Analysis of India's Air Quality Index using R

Project Report



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Course: Statistical Methods using R (SC - 612)

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<u>Index</u>

Sr. No	Particulars	Page No.
1	Problem Definition	03
2	Project Description	03
3	Methodology and Data Cleaning	03
4	Code and Analysis with Inference	04 - 20
5	Conclusion	21
6	References	21

Problem Definition:

The air quality index (AQI) is an index for reporting air quality on a daily basis. It is a measure of how air pollution affects one's health within a short time period. The purpose of the AQI is to help people know how the local air quality impacts their health. The higher the AQI value, the greater the level of air pollution and the greater the health concerns.

Project Description:

In this project, we have analyzed the Air Quality Index of a few selected states and their cities to gain inference on the various factors affecting the Air Quality of any given region.

Analyzing the exact Air Pollutant's like "NO2", "PM2.5", "PM10", "CO", "CO2", "Benzene", "O3", "Xylene", "Toluene", etc. by using various statistical methods like:

- a. Normality Test
- b. Summary Statistic
- c. Correlation
- d. Linear Regression
- e. Multiple Linear Regression And a few Visualization techniques like Bar Plot, Pie Chart, Scatter Plot, etc.

to analyze the given dataset.

Methodology and Data Cleaning:

We acquired this dataset from Kaggle.com. This data consisted of 15 columns and around 10,000 rows.

We performed data cleaning in R Studio & MS Excel and removed Null values along with unwanted rows and columns.

Further we performed various above mentioned statistical tools on this data to get reliable inferences.

We have taken the AQI Range and AQI Status by using the scale:

0-50-Good 50-100-Satisfactory

100 – 200 – Moderate 200 – 300 – Poor

300 – 400 – Very Poor More than 400 -- Severe

Code and Analysis:

Phase - 1: Importing the csv file into R studio:

```
getwd()
## [1] "E:/DA-IICT/Study/Sem 1/R Project/Indian_Air_Pollution_Data"

AQI_data = read.csv('AQIdatasetmain.csv')

getwd(): -To get or set Working directory.
read.csv(): -To read the comma seperated value files.
```

> Phase - 2: To see the content that the data holds:

```
head(AQI_data)
##
         StationId
                                                                        S
                        City PM2.5
                                            NO
                                                 NO2
                                                       NOx
                                                             NH3
                                                                  CO
                                     PM10
## 1 Andhra_Pradesh Amaravati 71.36 115.75 1.75 20.65 12.40 12.19 0.10 10.
76
## 2 Andhra_Pradesh Amaravati 81.40 124.50 1.44 20.50 12.08 10.72 0.12 15.
24
## 3 Andhra Pradesh Amaravati 78.32 129.06 1.26 26.00 14.85 10.28 0.14 26.
96
## 4 Andhra_Pradesh Amaravati 73.96 113.56 4.58 19.29 13.97 10.95 0.10 13.
## 5 Andhra_Pradesh Amaravati 89.90 140.20 7.71 26.19 19.87 13.12 0.10 19.
## 6 Andhra Pradesh Amaravati 87.14 130.52 0.97 21.31 12.12 14.36 0.15 11.
41
        03 Benzene Toluene Xylene AQI AQI_Bucket
##
## 1 109.26
              0.17
                      5.92
                             0.10 132
                                        Moderate
## 2 127.09
              0.20
                      6.50
                             0.06 184
                                        Moderate
## 3 117.44
              0.22
                      7.95
                             0.08 197
                                        Moderate
## 4 123.80
              0.17
                      2.85
                             0.04 191
                                        Moderate
## 5 128.73
                             0.07 191
              0.25
                      2.79
                                        Moderate
0.04 227
                                            Poor
```

head(data): – Return the first parts of the matrix, table or dataframe

Normality Check: To check whether the data follows normal distribution.

For Normality check, we use Shapiro-Wilk test on all the columns in our data.

