

CL-I 11 DMV

July 20, 2025

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ROLL NO.01  
COURSE: AI&DS  
CLASS: BE  
SUB:Computer Laboratory-I (DMV) '''
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[ ]: Data Visualization using matplotlib  
Problem Statement: Analyzing Air Quality Index (AQI) Trends in a City  
Dataset: "City_Air_Quality.csv"  
Description: The dataset contains information about air quality measurements in  
a specific  
city over a period of time. It includes attributes such as date, time,  
pollutant levels (e.g., PM2.5,  
PM10, CO), and the Air Quality Index (AQI) values. The goal is to use the  
matplotlib library  
to create visualizations that effectively represent the AQI trends and patterns  
for different  
pollutants in the city.  
Tasks to Perform:  
1. Import the "City_Air_Quality.csv" dataset.  
2. Explore the dataset to understand its structure and content.  
3. Identify the relevant variables for visualizing AQI trends, such as date,  
pollutant levels,  
and AQI values.  
4. Create line plots or time series plots to visualize the overall AQI trend  
over time.  
5. Plot individual pollutant levels (e.g., PM2.5, PM10, CO) on separate line  
plots to  
visualize their trends over time.  
6. Use bar plots or stacked bar plots to compare the AQI values across  
different dates or  
time periods.  
7. Create box plots or violin plots to analyze the distribution of AQI values  
for different  
pollutant categories.  
8. Use scatter plots or bubble charts to explore the relationship between AQI  
values and
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pollutant levels.
9. Customize the visualizations by adding labels, titles, legends, and appropriate color schemes

[5]: #1. Import the "City_Air_Quality.csv" dataset.

```
[33]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

[2]: df=pd.read_csv("city_day.csv")

[7]: df

```
[7]:
```

	City	Date	PM2.5	PM10	NO	NO2	NOx	NH3	\
0	Ahmedabad	2015-01-01	NaN	NaN	0.92	18.22	17.15	NaN	
1	Ahmedabad	2015-01-02	NaN	NaN	0.97	15.69	16.46	NaN	
2	Ahmedabad	2015-01-03	NaN	NaN	17.40	19.30	29.70	NaN	
3	Ahmedabad	2015-01-04	NaN	NaN	1.70	18.48	17.97	NaN	
4	Ahmedabad	2015-01-05	NaN	NaN	22.10	21.42	37.76	NaN	
...	
29526	Visakhapatnam	2020-06-27	15.02	50.94	7.68	25.06	19.54	12.47	
29527	Visakhapatnam	2020-06-28	24.38	74.09	3.42	26.06	16.53	11.99	
29528	Visakhapatnam	2020-06-29	22.91	65.73	3.45	29.53	18.33	10.71	
29529	Visakhapatnam	2020-06-30	16.64	49.97	4.05	29.26	18.80	10.03	
29530	Visakhapatnam	2020-07-01	15.00	66.00	0.40	26.85	14.05	5.20	
	CO	S02	O3	Benzene	Toluene	Xylene	AQI	AQI_Bucket	
0	0.92	27.64	133.36	0.00	0.02	0.00	NaN	NaN	
1	0.97	24.55	34.06	3.68	5.50	3.77	NaN	NaN	
2	17.40	29.07	30.70	6.80	16.40	2.25	NaN	NaN	
3	1.70	18.59	36.08	4.43	10.14	1.00	NaN	NaN	
4	22.10	39.33	39.31	7.01	18.89	2.78	NaN	NaN	
...	
29526	0.47	8.55	23.30	2.24	12.07	0.73	41.0	Good	
29527	0.52	12.72	30.14	0.74	2.21	0.38	70.0	Satisfactory	
29528	0.48	8.42	30.96	0.01	0.01	0.00	68.0	Satisfactory	
29529	0.52	9.84	28.30	0.00	0.00	0.00	54.0	Satisfactory	
29530	0.59	2.10	17.05	NaN	NaN	NaN	50.0	Good	

[29531 rows x 16 columns]

[9]: #2. Explore the dataset to understand its structure and content.

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[9]: df.head(5)
```

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[9]:      City        Date  PM2.5  PM10    NO   NO2   NOx  NH3    CO   SO2  \
0  Ahmedabad  2015-01-01    NaN    NaN  0.92  18.22  17.15  NaN  0.92  27.64
1  Ahmedabad  2015-01-02    NaN    NaN  0.97  15.69  16.46  NaN  0.97  24.55
2  Ahmedabad  2015-01-03    NaN    NaN  17.40  19.30  29.70  NaN  17.40  29.07
3  Ahmedabad  2015-01-04    NaN    NaN  1.70  18.48  17.97  NaN  1.70  18.59
4  Ahmedabad  2015-01-05    NaN    NaN  22.10  21.42  37.76  NaN  22.10  39.33

