

Phase 5

Topic: Air Quality Analysis in Tamil Nadu

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Steps in Data Preprocessing Step 1: Import the necessary libraries

```
# importing libraries
import pandas as pd
import scipy
import numpy as np
from sklearn.preprocessing import MinMaxScaler
import seaborn as sns
import matplotlib.pyplot as plt
```

Step 2: Load the dataset

```
from google.colab import drive
drive.mount('/content/gdrive')
```

Mounted at /content/gdrive

```
# Load the dataset
df = pd.read_csv('/content/gdrive/MyDrive/ADS_PHASE 3.csv')
print(df.head())
```

	Benzene	Eth-Benzene	MP-Xylene	BP	Ø Xylene	PM10	PM2.5	RH	\
0	1.08	0.04	0.00	754.05	2.50	140.23	90.62	42.51	
1	0.83	0.03	0.00	754.28	1.74	124.91	61.11	28.34	
2	1.38	0.27	0.06	754.49	2.39	114.27	70.89	36.48	
3	1.97	0.47	0.12	754.28	3.51	128.15	78.52	43.68	
4	1.80	0.75	0.20	754.00	4.00	122.36	70.48	51.57	

	SR	Temp	WD	WS	CO	NH3	NO	NO2	NOx	Ozone	SO2
0	125.03	17.90	119.19	0.97	0.51	20.53	4.29	22.95	27.24	44.36	4.97
1	148.95	20.04	71.50	1.21	0.53	17.37	2.80	25.59	28.38	53.04	5.59
2	131.87	18.31	147.10	1.00	0.78	18.45	6.85	30.91	37.75	41.94	10.22
3	129.32	18.56	182.79	1.06	0.81	22.52	7.36	29.05	36.41	44.15	30.99
4	145.73	18.81	183.47	0.91	0.96	19.14	13.15	28.60	41.74	37.13	15.78

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 298 entries, 0 to 297
Data columns (total 19 columns):
#   Column      Non-Null Count  Dtype
0   Benzene      298 non-null    float64
1   Eth-Benzene  298 non-null    float64
2   MP-Xylene    298 non-null    float64
3   BP           298 non-null    float64
4   Ø Xylene     298 non-null    float64
5   PM10         298 non-null    float64
6   PM2.5        298 non-null    float64
7   RH           298 non-null    float64
8   SR           298 non-null    float64
9   Temp         298 non-null    float64
10  WD           298 non-null    float64
11  WS           298 non-null    float64
12  CO           298 non-null    float64
13  NH3          298 non-null    float64
14  NO           298 non-null    float64
15  NO2          298 non-null    float64
16  NOx          298 non-null    float64
17  Ozone        298 non-null    float64
18  SO2          298 non-null    float64
float64 dtypes: float64(19)
memory usage: 44.4 KB
```

```
df.head()
```

Benzene Eth- NP-

en e

```
df.tail()
```

	Ethene	Eth - Benzene	MP-Xylene	PM10	Xylene	PM2.5	RH	SR	Temp	WD	WS	CO	NH3	NO	NO2	NOX	
293	1.31	0.22	0.08	757.14	2.57	178.71	81.02	29.44	207.84	28.43	210.70	1.01	1.01	23.42	6.62	32.33	38.13
294	0.64	0.01	0.00	756.64	2.24	158.81	76.39	32.40	214.16	28.21	132.54	1.20	0.62	21.50	8.26	25.42	33.68
295	0.71	0.07	0.03	756.01	2.03	138.19	63.19	30.18	202.04	28.75	172.23	1.18	0.72	19.76	8.66	27.63	36.29
296	0.89	0.10	0.04	755.72	2.51	144.54	58.81	29.82	206.48	29.82	214.90	1.36	0.83	19.35	9.01	27.79	36.80
297	1.10	0.20	0.08	756.84	2.82	152.39	62.50	34.85	167.95	28.89	169.89	0.97	0.91	17.60	8.92	33.32	42.24

STEP 3: check the null values

