

Phase 1 :

Problem Definition:

The project aims to analyze and visualize air quality data from monitoring stations in Tamil Nadu. The objective is to gain insights into air pollution trends, identify areas with high pollution levels, and develop a predictive model to estimate RSPM/PM10 levels based on SO2 and NO2 levels. This project involves defining objectives, designing the analysis approach, selecting visualization techniques, and creating a predictive model using Python and relevant libraries.

Design Thinking:

Project Objectives: Define objectives such as analyzing air quality trends, identifying pollution hotspots, and building a predictive model for RSPM/PM10 levels.

Analysis Approach: Plan the steps to load, preprocess, analyze, and visualize the air quality data.

Visualization Selection: Determine visualization techniques (e.g., line charts, heatmaps) to effectively represent air quality trends and pollution levels.

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Phase 2 :

Project Name : Air Quality Analysis and prediction in Tamil Nadu

Project Description : Develop a machine learning model predict a Air quality such as So2 ,No2 and RSPM .This prediction make a Air quality in Normal Flow .

Phase 2 : Innovation

Description : Using a Random Forest algorithm to predict the air Quality and Find a accuracy for a prediction.

Air Quality Analysis and Prediction in Tamilnadu

1. Collect and prepare data :

Gather a dataset of Air quality with their So2,NO2 And RSPM rates. Preprocess the data by cleaning the text, removing stop words, and stemming or lemmatizing the words.

2. Choose a Random Forest Technique :

There are many random forest techniques used to predict the value of Air quality.

3. Train Random Forest model.

Feed the prepared data to the model and allow it to learn the relationships between the features and the target variable (Accuracy of SO2 and NO2).

4. Evaluate the model.

Evaluate the performance of the model on a held-out test set. This will give an idea of how well about the model will generalize to new data.

5. Deploy the model.

Once you are satisfied with the performance of the model, it can deploy it to production. This may involve saving the model to a file or deploying it to a cloud-based platform.

Here is an example of how to train a simple feedforward Machine Learning Algorithms such as Random Forest Technique

Random forest in Python :

```
Import pandas as pd
```

```
From sklearn.model_selection import train_test_split
```

```
From sklearn.ensemble import RandomForestRegressor
```

```
From sklearn.metrics import mean_squared_error
```

```
Import matplotlib.pyplot as plt
```

```
# Load your dataset
```

```
Data = pd.read_csv('cpcb_dly_aq_tamil_nadu-2014.csv')
```

```
# Split the dataset into features (X) and target (y)
```

```
X = data.drop('AQI', axis=1)
```

```
Y = data['AQI']
```

```
# Split the data into training and testing sets
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
# Create and train the Random Forest model
```

```
Rf_model = RandomForestRegressor(n_estimators=100, random_state=42)
```

```
Rf_model.fit(X_train, y_train)
```

```
# Make predictions on the test set
```

```

Y_pred = rf_model.predict(X_test)

# Calculate the Mean Squared Error (MSE) to evaluate the model
Mse = mean_squared_error(y_test, y_pred)
Print(f'Mean Squared Error: {mse}')
```



```

# Visualize the feature importances
Feature_importances = rf_model.feature_importances_
Feature_names = X.columns
Plt.barh(feature_names, feature_importances)
Plt.xlabel('Feature Importance')
Plt.ylabel('Feature Name')
Plt.show()
\.....'
```

Phase 3 :

Applied DataScience :

Project Name: Air Quality Analysis and Prediction in Tamilnadu

Project Description: To Develop a Machine learning algorithms like random forest and Using Pandas and Numpy Libraries to predict and calculate the air quality in Tamilnadu.

Phase 3: Development Part 1

Description :

Begin building the Air quality prediction model by loading and preprocessing the dataset.

Load the Air Quality dataset and preprocess the data for analysis.

Dataset Link:

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

Working Procedure :

To load and preprocess the Air Quality dataset in 2014 from Kaggle, we can use the following steps:

Step 1:

Install the necessary Python libraries

Step 2 :

Load the dataset

Load the dataset from the Kaggle website

Step 3 :

Explore the dataset

Print the first 5 rows of the dataset

Print the basic information about the dataset

Step 4 :

Preprocess the data

Handle missing values: There are no missing values in the dataset.

Convert categorical features to numerical features:

Define a function to convert categorical features to numerical features

Encode the Genre feature

Encode the Language feature

Step 5 :

Scale the numerical features

Define a function to scale numerical features

Step 6 :

Split the dataset into training and test sets.