      O3  Benzene  Toluene  Xylene  AQI  AQI_Bucket
0  133.36     0.00     0.02     0.00  NaN      NaN
1   34.06     3.68     5.50     3.77  NaN      NaN
2   30.70     6.80    16.40     2.25  NaN      NaN
3   36.08     4.43    10.14     1.00  NaN      NaN
4   39.31     7.01    18.89     2.78  NaN      NaN
```

```
[11]: df.info
```

```
[11]: <bound method DataFrame.info of
      City          Date  PM2.5  PM10
NO   NO2   NOx  NH3  \
0    Ahmedabad  2015-01-01    NaN    NaN  0.92  18.22  17.15  NaN
1    Ahmedabad  2015-01-02    NaN    NaN  0.97  15.69  16.46  NaN
2    Ahmedabad  2015-01-03    NaN    NaN  17.40  19.30  29.70  NaN
3    Ahmedabad  2015-01-04    NaN    NaN  1.70  18.48  17.97  NaN
4    Ahmedabad  2015-01-05    NaN    NaN  22.10  21.42  37.76  NaN
...
...
29526  Visakhapatnam  2020-06-27  15.02  50.94  7.68  25.06  19.54  12.47
29527  Visakhapatnam  2020-06-28  24.38  74.09  3.42  26.06  16.53  11.99
29528  Visakhapatnam  2020-06-29  22.91  65.73  3.45  29.53  18.33  10.71
29529  Visakhapatnam  2020-06-30  16.64  49.97  4.05  29.26  18.80  10.03
29530  Visakhapatnam  2020-07-01  15.00  66.00  0.40  26.85  14.05  5.20

      CO   SO2          O3  Benzene  Toluene  Xylene  AQI  AQI_Bucket
0   0.92  27.64  133.36     0.00     0.02     0.00  NaN      NaN
1   0.97  24.55  34.06     3.68     5.50     3.77  NaN      NaN
2   17.40  29.07  30.70     6.80    16.40     2.25  NaN      NaN
3   1.70  18.59  36.08     4.43    10.14     1.00  NaN      NaN
4   22.10  39.33  39.31     7.01    18.89     2.78  NaN      NaN
...
...
29526   0.47   8.55  23.30     2.24    12.07     0.73  41.0      Good
29527   0.52  12.72  30.14     0.74     2.21     0.38  70.0  Satisfactory
29528   0.48   8.42  30.96     0.01     0.01     0.00  68.0  Satisfactory
29529   0.52   9.84  28.30     0.00     0.00     0.00  54.0  Satisfactory
29530   0.59   2.10  17.05     NaN     NaN     NaN  50.0      Good
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[29531 rows x 16 columns]>
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```
[13]: df.describe()
```

```
[13]:          PM2.5        PM10         NO        NO2        NOx \
count  24933.000000  18391.000000  25949.000000  25946.000000  25346.000000
mean   67.450578    118.127103   17.574730    28.560659    32.309123
std    64.661449    90.605110   22.785846    24.474746    31.646011
min    0.040000    0.010000   0.020000    0.010000    0.000000
25%   28.820000    56.255000   5.630000   11.750000   12.820000
50%   48.570000    95.680000   9.890000   21.690000   23.520000
75%   80.590000   149.745000  19.950000  37.620000  40.127500
max   949.990000  1000.000000  390.680000  362.210000  467.630000

          NH3         CO        SO2         O3      Benzene \
count  19203.000000  27472.000000  25677.000000  25509.000000  23908.000000
mean   23.483476    2.248598    14.531977   34.491430    3.280840
std    25.684275    6.962884    18.133775   21.694928   15.811136
min    0.010000    0.000000    0.010000   0.010000    0.000000
25%   8.580000    0.510000    5.670000   18.860000   0.120000
50%   15.850000   0.890000   9.160000   30.840000   1.070000
75%   30.020000   1.450000   15.220000  45.570000   3.080000
max   352.890000  175.810000  193.860000  257.730000  455.030000

