Assignment 2 - Group B

Problem Statement

Perform the following operations using Python on the Air quality data sets

- a. Data cleaning
- b. Data integration
- c. Data transformaton
- d. Error correcting
- e. Data model building

In [175...

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

Reading data from CSV file

| In [176 | <pre>Tej = pd.read_csv("C:\\Users\\Shree\\Desktop\\dsbdl_lab\\airquality.csv")</pre> | |
|---------|--|--|
| In [177 | Теј | |

| Out[177]: | | Unnamed: 0 | Ozone | Solar.R | Wind | Temp | Month | Day | Humidity |
|-----------|-----|------------|-------|---------|------|------|-------|-----|----------|
| _ | 0 | 1 | 41.0 | 190.0 | 7.4 | 67 | 5 | 1 | High |
| | 1 | 2 | 36.0 | 118.0 | 8.0 | 72 | 5 | 2 | Low |
| | 2 | 3 | 12.0 | 149.0 | 12.6 | 74 | 5 | 3 | High |
| | 3 | 4 | 18.0 | 313.0 | 11.5 | 62 | 5 | 4 | Medium |
| | 4 | 5 | NaN | NaN | 14.3 | 56 | 5 | 5 | High |
| | ••• | | | | | | | | |
| | 148 | 149 | 30.0 | 193.0 | 6.9 | 70 | 9 | 26 | High |
| | 149 | 150 | NaN | 145.0 | 13.2 | 77 | 9 | 27 | Low |
| | 150 | 151 | 14.0 | 191.0 | 14.3 | 75 | 9 | 28 | High |
| | 151 | 152 | 18.0 | 131.0 | 8.0 | 76 | 9 | 29 | Medium |
| | 152 | 153 | 20.0 | 223.0 | 11.5 | 68 | 9 | 30 | High |

153 rows × 8 columns

```
Tej.head
In [178...
           <bound method NDFrame.head of</pre>
                                                Unnamed: 0 Ozone Solar.R Wind Temp Month
Out[178]:
           Day Humidity
                                                               5
                          1
                              41.0
                                       190.0
                                               7.4
                                                       67
                                                                     1
                                                                           High
           1
                          2
                              36.0
                                       118.0
                                               8.0
                                                       72
                                                               5
                                                                     2
                                                                            Low
           2
                          3
                              12.0
                                       149.0 12.6
                                                       74
                                                               5
                                                                     3
                                                                           High
           3
                                                               5
                          4
                              18.0
                                       313.0 11.5
                                                       62
                                                                     4
                                                                         Medium
           4
                          5
                                                               5
                                                                     5
                               NaN
                                         NaN 14.3
                                                       56
                                                                           High
                               . . .
                                         . . .
                        . . .
                                               . . .
                                                                            . . .
           148
                        149
                              30.0
                                       193.0
                                               6.9
                                                       70
                                                               9
                                                                   26
                                                                           High
           149
                        150
                               NaN
                                       145.0 13.2
                                                       77
                                                               9
                                                                   27
                                                                            Low
           150
                        151
                              14.0
                                       191.0 14.3
                                                       75
                                                                   28
                                                                           High
                                                       76
                                                                   29
                                                                         Medium
           151
                        152
                              18.0
                                       131.0
                                              8.0
                                                               9
           152
                        153
                              20.0
                                       223.0 11.5
                                                       68
                                                               9
                                                                   30
                                                                           High
           [153 rows x 8 columns]>
In [179...
           Tej.shape
           (153, 8)
Out[179]:
In [180...
           Tej.isnull().sum()
           Unnamed: 0
                           0
Out[180]:
           0zone
                          37
           Solar.R
                           7
           Wind
                           0
           Temp
                           0
           Month
                           0
                           0
           Day
           Humidity
                           4
           dtype: int64
```

Data Cleaning

Removing unwanted columns

```
In [181... Tej.drop(Tej.iloc[: , [0]], axis=1 , inplace=True)
In [182... Tej
```

| Out[182]: | | Ozone | Solar.R | Wind | Temp | Month | Day | Humidity |
|-----------|-----|-------|---------|------|------|-------|-----|----------|
| - | 0 | 41.0 | 190.0 | 7.4 | 67 | 5 | 1 | High |
| | 1 | 36.0 | 118.0 | 8.0 | 72 | 5 | 2 | Low |
| | 2 | 12.0 | 149.0 | 12.6 | 74 | 5 | 3 | High |
| | 3 | 18.0 | 313.0 | 11.5 | 62 | 5 | 4 | Medium |
| | 4 | NaN | NaN | 14.3 | 56 | 5 | 5 | High |
| | ••• | | | | | | | |
| | 148 | 30.0 | 193.0 | 6.9 | 70 | 9 | 26 | High |
| | 149 | NaN | 145.0 | 13.2 | 77 | 9 | 27 | Low |
| | 150 | 14.0 | 191.0 | 14.3 | 75 | 9 | 28 | High |
| | 151 | 18.0 | 131.0 | 8.0 | 76 | 9 | 29 | Medium |
| | 152 | 20.0 | 223.0 | 11.5 | 68 | 9 | 30 | High |

