

# GRT INSTITUTE OF ENGINEERING AND TECHNOLOGY, TIRUTTANI - 631209



Approved by AICTE, New Delhi Affiliated to Anna University, Chennai

## DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING

### Phase4

# **PROJECT TITLE**

Air quality analysis and prediction in tamilnadu

**COLLEGE CODE:1103** 

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3rd yr, 5th sem

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#### Program:

```
import pandas as pd
import matplotlib.pyplot as plt
dtype = {
    'sampling date': str,
data= pd.read_csv('C:/project/data.csv', sep=',', encoding='ISO-8859-1',
dtype=dtype)
data=data.sample(500)
print(data)
average_so2 = data['so2'].mean()
average_no2 = data['no2'].mean()
average_rspm = data['rspm'].mean()
# You may want to add more analysis here
# Create visualizations
plt.figure(figsize=(10, 6))
plt.bar(['so2', 'no2', 'rspm'], [average_so2, average_no2, average_rspm])
plt.xlabel('Pollutant')
plt.ylabel('Average Level')
plt.title('Average Air Quality Levels')
plt.show()
pollutants = ['SO2', 'NO2', 'RSPM']
averages = [average_so2,average_no2 , average_rspm]
plt.bar(pollutants, averages)
plt.xlabel('Pollutant')
plt.ylabel('Average Level')
plt.title('Average Air Quality Levels')
plt.show()
# Calculate average levels for SO2, NO2, and RSPM/PM10
average_so2 = data['so2'].mean()
average_no2 = data['no2'].mean()
average_rspm_pm10 = data['rspm'].mean()
print(f"Average SO2 level: {average_so2}")
print(f"Average NO2 level: {average_no2}")
print(f"Average RSPM/PM10 level: {average_rspm}")
```

```
plt.figure(figsize=(12, 6))
plt.xlabel('sampling_date')
plt.ylabel('S02 Levels')
plt.title('S02 Levels in Tamil Nadu')
plt.legend()
plt.grid()
plt.show()
plt.figure(figsize=(12, 6))
plt.plot(data['sampling_date'], data['no2'], label='no2 Levels',
plt.xlabel('sampling_date')
plt.ylabel('no2 Levels')
plt.title('no2 Levels in Tamil Nadu')
plt.legend()
plt.grid()
plt.show()
plt.figure(figsize=(12, 6))
plt.plot(data['sampling_date'], data['rspm'], label='rspm Levels',
plt.xlabel('sampling_date')
plt.ylabel('rspm Levels')
plt.title('rspm Levels in Tamil Nadu')
plt.legend()
plt.grid()
plt.show()
import seaborn as sns
plt.figure(figsize=(12, 6))
plt.subplot(131)
sns.histplot(data['so2'], bins=20, kde=True)
plt.title('SO2 Distribution')
plt.subplot(132)
sns.histplot(data['no2'], bins=20, kde=True)
plt.title('NO2 Distribution')
plt.subplot(133)
sns.histplot(data['rspm'], bins=20, kde=True)
plt.title('RSPM/PM10 Distribution')
plt.tight_layout()
plt.show()
average_so2 = data.groupby('location_monitoring_station')['so2'].mean()
average_no2 = data.groupby('location_monitoring_station')['no2'].mean()
average_rspm_pm10 =
data.groupby('location_monitoring_station')['rspm'].mean()
# Example: Bar chart to visualize average SO2 levels by city
plt.figure(figsize=(10, 6))
plt.bar(average_so2.index, average_so2.values)
plt.xlabel('City')
plt.ylabel('Average SO2 Level')
```

```
plt.title('Average SO2 Levels by City')
plt.xticks(rotation=45)
plt.show()
avg_so2 = data.groupby('location_monitoring_station')['so2'].mean()
avg_no2 = data.groupby('location_monitoring_station')['no2'].mean()
avg_rspm_pm10 = data.groupby('location_monitoring_station')['rspm'].mean()
high_so2_areas = avg_so2[avg_so2 > 2] # Replace 'threshold' with your
desired threshold value
high_no2_areas = avg_no2[avg_no2 > 200]
high_rspm_pm10_areas = avg_rspm_pm10[avg_rspm_pm10 >60 ]
print("Average SO2 levels by location:")
print(avg_so2)
print("\nAverage NO2 levels by location:")
print(avg_no2)
print("\nAverage RSPM/PM10 levels by location:")
print(avg_rspm_pm10)
print("\nAreas with high SO2 levels:")
print(high_so2_areas)
print("\nAreas with high NO2 levels:")
print(high_no2_areas)
print("\nAreas with high RSPM/PM10 levels:")
print(high_rspm_pm10_areas)
# Visualize pollution trends over time
import matplotlib.pyplot as plt
plt.figure(figsize=(12, 6))
plt.plot(high_so2_areas.index, high_so2_areas.values, label='SO2')
plt.plot(high_no2_areas.index, high_no2_areas.values, label='NO2')
plt.plot(high_rspm_pm10_areas.index,high_rspm_pm10_areas.values,
plt.xlabel('Time')
plt.ylabel('Average Levels')
plt.title('Air Quality Trends Over Time')
plt.legend()
plt.show()
```