          Toluene       Xylene       AQI
count  21490.000000  11422.000000  24850.000000
mean   8.700972     3.070128    166.463581
std    19.969164    6.323247    140.696585
min    0.000000    0.000000   13.000000
25%   0.600000    0.140000   81.000000
50%   2.970000    0.980000  118.000000
75%   9.150000    3.350000  208.000000
max   454.850000  170.370000  2049.000000
```

```
[15]: df['Date'] = pd.to_datetime(df['Date'])
```

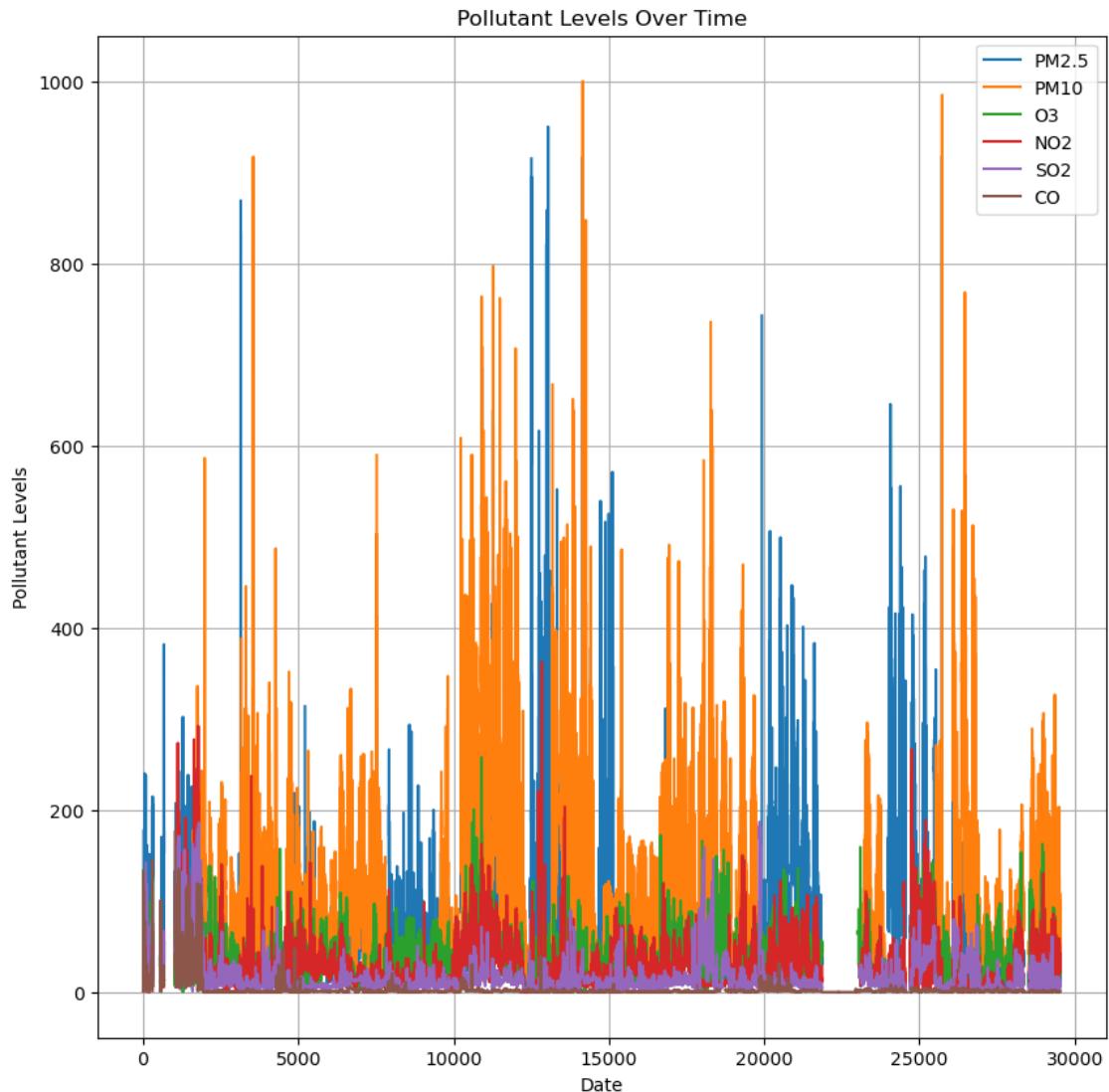
```
[14]: #3. Identify the relevant variables for visualizing AQI trends, such as date,  
      ↪pollutant levels, and AQI values.
```