153 rows × 7 columns

Replacing Numerical Null values

```
In [183... Tej['Ozone']=Tej['Ozone'].fillna(Tej['Ozone'].mean())
    Tej['Solar.R']=Tej['Solar.R'].fillna(Tej['Solar.R'].mean())
    Tej["Wind"] = Tej["Wind"].fillna(Tej["Wind"].mean())
```

Replacing Categorical Null values

```
In [184...
           Tej['Humidity']=Tej['Humidity'].fillna(Tej['Humidity'].mode()[0])
           Tej.isnull().sum()
          0zone
Out[184]:
           Solar.R
                       0
          Wind
                       0
           Temp
          Month
          Day
          Humidity
          dtype: int64
In [185...
          Tej.dtypes
          0zone
                       float64
Out[185]:
                       float64
           Solar.R
          Wind
                       float64
           Temp
                         int64
          Month
                         int64
                         int64
          Day
          Humidity
                        object
          dtype: object
```

Data Transformation

Using Label Encoding on Humidity column

```
In [186...
          from sklearn.preprocessing import LabelEncoder
          label=LabelEncoder()
          Tej['Humidity']=label.fit_transform(Tej['Humidity'])
          Tej['Humidity'].unique()
          array([0, 1, 2])
Out[186]:
          Tej.dtypes
In [187...
                       float64
          0zone
Out[187]:
          Solar.R
                      float64
          Wind
                       float64
                         int64
          Temp
          Month
                         int64
          Day
                         int64
          Humidity
                         int32
          dtype: object
```

Data Integration

Subset Creation (Row-wise)

```
In [188...
           #Subset-1
           s1 = Tej.iloc[[1,2,3,6,12,28],:]
           s1
Out[188]:
               Ozone Solar.R Wind Temp Month Day Humidity
            1
                 36.0
                        118.0
                                8.0
                                       72
                                                5
                                                               1
            2
                 12.0
                        149.0
                                12.6
                                       74
                                                5
                                                     3
                                                               0
            3
                 18.0
                        313.0
                               11.5
                                       62
                                                5
                                                     4
                                                               2
                                                   7
                                                               0
            6
                 23.0
                        299.0
                                8.6
                                       65
                                                5
           12
                 11.0
                        290.0
                                9.2
                                       66
                                                    13
                                                               0
                 45.0
                                                               0
           28
                        252.0
                               14.9
                                       81
                                                5
                                                    29
In [189...
           #Subset-2
           s2 = Tej.iloc[[70,81,95,105,123,137,149],:]
           s2
```

| Out[189]: | | Ozone | Solar.R | Wind | Temp | Month | Day | Humidity |
|-----------|-----|----------|------------|------|------|-------|-----|----------|
| | 70 | 85.00000 | 175.000000 | 7.4 | 89 | 7 | 10 | 0 |
| | 81 | 16.00000 | 7.000000 | 6.9 | 74 | 7 | 21 | 1 |
| | 95 | 78.00000 | 185.931507 | 6.9 | 86 | 8 | 4 | 2 |
| | 105 | 65.00000 | 157.000000 | 9.7 | 80 | 8 | 14 | 1 |
| | 123 | 96.00000 | 167.000000 | 6.9 | 91 | 9 | 1 | 2 |
| | 137 | 13.00000 | 112.000000 | 11.5 | 71 | 9 | 15 | 1 |
| | 149 | 42.12931 | 145.000000 | 13.2 | 77 | 9 | 27 | 1 |

Merging Subsets

merge = pd.concat([s1,s2]) In [190...

In [191...

merge

Out[191]:

| | Ozone | Solar.R | Wind | Temp | Month | Day | Humidity |
|-----|----------|------------|------|------|-------|-----|----------|
| 1 | 36.00000 | 118.000000 | 8.0 | 72 | 5 | 2 | 1 |
| 2 | 12.00000 | 149.000000 | 12.6 | 74 | 5 | 3 | 0 |
| 3 | 18.00000 | 313.000000 | 11.5 | 62 | 5 | 4 | 2 |
| 6 | 23.00000 | 299.000000 | 8.6 | 65 | 5 | 7 | 0 |
| 12 | 11.00000 | 290.000000 | 9.2 | 66 | 5 | 13 | 0 |
| 28 | 45.00000 | 252.000000 | 14.9 | 81 | 5 | 29 | 0 |
| 70 | 85.00000 | 175.000000 | 7.4 | 89 | 7 | 10 | 0 |
| 81 | 16.00000 | 7.000000 | 6.9 | 74 | 7 | 21 | 1 |
| 95 | 78.00000 | 185.931507 | 6.9 | 86 | 8 | 4 | 2 |
| 105 | 65.00000 | 157.000000 | 9.7 | 80 | 8 | 14 | 1 |
| 123 | 96.00000 | 167.000000 | 6.9 | 91 | 9 | 1 | 2 |
| 137 | 13.00000 | 112.000000 | 11.5 | 71 | 9 | 15 | 1 |
| 149 | 42.12931 | 145.000000 | 13.2 | 77 | 9 | 27 | 1 |