#### Output:

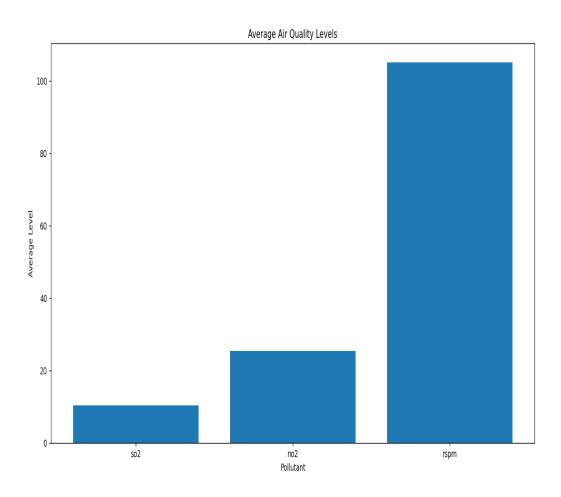
```
stn_code sampling_date ... Unnamed: 11
                                      date
10031
             17/4/2009 ...
                             NaN 4/17/2009
        NaN
227291
        165 18/03/2011 ...
                             NaN 3/18/2011
367294
              1/1/2015 ...
                            NaN 1/1/2015
        581
21563
                            NaN 9/4/2014
        742
             4/9/2014 ...
72170
       144 20/10/2011 ... NaN 10/20/2011
```

... ... ... ... ... ...

351040 NaN 22/5/2009 ... NaN 5/22/2009 395146 6 17/02/2012 ... NaN 2/17/2012 275222 NaN 7/12/2007 ... NaN 12/7/2007 398317 1 14-03-13 ... NaN 3/14/2013 107544 NaN 27-10-05 ... NaN 10/27/2005

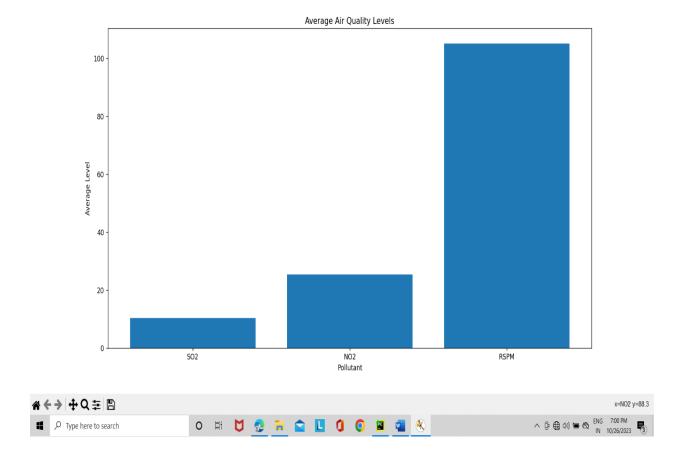
## [500 rows x 13 columns]