Shapiro Test:

```
apply(AQI_data[,c(3:14)],2,shapiro.test)
## $PM2.5
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.72028, p-value = 1.398e-11
##
##
## $PM10
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.88156, p-value = 9.438e-07
##
##
## $NO
##
##
   Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.7365, p-value = 3.39e-11
##
##
## $NO2
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.92744, p-value = 0.0001137
##
##
## $NOx
##
    Shapiro-Wilk normality test
##
##
## data: newX[, i]
## W = 0.77378, p-value = 3.015e-10
##
##
## $NH3
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.92673, p-value = 0.0001045
```

```
##
## $CO
##
    Shapiro-Wilk normality test
##
##
## data: newX[, i]
## W = 0.36576, p-value < 2.2e-16
##
##
## $SO2
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.78091, p-value = 4.704e-10
##
##
## $03
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.75265, p-value = 8.505e-11
##
##
## $Benzene
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.87525, p-value = 5.322e-07
##
##
## $Toluene
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.74148, p-value = 4.482e-11
##
##
## $Xylene
##
##
    Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.83072, p-value = 1.418e-08
```

For Normality check, if P-value is > Alpha (0.05), then we can say that the data follows normal distribution.

But here, for all the columns P-value < Alpha (0.05). So, we conclude that our data is not normal.

➤ Phase - 3: To know the Summary Statistic of data (mean, median and Quantile):

```
# Moderate , Poor , Satisfactory, Very Poor, Severe, Good
summary(AQI_data)
##
    StationId
                          City
                                            PM2.5
                                                              PM10
##
                      Length:87
                                                         Min. : 18.03
   Length:87
                                              : 11.04
##
   Class :character
                      Class :character
                                        1st Qu.: 26.41
                                                         1st Qu.: 70.61
                      Mode :character
##
   Mode :character
                                        Median : 64.01
                                                         Median :107.73
##
                                        Mean
                                                         Mean
                                              : 73.51
                                                              :126.63
##
                                         3rd Qu.: 90.66
                                                         3rd Qu.:179.50
##
                                              :412.34
                                                              :404.52
                                        Max.
                                                         Max.
##
         NO
                         NO2
                                         NO<sub>x</sub>
                                                         NH3
##
          : 0.570
                    Min. : 3.13
                                   Min. : 2.18
                                                    Min.
                                                          : 2.77
   Min.
   1st Qu.: 3.615
                    1st Ou.:12.73
                                   1st Ou.: 14.29
                                                    1st Ou.:12.12
##
##
                    Median :19.87
                                   Median : 18.21
                                                    Median :14.36
   Median : 7.250
   Mean
         :13.867
                    Mean :24.25
                                   Mean : 28.75
                                                    Mean :17.51
##
   3rd Qu.:15.985
                    3rd Qu.:32.09
                                    3rd Qu.: 34.78
                                                    3rd Qu.:22.60
##
   Max. :78.090
                    Max. :66.23
                                   Max. :112.44
                                                    Max.
                                                           :41.78
##
                         S02
                                          03
         CO
                                                        Benzene
##
   Min.
         : 0.100
                    Min. : 3.390
                                     Min.
                                          : 1.31
                                                     Min.
                                                           : 0.010
##
   1st Qu.: 0.575
                    1st Qu.: 9.315
                                     1st Qu.: 18.39
                                                     1st Qu.: 0.580
##
   Median : 0.850
                    Median :12.500
                                     Median : 26.34
                                                     Median : 2.940
        : 1.785
                                    Mean : 33.49
##
   Mean
                    Mean :15.611
                                                     Mean
                                                            : 3.871
##
   3rd Qu.: 1.130
                    3rd Qu.:18.995
                                     3rd Qu.: 36.52
                                                     3rd Qu.: 6.270
##
   Max.
        :27.140
                    Max. :79.470
                                     Max.
                                          :128.73
                                                     Max. :12.450
##
      Toluene
                        Xylene
                                         AQI
                                                     AQI_Bucket
                                     Min.
                                                    Length:87
          : 0.040
                    Min. : 0.000
                                          : 28.0
   Min.
##
   1st Qu.: 1.520
                    1st Qu.: 0.575
                                     1st Qu.: 95.0
                                                    Class :character
##
   Median : 2.850
                    Median : 5.060
                                    Median :136.0
                                                    Mode :character
   Mean : 6.290
##
                    Mean : 6.136
                                     Mean :178.6
##
   3rd Qu.: 9.975
                    3rd Qu.: 7.680
                                     3rd Qu.:227.0
   Max. :41.200 Max. :25.660
                                    Max. :653.0
```