Conclusion:

We have now loaded and preprocessed the Air Quality Analysis dataset for analysis. The next step is to build a machine learning model to predict Air quality

Program for an above steps :

```
In[1] : import pandas as pd
```

```
# Load the dataset from the Kaggle website
```

```
In [2] : air_quality=  
pd.read_csv(https://tn.data.gov.in/resource/location-wise-daily-ambient-air-quality-tamil-nadu-year-2014)
```

```
Out[2] :
```

Stn Code	Sampling Date	State	City/Town/Village/Area	Location of Monitoring Station	Agency
Type of Location	SO2	NO2	RSPM/PM10	PM 2.5	

38	1/2/2014	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai				Tamilnadu State Pollution Control Board	
NA				Industrial Area	11 17 55

38	1/7/2014	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai				Tamilnadu State Pollution Control Board	
NA				Industrial Area	13 17 45

38	21-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai				Tamilnadu State Pollution Control Board	
NA				Industrial Area	12 18 50

38	23-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai				Tamilnadu State Pollution Control Board	
NA				Industrial Area	15 16 46

38	28-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	13	14
NA				42

38	30-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	14	18
NA				43

... ..

In [3] : air_qualitu.head()

Out [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	

38	1/2/2014	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	11	17
NA				55

38	1/7/2014	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	13	17
NA				45

38	21-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	12	18
NA				50

38	23-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	15	16
NA				46

38	28-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	13	14
NA				42

38	30-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,
Chennai	Tamilnadu State Pollution Control Board	Industrial Area	14	18
NA				43

```
<class 'pandas.core.frame.DataFrame'
```

```
# Print the basic information about the dataset
```

```
In [4] : air_quality.info()
```

```
Out [4] :
```

```
Data columns (total 6 columns):
```

```
# Column Non-Null Count Dtype
```

```
--- ----
```

```
0 stncode 105 non-null object
```


1 date 105 non-null int64

2 State 105 non-null object

3 Village 105 non-null object

4 SO2 105 non-null int64

5 NO2e 105 non-null float64

Dtypes: float64(1), int64(2), object(3)

Memory usage: 5.0+ KB

Check for missing values

In [5] : netflix_originals.isnull().sum()

Out [5] :

0 stncode 0

1 date 0

2 State 0

3 Village 0

4 SO2 0

5 NO2. 0

This means that the training set contains SO2samples and the test set contains NO2 samples.

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Phase 4 :

Applied DataScience :

Project Name: Air Quality Analysis and Prediction in Tamilnadu

Project Description: To Develop a Machine learning algorithms like random forest and Using Pandas and Numpy Libraries to predict ,calculate and using matplotlib ,seaborn to visualize the air quality in Tamilnadu.

Phase 4 : Development Part 2

Description :

Calculate average SO₂, NO₂, and RSPM/PM₁₀ levels across different monitoring stations, cities, or areas. Identify pollution trends and areas with high pollution levels.

- Create visualizations

Create visualizations using data visualization libraries (e.g., Matplotlib, Seaborn).

Load the Air Quality dataset and preprocess the data for analysis.

Dataset Link:

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

Working Procedure :

To load and preprocess the Air Quality dataset in 2014 from Kaggle, we can use the following steps:

Step 1:

Install the necessary Python libraries

Step 2 :

Load the dataset

Load the dataset from the Device

Step 3 :

Explore the dataset

Print the first 5 rows of the dataset

Print the basic information about the dataset

Step 4 :

Preprocess the data

Handle missing values: There are no missing values in the dataset.

Convert categorical features to numerical features:

Define a function to convert categorical features to numerical features

Encode the Genre feature

Encode the Language feature

Step 5 :

Scale the numerical features

Define a function to scale numerical features

Step 6 :

Split the dataset into training and test sets.

Step 7 :

Using matplotlib to visualize the air quality with Histogram,lineplot,scatterplot and more than many plots.

Conclusion:

We have now loaded and preprocessed the Air Quality Analysis dataset for analysis. Data Visualisation libraries to visualize the SO₂,NO₂ and RSPM/PM Levels.

