

# ★Project 2 : Air Quality Analysis in TN Analysis(DAC\_Phase5)

## PHASE 5 : Project Documentation & Submission

### Title: Air Quality Analysis in TN

#### Abstract:

An index for reporting air quality is called the air quality index (AQI). It measures the impact of air pollution on a person's health over a short period of time. The purpose of the AQI is to educate the public on the negative health effects of local air pollution. The amount of air pollution in Indian cities has significantly increased. There are several ways to create a mathematical formula to determine the air quality index. Numerous studies have found a link between air pollution exposure and adverse health impacts in the population. Data mining techniques are one of the most interesting approaches to forecast AQI and analyze it. The aim of this paper is to find the most effective way for AQI prediction to assist in climate control. The most effective method can be improved upon to find the most optimal solution. Hence, the work in this paper involves intensive research and the addition of novel techniques such as SMOTE to make sure that the best possible solution to the air quality problem is obtained. Another important goal is to demonstrate and display the exact metrics involved in our work in such a way that it is educational and insightful and hence provides proper comparisons and assists future researchers.

#### 1. Introduction

Air quality is a measure of how clean or polluted the air is. Monitoring air quality is important because polluted air can be bad for our health—and the health of the environment. Air quality is measured with the Air Quality Index, or AQI. The AQI works sort of like a thermometer that runs from 0 to 500 degrees. However, instead of showing changes in the temperature, the AQI is a way of showing changes in the amount of pollution in the air

- To plan a comprehensive programme for the prevention, control and abatement of air pollution.
- To advise the State Government on any matter concerning the prevention, control or abatement of air pollution.
- To collect and disseminate information relating to air pollution and the prevention, control or abatement thereof.
- To inspect sewage and trade effluent treatment plants for their effectiveness and review plans, specifications for corrective measures.
- To inspect industrial plants or manufacturing process, any control equipment and to give directions to take steps for the prevention, control or abatement of air pollution.

## 2. Design Thinking Process

### Process Air Technology

#### **Cleaning Systems**

Cleaning systems are used to remove contaminants, clean the resulting fluid flows, and collect materials before discharge of exhaust air.

#### **Pneumatic Conveying Systems**

Conveying systems are used to transport captured pollutants from processes to a collection point.

#### **Drying System**

Drying systems are used to remove moisture, gases, and vapors from a product.

## 3. Development Phases

### 3.1 Data Collection

- Data are collected at a few locations taken to represent transport activity, travel movement and traffic flow across the study area or a sample of individual travellers.
- The dataset may include information on various transport

#### **DATASET LINK:**

<https://tn.data.gov.in/resource/location-wise-daily-ambient-air-quality-tamil-nadu-year-2014>

### 3.2 Data Preprocessing

- ❖ Import necessary packages: Here I have imported packages needed for preprocessing .

The dataset upon loading and observation, it's important and easy to process the data with smaller size.

- The data types are changed while loading.
- String data is converted into categorical values, float64 to float32 etc. \* \*
- This process place vital role while handling dataset with larger size.

3]:

|      | Stn Code | Sampling Date | State      | City/Town/Village/Area | Location of Monitoring Station                   | Agency                                  | Type of Location                   | SO2  | NO2  | RSPM/PM10 | PM 2.5 |
|------|----------|---------------|------------|------------------------|--|---|------------------------------------|------|------|-----------|--------|
| 0    | 38       | 01-02-14      | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area                    | 11.0 | 17.0 | 55.0      | NaN    |
| 1    | 38       | 01-07-14      | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area                    | 13.0 | 17.0 | 45.0      | NaN    |
| 2    | 38       | 21-01-14      | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area                    | 12.0 | 18.0 | 50.0      | NaN    |
| 3    | 38       | 23-01-14      | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area                    | 15.0 | 16.0 | 46.0      | NaN    |
| 4    | 38       | 28-01-14      | Tamil Nadu | Chennai                | Kathivakkam, Municipal Kalyana Mandapam, Chennai | Tamilnadu State Pollution Control Board | Industrial Area                    | 13.0 | 14.0 | 42.0      | NaN    |
| ...  | ...      | ...           | ...        | ...                    | ...  | ...                                     | ...                                | ...  | ...  | ...       | ...    |
| 2874 | 773      | 12-03-14      | Tamil Nadu | Trichy                 | Central Bus Stand, Trichy                        | Tamilnadu State Pollution Control Board | Residential, Rural and other Areas | 15.0 | 18.0 | 102.0     | NaN    |
| 2875 | 773      | 12-10-14      | Tamil Nadu | Trichy                 | Central Bus Stand, Trichy                        | Tamilnadu State Pollution Control Board | Residential, Rural and other Areas | 12.0 | 14.0 | 91.0      | NaN    |
| 2876 | 773      | 17-12-14      | Tamil Nadu | Trichy                 | Central Bus Stand, Trichy                        | Tamilnadu State Pollution Control Board | Residential, Rural and other Areas | 19.0 | 22.0 | 100.0     | NaN    |
| 2877 | 773      | 24-12-14      | Tamil Nadu | Trichy                 | Central Bus Stand, Trichy                        | Tamilnadu State Pollution Control Board | Residential, Rural and other Areas | 15.0 | 17.0 | 95.0      | NaN    |
| 2878 | 773      | 31-12-14      | Tamil Nadu | Trichy                 | Central Bus Stand, Trichy                        | Tamilnadu State Pollution Control Board | Residential, Rural and other Areas | 14.0 | 16.0 | 94.0      | NaN    |

2879 rows x 11 columns

### 3.3 Exploratory Data Analysis (EDA)

- ❖ Conduct summary statistics and visualizations to understand the dataset's characteristics.
- ❖ Identify correlations between different air quality parameters.
- ❖ Explore potential factors influencing potability.

### 3.4 Data Visualization

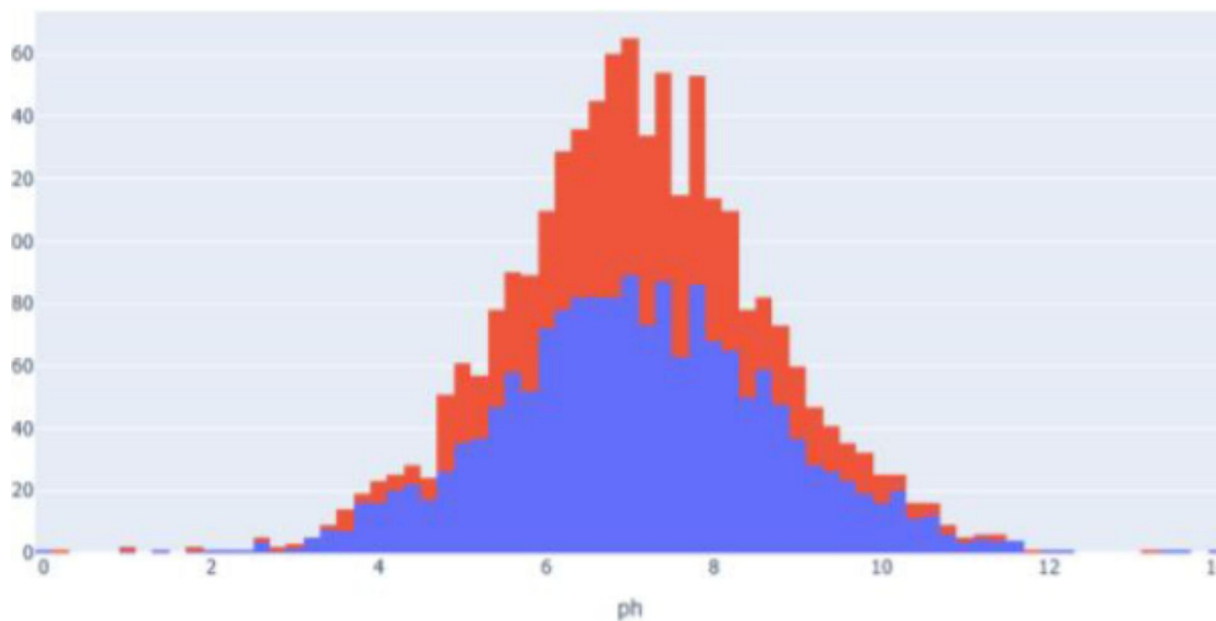
- ❖ Utilize various data visualization techniques, such as scatter plots, histograms, box plots, and heatmaps, to visually represent the data.
- ❖ Visualizations will help in understanding the distribution of air quality parameters and identifying patterns.

```

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error, r2_score
import warnings
warnings.filterwarnings("ignore")
import pandas.util.testing as tm

