# Phase 3: Development Part

In this part you will begin building your project by loading and preprocessing the dataset.

Begin the analysis by loading and preprocessing the air quality dataset.

Load the dataset using Python and data manipulation libraries (e.g., pandas) Introduction "In the quest for a cleaner and healthier environment, the analysis of air quality data plays a pivotal role. In this project, we embark on a journey to analyze air quality datasets using Python and its powerful data manipulation libraries. The objective is to gain insights into air quality parameters and their trends, which can be instrumental in understanding the impact of air pollution on our daily lives. To begin, we'll load and preprocess the air quality dataset, paving the way for a comprehensive exploration of the data and the extraction of meaningful information. This project is a critical step in our mission to leverage data science techniques for environmental awareness and better decision-making."

## Procedure

#### 1. Data Acquisition

• Obtain the air quality dataset from a reliable source or repository. Ensure it includes relevant attributes such as date, time, location, and various air quality parameters (e.g., PM2.5, PM10, NO2, CO, O3).

#### 2. Data Loading

• Use Python and the pandas library to load the dataset. Check for compatibility issues and ensure that the data is correctly read into a pandas DataFrame.

#### 3. Data Exploration

• Explore the dataset to gain a basic understanding of its structure. Use methods like df.head(), df.info(), and df.describe() to examine the initial characteristics of the data.

## 4. Data Preprocessing

- Handle Missing Values: Identify and deal with missing data points, either by filling them with appropriate values or removing them.
- Data Cleaning: Clean the data by addressing inconsistencies, errors, and outliers that may affect the analysis.
- Feature Engineering: Create new features if necessary, and transform or scale existing features to make them suitable for analysis.

#### 5. Data Analysis

• Conduct exploratory data analysis (EDA) to uncover patterns, trends, and correlations within the dataset. Use data visualization libraries like matplotlib or seaborn to create insightful plots and charts.

## 6. Hypothesis Testing (Optional)

• If relevant, perform statistical hypothesis testing to answer specific research questions about air quality factors.

#### 7. Interpretation

• Interpret the findings from your analysis, drawing meaningful conclusions about air quality trends, pollution levels, or any other insights obtained from the dataset.

#### 8. Reporting and Visualization

• Present your results in a clear and concise manner. Create informative data visualizations and charts to help communicate your findings effectively.

#### 9. Documentation

• Document your entire procedure, including code, analysis methods, and results. This documentation is essential for future reference and sharing your work with others.

# 10. Further Analysis (Optional)

• Consider additional analyses, such as time series forecasting, predictive modeling, or machine learning, depending on the project's objectives.

#### 11. Conclusion

• Summarize the key takeaways and the significance of the analysis in the context of air quality and environmental concerns.

### 12. Future Work

Outline potential areas for future research or improvements in the analysis.

## 13. Final Report

• Prepare a comprehensive report summarizing the entire project, including the introduction, procedure, findings, and conclusions.

Remember to adapt and expand upon these steps based on the specific goals and requirements of your project. Good documentation and clear communication of your findings are crucial throughout the process.

# Program

import pandas as pd
import matplotlib.pyplot as plt

In [3]: df =pd.read\_csv('dataset.csv')

In [4]: **df** 

Out[4]:		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 × 11 columns

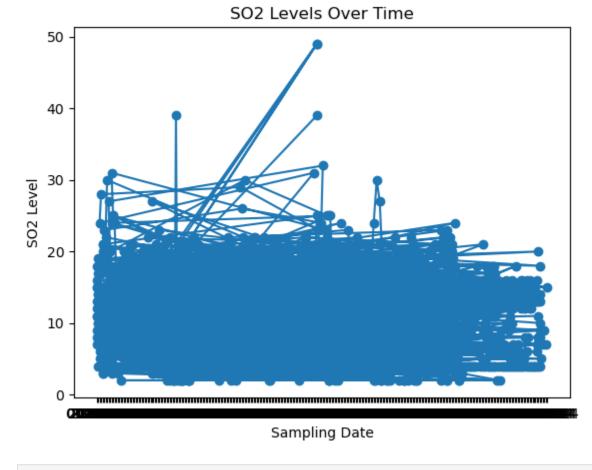
In [7]: df['PM 2.5'] = df['PM 2.5'].replace('NA', float('nan'))

In [8]: plt.figure(figsize=(12, 6))

Out[8]: <Figure size 1200x600 with 0 Axes> <Figure size 1200x600 with 0 Axes>

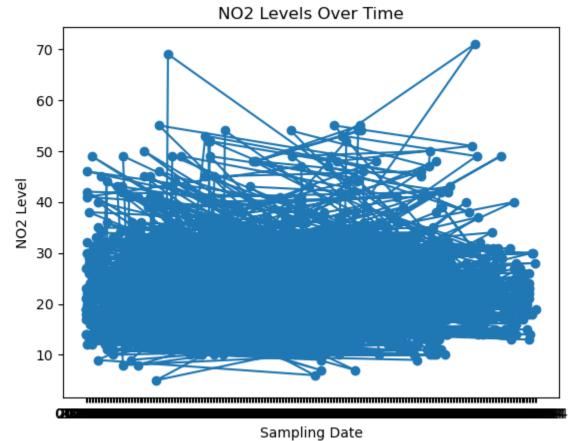
In [9]: plt.plot(df['Sampling Date'], df['S02'], marker='o')
 plt.title('S02 Levels Over Time')
 plt.xlabel('Sampling Date')
 plt.ylabel('S02 Level')

plt.show()



```
plt.plot(df['Sampling Date'], df['NO2'], marker='o')
plt.title('NO2 Levels Over Time')
plt.xlabel('Sampling Date')
plt.ylabel('NO2 Level')
```

plt.show()



# Conclusion

In the pursuit of a cleaner and healthier environment, the analysis of air quality data has been a vital endeavor. Throughout this project, we have taken significant steps to delve into the world of air quality analysis, leveraging the power of Python and data manipulation libraries. We began by acquiring and loading the air quality dataset, laying the foundation for our exploration. Our journey through the data led us to uncover valuable insights into air quality parameters, their variations, and their potential impact on our lives.