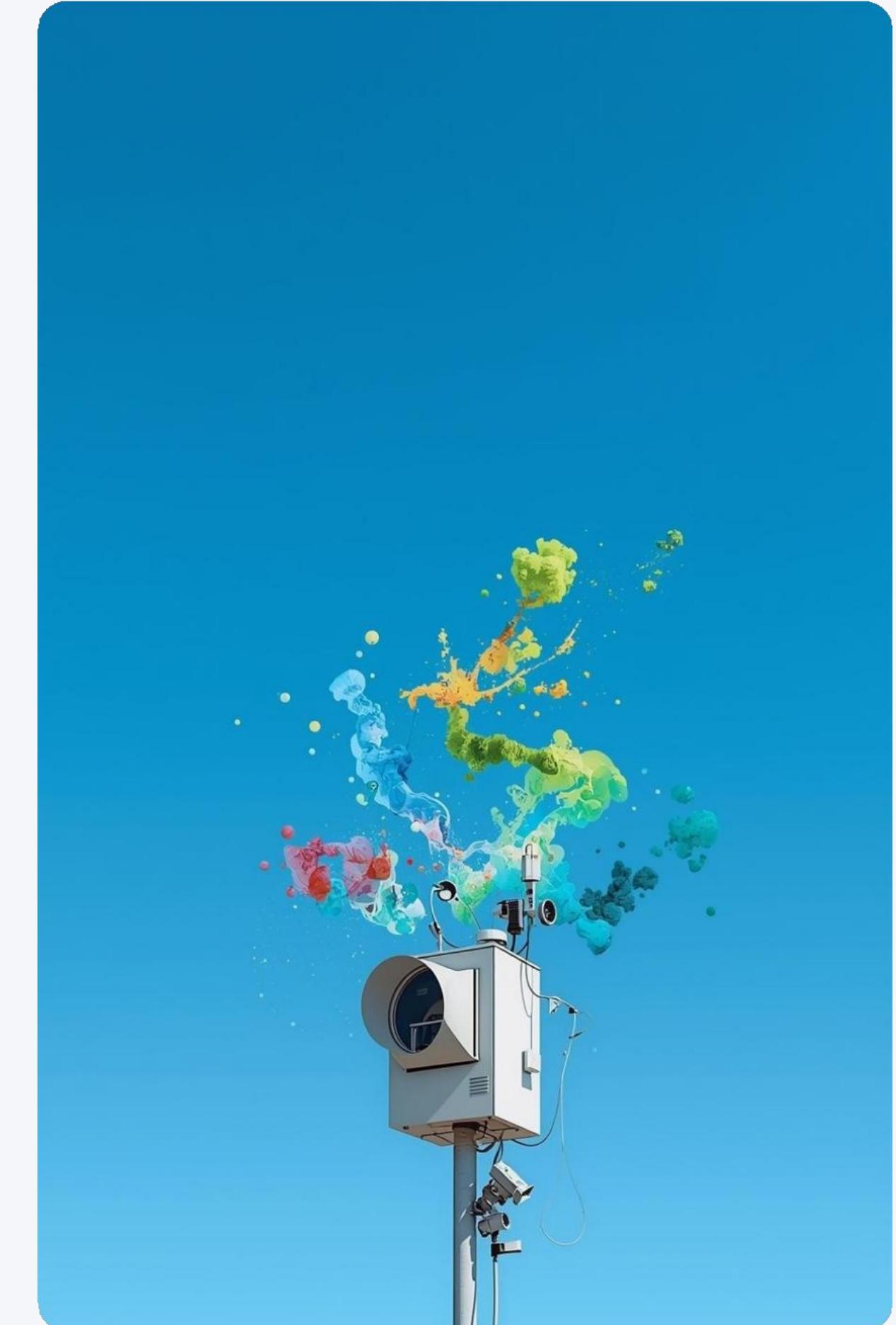


# AIR QUALITY INDEX PREDICTION - ML Model

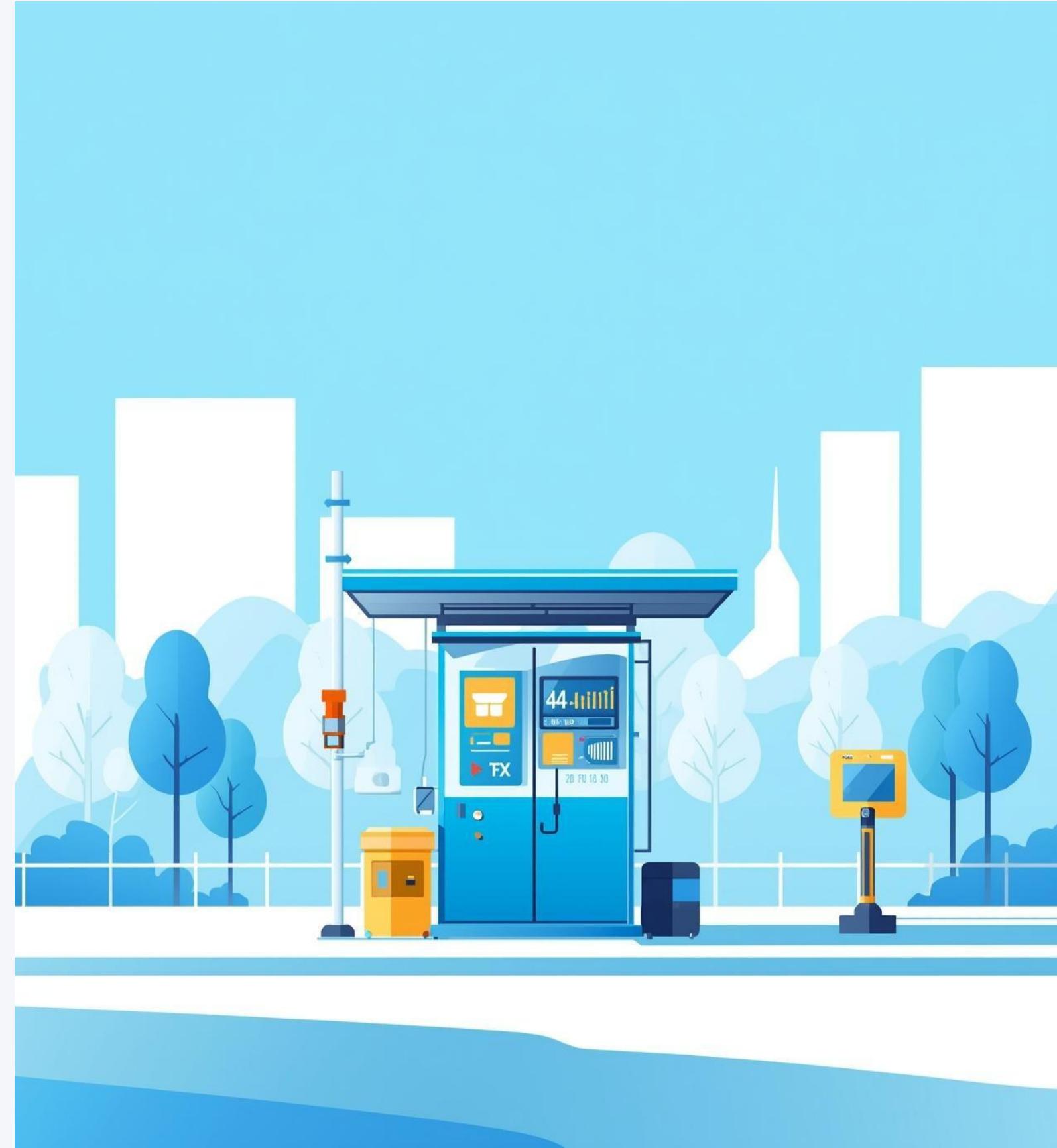
team a

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Chinthala Purnachander Rao



# Overview of AQI Prediction System

This project focuses on developing a Machine Learning-based system to predict the Air Quality Index (AQI) using key air pollutants such as PM<sub>2.5</sub>, PM<sub>10</sub>, NO<sub>2</sub>, SO<sub>2</sub>, CO, and O<sub>3</sub>. The goal is to provide an accurate and easy-to-understand prediction of air quality levels to promote environmental awareness.