```

factors affecting air quality



### 3.5 Predictive Modelling

- ❖ Split the dataset into training and testing sets.
- ❖ Choose appropriate machine learning algorithms (e.g., logistic regression, decision trees, random forests, or neural networks) for potability prediction.
- ❖ Train and evaluate the models using appropriate performance metrics.
- ❖ Fine-tune the models to achieve the best predictive accuracy.

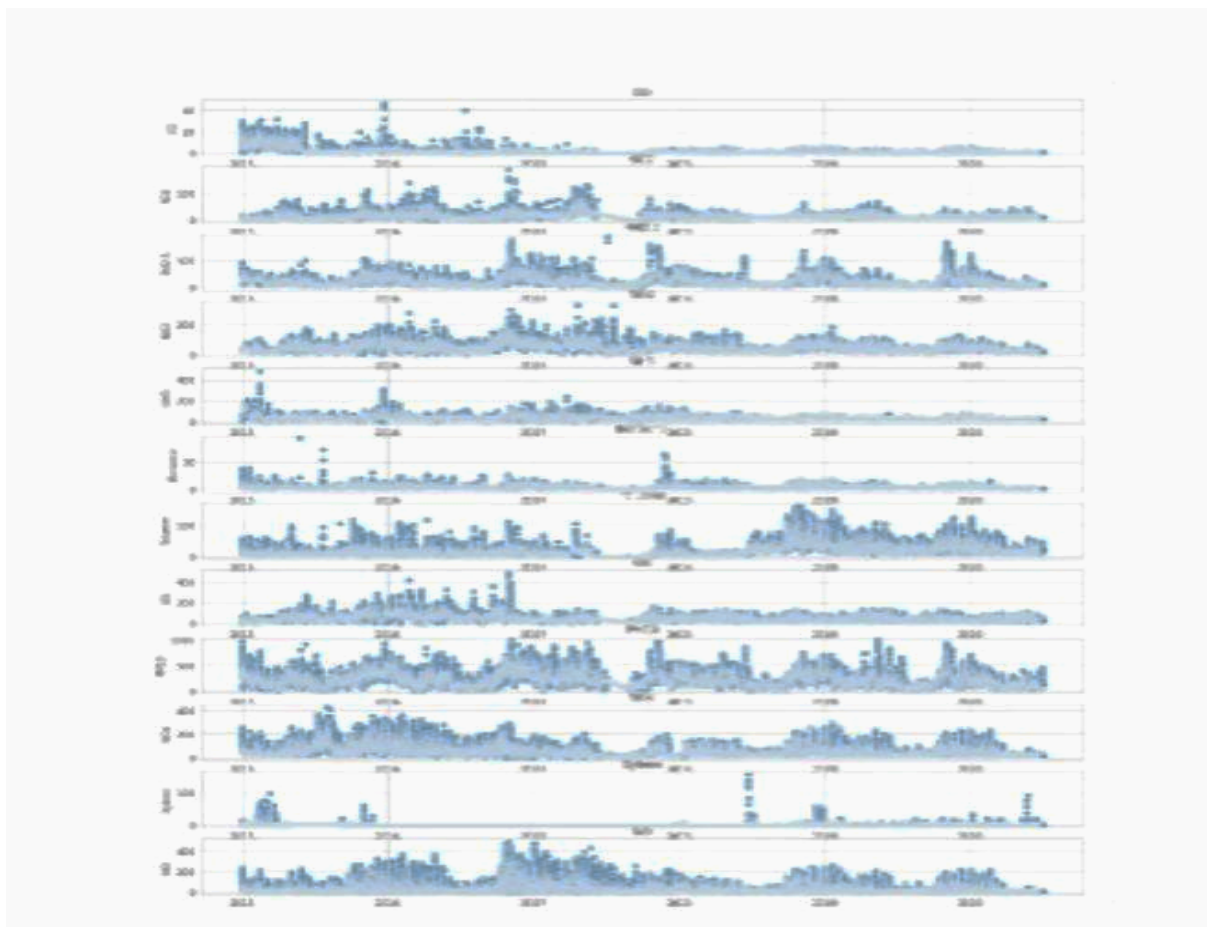
## 4. Analysis Objectives

### 4.1 Air Quality Assessment

- ❖ Determine the overall quality of air by analysing various parameters, including ozone(O<sub>3</sub>), ammonia(NH<sub>3</sub>),etc...