```
[17]: # Plotting pollutant levels
pollutants = ['PM2.5', 'PM10', 'O3', 'NO2', 'SO2', 'CO']

plt.figure(figsize=(10, 10))
for pollutant in pollutants:
    if pollutant in df.columns:
        plt.plot(df.index, df[pollutant], label=pollutant)

plt.xlabel('Date')
plt.ylabel('Pollutant Levels')
```

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plt.title('Pollutant Levels Over Time')
plt.legend()
plt.grid(True)
plt.show()
```



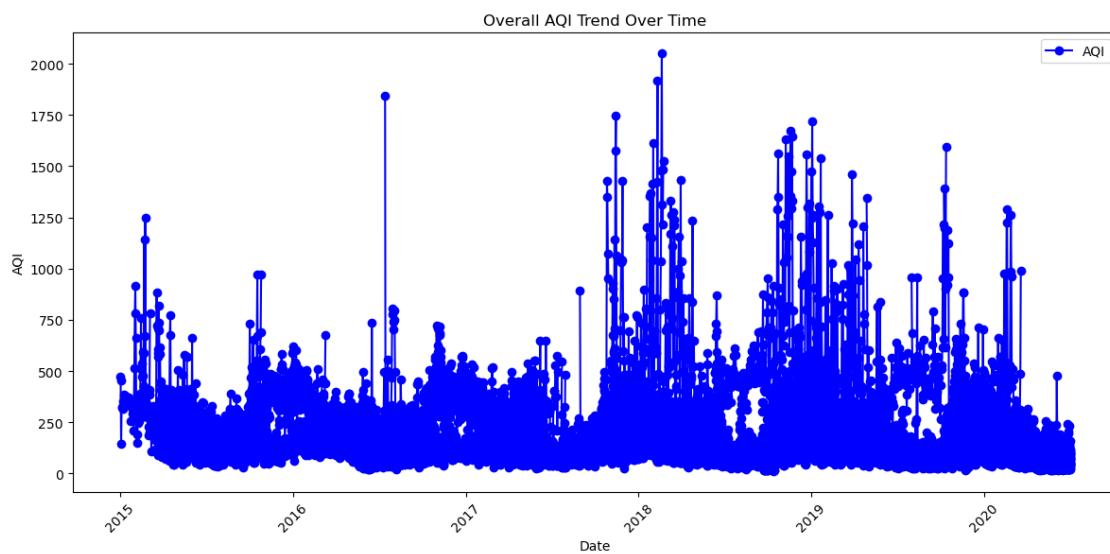
[19]: df.columns

[19]: Index(['City', 'Date', 'PM2.5', 'PM10', 'NO', 'NO2', 'NOx', 'NH3', 'CO', 'SO2', 'O3', 'Benzene', 'Toluene', 'Xylene', 'AQI', 'AQI_Bucket'],
dtype='object')

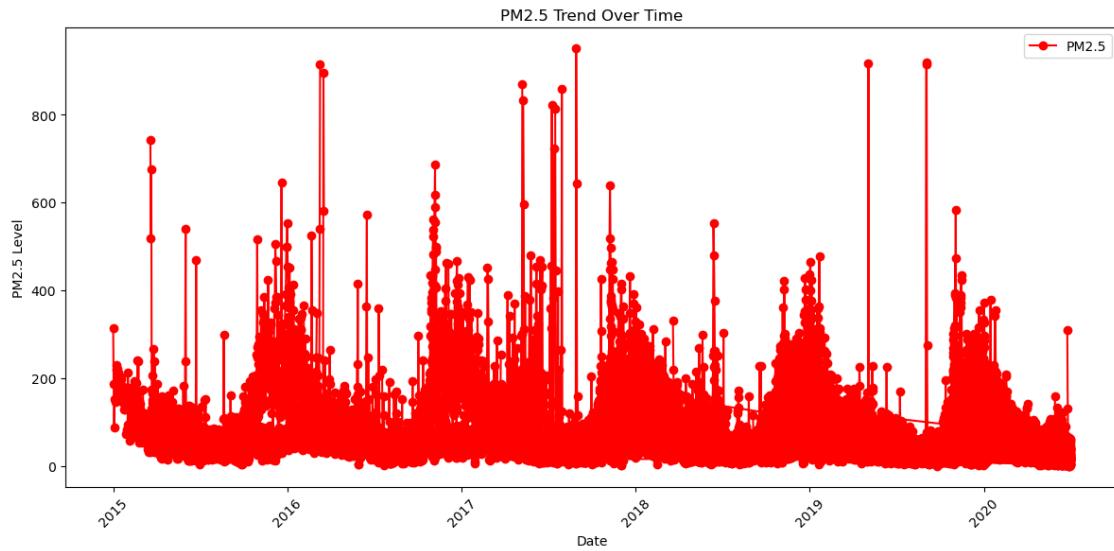
[17]: #4. Create line plots or time series plots to visualize the overall AQI trend over time.

```
#5. Plot individual pollutant levels (e.g., PM2.5, PM10, CO) on separate line plots to visualize their trends over time.
```