€ Figure 1
— □ X





® Figure 1 − □ X

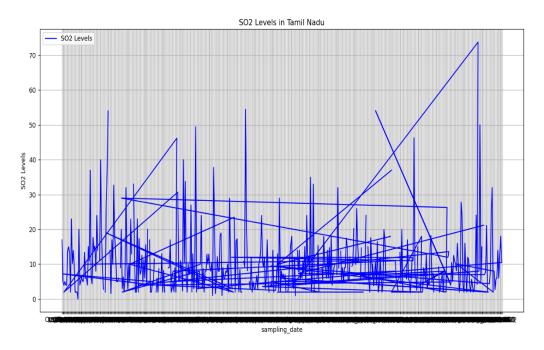


Average SO2 level: 10.347529112074236

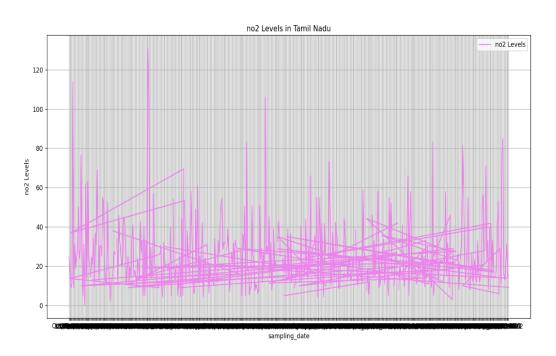
Average NO2 level: 25.39095238093789

Average RSPM/PM10 level: 105.0656695780131

® Figure 1 − σ ×

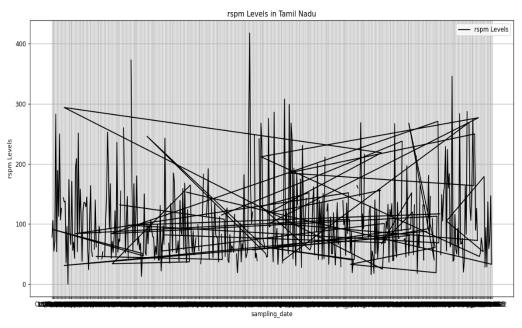


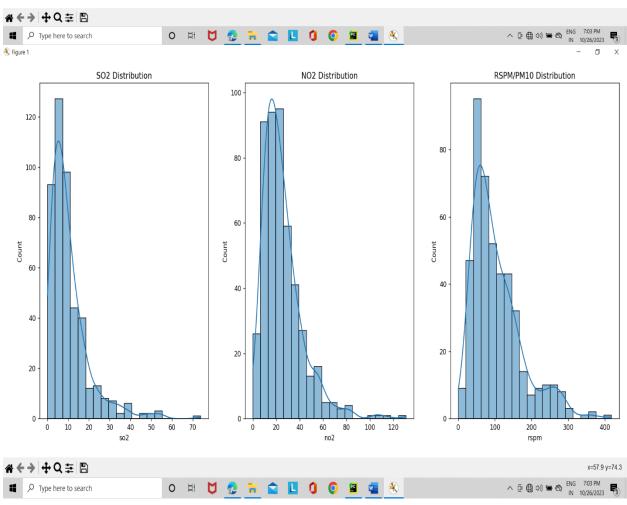




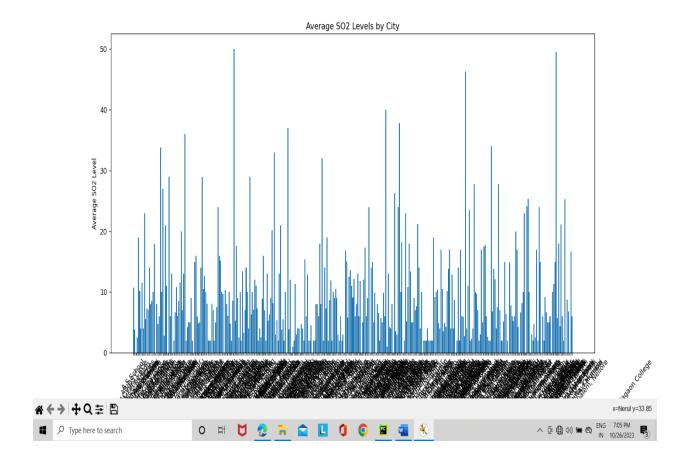


% Figure 1 — □ ×





% Figure 1 − □ □



## Average SO2 levels by location:

location\_monitoring\_station

A S School 10.75

ABIDS Circle 3.80

AC Office Bldg. NaN

AC Office Building, Parwanoo 2.50

AMCO Batteries 19.00

...

Visak Hostel, Sector-4, Bhilai 8.75

WBIIDC 6.80

WBIIDC, Haldia NaN

WIT Campus 16.70

Water Resources Division Office Campus, Christian Patty, near Nagaon College 6.00

Name: so2, Length: 359, dtype: float64

Average NO2 levels by location:

location\_monitoring\_station

A S School 30.35

ABIDS Circle 54.10

AC Office Bldg. 7.50

AC Office Building, Parwanoo 13.34

AMCO Batteries 37.90

...

Visak Hostel, Sector-4, Bhilai 21.25

WBIIDC 44.20

WBIIDC, Haldia 29.60

WIT Campus 36.30

Water Resources Division Office Campus, Christian Patty, near Nagaon College 21.00

Name: no2, Length: 359, dtype: float64

Average RSPM/PM10 levels by location:

location\_monitoring\_station

A S School 240.50

ABIDS Circle 49.00

AC Office Bldg. 50.00

AC Office Building, Parwanoo 71.00

AMCO Batteries 46.00

...

Visak Hostel, Sector-4, Bhilai 88.64

WBIIDC 33.00

WBIIDC, Haldia 73.40

WIT Campus 121.00

Water Resources Division Office Campus, Christian Patty, near Nagaon College 194.00

Name: rspm, Length: 359, dtype: float64

Areas with high SO2 levels:

location\_monitoring\_station

A S School 10.75

ABIDS Circle 3.80

AC Office Building, Parwanoo 2.50

AMCO Batteries 19.00

AS School, Khanna 10.20

...

Victoria Hospital 25.30

Visak Hostel, Sector-4, Bhilai 8.75

WBIIDC 6.80

WIT Campus 16.70

Water Resources Division Office Campus, Christian Patty, near Nagaon College 6.00

Name: so2, Length: 285, dtype: float64

Areas with high NO2 levels:

Series([], Name: no2, dtype: float64)

Areas with high RSPM/PM10 levels:

location\_monitoring\_station

A S School 240.50

AC Office Building, Parwanoo 71.00

AS School, Khanna 229.00

ASRAM Diagnostic Center, Eluru 88.00

AZL Behrampura, Ahmadabad 66.50

...

Vijay Nagar 157.00

Visak Hostel, Sector-4, Bhilai 88.64

WBIIDC, Haldia 73.40

WIT Campus 121.00

Water Resources Division Office Campus, Christian Patty, near Nagaon College 194.00

Name: rspm, Length: 247, dtype: float64



