### **Phase-1 Submission**

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#### 1. Problem Statement

Air pollution is a growing concern worldwide, especially in urban regions. Its adverse effects on human health, ecosystems, and climate make accurate forecasting of air quality a vital task. This project aims to leverage advanced machine learning algorithms to predict air quality levels (e.g., AQI), enabling authorities and citizens to make proactive decisions.

# 2. Objectives of the Project

- Predict the Air Quality Index (AQI) based on environmental and meteorological parameters.
- Identify key contributors (e.g., pollutants) to air pollution.
- Generate insights to help government agencies, researchers, and the public.
- Optionally develop a web-based dashboard for real-time AQI forecasting.

# 3. Scope of the Project

- Features: PM2.5, PM10, NO2, SO2, CO, O3, temperature, humidity, wind speed.
- Algorithms: Regression models, Ensemble methods (e.g., Random Forest, XGBoost), possibly time-series models like LSTM.
- Limitations: Real-time data availability, data quality from sensors, and computational constraints.

#### 4.Data Sources

- Source: Kaggle, UCI, OpenAQ API, Central Pollution Control Board (India).
- Type: Public datasets; both static (historical data) and dynamic (API-fed real-time data).
- Format: CSV, JSON.
- Data

  source:

  <a href="https://www.kaggle.com/datasets/waqi786/global-air-quality-dataset?resour">https://www.kaggle.com/datasets/waqi786/global-air-quality-dataset?resour</a>

  ce=download

## 5. High-Level Methodology

#### a. Data Collection

- Download datasets from platforms like Kaggle/UCI.
- Use APIs (e.g., OpenAQ) for dynamic updates.

### b. Data Cleaning

- Handle missing/null values.
- Remove duplicates and irrelevant features.

• Normalize pollutant concentrations.

#### c. Exploratory Data Analysis (EDA)

• Visualize pollutant trends, seasonal patterns, and location-wise variations using heatmaps, line plots, and box plots.

#### d. Feature Engineering

• Create pollutant ratios, categorize AQI levels, and add time-based features.

#### e. Model Building

- Test with Linear Regression, Random Forest, XGBoost, LSTM (if time-series).
- Optimize using GridSearchCV or RandomizedSearch.

#### f. Model Evaluation

- *Use MAE, RMSE, and R<sup>2</sup> Score.*
- Employ cross-validation.

## g. Visualization & Interpretation

- Use graphs and dashboards to show pollutant levels, AQI trends, and predictions.
- Tools: Matplotlib, Seaborn, Plotly.

# h. Deployment (Optional)

- Use Streamlit or Flask to create a web app for AQI prediction.
- Integrate with real-time API for live monitoring.

# **6.Tools and Technologies**

Category	Tools & Technologies		
Programming Language	Python		
IDE/Notebook	Jupyter Notebook / Google Colab		
Libraries	pandas, numpy, seaborn, matplotlib, scikit-learn, XGBoost, TensorFlow (optional)		
Deployment Tools	Deployment Tools		

# 7.Team Members and Roles

Team Member Name	Role/Responsibility	
Jeniliya	Data Collection, Model Building,	
Aadharsh	Visualization, Deployment	
Ashwinth	Evaluation, Presentation	
Jeevanandan	Documentation, EDA	
Ranjana sri	Reporting, , Feature Engineering	