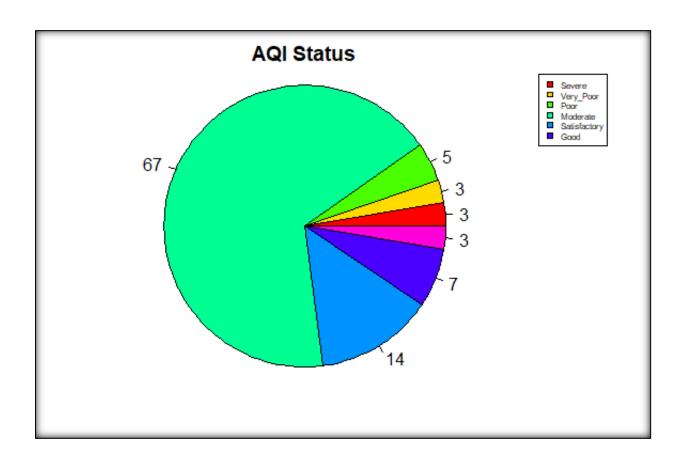
Phase - 4: Checking the Air Quality Status:

```
#Moderate, Very Poor, Poor , Severe , Good, Satisfactory
#Average = mean(AQI_data$AQI)
moderate = length(which(AQI_data$AQI_Bucket == 'Moderate'))
VeryPoor = length(which(AQI_data$AQI_Bucket == 'Very_Poor'))
Poor = length(which(AQI_data$AQI_Bucket == 'Severe'))
Severe = length(which(AQI_data$AQI_Bucket == 'Severe'))
Satisfactory = length(which(AQI_data$AQI_Bucket == 'Satisfactory'))
Good = length(which(AQI_data$AQI_Bucket == 'Good'))

pie_chart = cbind(Severe,VeryPoor,Poor,moderate,Satisfactory,Good)
percentage = round(100*pie_chart/sum(pie_chart))

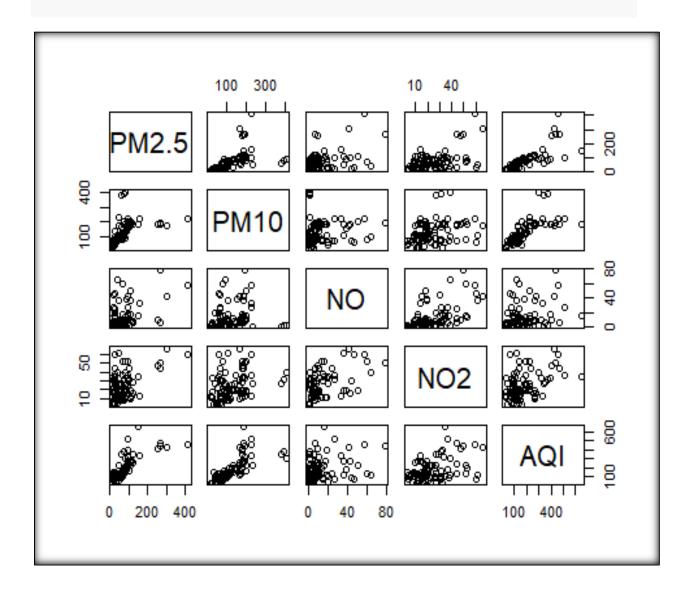
pie(pie_chart,labels = percentage,col = rainbow(length(pie_chart)),main = "AQI Status",radius = 1)

legend("topright",c("Severe","Very_Poor","Poor","Average","Moderate","Satisfactory","Good"),cex = 0.5,fill = rainbow(length(pie_chart)))
```

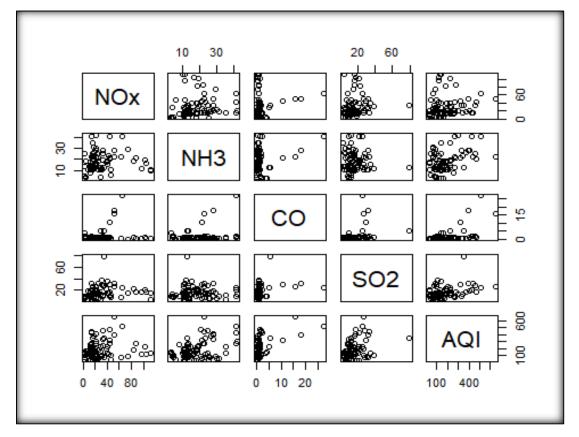


▶ Phase - 5: Correlation between different affecting factors affecting AQI:

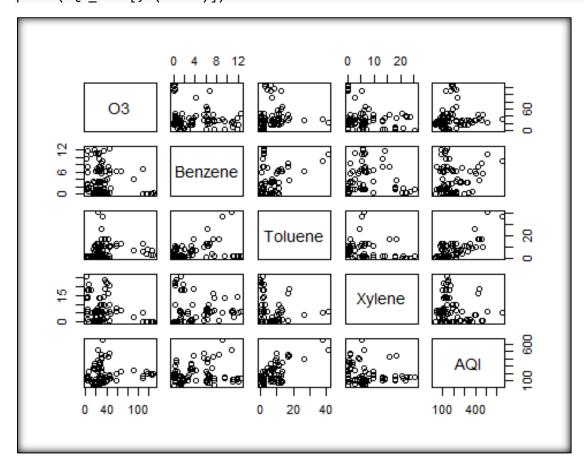
pairs(AQI_data[,c(3:6,15)])



pairs(AQI_data[,c(7:10,15)])



pairs(AQI_data[,c(11:15)])



> Phase - 6: Testing of Hypothesis

Correlation

o **1**:

H0: Correlation between Xylene and AQI is zero.

H1: Correlation between Xylene and AQI is not zero.

Inference: Here, we can see that the correlation coefficient is -0.015. i.e., Weak Negative Correlation. But the P-value > Alpha (0.05) and the Correlation coefficient is very close to 0, so we cannot reject H0 and the correlation is almost equal to 0.

o **2**:

H0: Correlation between PM2.5 and AQI is zero.

H1: Correlation between PM2.5 and AQI is not zero.

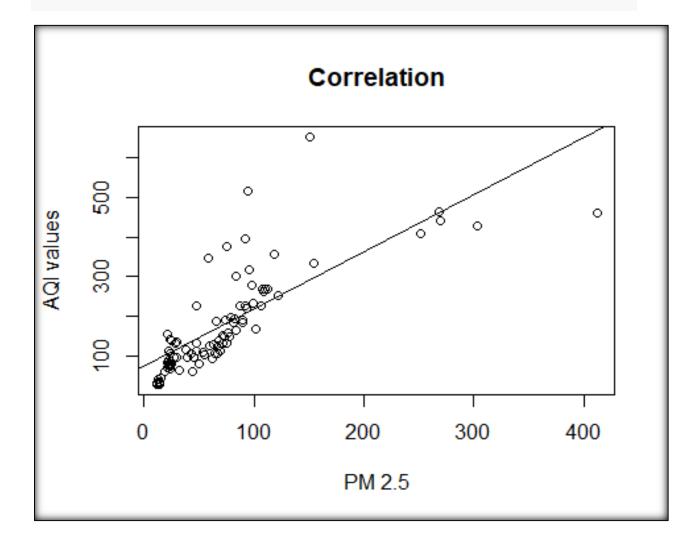
```
cor.test(AQI_data$PM2.5,AQI_data$AQI,method = 'spearman')
## Warning in cor.test.default(AQI_data$PM2.5, AQI_data$AQI, method = "spearman"):
## Cannot compute exact p-value with ties
##
## Spearman's rank correlation rho
##
## data: AQI_data$PM2.5 and AQI_data$AQI
## S = 12972, p-value < 2.2e-16
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
## rho
## 0.8817868</pre>
```

Inference: Here, we can see that the correlation coefficient is 0.881. i.e., Strong Positive Correlation. Also, the P-vale < Alpha (0.05), so we can reject H0.

We can conclude that there is Strong Correlation between PM2.5 and AQI.