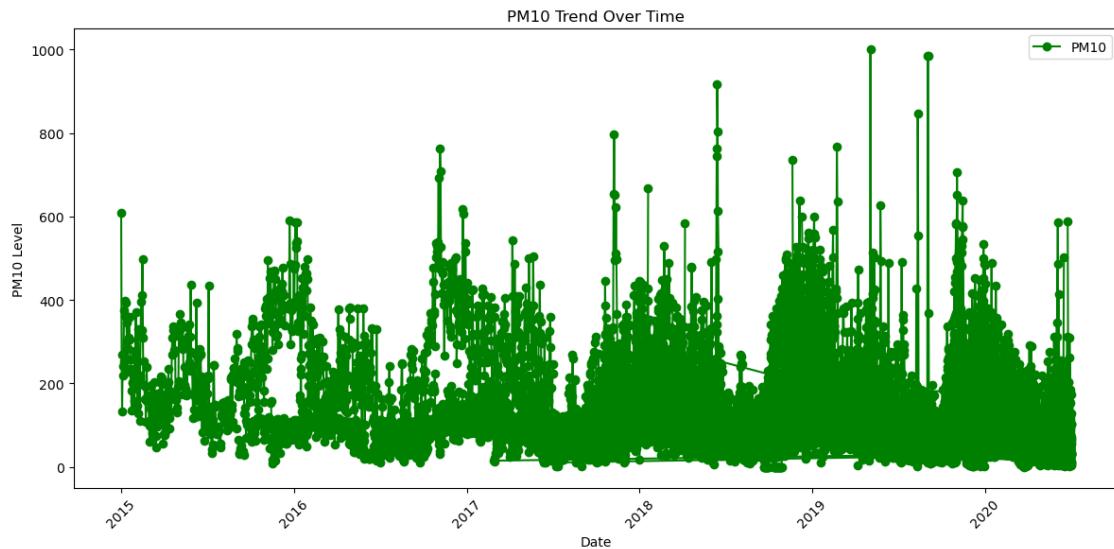
```
[21]: plt.figure(figsize=(12, 6))
plt.plot(df['Date'], df['AQI'], marker='o', linestyle='-', color='b', label='AQI')
plt.xlabel('Date')
plt.ylabel('AQI')
plt.title('Overall AQI Trend Over Time')
plt.xticks(rotation=45)
plt.legend()
plt.tight_layout()
plt.show()
```



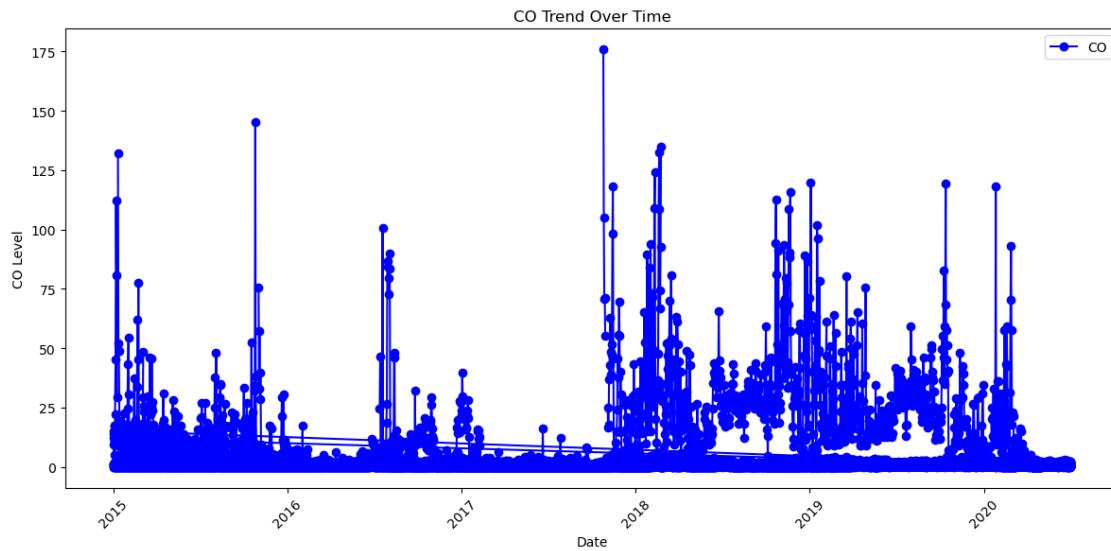
```
[19]: plt.figure(figsize=(12, 6))
plt.plot(df['Date'], df['PM2.5'], marker='o', linestyle='-', color='r', label='PM2.5')
plt.xlabel('Date')
plt.ylabel('PM2.5 Level')
plt.title('PM2.5 Trend Over Time')
plt.xticks(rotation=45)
plt.legend()
plt.tight_layout()
plt.show()
```



```
[23]: plt.figure(figsize=(12, 6))
plt.plot(df['Date'], df['PM10'], marker='o', linestyle='-', color='g', label='PM10')
plt.xlabel('Date')
plt.ylabel('PM10 Level')
plt.title('PM10 Trend Over Time')
plt.xticks(rotation=45)
plt.legend()
plt.tight_layout()
plt.show()
```