```
df.isnull().sum()
```

```
Benzene      0
Eth-Benzene   0
MP-Xylene     0
```

	Benzene	Xylene	PM10	PM2.5	RH	SR	Temp	WD	WS	CO	NH3	NO	NO2	NOx			
0	1.08	0.04	0.00	754.05	2.50	140.23	90.62	42.51	125.03	17.90	119.19	0.97	0.51	20.53	4.29	22.95	27.24
1	0.83	0.03	0.00	754.28	1.74	124.91	61.11	28.34	148.95	20.04	71.50	1.21	0.53	17.37	2.80	25.59	28.38
2	1.38	0.27	0.06	754.49	2.39	114.27	70.89	36.48	131.87	18.31	147.10	1.00	0.78	18.45	6.85	30.91	37.75
3	1.97	0.47	0.12	754.28	3.51	128.15	78.52	43.68	129.32	18.56	182.79	1.06	0.81	22.52	7.36	29.05	36.41
4	1.80	0.75	0.20	754.00	4.00	122.36	70.48	51.57	145.73	18.81	183.47	0.91	0.96	19.14	13.15	28.60	41.74
0	Xylene																
PM10		0															
PM2.5		0															
RH		0															
SR		0															
Temp		0															
WD		0															
NS		0															
CO		0															
NH		0															
3		0															
NO		0															
N0																	
2																	
NOx		0															
Ozon		0															
e		0															
SO2																	
dtype: Int64																	

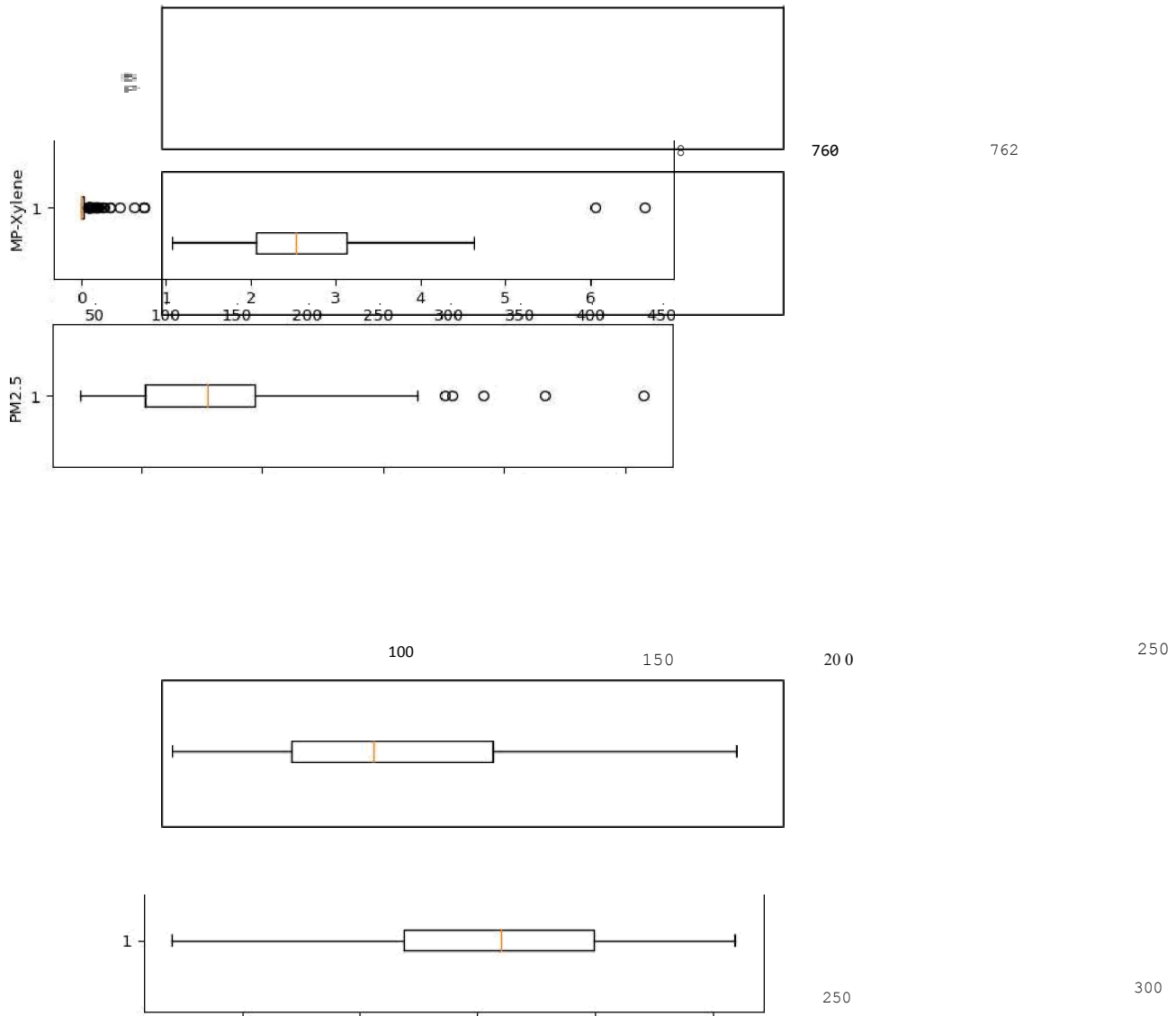
```
df.describe()
```

	SR	Temp
298.000000	298.00000	2
208.083020	30.42802	1
50.891946	5.45345	
69.960000	16.67000	
168.785000	28.14250	1
210.255000	30.86500	2
249.402500	34.15250	2
309.610000	40.07000	2

Step 5: Check the outliers