Graphical Representation:

```
plot(x = AQI_data$PM2.5, y = AQI_data$AQI,xlab ="PM 2.5",ylab = "AQI value
s",main = "Correlation")
abline(lm(AQI_data$AQI~AQI_data$PM2.5))
```

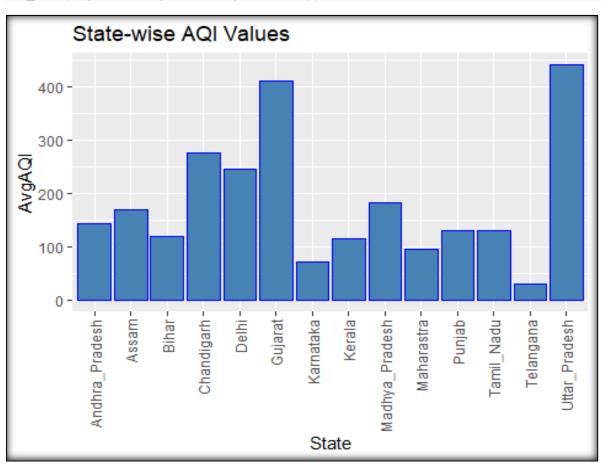


Inference: We can visualize that there is positive correlation between PM2.5 and AQI, which says that AQI values are dependent on PM2.5 (Higher the PM2.5 values, higher the AQI).

We can also see that in the Trend line.

Phase - 7: State-wise Average AQI index:

```
Lalu = aggregate(x = AQI_data$AQI,by = list(AQI_data$StationId),FUN = mean)
colnames(Lalu)[1] = "State"
colnames(Lalu)[2] = "AvgAQI"
print(Lalu)
##
               State
                        AvgAQI
## 1
      Andhra_Pradesh 142.25000
## 2
               Assam 169.66667
## 3
               Bihar 118.71429
## 4
          Chandigarh 276.33333
## 5
               Delhi 244.16667
## 6
             Gujarat 410.83333
## 7
           Karnataka 71.42857
## 8
              Kerala 113.80000
## 9
      Madhya_Pradesh 182.83333
## 10
          Maharastra 96.16667
## 11
              Punjab 130.80000
## 12
          Tamil_Nadu 129.80000
## 13
           Telangana 30.00000
## 14
       Uttar_Pradesh 439.80000
require(ggplot2)
## Loading required package: ggplot2
ggplot(Lalu,aes(x = State,y = AvgAQI)) + ggtitle("State-wise AQI Values") + geom_b
ar(stat = "identity",color ="blue", fill = "steelblue") + theme(axis.text.x = elem
ent_text(angle = 90,hjust = 1,vjust = 0.5))
```



Phase - 8: Regression:

Ho: There is no significant relationship between the variables.

H1: There is significant relationship between the variables.

Linear Regression:

```
linear_rega= lm(AQI ~ CO,AQI_data)
summary(linear rega)
##
## Call:
## lm(formula = AQI ~ CO, data = AQI_data)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -123.50 -74.59 -30.05
                            48.35 292.43
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
                          12.327 11.881 < 2e-16 ***
## (Intercept) 146.460
                                    6.236 1.67e-08 ***
## CO
                18.007
                            2.887
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 104.4 on 85 degrees of freedom
## Multiple R-squared: 0.3139, Adjusted R-squared: 0.3058
## F-statistic: 38.89 on 1 and 85 DF, p-value: 1.672e-08
# Predicting the Value of AQI for a given CO reading:
df1 = data.frame(CO = 1.69)
predict(linear_rega,df1)
## 176.8913
```

Inference: Here, P-value (1.67e-08) < Alpha (0.001) which shows that the regression model fits properly and also has a high F-statistic (38.89).

So, we reject H0.

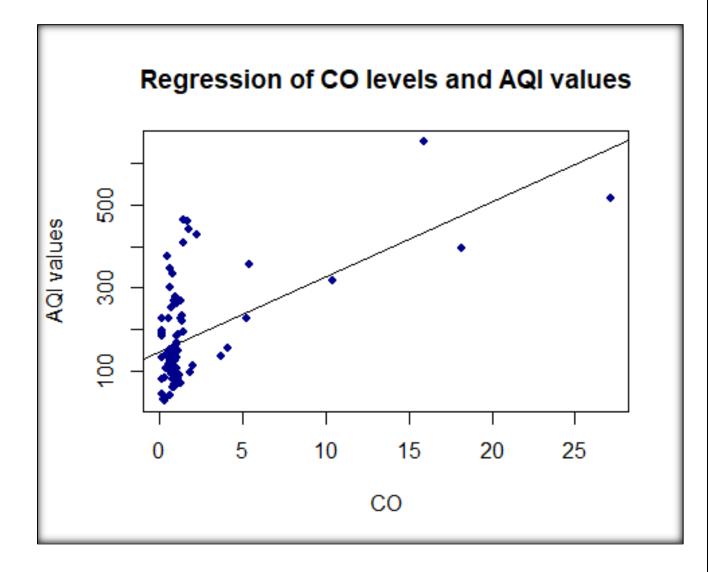
This shows that AQI and CO levels are highly related to each other.