Program for an above steps :

In[1] : import pandas as pd

Load the dataset from the Kaggle website

**In [2] : air_quality=
pd.read_csv(<https://tn.data.gov.in/resource/location-wise-daily-ambient-air-quality-tamil-nadu-year-2014.csv>)**

Out[2] :

Stn Code	Sampling Date	State	City/Town/Village/Area	Location of Monitoring Station	Agency
Type of Location	SO2	NO2	RSPM/PM10	PM 2.5	

38	1/2/2014	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai		Tamilnadu State Pollution Control Board		Industrial Area 11	17 55
NA					

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Chennai		Tamilnadu State Pollution Control Board		Industrial Area 13	17 45
NA					

38	21-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai		Tamilnadu State Pollution Control Board		Industrial Area 12	18 50
NA					

38	23-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai		Tamilnadu State Pollution Control Board		Industrial Area 15	16 46
NA					

38	28-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai		Tamilnadu State Pollution Control Board		Industrial Area 13	14 42
NA					

38	30-01-14	Tamil Nadu	Chennai	Kathivakkam, Municipal Kalyana Mandapam,	
Chennai		Tamilnadu State Pollution Control Board		Industrial Area 14	18 43
NA					

... ..

In [3] : air_qualitu.head()

Out [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	

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Chennai	Tamilnadu State Pollution Control Board		Industrial Area	13	14
NA					42

38 30-01-14 Tamil Nadu Chennai Kathivakkam, Municipal Kalyana Mandapam,
Chennai Tamilnadu State Pollution Control Board Industrial Area 14 18 43
NA

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<class 'pandas.core.frame.DataFrame'
```

```
# Print the basic information about the dataset
```

```
In [4] : air_quality.info()
```

```
Out [4] :
```

```
Data columns (total 6 columns):
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# Column      Non-Null Count  Dtype
```

```
---  -----  -
```

```
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```
1 date          105 non-null   int64
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```
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```
3 V
```

```
llage        105 non-null   object
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5 NO2e 105 non-null float64

Dtypes: float64(1), int64(2), object(3)

Memory usage: 5.0+ KB

Check for missing values

In [5] : netflix_originals.isnull().sum()

Out [5] :

Import matplotlib.pyplot as plt

Create a figure and axis

Ax = plt.subplots()

Plot PM2.5 data

Ax.plot(dates, pm25_values, label='PM2.5', marker='o')

Plot PM10 data

Ax.plot(dates, pm10_values, label='PM10', marker='x')

Set labels and title

Ax.set_xlabel('Date')

Ax.set_ylabel('Air Quality Index')

Ax.set_title('Air Quality Over Time')

Ax.legend()

Rotate x-axis labels for better readability

Plt.xticks(rotation=45)

Show the plot

Plt.tight_layout()

Plt.show()

Histogram :

Plt.hist(air_quality_data, bins=10, edgecolor='k', alpha=0.7)

Set labels and title

Plt.xlabel('PM2.5 Levels')

Plt.ylabel('Frequency')

```
Plt.title('Air Quality Histogram')
```

```
# Show the plot
```

```
Plt.show()
```

Scatterplot :

```
Plt.scatter(pm25_values, pm10_values, c='b', marker='o', label='Air Quality Data')
```

```
# Set labels and title
```

```
Plt.xlabel('PM2.5 Levels')
```

```
Plt.ylabel('PM10 Levels')
```

```
Plt.title('Air Quality Scatterplot')
```

```
# Show the plot
```

```
Plt.legend()
```

```
Plt.grid(True)
```

```
Plt.show()
```

Coding :

Import pandas as pd

From sklearn.model_selection import train_test_split

From sklearn.ensemble import RandomForestRegressor

From sklearn.metrics import mean_squared_error

Import matplotlib.pyplot as plt

Load your dataset

Data = pd.read_csv('cpcb_dly_aq_tamil_nadu-2014.csv')

Split the dataset into features (X) and target (y)

X = data.drop('AQI', axis=1)

Y = data['AQI']

Split the data into training and testing sets

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

Create and train the Random Forest model

Rf_model = RandomForestRegressor(n_estimators=100, random_state=42)

Rf_model.fit(X_train, y_train)

Make predictions on the test set

Y_pred = rf_model.predict(X_test)

Calculate the Mean Squared Error (MSE) to evaluate the model

Mse = mean_squared_error(y_test, y_pred)

Print(f'Mean Squared Error: {mse}')

Visualize the feature importances

```
Feature_importances = rf_model.feature_importances_
```

```
Feature_names = X.columns
```

```
Plt.barh(feature_names, feature_importances)
```

```
Plt.xlabel('Feature Importance')
```

```
Plt.ylabel('Feature Name')
```

```
Plt.show()
```

```
.....-----
```