```
# Box Plots
fig, axs = plt.subplots(9,1,dpi=95,
figsize=(7,17)) i = 0
for col in df.columns:
    axs[i].boxplot(df[col], vert=False)
```

```
axs[i].set_label(col)
1+=1
plt.show()
```



STEP 1: create visualization for SO2

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

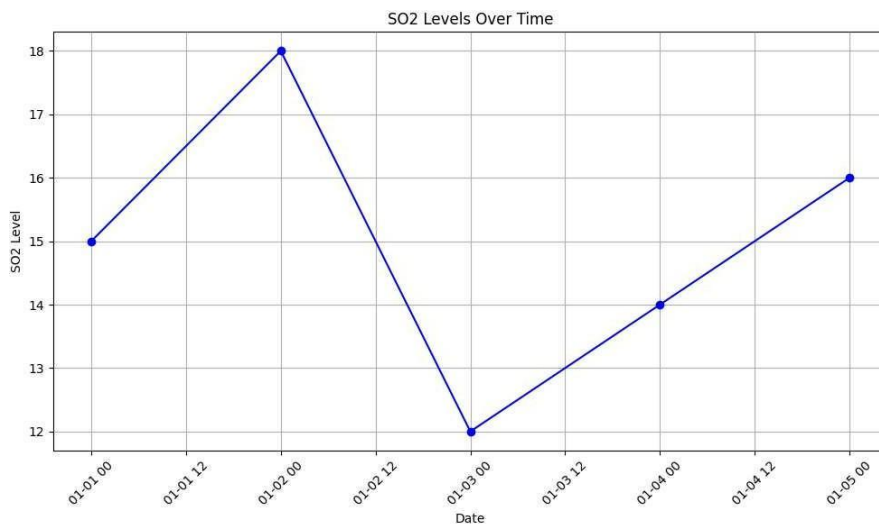
# Sample data (replace this with your actual SO2 dataset)
data = {
    'Date': ['2023-01-01', '2023-01-02', '2023-01-03', '2023-01-04', '2023-01-05'],
    'SO2_Level' : [15, 18, 12, 14, 16] # Replace with your SO2 data

# Create a DataFrame from the data
df = pd.DataFrame(data)
df['Date'] = pd.to_datetime(df['Date']) # Convert Date column to datetime format

# Plot the SO2 levels
plt.figure(figsize=(10, 6))
plt.plot(df['Date'], df['SO2_Level'], marker='o', linestyle='-',
color='b') plt.title('SO2 Levels Over Time')
plt.xlabel('Date')
plt.ylabel('SO2 Level')
plt.grid(True)

# Format the x-axis to display dates nicely
plt.xticks(rotation=45)

# Show the
plot
plt.tight_layout()
plt.show()
```



STEP 2: create visualization for NO2

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

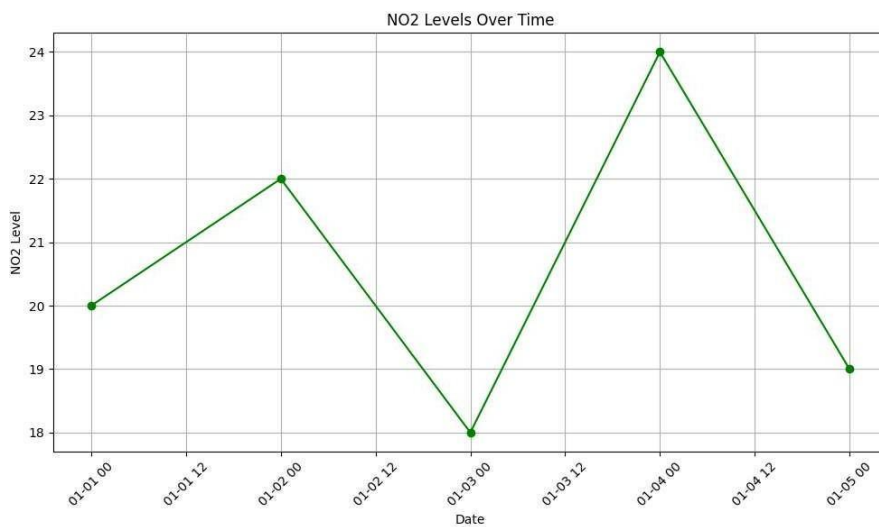
# Sample data (replace this with your actual NO2 dataset)
data = {
    'Date': ['2023-01-01', '2023-01-02', '2023-01-03', '2023-01-04', '2023-01-05'],
    'NO2_Level' : [28, 22, 18, 24, 19] # Replace with your NO2 data
```