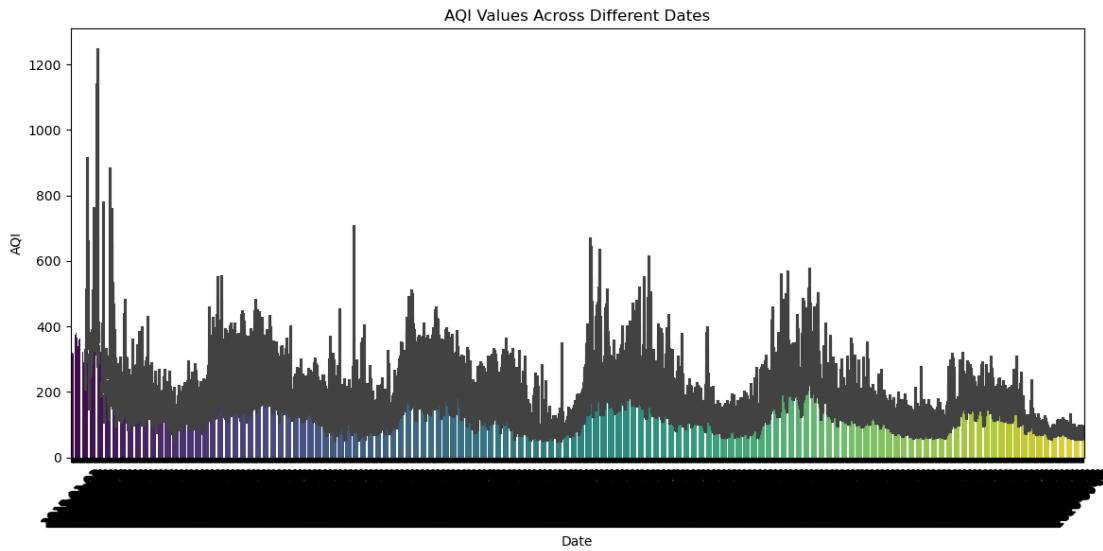


```
[25]: plt.figure(figsize=(12, 6))
plt.plot(df['Date'], df['CO'], marker='o', linestyle='-', color='b', label='CO')
plt.xlabel('Date')
plt.ylabel('CO Level')
plt.title('CO Trend Over Time')
plt.xticks(rotation=45)
plt.legend()
plt.tight_layout()
plt.show()
```



[22]: #6. Use bar plots or stacked bar plots to compare the AQI values across different dates or time periods.

