## **PROJECT-2**

# **AIR QUALITY ASSESSMENT TN**

#### **INTRODUCTION**

Air quality refers to the condition or cleanliness of the air in a specific location, typically measured in terms of the presence and concentration of various air pollutants.

The quality of the air we breathe is of paramount importance, as it directly impacts human health, the environment, and overall well-being. This document aims to provide a comprehensive description of air quality, including the factors influencing it, key pollutants, and the methods used for measuring and assessing air quality.

Air quality depends on a combination of natural and human-made factors. These factors include emissions from various sources, such as industrial processes and energy production, which significantly affect air quality. Additionally, weather patterns, climate change, and chemical reactions in the atmosphere can also lead to the formation of secondary pollutants, such as harmful gases like ozone, which have a substantial impact on air quality.

#### **ABSTRACT**

Air quality data analysis provides a powerful and unique method for extracting valuable insights from the extensive information collected through air quality monitoring stations and sensors in various regions. This analysis allows for the identification of patterns in air pollution data and informs decision-making processes related to pollution control measures.

In this document, we will explore the data analysis process for air quality and the associated methods.

#### **DATA SOURCES**

The data required for this analysis consists of historical air quality information for a specific region or over a set of years. These data sources include data from air quality monitoring stations and satellite remote sensing equipment for visualization.

## **ANALYSIS OF THE PROBLEM - STEPS:**

 To initiate the data analysis, air quality data for the Tamil Nadu region was collected. The dataset underwent preprocessing, which involved handling missing data, removing outliers, and ensuring data consistency.

- 2. Subsequently, the air pollution dataset was analyzed through various visualization techniques, such as bar charts, pie charts, line plots, and bar plots. This aided in understanding the relationships between air quality parameters and factors like weather and climate change.
- Spatial distribution of air quality was achieved by identifying pollution hotspots, predicting air
  quality levels, and pinpointing pollution sources. This information facilitated the optimization of
  pollution control strategies, incorporating methods like clustering and neural networks.
- 4. During the decision-making process using the dataset, trend analysis was a crucial consideration. This involved identifying long-term trends in air quality data to assess whether air quality was improving or deteriorating over time. Additionally, forecasting was employed to predict future patterns based on historical data and current information, aiding in more effective decision-making for areas affected by pollution.
- 5. This comprehensive data analysis process contributes to a deeper understanding of air quality and supports informed decisions for pollution control and public health protection.

```
# Importing the necessary python libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Importing the csv file
df = pd.read_csv('Air_Quality.csv')
df.head()
```

|   | Stn Code | Sampling Date | State      | City/Town/Village/Area | Location of Monitoring Station                   | Agency                                  | Type of Location | 502   | NO2       | RSPM/PM10  | SPM        |
|---|----------|---------------|------------|------------------------|--|---|------------------|-------|-----------|------------|------------|
| 0 | 38       | 5/1/2010      | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area  | 9.60  | 17.166667 | 73.333333  | 149.666667 |
| 1 | 38       | 7/1/2010      | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area  | 11.15 | 20.283333 | 61.333333  | 150.333333 |
| 2 | 38       | 12/1/2010     | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area  | 12.45 | 20.516667 | 75.000000  | 114.666667 |
| 3 | 38       | 1/19/2010     | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area  | 10.75 | 18.183333 | 120.000000 | 197.666667 |
| 4 | 38       | 1/21/2010     | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area  | 9.78  | 17.320000 | 98.500000  | 216.000000 |

## df.shape

#### Output:

(12351, 10)

## df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12351 entries, 0 to 12350
Data columns (total 10 columns):
# Column
                                     Non-Null Count Dtype
--- -----
                                     -----
                                   12351 non-null int64
Ø Stn Code
1 Sampling Date 12351 non-null object 2 City/Town/Village/Area 12351 non-null object
3 Location of Monitoring Station 12351 non-null object
                         12351 non-null object
12351 non-null object
    Agency
5 Type of Location
                                   12149 non-null float64
12153 non-null float64
6 S02
    NO2
8 RSPM/PM10
                                    12303 non-null float64
9 SPM
                                     1902 non-null float64
dtypes: float64(4), int64(1), object(5)
memory usage: 965.0+ KB
```

```
# Removing duplicate values
df.drop_duplicates()
```

## Output:

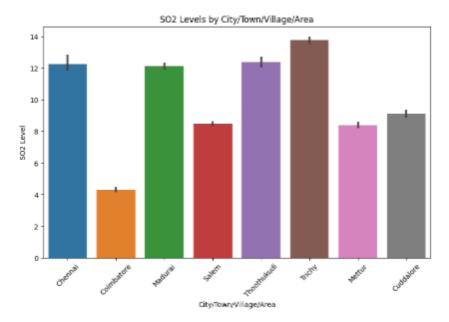
1

```
# Check for null values
df.isna().sum()
#pd.isnull(df).sum()
```

## Output:

```
Stn Code
Sampling Date
                                         0
City/Town/Village/Area
                                        Θ
Location of Monitoring Station
                                        Θ
Agency
                                        Θ
S02
                                      202
NO<sub>2</sub>
                                      198
RSPM/PM10
                                       48
SPM
                                    10449
dtype: int64
```

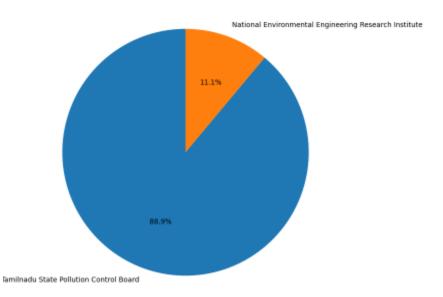
```
# Create the line chart
dat = pd.DataFrame(df)
plt.figure(figsize=(10, 6))
sns.barplot(x='City/Town/Village/Area', y='S02', data=dat)
plt.title('S02 Levels by City/Town/Village/Area')
plt.xticks(rotation=45)
plt.xlabel('City/Town/Village/Area')
plt.ylabel('S02 Level')
plt.show()
```



```
# Pie chart for 'Agency' column
agency_counts = df['Agency'].value_counts()
plt.figure(figsize=(8, 8))
plt.pie(agency_counts, labels=agency_counts.index, autopct='%1.1f%%',
startangle=90)
plt.title('Agency Distribution')
plt.show()
```

## Output:

Agency Distribution



```
# Scatter plot for 'S02' vs 'N02'
plt.figure(figsize=(10, 6))
plt.scatter(df['S02'], df['N02'], c='blue', alpha=0.5)
plt.title('S02 vs N02 Scatter Plot')
plt.xlabel('S02')
plt.ylabel('N02')
plt.grid(True) # Add gridlines (optional)
plt.show()
```

