



Project Progress Review #1

Project Title: Monitoring the concentration of air pollutants and its health hazards using Machine Learning models.

Project ID : 102

Project Guide: Prof. Saritha

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Outline

- Abstract and Scope of the Project
- Capstone Project Phase 1
 - Summary of work
 - Suggestions
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- Approach using simulated imitated data
- Expected Deliverables
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Abstract

- Introducing a continuous air quality monitoring system that tracks local air quality to anticipate higher chances of lung cancer. A blend of Adaptive LSTM and ARIMA ML models will be employed and hosted on a cloud platform.
- The system's focus is to constantly gauge PM2.5, PM10, NO2, and CO levels, offering users real-time insights into their environmental air quality.
- Utilizing this data, the system will evaluate potential lung cancer risks, proactively alerting users to changing air quality and potential health concerns via IoT sensors.
- Our vision is to provide users with timely information, helping them make informed decisions for their well-being, driven by cutting-edge technology and real-world impact.



Scope

- We are proposing a system that continuously monitors air quality in the user's area, predicting and alerting about increased lung cancer risk. Using a hybrid of Adaptive LSTM and ARIMA ML models, this will be hosted on a cloud platform.
- Our focus is on tracking PM2.5, PM10, NO2, and CO levels in real-time, offering users insights into local air quality.
- By analyzing this data, the system will gauge lung cancer risk, proactively notifying users via IoT sensors if air quality changes.
- Our aim is to empower users with timely information for informed choices, employing cutting-edge tech to make a real-world impact.



Summary of Work Done in Capstone Project Phase - 1

Work done in Phase 1:

- Performed literature review.
- Finalized the ML model.
 - bidirectional adaptive LSTM + ARIMA
- Created architectural and structural diagrams.
- Chosen cloud platforms
 - ThingSpeak
 - Hugging face, Streamlit
- Exploratory Data Analysis

Suggestions given:

- Find datasets for our city
- Make visualisation better
- Stay on track of the proposed timeline

Improvements:

- Tackle the dataset issue
- Fine tuned the modules for the IoT sensor station
- Improved data visualisation & data preprocessing



Approach using simulated imitated data

Approach if data is scarce:

 Addressing the issue without actual datasets by outlining the methodology to solve it effectively.

• Methodology Presentation:

 Illustrating step-by-step algorithm for solving the issue using a well-structured approach.

Simulated Environment:

- Generating data in a controlled, simulated environment for demonstration purposes.
- Data will be generated randomly to showcase the methodology.

This approach is a conceptual approach we will take in case proper datasets are not available.

• Simulation Tool or Manual Creation:

- Utilizing a simulation tool to generate data or manually creating representative data points.
- Ensuring the simulated data mimics realworld scenarios.

Focused Presentation:

- Focus is on showcasing the thought process, algorithmic solutions, and data generation approach.
- Emphasizing methodology's feasibility and potential real-world impact.



- Sourcing data
- Data preprocessing
- Data visualisation
- Data input
- Model training + Optimisation
- Generating results
- Real time data acquisition via IoT
- Deployment to cloud



Sourcing data

Sourced original lung cancer dataset from Harvard website Tried to source dataset from National Cancer Institute

Data pre-processing

Feature Selection Cleaned the data for NULL values and duplicate entries Data transformation



Data visualisation

Checking outliers, for each column and removing them Checking the data range via graphs Histograms, count plots, skew graphs

Data input

Automated Data Collection via IoT sensors File Upload .csv file for the already acquired data Data Streaming for continous data via IoT sensors



Model training + Optimisation

Scaling data and fitting it for the model Apply the fitted data to the model Hyperparameter Tuning to increase optimisation Model Evaluation

Generating results

Evaluate Performance
Visualize Results as different graphs
Handle Edge Cases
Deploy the model



Real time data acquisition via IoT

Sensor data generation (like MQ9/7, SharpGP2Y10, etc.) Data Retention and Archival Feedback Loop and Improvements

Deployment to cloud

Connecting the IoT sensors to ThingSpeak Deployment of ML model to Streamlit/Hugging Face



Contribution

Modules	Members	Development
Sourcing data	Everyone	Time - 1 week
Data pre-processing	Aditi Jain Ananya Adiga	Time - 3 days Lines of code - 40 lines
Data visualisation	Aditi Jain Ananya Adiga	Time - 3 days Lines of code - 50 lines



Contribution

Modules	Members	Development
Building the IoT sensor station	Anirudha Anekal Aditya R. Shenoy	Time - 2 weeks
Calibrating and code for the IoT sensors	Anirudha Anekal Aditya R. Shenoy	Time - 4 days Lines of code - 37 lines
Uploading IoT sensor readings to ThingSpeak	Anirudha Anekal Aditya R. Shenoy	Time - 3 days



Contribution

Modules	Members	Development
Coding the model	Aditi Jain Ananya Adiga	Time - 3 weeks Lines of code - 58 lines
Hyperparameter Tuning to increase optimisation Model Evaluation	Everyone	Time - 3 days Lines of code - 20 lines

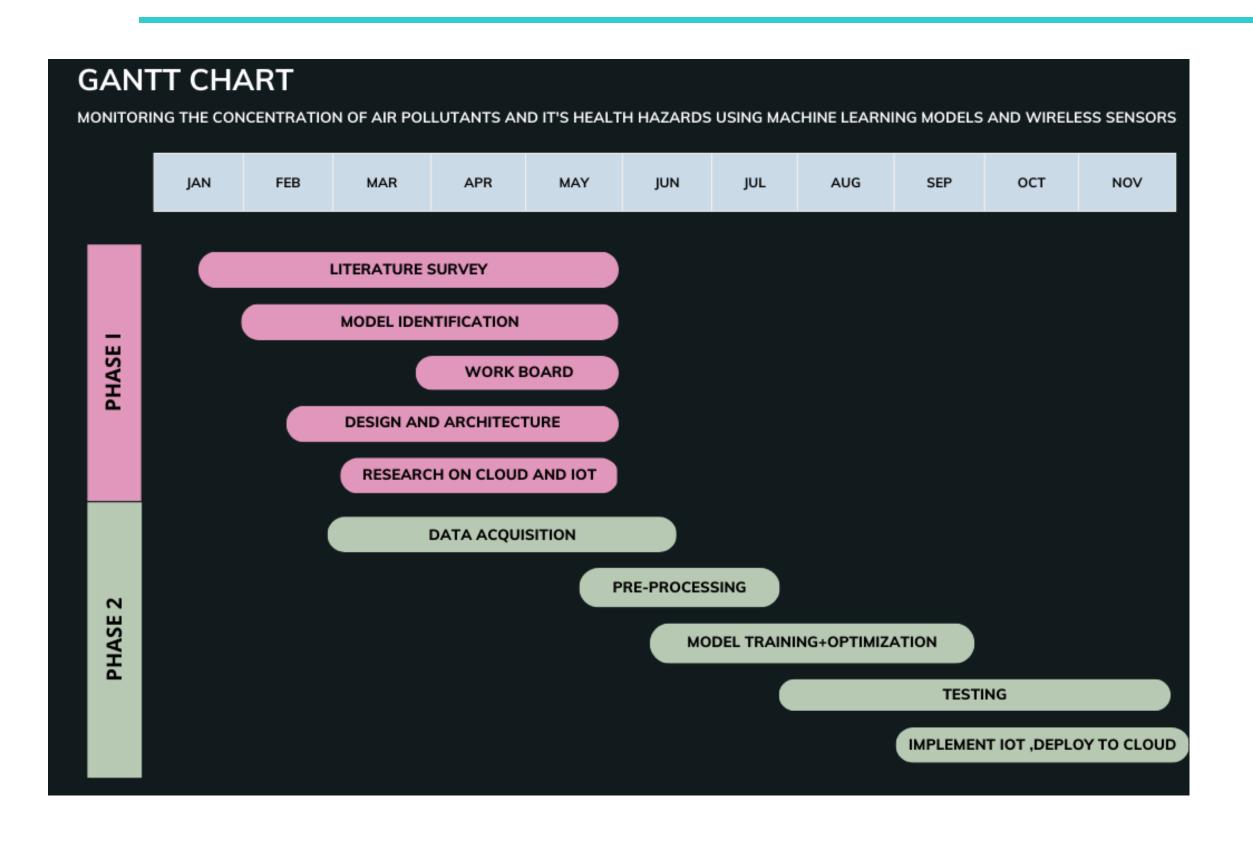


Technologies Used so far

- Google Colab
- ARIMA + enhanced LSTM
- Sensors: MQ7/9(Carbon Monoxide), GP2Y1010F (Dust)
- ESP8266
- ThingSpeak
- Hugging Face / Streamlit



Gantt Chart





References

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By: Kung-Min Wang, Kun-Huang Chen, Shieh-Hsen Tseng

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By: Samiran Rana

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By: Cosimo Magazzino , Marco Mele , Samuel Asumadu Sarkodie

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Thank You

We open the floor for questions.