### 4.2 Factors Influencing air quality

- ❖ Identify the most influential factors affecting air , providing actionable insights for air quality improvement.



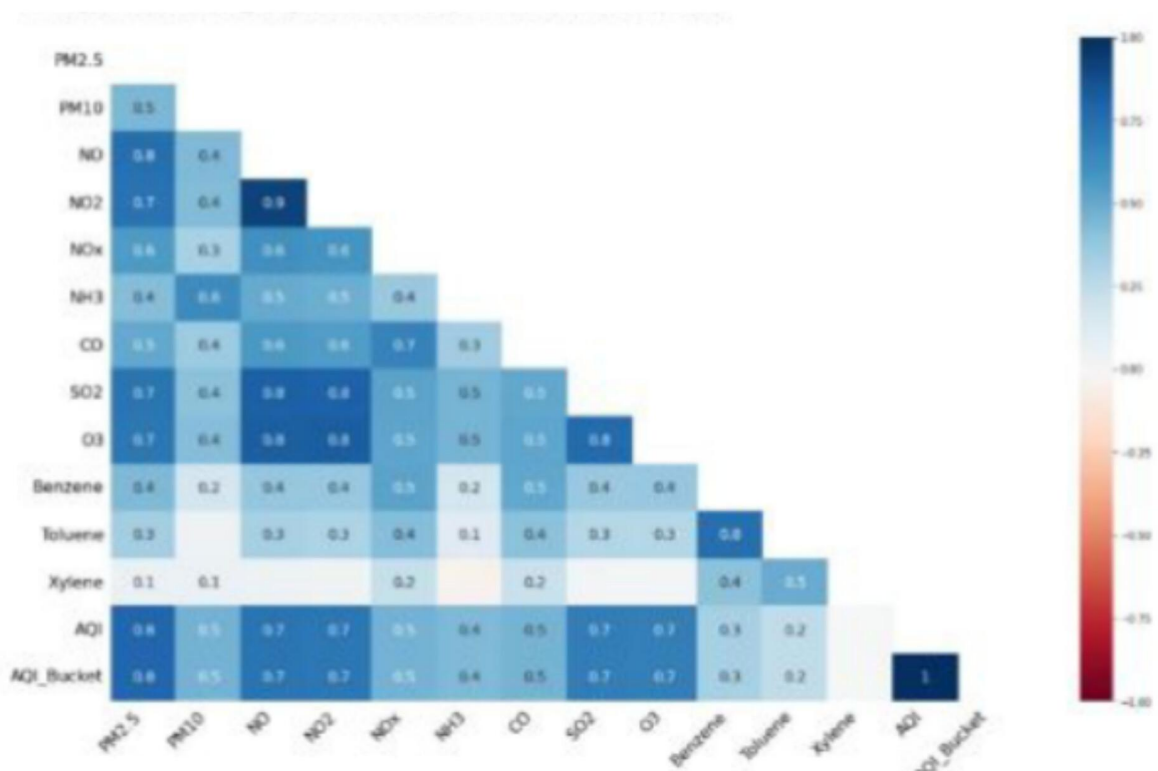
## 5. Data Preprocessing

Data preprocessing is a crucial phase to ensure the reliability and accuracy of the analysis. This phase includes the following steps:

## 5.1 Handling Missing Values

- ❖ Identify and address missing data points by either imputing values or removing incomplete records.