Deriving correlation between Columns

In [192... corr = Tej.corr()

In [193...

corr

| Out[193]: | | | r.R | Wind | ٠ | Temp | Мо | nth | Day | Humidity | | |
|-----------|---|--------------------------------|------------------|--|---------|------------------|--------|---------|--------|-----------|-----------|-----------|
| | Ozone | Ozone 1.000000 0.302970 | | | | | 0.60 | 08742 | 0.149 | 081 | -0.011355 | -0.012681 |
| | Solar.R | 0.30 | 2970 | 1.000000 -0.0 | | .055245 0.262569 | | 52569 | -0.072 | 904 | -0.145621 | -0.020428 |
| | Wind | Wind -0.530936 Temp 0.608742 | | -0.0552 | 245 1 | .000000 | 0.45 | 7988 | -0.178 | 293 | 0.027181 | 0.090264 |
| | Temp | | | 0.2625 | 669 -0 | .457988 | 3 1.00 | 00000 | 0.420 | 947 | -0.130593 | 0.008397 |
| | Month 0.149081 Day -0.011355 | | 081 -0.072904 -0 | | .178293 | 0.42 | 20947 | 1.000 | 000 | -0.007962 | 0.043569 | |
| | | | 1355 | -0.1456 | 521 0 | .027181 | -0.13 | 30593 | -0.007 | 962 | 1.000000 | -0.038271 |
| | Humidity | -0.01 | 2681 | -0.0204 | 128 0 | .090264 | 0.00 | 08397 | 0.043 | 569 | -0.038271 | 1.000000 |
| Out[194]: | <pre>sns.heatmap(corr, <axessubplot:></axessubplot:></pre> | | | <pre>s.heatmap(corr, vmin = -1, vmax = 1, a xesSubplot:></pre> | | | | annot= | • | | | |
| | Ozone - | 1 | 0.3 | -0.53 | 0.61 | 0.15 | -0.011 | -0.013 | | 1.00 | | |
| | Solar.R - | 0.3 | 1 | -0.055 | 0.26 | -0.073 | -0.15 | -0.02 | | 0.75 | | |
| | Wind - | -0.53 | -0.055 | 1 | -0.46 | -0.18 | 0.027 | 0.09 | | 0.25 | | |
| | Temp - | 0.61 | 0.26 | -0.46 | 1 | 0.42 | -0.13 | 0.0084 | - | 0.00 | | |
| | Month - | 0.15 | -0.073 | -0.18 | 0.42 | 1 | -0.008 | 0.044 | - | -0.25 | 5 | |
| | Day - | -0.011 | -0.15 | 0.027 | -0.13 | -0.008 | 1 | -0.038 | | -0.50 | | |
| | Humidity - | -0.013 | -0.02 | 0.09 | 0.0084 | 0.044 | -0.038 | 1 | | -0.75 | | |
| | ' | Ozone | Solar.R | Wind | Temp | Month | Day I | lumidit | | -1.00 |) | |

Building Data Model

Using Linear Regression

```
In [195... x = Tej[["Ozone"]]
y = Tej[["Temp"]]

In [196... from sklearn.model_selection import train_test_split

In [197... x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2)

In [198... from sklearn.linear_model import LinearRegression

In [199... lr = LinearRegression()

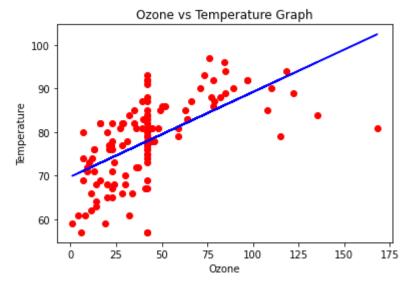
In [200... model = lr.fit(x_train, y_train)
```

```
In [201... y_predict = model.predict(x_test)
```

Plotting Graph

```
In [202... import matplotlib.pyplot as plt

In [203... plt.scatter(x_train, y_train, color="red")
    plt.plot(x_train, lr.predict(x_train), color="blue")
    plt.xlabel("Ozone")
    plt.ylabel("Temperature")
    plt.title("Ozone vs Temperature Graph")
    plt.show()
```



Calculating Metrics

```
In [213...
          from sklearn.metrics import mean_absolute_error,mean_squared_error,r2_score
          import numpy as np
In [214...
          MSE = mean_squared_error(y_test,y_predict)
          MAE = mean_absolute_error(y_test,y_predict)
          r2_score = r2_score(y_test,y_predict)
          RMSE = np.sqrt(MSE)
          print("MSE : {} \nMAE : {} \nMAE : {} \nR2 Score : {}".format(MSE, RMSE, MAE, r2_s
In [215...
          MSE: 58.75683748803407
          RMSE: 7.665300873940571
          MAE : 5.3610432690122485
          R2 Score: 0.33559267613488397
 In [ ]:
 In [ ]:
```