Regression Equation: y = 146.460 + 18.007 * x

Thereafter, we have also predicted the value of AQI for a given CO value by using the *predict()* function.

Graphical Representation:

```
# Plotting scatter plot with Regression Line:
plot(x = AQI_data$CO, y = AQI_data$AQI,xlab = "CO",ylab = "AQI values",main
= "Regression of CO levels and AQI values",abline(linear_rega),cex = 1.3,p
ch=20,col="Dark Blue")
```



Inference: We can visualize the regression and regression line.

Multiple Regression:

```
rega = lm(AQI \sim PM2.5 + PM10 + NO + NO2 + NOx + NH3 + CO + SO2 + O3 + Benz
ene + Toluene + Xylene ,AQI_data)
summary(rega)
##
## Call:
## lm(formula = AQI \sim PM2.5 + PM10 + NO + NO2 + NOx + NH3 + CO +
##
       SO2 + O3 + Benzene + Toluene + Xylene, data = AQI_data)
##
## Residuals:
##
        Min
                  10
                       Median
                                     30
                                             Max
## -106.054 -24.209
                        0.804
                                 23.109
                                        154.998
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
                          17.92562
                                    -0.103
                                              0.9180
## (Intercept) -1.85229
## PM2.5
                0.93116
                           0.09160
                                   10.166 1.08e-15 ***
                                     8.394 2.33e-12 ***
## PM10
                0.63656
                           0.07584
## NO
               -0.09341
                           0.44768
                                   -0.209
                                              0.8353
## NO2
                           0.42835
                                    -1.246
               -0.53375
                                              0.2167
## NOx
                0.10776
                           0.26438
                                     0.408
                                              0.6847
                                   -0.344
## NH3
               -0.22906
                           0.66572
                                              0.7318
                                     5.101 2.52e-06 ***
## CO
                9.62891
                           1.88765
## S02
                                              0.5006
                0.34155
                           0.50460
                                     0.677
## 03
                0.18276
                           0.16717
                                     1.093
                                              0.2778
## Benzene
                1.25150
                           1.50435
                                      0.832
                                              0.4081
## Toluene
                2.21884
                           1.23219
                                     1.801
                                              0.0758 .
## Xylene
               -0.14488
                           0.69765 -0.208
                                              0.8361
## ---
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
## Residual standard error: 38.26 on 74 degrees of freedom
## Multiple R-squared: 0.9199, Adjusted R-squared: 0.9069
## F-statistic: 70.78 on 12 and 74 DF, p-value: < 2.2e-16
# Predicting the value of AQI:
xyz = data.frame(PM2.5=25.2,PM10=69,N0=28.5,N02=37,N0x=24.3,NH3=21.3,C0=0.
89, SO2=22.9, O3=108, Benzene=3.5, Toluene=6.9, Xylene=4)
predict(rega,xyz)
## 96.10458
```

Here, we have taken the Multiple Linear regresion of AQI (dependent variable) with respect to all the other Independent variables.

We can see that there is a very high F-statistic value (70.78) and also P-value (2.2e-16) < Alpha (0.05) so we can conclude that AQI is highly dependent on all these independent variables.

After that, we have also predicted the AQI for given values of all the other variables.