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error, r2_score
import warnings
warnings.filterwarnings("ignore")
import pandas.util.testing as tm
```



## 5.2 Outlier Detection and Treatment

- ❖ Detect and handle outliers that may skew the analysis.

- ❖ Determine whether to remove or transform outlier .

### 5.3 Data Normalization

- ❖ Normalize the data if necessary to bring all variables to the same scale for accurate modelling.

## 6.Exploratory Data Analysis (EDA)

EDA is a critical step to understand the dataset and uncover insights. The following EDA tasks will be performed:

### 6.1 Summary Statistics

- ❖ Calculate descriptive statistics for all air quality parameters, including mean, median, standard deviation, and percentiles.

### 6.2 Data Distribution

- ❖ Create histograms, density plots, and box plots to visualize the distribution of each parameter.

### 6.3 Correlation Analysis

- ❖ Explore the relationships between different parameters by calculating correlation coefficients and creating correlation matrices.

```
dtypes1 = {  
    'City': 'category',  
    'Datetime': 'category',  
    'PM2.5': 'float32',  
    'PM10': 'float32',  
    'NO': 'float32',  
    'NO2': 'float32',  
    'NOx': 'float32',  
    'NH3': 'float32',  
    'CO': 'float32',  
    'SO2': 'float32',  
    'O3': 'float32',  
    'Benzene': 'float32',  
    'Toluene': 'float32',  
    'Xylene': 'float32',  
    'AQI': 'float32',  
    'AQI_Bucket': 'category'  
}
```

## **7. Data Visualization**

Data visualization is essential for conveying information and patterns in the data. The following visualizations will be used:

### **7.1 Scatter Plots**

- ❖ Visualize the relationships between two continuous variables to identify patterns and trends.

### **7.2 Box Plots**

- ❖ Use box plots to compare the distribution of various parameters for potable and non-potable water samples.

### **7.3 Time Series Plots (if applicable)**

- ❖ If the dataset includes time-related data, create time series plots to visualize trends over time.

## **DATA ANALYTICS WITH IBM COGNOS**

### **I. IBM Cognos Introduction**

Introduce IBM Cognos as a tool for data analytics.

### **II. Data Exploration**

Showcase how IBM Cognos aids in exploring and understanding the dataset.

### **III. Visualization**

Demonstrate the creation of visualizations in IBM Cognos.



Tab 1



## DATA VISUALIZATION WITH JUPYTER NOTEBOOK

### i. Jupyter Notebook Introduction

Present Jupyter Notebook as a tool for data analysis and visualization.

### ii. Visualizing Air Quality Parameters

Use Jupyter Notebook to create visualizations of air quality parameters.

### iii. Geographic Mapping

Visualize air quality by location using Jupyter Notebook.

### iv. Time Series Analysis

Analyze temporal changes in water quality using Jupyter Notebook.

## 8. Predictive Modelling

The predictive modelling phase aims to develop a model that can classify air samples as based on the analysed parameters. This phase includes the following steps:

## 8.1 Data Splitting

- ❖ Divide the dataset into a training set and a testing set for model training and evaluation.

## 8.2 Model Selection

- ❖ Choose appropriate machine learning algorithms for binary classification.
- ❖ Evaluate multiple models to select the most suitable one.

## 8.3 Model Training

- ❖ Train the selected model on the training data using the water quality parameters as features as the target variable.

## 8.4 Model Evaluation

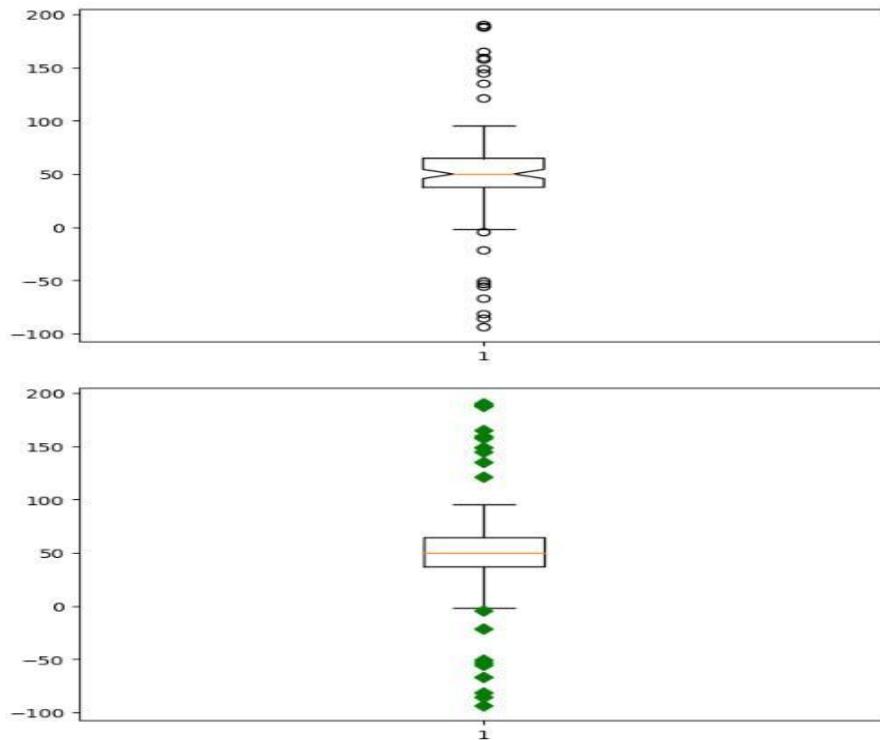
- ❖ Assess the model's performance on the testing set using relevant metrics such as accuracy, precision, recall, F1-score, and the ROC curve.

## 8.5 Model Optimization

- ❖ Fine-tune the model parameters, perform feature selection, and optimize hyperparameters to improve model accuracy.