# Data Collection Setup

The project begins with **data collection** from various sources, including governmental and environmental organizations, focusing on key pollutants like PM<sub>2.5</sub>, PM<sub>10</sub>, NO<sub>2</sub>, SO<sub>2</sub>, CO, and O<sub>3</sub> to ensure accurate AQI predictions.

	AQI Value	CO AQI Value	Ozone AQI Value	NO2 AQI Value	PM2.5 AQI Value	lat	lon
0	51	1	36	0	51	44.7444	-51.2900
1	41	1	5	1	41	-5.2900	-11.2958
2	41	1	5	1	41	37.1667	53.0167
3	66	1	39	2	66		
4	34	1	34	0	20		

# Model Training and Accuracy Criteria

The model training phase focused on evaluating various regression algorithms. Selection criteria emphasized **accuracy and reliability**, ensuring optimal predictions for AQI based on collected pollutant data and environmental factors.

	MAE	MSE	RMSE	R2 Score	Accuracy (%)	
Linear Regression	4.126361	41.797808	6.465122	0.974798	97.479807	
Random Forest	0.090036	2.388645	1.545524	0.998560	99.855977	
Gradient Boosting	0.827992	3.049675	1.746332	0.998161	99.816120	
SVR	2.920637	278.538232	16.689465	0.832056	83.205574	
KNN	1.316502	6.643031	2.577408	0.995995	99.599459	

# User Interface Development Overview

Air Quality Index Prediction

Select your city, enter pollutant values, and get the AQI prediction + category.

