

Name of Project: EV Vehicle/Charging Demand Prediction

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Learning Objectives

Analyze real-world EV adoption data to forecast charging demand using a full ML pipeline.

- Apply time series forecasting techniques.
- Perform data cleaning, transformation, and visualization.
- Engineer features (lags, trends, rolling averages).
- Build and evaluate regression-based prediction models.
- Interpret results using forecasting metrics.
- Deploy an interactive dashboard with Streamlit.
- Compare EV trends across regions.
- Gain end-to-end AI/ML project experience.



Tools and Technology used

Languages & Libraries

- **Python** – Core programming language
- **pandas, numpy** – Data manipulation & processing
- **scikit-learn** – Machine learning models
- **matplotlib, statsmodels** – Visualization & statistical analysis
- **joblib, tensorflow** – Model saving & support

Development & Deployment

- **Jupyter Notebook** – Model development & experimentation
- **Streamlit** – Interactive dashboard deployment
- **VS Code** – Code editing & debugging

Environment

- **Virtual environment** managed via requirements.txt

Data Source

- EV registration data from **Washington State, USA**

Methodology

1. Data Collection

Collected county-wise EV registration data from Washington State.

2. Data Preprocessing

Handled missing values, parsed dates, and encoded categorical features.

3. Feature Engineering

Generated lag features, rolling averages, growth rates, and trend indicators.

4. Model Building

Trained a **Random Forest** model using the engineered features.

5. Forecasting

Predicted monthly EV adoption for the next 3 years.

6. Visualization

Plotted EV adoption trends for individual and multiple counties.

7. Deployment

Created an interactive dashboard using **Streamlit** for real-time insights.

Problem Statement:

Despite rising interest in electric vehicles, government planners and stakeholders lack effective tools to:

- Understand current EV adoption trends at the regional (county) level
- Predict future EV growth for infrastructure and policy planning
- Visualize adoption patterns across different counties

To address this, we built a machine learning model that forecasts EV adoption for the next 3 years and displays the results through an **interactive, user-friendly dashboard**.

Solution:

- Developed a machine learning model using historical EV registration data.
- Engineered time-based features to capture trends, seasonality, and growth patterns.
- Applied **Random Forest Regression** to forecast EV adoption at the county level.
- Built an interactive **Streamlit web app** for easy visualization and analysis.

Dashboard Features

- Select a county to view its EV growth forecast
- Compare adoption trends across up to **three counties**
- Analyze projected EV growth over a **3-year period**

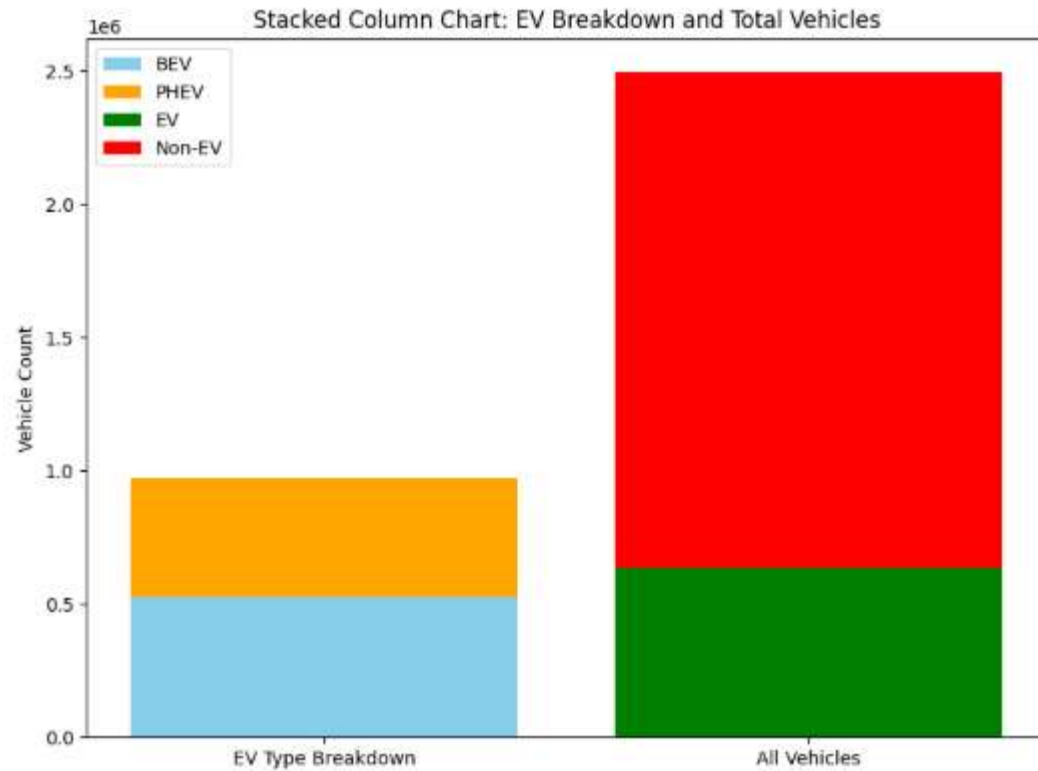
GitHub Repository Link:

https://github.com/ujjwalgupta2021/EV_Vehicle_Charge_Demand

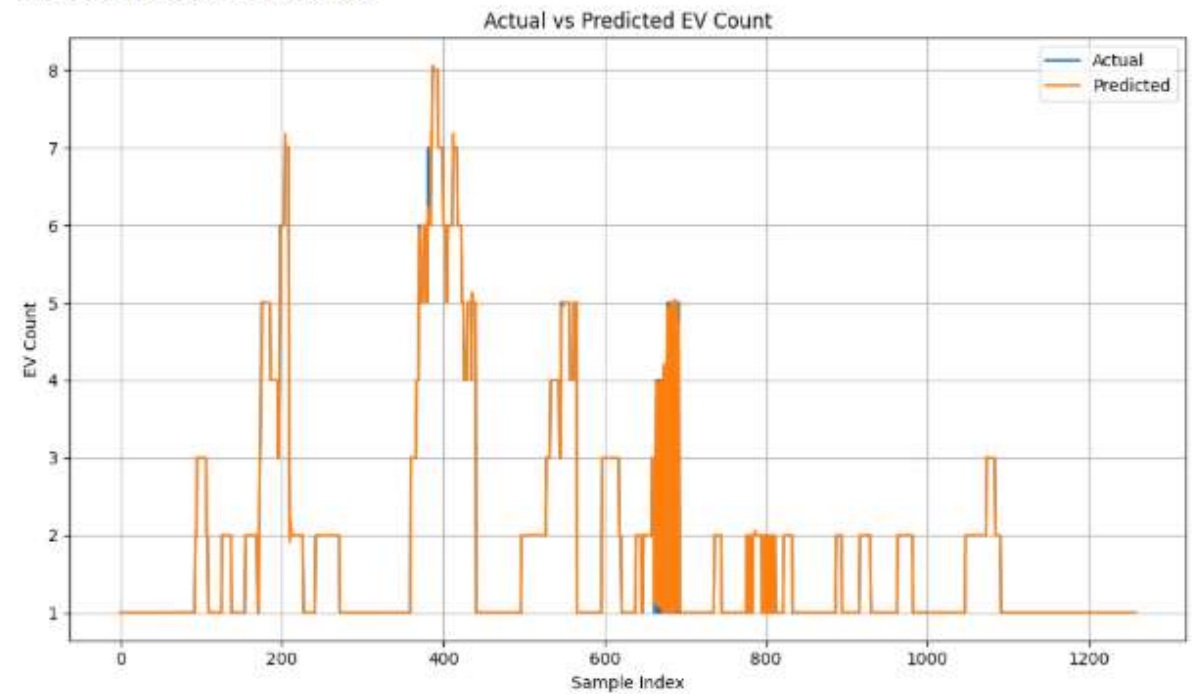
Streamlit Deployment Link:

<https://ev-demand.streamlit.app/>

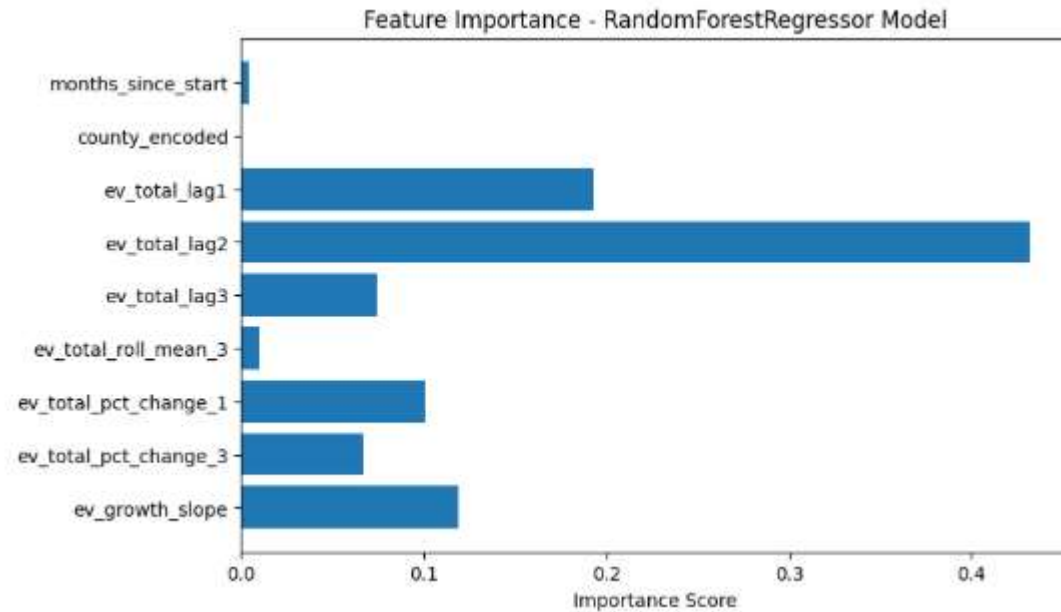
Screenshot of Output:



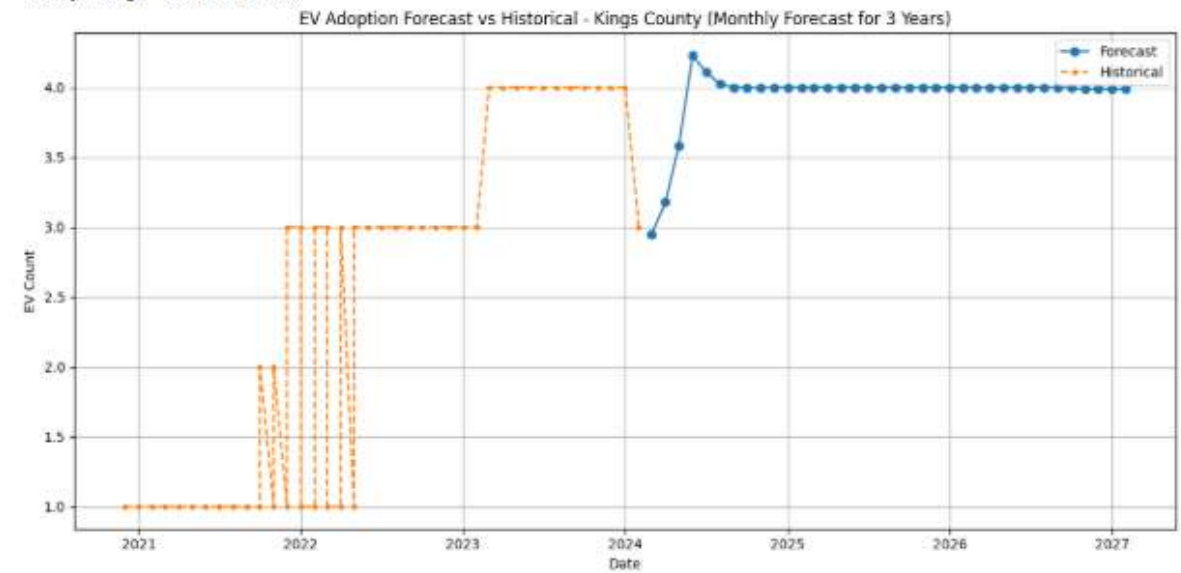
MAE: 0.01, RMSE: 0.06, R2 Score: 1.00



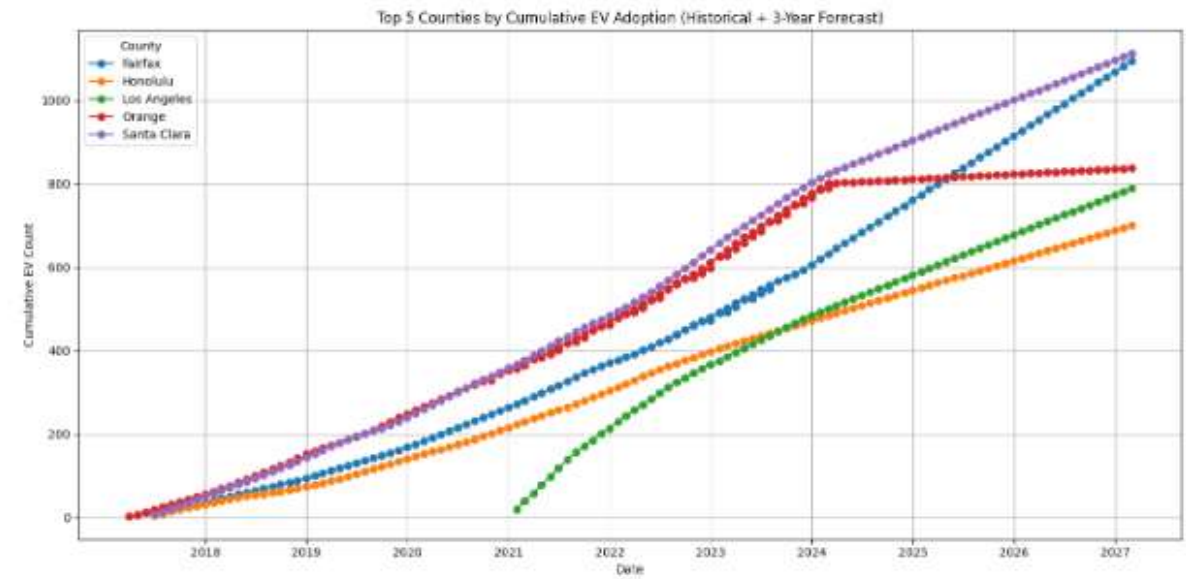
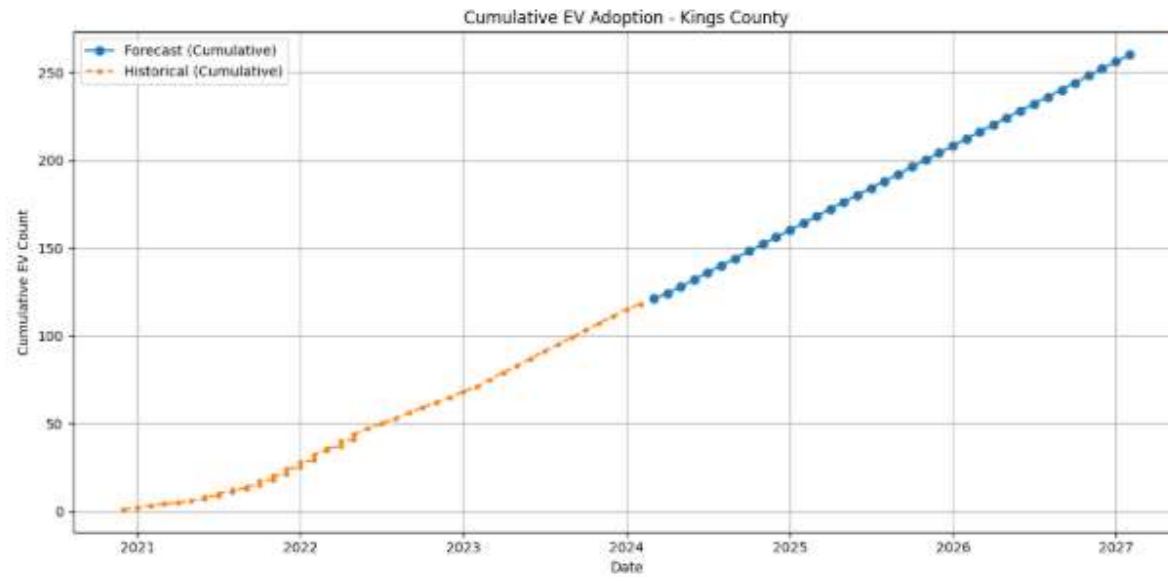
Screenshot of Output:



County 'Kings' encoded as 136.