```
In [21]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
spread = np.random.rand(50) * 100
center = np.ones(25) * 50
flier_high = np.random.rand(10) * 100 + 100
flier_low = np.random.rand(10) * -100
data = np.concatenate((spread, center, flier_high, flier_low), 0)
print (data)
plt.figure(figsize = (7, 5))
plt.boxplot(data, 1)
plt.show()
plt.figure(figsize = (7, 5))
plt.boxplot(data, 0, 'gD')
plt.show()
plt.figure(figsize = (7, 5))
plt.boxplot(data, 0, 'rs', 0, 0.75)
plt.show()
```

|              |              |              |              |              |
|--------------|--------------|--------------|--------------|--------------|
| 76.00992876  | 65.91711212  | 68.18394798  | 50.13641895  | 91.8789882   |
| 49.30853186  | 36.71959485  | 53.64321859  | 54.68554503  | 46.12751574  |
| 1.98824902   | 94.19763425  | 70.15024029  | 39.48739256  | 40.27811781  |
| 63.7901335   | 1.51570733   | 35.8664904   | 66.20022493  | 95.40368517  |
| 33.23315008  | 26.50091155  | 58.3549899   | 55.29389813  | 83.63424892  |
| 85.93968355  | 1.41949078   | 50.28720052  | 17.89504118  | 51.17823429  |
| 38.13887785  | 31.33908749  | 63.80073629  | 4.93058491   | 15.25825846  |
| 54.78652661  | 49.32384777  | 69.18585901  | 56.66409472  | 42.26304259  |
| 57.08504526  | 75.30971113  | 25.95235052  | 51.40094291  | 41.79676876  |
| 26.8334635   | 4.42275024   | 50.51908669  | 91.16687873  | 80.61094611  |
| 50.          | 50.          | 50.          | 50.          | 50.          |
| 50.          | 50.          | 50.          | 50.          | 50.          |
| 50.          | 50.          | 50.          | 50.          | 50.          |
| 50.          | 50.          | 50.          | 50.          | 50.          |
| 50.          | 50.          | 50.          | 50.          | 50.          |
| 148.63433008 | 190.45391435 | 188.31602969 | 187.62505984 | 121.7429514  |
| 135.28761195 | 145.15853198 | 159.7340601  | 165.47596769 | 158.17283819 |
| -93.26456687 | -80.91345169 | -50.20532417 | -66.38054113 | -52.98736074 |
| -3.91547163  | -21.35458017 | -85.58189656 | -55.23019443 | -1.70216355] |



## 9. Insights from Analysis

The insights derived from the analysis will provide valuable information for assessing air quality . Here are some of the key insights that can be obtained:

### 9.1 Identification of Critical Parameters

- ❖ Determine which air quality parameters have the most significant impact on pollutants.

### 9.2 Pattern Recognition

- ❖ Discover patterns and trends in the data that may indicate specific factors affecting air quality.

### 9.3 Data-Driven Recommendations

- ❖ Provide data-driven recommendations for improving air quality based on the identified factors.

## 9.4 Decision Support

- ❖ Offer decision support tools for stakeholders, such as air treatment plants or regulatory authorities, to make informed decisions about air quality management.

In [22]:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.ensemble import RandomForestRegressor
data = pd.read_csv(r"C:\Users\divya\OneDrive\Documents\Untitled Folder\water_potability.csv")
print(data)
```