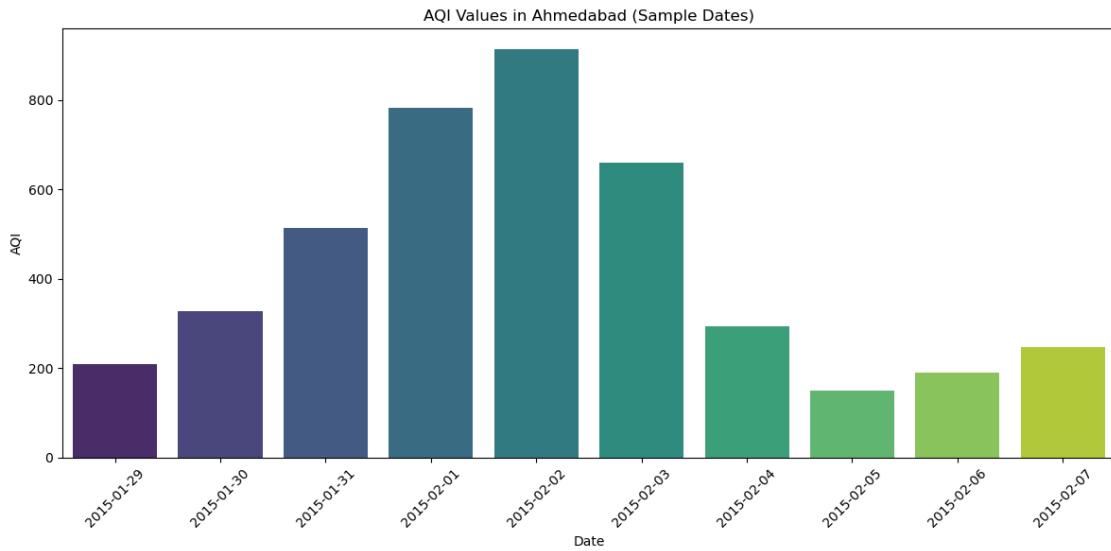
```
[37]: plt.figure(figsize=(12, 6))
# Assign 'Date' to hue to avoid deprecation warning, and suppress legend
sns.barplot(x='Date', y='AQI', data=df, hue='Date', palette='viridis', dodge=False, legend=False)
plt.xlabel('Date')
plt.ylabel('AQI')
plt.title('AQI Values Across Different Dates')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



```
[45]: df['Date'] = pd.to_datetime(df['Date'])
df = df.dropna(subset=['AQI'])

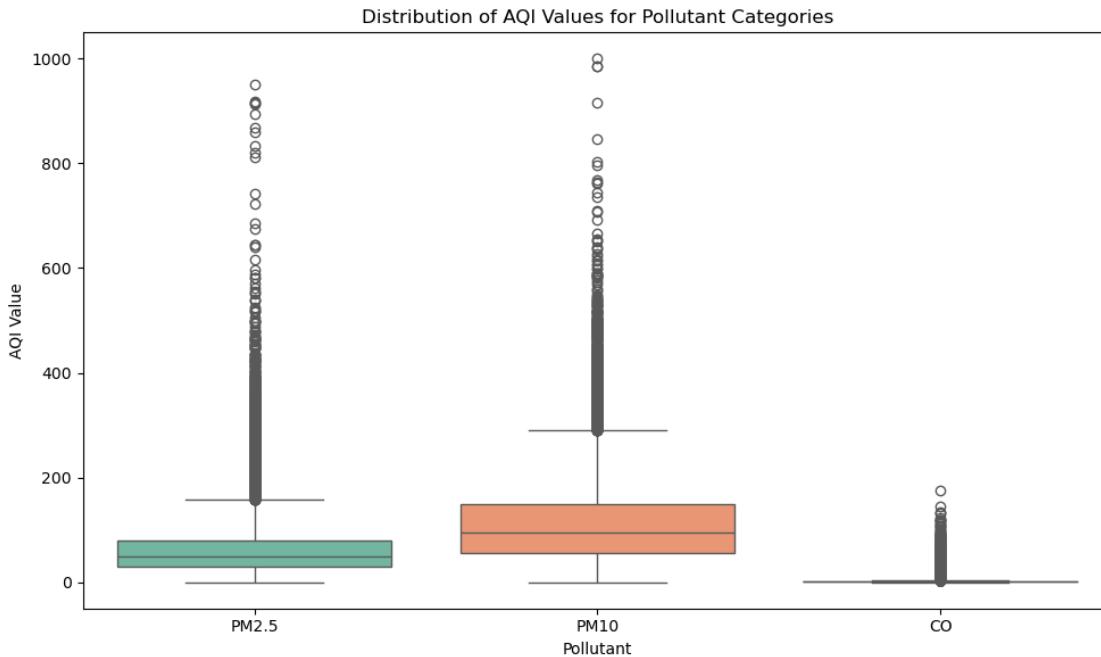
# Sample: Take data for a single city and 10 dates
df_sample = df[df['City'] == 'Ahmedabad'].sort_values('Date').head(10)

plt.figure(figsize=(12, 6))
sns.barplot(x='Date', y='AQI', data=df_sample, hue='Date', palette='viridis',  
            dodge=False, legend=False)
plt.xlabel('Date')
plt.ylabel('AQI')
plt.title('AQI Values in Ahmedabad (Sample Dates)')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