Correlation Matrix:

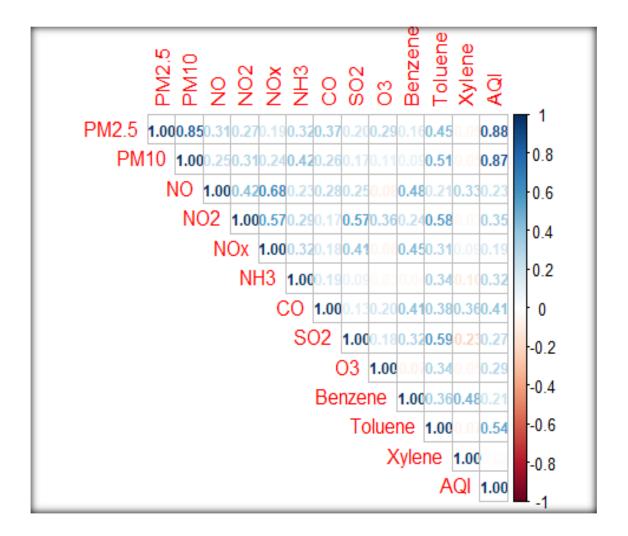
```
cor_matrix = cor(AQI_data[,c(3:15)],method = 'spearman')
round(cor_matrix,2)
##
           PM2.5
                 PM10
                         NO
                              N<sub>0</sub>2
                                    NOx
                                          NH3
                                                CO
                                                      S02
                                                            03 Benzene To
luene
                             0.27
                                   0.19 0.32 0.37
                                                                   0.16
## PM2.5
           1.00
                 0.85
                       0.31
                                                    0.20
                                                          0.29
0.45
## PM10
                                   0.24 0.42 0.26 0.17 0.11
           0.85
                 1.00
                       0.25
                             0.31
                                                                  0.09
0.51
## NO
                                   0.68 0.23 0.28 0.25 -0.08
            0.31
                 0.25
                       1.00
                             0.42
                                                                   0.48
0.21
## NO2
            0.27
                 0.31
                       0.42
                             1.00
                                   0.57 0.29 0.17 0.57 0.36
                                                                   0.24
0.58
## NOx
           0.19
                 0.24
                       0.68
                             0.57
                                   1.00 0.32 0.18 0.41 -0.06
                                                                   0.45
0.31
## NH3
           0.32
                 0.42
                       0.23
                             0.29
                                   0.32 1.00 0.19 0.09 -0.03
                                                                  -0.04
0.34
## CO
           0.37
                 0.26
                       0.28
                             0.17
                                   0.18 0.19 1.00 0.13
                                                         0.20
                                                                  0.41
0.38
## S02
                 0.17
                       0.25
                             0.57 0.41 0.09 0.13 1.00
           0.20
                                                          0.18
                                                                  0.32
0.59
## 03
                 0.11 -0.08 0.36 -0.06 -0.03 0.20 0.18
           0.29
                                                                  -0.07
                                                          1.00
0.34
## Benzene
           0.16
                 0.09
                       0.48
                             0.24
                                   0.45 -0.04 0.41 0.32 -0.07
                                                                   1.00
0.36
## Toluene 0.45 0.51
                       0.21 0.58 0.31 0.34 0.38 0.59 0.34
                                                                   0.36
1.00
## Xylene -0.05 -0.05 0.33 -0.03 0.09 -0.10 0.36 -0.23 -0.05
                                                                   0.48
-0.07
           0.88 0.87 0.23 0.35 0.19 0.32 0.41 0.27 0.29
## AQI
                                                                   0.21
0.54
##
          Xylene
                   AQI
            -0.05
## PM2.5
                  0.88
## PM10
            -0.05 0.87
            0.33 0.23
## NO
## NO2
            -0.03
                  0.35
## NOx
            0.09 0.19
## NH3
            -0.10 0.32
## CO
            0.36
                  0.41
## S02
            -0.23 0.27
## 03
            -0.05 0.29
## Benzene
            0.48 0.21
## Toluene
           -0.07 0.54
## Xylene
             1.00 -0.02
            -0.02 1.00
## AQI
```

Inference: Here, we can see the correlation matrix for the data.

Corrplot:

```
require(corrplot)
## Loading required package: corrplot
## corrplot 0.92 loaded

corrplot(cor(AQI_data[,c(3:15)],method = 'spearman'), method = 'number',ty
pe = 'upper',number.cex = 0.75, )
```

