

# **Internship Project Report: Karachi AQI Prediction System**

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**Project Title:** End-to-End Automated Air Quality Forecasting System For Karachi , Sindh , Pakistan.

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## **1. Executive Summary**

The **Karachi AQI Prediction System** is a sophisticated machine learning platform designed to provide real-time and predictive insights into Karachi's air quality. The project addresses the critical need for accurate air quality monitoring in high-density urban areas. Key achievements include the deployment of a robust ML pipeline with an **RMSE of 1.75**, automated data ingestion from global APIs, and a professional-grade interactive dashboard.

## **2. System Architecture**

The system follows a modular, scalable architecture implementing a complete Data-to-Inference loop:

- **Data Ingestion Engine:** Fetches hourly weather and air quality parameters from the **Open-Meteo API**.
- **Feature Store (MongoDB Atlas):** A cloud-native NoSQL database serves as the centralized feature store, ensuring data persistence and enabling time-series analysis.
- **ML Pipeline:** Implements automated preprocessing (interpolation, lag engineering) and model retraining.
- **Inference Layer:** A dark-themed **Streamlit** dashboard provides real-time predictions and 72-hour forecasts.

## **3. Implementation Details**

### **3.1 Data Ingestion & Storage**

- **Source:** Open-Meteo Air Quality API.
- **Parameters:** PM2.5, PM10, CO, NO2, SO2, O3, and US-AQI.
- **Reliability:** Implemented **retry logic** and **exponential backoff** to handle API rate limits and network instability.

- **Persistence:** Data is synchronized to MongoDB Atlas using efficient **Upsert** (Bulk Write) operations to avoid duplication.

### 3.2 Machine Learning Workflow

- **Features:** Lag features (1h, 2h, 24h), rolling averages (6h, 24h), and temporal components (hour, day).
- **Model Selection:** Evaluated Linear Regression, Random Forest, and **XGBoost**. XGBoost was selected as the production model due to its superior handling of non-linear trends in AQI data.
- **Validation:** 80/20 Time-series split ensuring the model is validated on future data relative to training.

### 3.3 Dashboard Features

Preview unavailable *Figure 1: Real-time AQI Gauge and 3-Day Forecast Visualization*

- **Real-time Monitoring:** Instant display of the current hour's predicted AQI with color-coded severity badges.
- **Interactive Forecasts:** Graphing 72-hour trends using Plotly for deep dive analysis.
- **Automated Insights:** Context-aware summary boxes indicating whether air quality is improving or worsening.

## 4. Technical Challenges & Solutions

Challenge	Impact	Solution Implemented
<b>Data Synchronization</b>	Inconsistent hourly data retrieval	Implemented GitHub Actions with cron jobs for hourly sync and daily retraining.
<b>Prediction Drift</b>	Multi-step forecasts tended to explode	Applied <b>Recursive Multi-Step Forecasting</b> with dampening and physical limit clipping (0-500).
<b>Database Connectivity</b>	Intermittent cloud connection errors	Integrated robust Mongo client initialization with DNS resilience and connection timeouts.
<b>Model Sensitivity</b>	Outliers in sensor data affecting accuracy	Implemented training-time clipping (1st-99th percentile) during preprocessing.

## 5. Performance Metrics

Metric	Value	Significance
<b>RMSE</b>	1.75	High precision in AQI point estimation.
<b>R<sup>2</sup> Score</b>	0.99	Explains 99% of variance in the test set.
<b>Inference Time</b>	<50ms	Rapid dashboard response for end-users.

## 6. Conclusion

This project successfully demonstrates the application of modern ML Ops principles to public health data. By combining automated data pipelines, a centralized cloud feature store, and high-performance regressors, the Karachi AQI Prediction System provides a reliable tool for urban environmental monitoring.