City  
Select the city  
Hyderabad

PM2.5  
2

PM10  
4

NO2  
3

SO2  
2

CO  
1

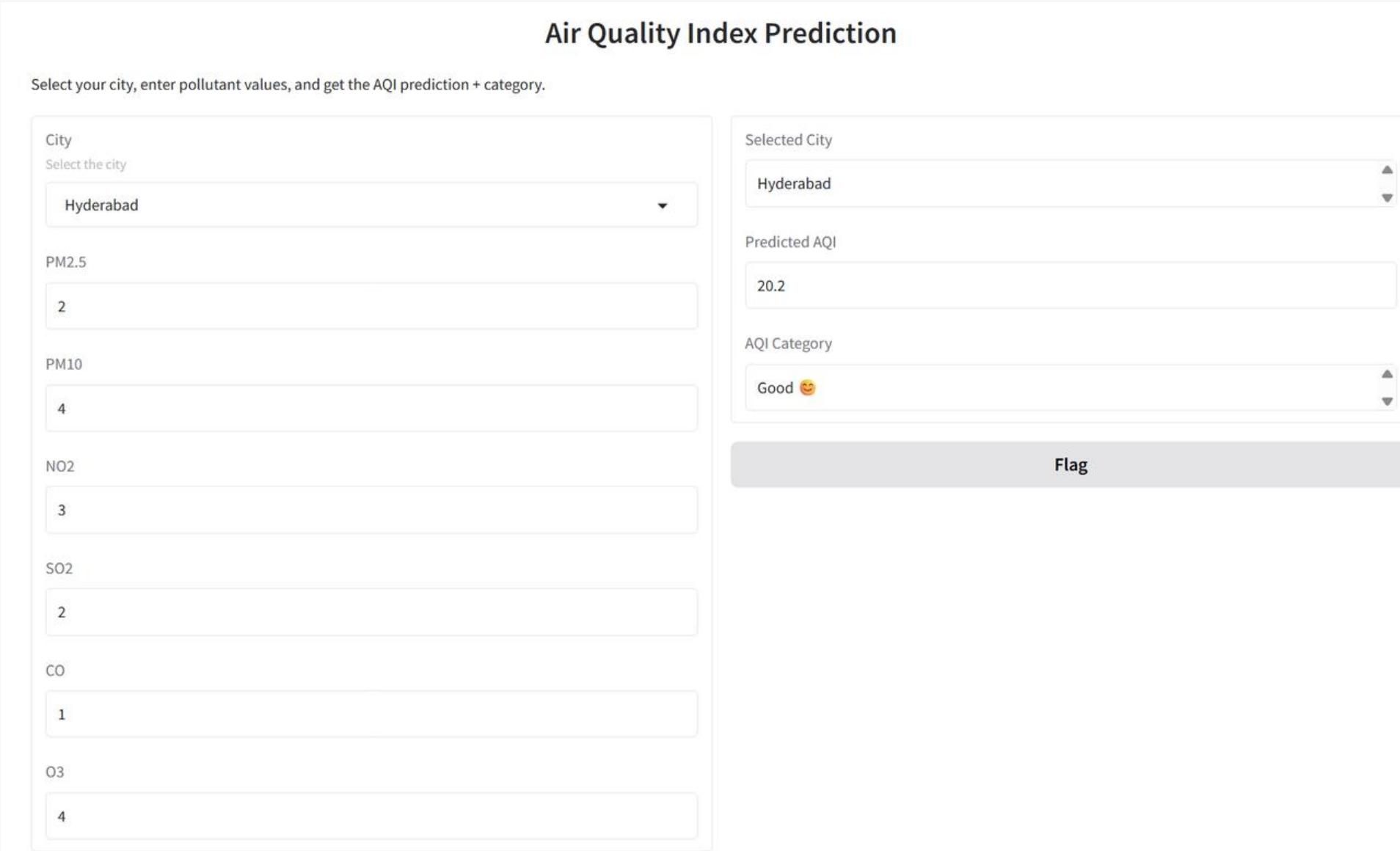
O3  
4

Selected City  
Hyderabad

Predicted AQI  
20.2

AQI Category  
Good 😊

Flag



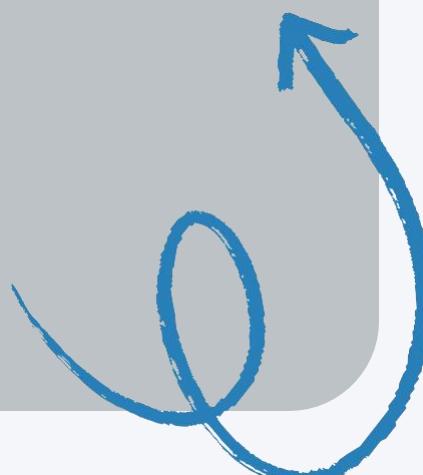
```
def predict_aqi(city, PM2_5, PM10, NO2, SO2, CO, O3):  
    input_data = np.array([[PM2_5, PM10, NO2, SO2, CO, O3]])  
    prediction = model.predict(input_data)[0]  
    prediction = round(prediction, 2)
```

```
# AQI Category  
if prediction <= 50:  
    category = "Good 😊"  
  
elif prediction <= 100:  
    category = "Satisfactory ☺️"  
  
elif prediction <= 200:  
    category = "Moderate ☹️"  
  
elif prediction <= 300:  
    category = "Poor ☹️"  
  
elif prediction <= 400:  
    category = "Very Poor ☹️"
```

# Progress Milestones

## **Achieving steady and timely completion**

Throughout the 8-week project, we maintained a **consistent pace**, successfully meeting all key milestones while adapting to challenges, ensuring the AQI prediction system was developed efficiently and on schedule.



# Challenges Faced

## Encountering obstacles in AQI project

### Data Inconsistencies

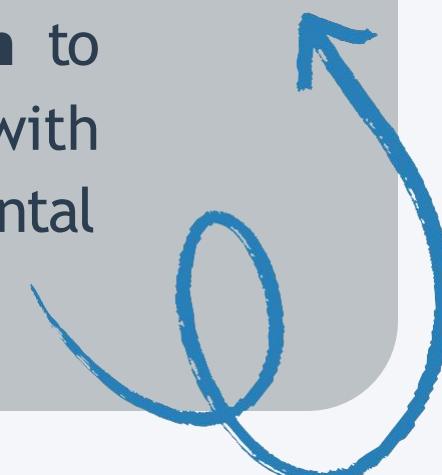
Inconsistent and incomplete air quality data presented significant challenges, requiring **extensive data cleaning** and validation processes to ensure reliability and accuracy in model training and predictions.

### Model Selection

Selecting the best regression model involved evaluating multiple algorithms, requiring **careful analysis of performance metrics** and adjustments to ensure optimal accuracy in AQI predictions.

