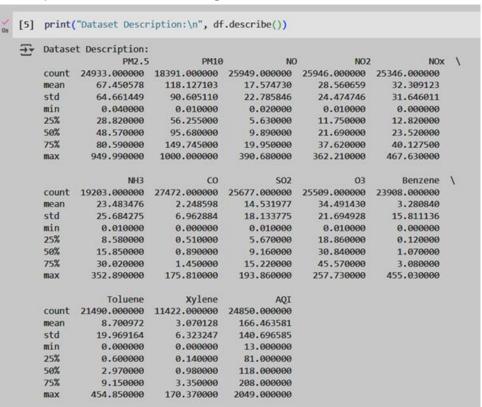
· Reading data with the help of pandas library.

```
[2] import pandas as pd
import numpy as np

(4) df = pd.read_csv('data.csv')
```

Description of the data set using describe method.



Dropped unnecessary column

```
columns_to_drop = ['Xylene']
df.drop(columns=columns_to_drop, inplace=True)
```

• Dropped rows with maximum number of missing values.

```
df.dropna(thresh=df.shape[1] - 1, inplace=True)
```

• Take care of missing data.

```
df.fillna(df.select_dtypes(include=['number']).mean(), inplace=True)
```

• Create dummy variables.

```
+ + + 0 0 0 B I
o df_encoded = pd.gst_dummles(df, columns=['AQL_Sucket'], drop_first=True)
              print(df_encoded)

        City
        Dute
        PPC.5
        PPDB
        NO
        NO2
        NO

        Absordabad
        15-05-2019
        37.55
        122.41
        15.08
        05.12
        56.72

        Absordabad
        16-05-2019
        37.97
        116.32
        14.67
        29.71
        55.63

        Absordabad
        17-05-2019
        35.48
        130.07
        18.02
        77.41
        56.41

        Absordabad
        18-05-2019
        34.11
        131.31
        11.27
        75.23
        51.43

        Absordabad
        18-05-2019
        37.40
        131.73
        34.56
        68.00
        69.77

3
              29525 Visakhapatnam 26-06-2020 7.63 32.27 5.91 23.27 17.19
              29506 Visakhapatnam 27-06-2020 15.02 50.94 7.68 25.06 19.54 29527 Visakhapatnam 28-06-2020 24.08 74.09 3.47 26.06 18.53 29528 Visakhapatnam 29-06-2020 22.91 65.73 3.45 29.53 18.33
              29529 Visakhapatnam 30-06-2028 16.64 49.97 A.95 29.26 18.60
            | MeS | 10 | 502 | 00 | Bentere | Tellums | AGS | 1595 | 25.269129 | 15.08 | 163.61 | 46.23 | 15.44 | 85.54 | 281.0 | 1596 | 25.249129 | 14.67 | 97.26 | 31.08 | 15.55 | 81.09 | 336.4 | 1597 | 25.249129 | 34.92 | 98.05 | 48.22 | 15.93 | 82.73 | 356.4 | 3598 | 25.249129 | 34.36 | 60.56 | 64.22 | 15.93 | 82.73 | 359.8 | 359.92 | 35.249129 | 34.36 | 60.56 | 60.59 | 35.53 | 64.17 | 547.8
             29525 11.150000 0.40 6.87 19.90 1.85 5.37 47.0
29526 12.470000 0.47 0.55 23.30 2.24 12.07 41.0
29527 13.99000 0.52 12.72 90.14 0.74 2.71 79.0
29528 10.730000 0.48 0.42 20.90 0.01 0.46 08.0
29529 10.430000 0.52 3.04 28.30 0.00 0.00 0.40
                                AQC_Bucket_Moderate AQC_Bucket_Poor AQC_Bucket_Satisfactory \
false false False False
false false false
             1595
1596
1597
              1598
1599
                                                                                                                            False
              29575
                                                                          False
                                                                                                                           False
                                                                                                                           False
False
False
False
              29526
29527
              29528
29529
                                                                           False
              1595
1596
1597
```

• Finding Outlier with the method of IQR:

```
df_numeric = df.select_dtypes(include=[float, int])
    Q1 = df numeric.quantile(0.25)
    Q3 = df_numeric.quantile(0.75)
    IOR = Q3 - Q1
    outliers = ((df numeric < (Q1 - 1.5 * IQR)) | (df numeric > (Q3 + 1.5 * IQR)))
    print("Outliers:\n", df_numeric[outliers.any(axis=1)])
→ Outliers:
            PM2.5
                  PM10
                          NO NO2
                                        NOx
                                                   NH3
                                                          co
                                                                 S02
                                                                        03 \
    1595
           37.55 122.41 15.08 85.12 58.72 25.249129 15.08 163.01 48.23
           33.97 116.32 14.67 79.71 55.61 25.249129 14.67 91.26 51.86
    1596
    1597
           35.48 130.07 18.02 77.61 58.41 25.249129 18.02
                                                              98.35 38.99
    1598
           34.11 138.31 13.27 75.23 51.83 25.249129 13.27
                                                              88.66 42.22
           33.69 111.73 34.56 68.90 69.77 25.249129 34.56 80.90 36.95
    1599
                                        ...
          73.83 125.02 2.93 30.68 18.71 11.440000
                                                       1.03 10.25 83.49
    29359
    29361 177.20 326.40 37.86 79.29 72.95 22.010000 2.08 16.45 44.07
          53.30 128.35 6.90 59.38 37.20 13.410000
36.68 76.40 2.55 35.01 20.31 11.700000
    29373
                                                        0.94
                                                               8.17 16.86
                                                       1.22
                                                               5.77 20.25
    29403
    29404 43.59 107.91 2.08 39.80 22.62 12.820000 1.29
                                                             4.83 21.39
          Benzene Toluene
            16.44 85.54 281.0
    1595
    1596
            15.55
                    83.89 330.0
    1597
                   83.83 356.0
            15.88
            15.93
                    82.73 359.0
    1598
                   84.17 547.0
    1599
            15.53
    29359
             3.87
                     7.65 181.0
             9.40
                   15.42 326.0
    29361
    29373
             5.99
                    31.46 113.0
    29403
             4.01
                    20.22
                           90.0
             5.23
                    24.81 102.0
    29404
    [6081 rows x 12 columns]
```

