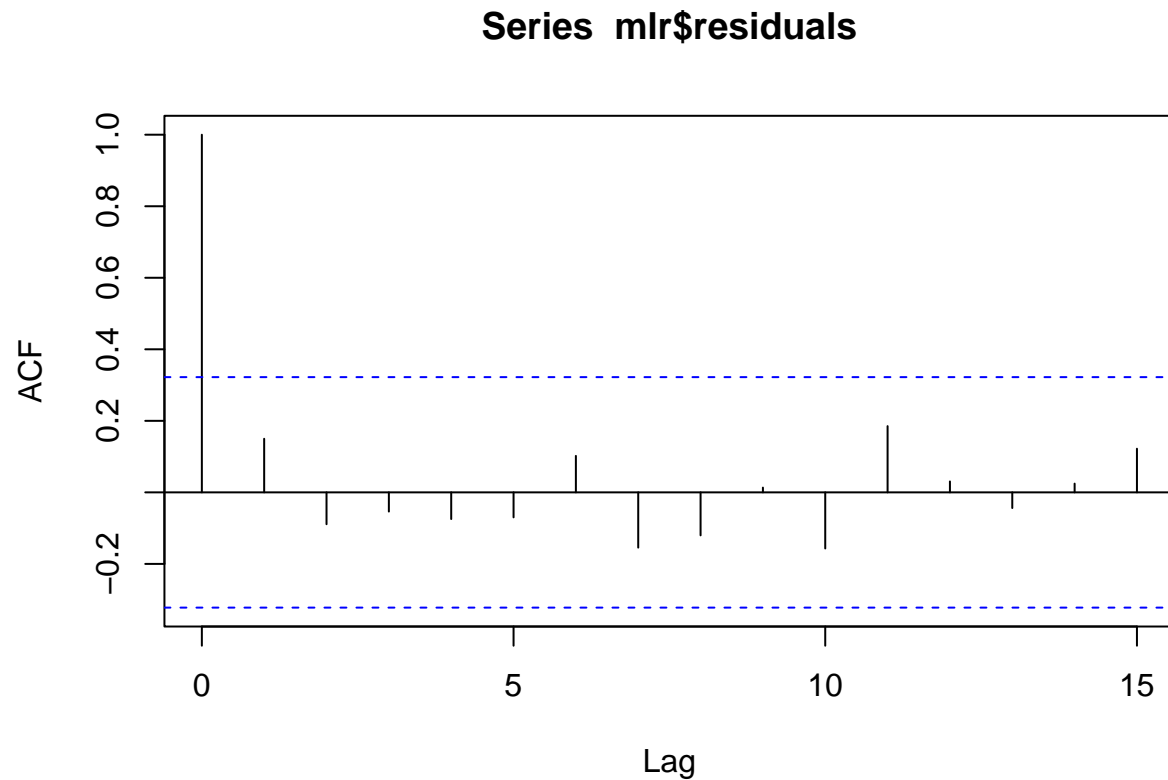


```
### Confirmative  
summary(mlr$residuals) #Mean = 0 so residuals are linear
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.  
## -4.8132 -1.7164 -0.5668 0.0000 1.2990 7.1819
```

I = Independence of residuals

```
### Suggestive
acf(mlr$residuals)      # Show Up and Down so no autocorrelation
```

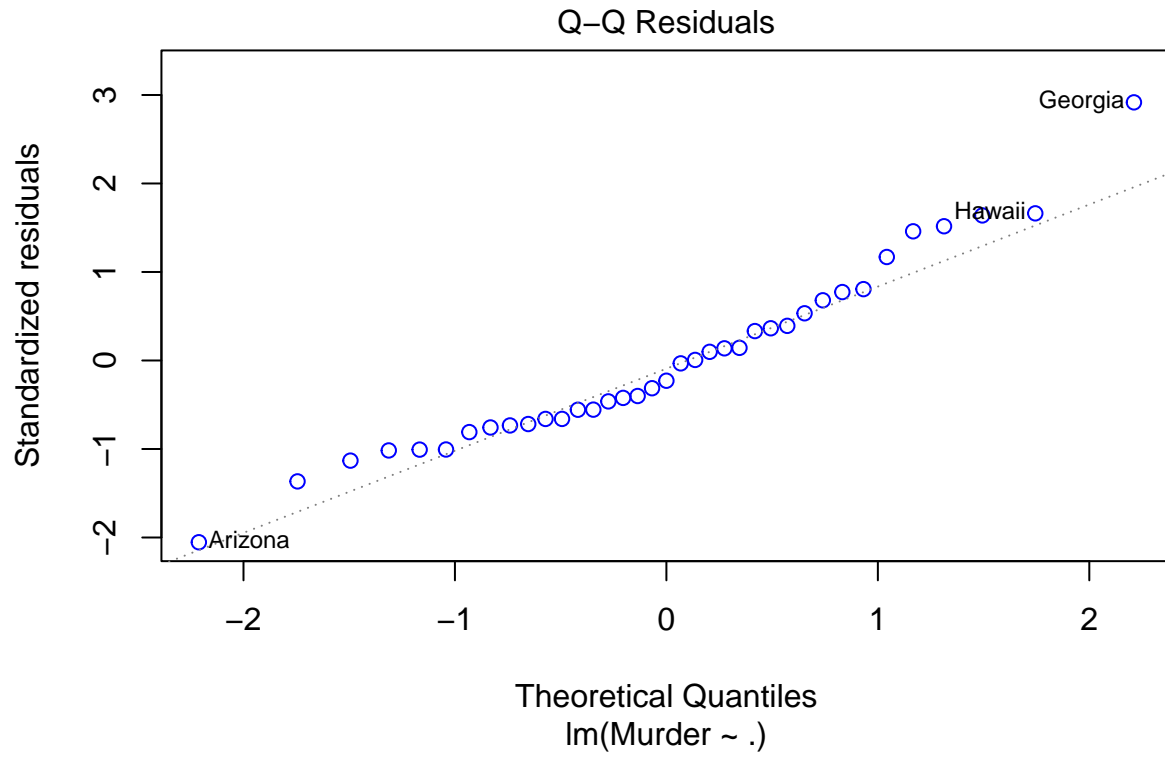


```
### Confirmative
library(car)
durbinWatsonTest(mlr)      # p-value>0.05 no autocorrelation
```

```
## lag Autocorrelation D-W Statistic p-value
## 1 0.1498789 1.654913 0.296
## Alternative hypothesis: rho != 0
```