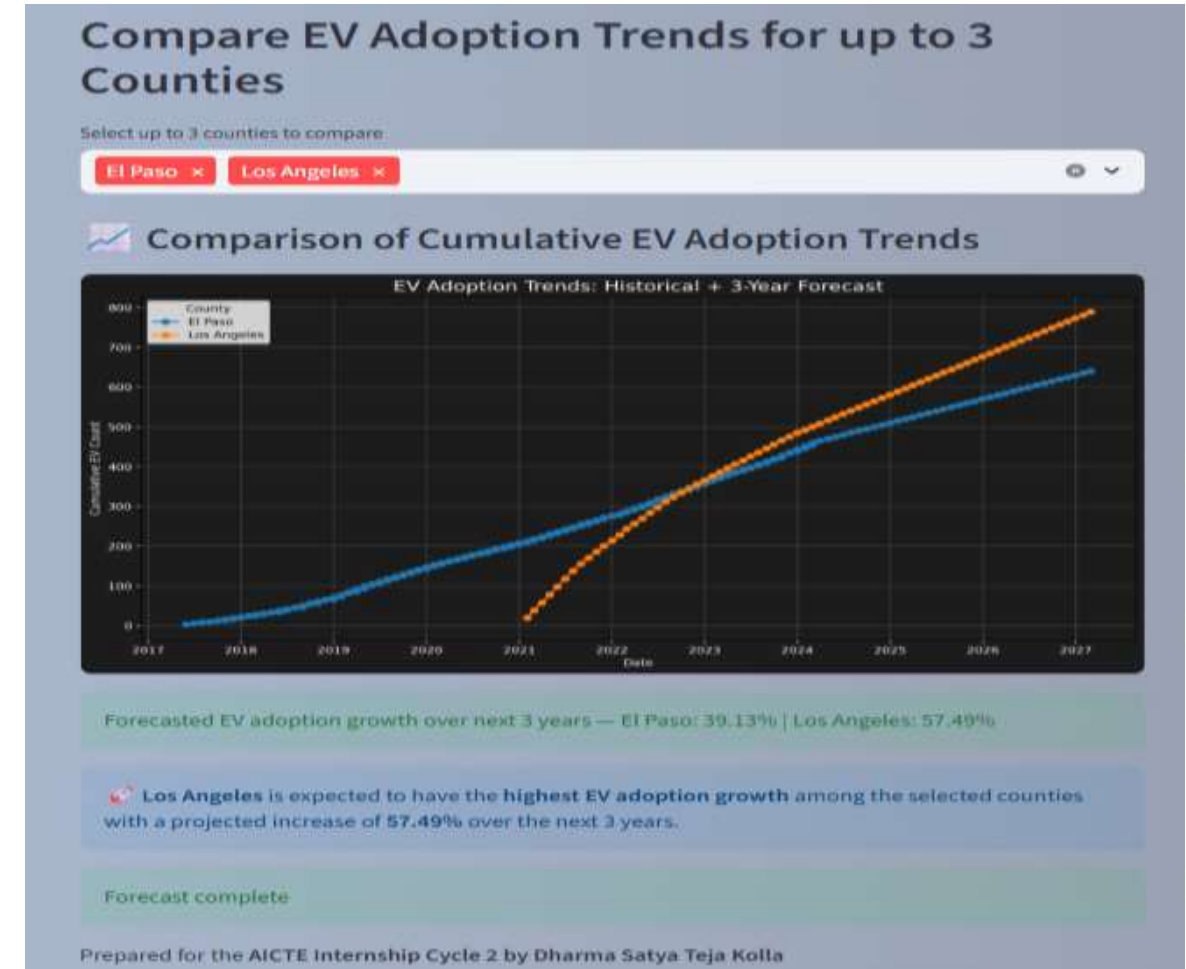
Screenshot of Output:



Screenshot of Output:

EV Adoption Forecaster for a County in Washington State

Welcome to the Electric Vehicle (EV) Adoption Forecast tool.



Conclusion:

- Successfully developed a machine learning model to forecast EV adoption using real-world data.
- Applied essential ML techniques including preprocessing, feature engineering, and regression modeling.
- Deployed an interactive **Streamlit dashboard** to visualize and explore predictions.
- Enabled data-driven insights to support regional planning and policy decisions.
- Demonstrated the practical value of AI/ML in addressing environmental and policy challenges.