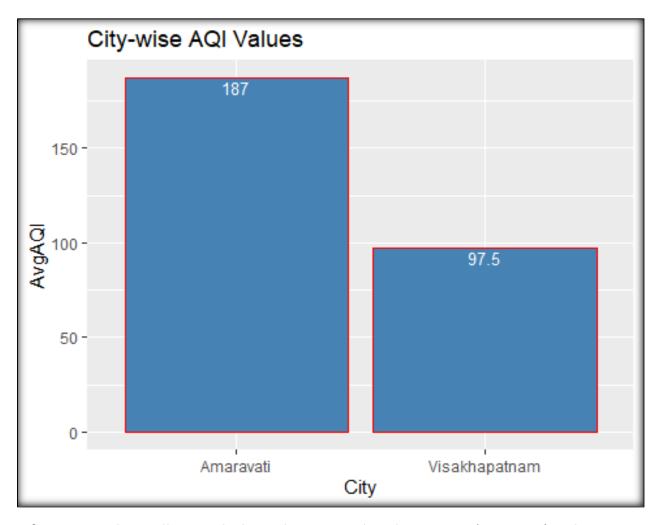


Inference: Here, we can see the correlation matrix using corrplot for the data.

With reference to the multiple regression in the above section, the independent variables namely "PM2.5", "PM10", "CO", "Toluene" are perfectly regressed wrt AQI and we can see in the regression matrix, these variables have the highest correlation with AQI.

▶ Phase - 9: Average AQI for given cities in a state:

```
City_avg = aggregate(x=AQI_data$AQI,by=list(AQI_data$City),FUN = mean)
City_avg_1 = City_avg[c(2,15),]
colnames(City_avg_1)[1] = "City"
colnames(City_avg_1)[2] = "AvgAQI"
ggplot(City_avg_1,aes(x = City, y = AvgAQI)) + ggtitle("City-wise AQI Values") + geom_bar(stat = "identity",color = "red", fill = "steelblue") + geom_text(aes(label = AvgAQI),vjust = 1.2,color = "white", position = position_dodge(0.9),size = 3.5 )
```



Inference: In the Andhra Pradesh, we have considered two cities 'Amravati' and 'Vishakhapatnam', looking at the bar plot, we can say that the mean Air Quality Index of 'Amravati' is higher than 'Vishakhapatnam', therefore Vishakhapatnam has safer air than Amravati.

➤ Phase – 9: Chi Square Test:

In this section, we took the AQI Status that was given as 'Good', 'Satisfactory', 'Moderate', 'Poor', 'Very Poor' and 'Severe' and converted them into numeric data i.e., 1, 2, 3, 4, 5 and 6 respectively.

```
getwd()
## [1] "E:/DA-IICT/Study/Sem 1/R Project/Indian_Air_Pollution_Data"
chi_sq = read.csv("Column_15_16.csv")
head(chi_sq)
##
     Toluene AQI Bucket
## 1
        5.92
                      3
## 2
        6.50
                      3
## 3
        7.95
## 4
       2.85
                      3
## 5
        2.79
                      3
                      4
## 6
        3.82
# Chi-Square test:
data1 = data.matrix(chi_sq)
chisq.test(data1)
## Warning in chisq.test(data1): Chi-squared approximation may be incorrec
t
##
## Pearson's Chi-squared test
##
## data: data1
## X-squared = 146.95, df = 86, p-value = 4.708e-05
```

Inference: Here, we have performed Chi-Square test between two variables CO and AQI Status.

We can see that the p-value (4.708e-05) < Alpha (0.05), and also observed Chisquared value (146.95) > Critical Chi-squared value (101.87) for 86 degrees of freedom.

So, we reject the H0 and can conclude that there is a significant relationship between these variables CO and AQI status.

Conclusion:

- We found that all the factors i.e., "NO2", "PM2.5", "PM10", "CO", "CO2", "Benzene", "O3", "Xylene", "Toluene", etc. are the pollutants that generally affect the AQI of any place.
- From the above-mentioned factors, "PM2.5", "PM10", "CO" and "Toluene" are the major affecting factors for AQI and also have very high Correlation with it.
- "O3" and "Xylene" are least corelated and don't have much effect on the AQI values.
- "Uttar-Pradesh" and "Gujarat" have the highest mean AQI which means that these states have high air pollution and lie in the "Severe" category.
- "Telangana" has a lower mean AQI and comes in the "Good" AQI category.
 And on the second number is "Karnataka" with "Satisfactory" AQI values.

* References:

Dataset: Air Quality Data India.

Link: https://www.kaggle.com/rohanrao/air-quality-data-in-india?select=station_day.csv