N = Normality of residuals

```
## Suggestive
plot(mlr,which = 2,col=c("blue"))
```

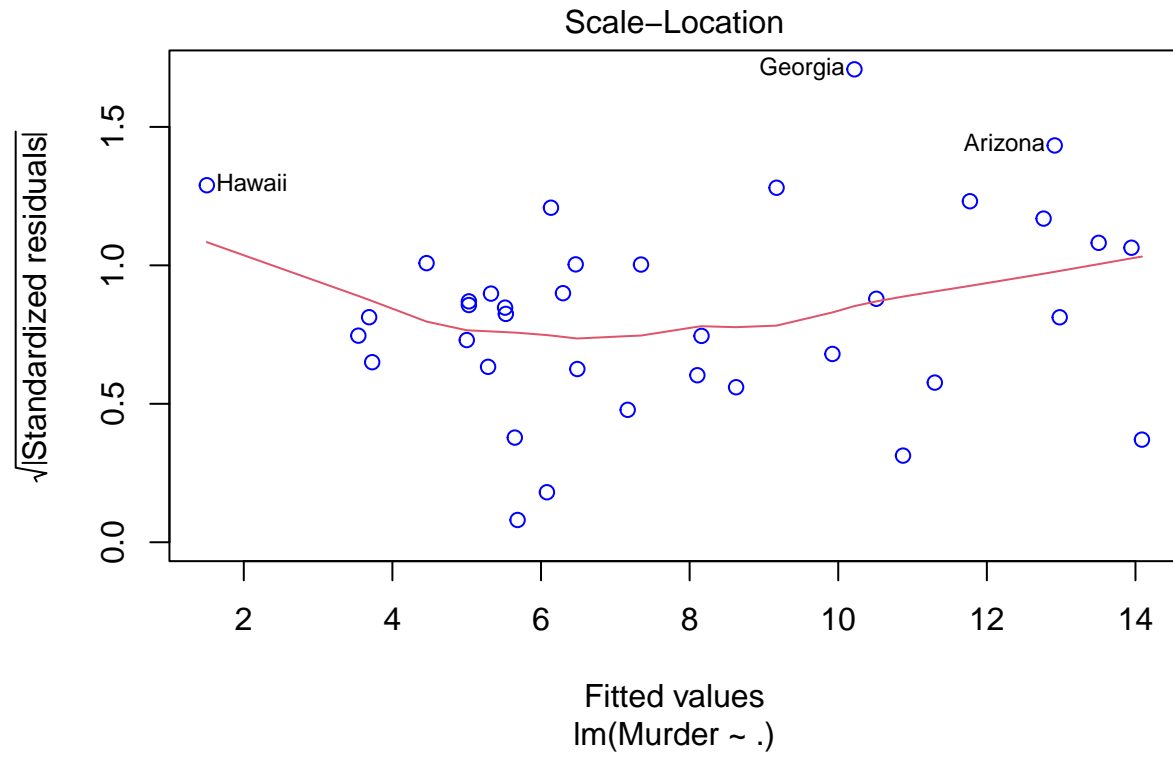


```
### Confirmative
shapiro.test(mlr$residuals) # p-value > 0.05, residuals follow normal distribution
```

```
##
## Shapiro-Wilk normality test
##
## data:  mlr$residuals
## W = 0.94906, p-value = 0.09001
```

E = Equal variance of residuals

```
### Suggestive
plot(mlr, which = 3, col=c("blue"))
```



```
### Confirmative
library(lmtest)
```

```
## Loading required package: zoo
```

```
##
```

```
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
## as.Date, as.Date.numeric
```

```
bptest(mlr) # p-value > 0.05, so residual variances are equal (homoscedasticity)
```

```
##
```

```
## studentized Breusch-Pagan test
```

```
##
```

```
## data: mlr
```

```
## BP = 1.7374, df = 3, p-value = 0.6287
```

G) Predict the Murder in the testing dataset using the fitted model


```
prediction <- predict(mlr, test.crime)
prediction
```

```
##      Alabama      California      Delaware      Florida      Idaho
##      11.605533      11.256332      11.119632      14.797378      6.598073
##      Illinois      Maine North Carolina      North Dakota      Rhode Island
##      10.849025      5.199378      17.109191      3.804866      7.566359
##      Texas      Virginia      Wisconsin
##      8.730502      7.638550      2.951795
```

H) Report R-square and RMSE of predicted model and interpret them carefully

```
library(caret)
```

```
## Warning: package 'caret' was built under R version 4.3.3
```

```
## Loading required package: lattice
```

```
## Warning: package 'lattice' was built under R version 4.3.3
```

```
data.frame(R2 = R2(prediction, test.crime$Murder),
           RMSE = RMSE(prediction, test.crime$Murder))
```

```
##      R2      RMSE
## 1 0.7008134 3.049961
```