|      | ph       | Hardness   | Solids      | Chloramines | Sulfate    | \   |
|------|----------|------------|-------------|-------------|------------|-----|
| 0    | NaN      | 204.890456 | 20791.31898 | 7.300212    | 368.516441 |     |
| 1    | 3.716080 | 129.422921 | 18630.05786 | 6.635246    | NaN        |     |
| 2    | 8.099124 | 224.236259 | 19909.54173 | 9.275884    | NaN        |     |
| 3    | 8.316766 | 214.373394 | 22018.41744 | 8.059332    | 356.886136 |     |
| 4    | 9.092223 | 181.101509 | 17978.98634 | 6.546600    | 310.135738 |     |
| ...  | ...      | ...        | ...         | ...         | ...        | ... |
| 3271 | 4.668102 | 193.681736 | 47580.99160 | 7.166639    | 359.948574 |     |
| 3272 | 7.808856 | 193.553212 | 17329.80216 | 8.061362    | NaN        |     |
| 3273 | 9.419510 | 175.762646 | 33155.57822 | 7.350233    | NaN        |     |
| 3274 | 5.126763 | 230.603758 | 11983.86938 | 6.303357    | NaN        |     |
| 3275 | 7.874671 | 195.102299 | 17404.17706 | 7.509306    | NaN        |     |

|      | Conductivity | Organic_carbon | Trihalomethanes | Turbidity | Potability |
|------|--------------|----------------|-----------------|-----------|------------|
| 0    | 564.308654   | 10.379783      | 86.990970       | 2.963135  | 0          |
| 1    | 592.885359   | 15.180013      | 56.329076       | 4.500656  | 0          |
| 2    | 418.606213   | 16.868637      | 66.420093       | 3.055934  | 0          |
| 3    | 363.266516   | 18.436525      | 100.341674      | 4.628771  | 0          |
| 4    | 398.410813   | 11.558279      | 31.997993       | 4.075075  | 0          |
| ...  | ...          | ...            | ...             | ...       | ...        |
| 3271 | 526.424171   | 13.894419      | 66.687695       | 4.435821  | 1          |
| 3272 | 392.449580   | 19.903225      | NaN             | 2.798243  | 1          |
| 3273 | 432.044783   | 11.039070      | 69.845400       | 3.298875  | 1          |
| 3274 | 402.883113   | 11.168946      | 77.488213       | 4.708658  | 1          |
| 3275 | 327.459761   | 16.140368      | 78.698446       | 2.309149  | 1          |

[3276 rows x 10 columns]

## Conclusion

This project's objective is to analyze air quality data and predict pollutants using a dataset, following the design thinking process and various development phases. By conducting data preprocessing, exploratory data analysis, data visualization, and predictive modelling, we aim to provide valuable insights for assessing air quality. The insights obtained from this analysis can have a significant impact on air management, public health, and environmental conservation.

In summary, this comprehensive analysis will not only assess the quality of air but also empower decision-makers with the tools and knowledge needed to safeguard and enhance the safety of air sources. Air quality analysis and pollutants prediction play a vital role in ensuring access to clean and pure air, a fundamental human right.

**LINK FOR JUPYTER NOTEBOOK (ipynb) :**

<https://github.com/HarshiniSivakumar30/AirQassesment.git>

**LINK FOR JUPYTER NOTEBOOK (pdf) :**

<https://github.com/HarshiniSivakumar30/AirQassesment.git>

**LINK FOR IBM COGNOS VISUALIZATION (pdf) :**

<https://github.com/HarshiniSivakumar30/AirQassesment.git>