Standardization using sk learn library

```
◆ ◆ ◆ ⊕ ♥ 图 目 :
    df_numeric_scaled = pd.DataFrame(scaler.fit_transform(df_numeric), columns=df_numeric.columns)
    print("Standardized Dataframe:\n", df_numeric_scaled)
Standardized Dataframe:
          PM2.5 PM10 NO NO2 NOX NH3 CC
-0.447850 0.017618 -0.138444 2.434413 0.805041 0.000000 3.749043
-0.513731 -0.055051 -0.156647 2.195051 0.702156 0.000000 3.636224
           -0.485943 0.109020 -0.007911 2.102138 0.794786 0.000000 4.558037 -0.511155 0.207344 -0.218806 1.996837 0.577105 0.000000 3.250989
           -0.518884 -0.109821 0.726448 1.716770 1.170598 0.000000 9.109317
    16349 -0.998455 -1.057974 -0.545582 -0.302097 -0.568858 -0.578799 -0.273914
    16350 -0.862460 -0.835195 -0.466996 -0.222899 -0.491115 -0.524610 -0.271162
    16351 -0.690212 -0.558958 -0.656136 -0.178655 -0.590692 -0.544315 -0.257404
    16352 -0.717264 -0.658714 -0.654804 -0.025127 -0.531145 -0.596862 -0.268411
    16353 -0.832648 -0.846769 -0.628164 -0.037073 -0.515596 -0.624778 -0.257404
                                   Benzene
                                              Toluene
                  502
           11.598630 0.675192 0.707820 3.814665 1.257839
            6.076148 0.850690 0.658131 3.732573 1.718807
             6.621853 0.228470 0.676555 3.729588 1.963403
             5.876030 0.384629 0.679346 3.674859 1.991625
            5.278755 0.129843 0.657014 3.746504 3.760238
    16349 -0.419216 -0.694466 -0.129073 -0.174045 -0.943520
    16350 -0.289909 -0.530088 -0.084968 0.159301 -0.999965
    16351 0.031049 -0.199397 -0.168713 -0.331265 -0.727147
           -0.299915 -0.159753 -0.209469 -0.440722 -0.745962
    16353 -0.190620 -0.288355 -0.210027 -0.441219 -0.877667
    [16354 rows x 12 columns]
```

Normalization:

```
normalizer = MinMaxScaler()
      df_numeric_normalized = pd.DataFrame(normalizer.fit_transform(df_numeric_scaled), columns=df_numeric.co
print("Standardized and Normalized Dataframe:\n", df_numeric_normalized)

→ Standardized and Normalized Dataframe:
              PRZ-5 PR10 NO NO2 NOX NH3 CC
0.053421 0.133280 0.055728 0.306924 0.209341 0.076743 0.162220
0.048190 0.126637 0.054210 0.287414 0.189730 0.076743 0.157810
                                                                                                           co \
               0.050397 0.141634 0.066615 0.279841 0.199284 0.076743 0.193847

        0.048395
        0.150621
        0.049926
        0.271259
        0.176834
        0.076743
        0.142750

        0.047781
        0.121631
        0.127860
        0.248431
        0.238042
        0.076743
        0.371773

      4
      16349 0.009702 0.034967 0.021773 0.083880 0.058649 0.033873 0.004948
16350 0.020501 0.055330 0.028327 0.090335 0.066667 0.037886 0.005056
      16351 0.034177 0.080578 0.012553 0.093942 0.056397 0.036427 0.005594
      16352 0.032029 0.071461 0.012664 0.106455 0.062538 0.032535 0.005164 16353 0.022868 0.054272 0.014886 0.105481 0.064142 0.030467 0.005594
               SO2 O3 Benzene Toluene AQI
0.876014 0.187102 0.036129 0.188062 0.194182
               0.490407 0.201187 0.034174 0.184434 0.229818
               0.528511 0.151249 0.034899 0.184303 0.248727
               0.476434 0.163782 0.035009 0.181884 0.250909
               0.434729 0.143334 0.034130 0.185050 0.387636
      16350 0.045897 0.090369 0.004923 0.026536 0.019636 16351 0.068308 0.116910 0.001626 0.004859 0.040727
      16352 0.045198 0.120092 0.000022 0.000022 0.039273
      16353 0.052830 0.109770 0.000000 0.000000 0.029091
      [16354 rows x 12 columns]
```