

Predictive Modeling for Traffic Volume Forecasting

*Using Machine Learning (XGBoost) on Real-World
Traffic Data*

Introduction:

- Traffic congestion is a critical urban challenge.
- Accurate prediction of traffic volume helps in better road planning and congestion management.
- This project builds a machine learning model to predict traffic volume using historical and contextual data (like time, weather, incidents).
- The XGBoost model is used for robust regression performance.

Dataset Overview:

- ◆ Dataset columns:{

Timestamp, location, traffic_volume, temperature, weather, is_holiday, incident, event}

- ◆ Target Variable:{

traffic_volume }

- ◆ Features include both numerical and categorical data.
- ◆ Preprocessing was done to extract useful features from the timestamp and other variables.

Feature Engineering:

- ◆ Extracted time-based features:

{hour, day_of_week}

- ◆ Created lag features:

{lag_1, lag_2, lag_3}

- ◆ Created rolling mean features:

{rolling_mean_3, rolling_mean_6}

- ◆ Encoded categorical variables:

{weather, event using one-hot encoding}

Data Preprocessing:

- ◆ Checked for missing values and cleaned the data.
- ◆ Converted timestamp to datetime and extracted useful components.
- ◆ Applied one-hot encoding for categorical features.
- ◆ Prepared the final feature set for modeling.

Model Training:

- ❖ Model used: XGBoost Regressor
- ❖ Parameters: n_estimators=100, learning_rate=0.1
- ❖ Split data: 80% training, 20% testing
- ❖ Target: traffic_volume

Model Evaluation:

- ✓ Evaluation Metrics:
 - ❖ MAE (Mean Absolute Error)
 - ❖ RMSE (Root Mean Squared Error)
 - ❖ R^2 Score
- ✓ Predicted traffic volume vs actual shown using line plot.
- ✓ Visualized performance over first 100 samples.

Feature Importance:

- ❖ XGBoost provides feature importance scores.
- ❖ Most important features were:
- ❖ hour, day_of_week
- ❖ recent lag values and rolling_mean
- ❖ contextual info like weather and incident
- ❖ Bar chart used to display importance scores.

Conclusion & Future Work

- ❖ XGBoost model successfully predicts traffic volume with good accuracy.
- ❖ Useful for real-time traffic monitoring and planning.
- ❖ Future enhancements:
 - ❖ Include real-time sensor data
 - ❖ Use deep learning (e.g., LSTM)
 - ❖ Deploy as a live dashboard or API

THANK YOU