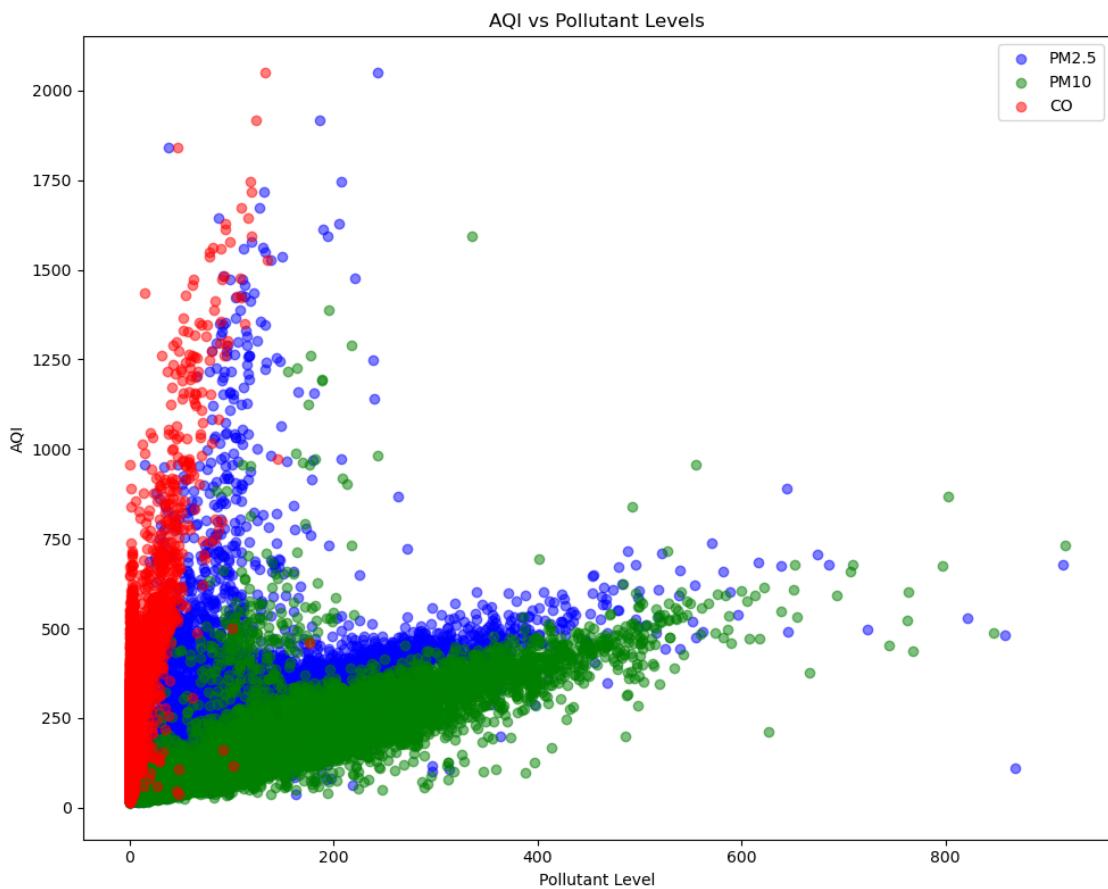
[29]: #7. Create box plots or violin plots to analyze the distribution of AQI values for different pollutant categories.

```
[30]: plt.figure(figsize=(10, 6))
sns.boxplot(data=df[['PM2.5', 'PM10', 'CO']], palette='Set2')
plt.xlabel('Pollutant')
plt.ylabel('AQI Value')
plt.title('Distribution of AQI Values for Pollutant Categories')
plt.tight_layout()
plt.show()
```



[]: #8. Use scatter plots or bubble charts to explore the relationship between AQI values and pollutant levels.

```
[31]: plt.figure(figsize=(10, 8))
plt.scatter(df['PM2.5'], df['AQI'], color='b', alpha=0.5, label='PM2.5')
plt.scatter(df['PM10'], df['AQI'], color='g', alpha=0.5, label='PM10')
plt.scatter(df['CO'], df['AQI'], color='r', alpha=0.5, label='CO')
plt.xlabel('Pollutant Level')
plt.ylabel('AQI')
plt.title('AQI vs Pollutant Levels')
plt.legend()
plt.tight_layout()
plt.show()
```



[]: