1. INTRODUCTION

1.1 Project Overview

TrafficTelligence is a web-based application that uses Machine Learning to predict traffic volume based on weather and time conditions. It leverages historical traffic data and contextual features such as temperature, weather, holidays, and time of day to provide accurate predictions, assisting in dynamic traffic management and urban planning.

1.2 Purpose

The purpose of this project is to harness AI and ML capabilities to enhance urban traffic forecasting systems. Accurate traffic predictions can help in reducing congestion, improving commuter experience, and supporting city planning initiatives.

2. IDEATION PHASE

2.1 Problem Statement

With rapid urbanization, traffic congestion has become a critical issue. The existing static models fail to incorporate dynamic factors like weather and public holidays, making traffic prediction less reliable.

2.2 Empathy Map Canvas

- Think & Feel: Citizens desire a smooth commute and real-time traffic information.
- See: Crowded roads, delays, and frustration.
- Say & Do: Complain about poor traffic systems.
- Hear: Suggestions for smart traffic management.
- Pain: Time loss, fuel consumption, increased stress.
- Gain: Reliable ETA, optimized routes, reduced travel time.

2.3 Brainstorming

- Integrate real-time weather data
- Use historical data for model training
- Deploy model via web application
- Visualize traffic prediction results for easy interpretation

3. REQUIREMENT ANALYSIS

3.1 Customer Journey Map

User visits the web app \rightarrow Inputs weather & time details \rightarrow Receives traffic prediction \rightarrow Plans journey accordingly.

3.2 Solution Requirement

Dataset (traffic_volume.csv)

• ML Model: Random Forest Regressor

• Web Framework: Flask

• Frontend: HTML, CSS

• Libraries: pandas, numpy, matplotlib, seaborn, sklearn, joblib

3.3 Data Flow Diagram

[User Input] → [Preprocessing] → [Model Prediction] → [Display Output]

3.4 Technology Stack

• Frontend: HTML, CSS

Backend: Python (Flask)

• ML Libraries: scikit-learn, pandas, numpy, joblib

Deployment: Localhost / Render / Heroku

4. PROJECT DESIGN

4.1 Problem Solution Fit

Integrating weather and time into traffic prediction creates a dynamic and context-aware system, reducing guesswork in travel planning.

4.2 Proposed Solution

A trained ML model predicts traffic volume based on inputs, and a Flask-based interface makes it user-friendly and accessible.

4.3 Solution Architecture

- Input Layer (Form)
- Data Preprocessing
- ML Prediction Layer

Output Display

5. PROJECT PLANNING & SCHEDULING

5.1 Project Planning

- Week 1: Requirement Gathering & Ideation
- Week 2: Data Preprocessing
- Week 3: Model Training & Testing
- Week 4: Web App Development
- Week 5: Integration & Deployment

6. FUNCTIONAL AND PERFORMANCE TESTING

6.1 Performance Testing

- Model evaluated using:
 - o R² Score: High accuracy
 - MAE & RMSE: Low error rates
- Real-time inputs show consistent performance

7. RESULTS

7.1 Output Screenshots

- Sample Inputs: Temperature, Weather, Hour
- Output: Predicted Traffic Volume

8. ADVANTAGES & DISADVANTAGES

Advantages

- Real-time prediction
- Easy to use

High accuracy

Disadvantages

- Dependent on quality of data
- No live data fetching in current version

9. CONCLUSION

TrafficTelligence demonstrates how ML can be effectively used for urban traffic volume prediction. The project achieved high prediction accuracy and provides a usable interface for end users.

10. FUTURE SCOPE

- Integration with live traffic APIs
- Mobile app version
- Incorporate GPS data
- Advanced deep learning models

11. APPENDIX

Source Code

Included in the submitted .ipynb and Flask app files

Dataset Link

Available as traffic volume.csv

GitHub & Project Demo Link

(Include your GitHub link and deployed app URL here)