### AQI Mapping

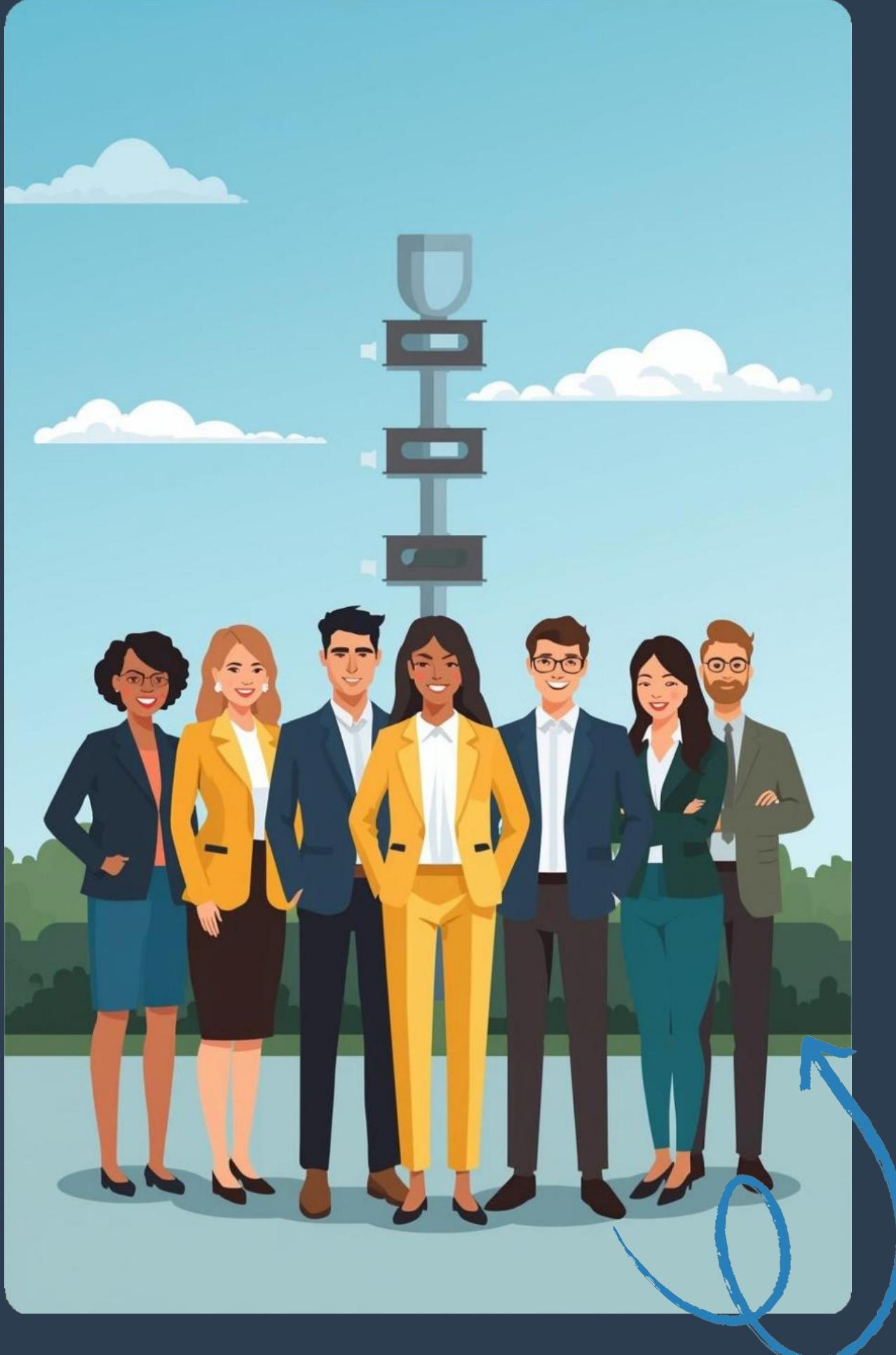
Correctly mapping predicted AQI values to standard categories posed a challenge, necessitating a **robust classification approach** to align predictions with established environmental standards effectively.



# Key Learnings from the Project

Throughout the project, I gained insights into effective **data cleaning techniques**, learned the importance of model evaluation metrics, and developed a deeper understanding of how pollutants affect air quality.





# Thank You

We appreciate your attention and  
questions!