```
# Create a DataFrame from the
data df = pd.DataFrame(data)
df['Date'] = pd.to_datetime(df['Date']) # Convert Date column to datetime format

plt.figure(figsize=(10, 6))
plt.plot(df['Date'], df['NO2_Level'], marker='o',
         linestyle='-', color='g') plt.title('NO2 Levels Over Time')
plt.xlabel('Date')
plt.ylabel('NO2 Level') plt.grid(True)

# Format the x-axis to display dates nicely
plt.xticks(rotation=45)

# Show the
plot
plt.tight_layout()
plt.show()
```



STEP 3: create visualization for RSPM/PM10

```
import matplotlib.pyplot as plt

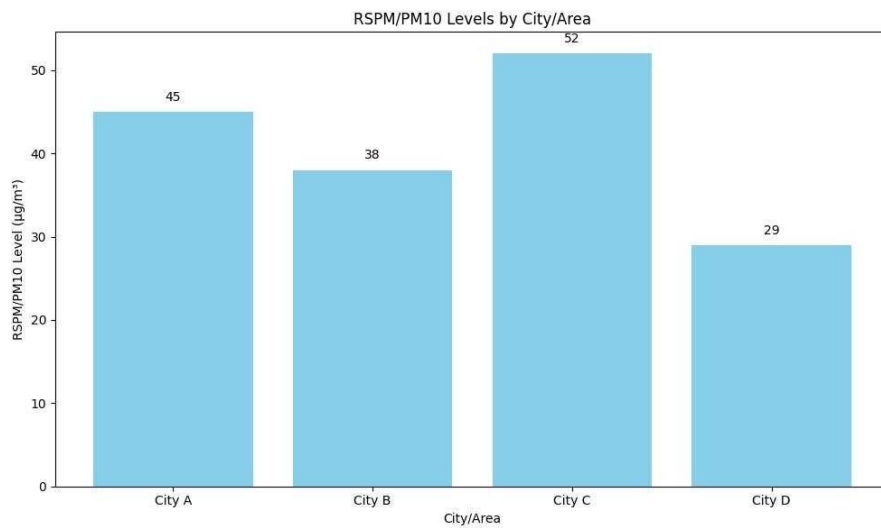
# Sample data (replace this with your actual RSPM/PM10
dataset) categories = ['City A', 'City B', 'City C', 'City
D']
rspm_pm10_levels = [45, 38, 52, 29] # Replace with your RSPM/PM10 data

# Create a bar chart to visualize RSPM/PM10 levels
plt.figure(figsize=(10, 6))
plt.bar(categories, rspm_pm10_levels,
        color='skyblue') plt.title('RSPM/PM10 Levels by
City/Area') plt.xlabel('City/Area')
plt.ylabel('RSPM/PM10 Level (pg/m³)')

for i, level in enumerate(rspm_pm10_levels):
    plt.text(i, level + 1, str(level), ha='center', va='bottom')

# Show the
plot
plt.tight_layout()
plt.show()
```

```
plt.show()
```



```
from google.colab import
drive
drive.mount('/content/dri
ve')

Mounted at /content/drive
```

STEP 4: grouping data

```
import pandas as pd

# Load your data into a Pandas DataFrame (replace 'data.csv' with your data file)
df = pd.read_csv('/content/drive/MyDrive/Excel 4.csv')

# Group the data by the desired column (e.g., 'City' or 'Monitoring
Station') grouped_data = df.groupby('CITIES')
```

STEP 5: calculating average

```
# Calculate the average SO2, NO2, and RSPM/PM10 levels for
each group
averages = grouped_data[['SO2', 'NO2',
'RSPM/PM10']].mean()

# Display the calculated averages
print(averages)
```

CITIES	SO2	NO2	RSPM/PM10
Chennai	9.433333	23.036667	145.640000
Dindigul	22.880000	23.753333	129.780000
Erode	18.900000	27.065800	165.580000
Kauniyakumari	27.790000	34.330000	148.680000
Kodaikanal	12.130000	38.050000	171.910000
Madurai	13.897500	26.727500	156.572500
Salem	11.100000	30.060000	160.976667
Thanjavur	3.650000	30.790000	145.180000
Vellore	16.726667	23.116667	141.240000
madurai	10.220000	30.910000	114.270000
salem	30.990000	29.050000	128.150000

```
# Calculate the average S02, N02, and RSPM/PM10 levels for each
group averages = grouped_data[ ['RSPM/PM10']].mean()
```

```
# Display the calculated
averages print(averages)
```

CITIES	RSPM/PM10
Chennai	145.640000
Dindigul	129.780666
Erode	165.580000
Kanniyakumari	148.680000
Kodaikanal	171.910000
Madurai	156.572500
Salem	160.976667
Thanjavur	145.180000
Vellore	141.240000
madurai	114.270000
salem	128.150000