

```
# Import necessary libraries
import pandas as pd
import numpy as np
```

✓ Step 1: Data Collection

Start by loading the dataset. For this example, let's assume the dataset is in CSV format.

```
# Load the dataset
df = pd.read_csv('Electric Vehicle Sales by State in India.csv')
```

```
# Display the first few rows of the dataset
df.head()
```

	Year	Month_Name	Date	State	Vehicle_Class	Vehicle_Category	Vehicle_Type	EV_Sales_Quantity
0	2014.0	jan	1/1/2014	Andhra Pradesh	ADAPTED VEHICLE	Others	Others	0.0
1	2014.0	jan	1/1/2014	Andhra Pradesh	AGRICULTURAL TRACTOR	Others	Others	0.0
2	2014.0	jan	1/1/2014	Andhra Pradesh	AMBULANCE	Others	Others	0.0
3	2014.0	jan	1/1/2014	Andhra Pradesh	ARTICULATED	Others	Others	0.0

✓ Data Preprocessing

Handle missing values and convert the date column to a proper datetime format.

```
# Convert 'Date' column to datetime format
df['Date'] = pd.to_datetime(df['Date'])
```

```
df.head()
```

	Year	Month_Name	Date	State	Vehicle_Class	Vehicle_Category	Vehicle_Type	EV_Sales_Quantity
0	2014.0	jan	2014-01-01	Andhra Pradesh	ADAPTED VEHICLE	Others	Others	0.0
1	2014.0	jan	2014-01-01	Andhra Pradesh	AGRICULTURAL TRACTOR	Others	Others	0.0
2	2014.0	jan	2014-01-01	Andhra Pradesh	AMBULANCE	Others	Others	0.0
3	2014.0	jan	2014-01-01	Andhra Pradesh	ARTICULATED	Others	Others	0.0

```
# Check for missing values
print(df.isnull().sum())
```

```
Year      0
Month_Name 0
Date      0
State     0
Vehicle_Class 0
Vehicle_Category 0
Vehicle_Type 0
EV_Sales_Quantity 0
dtype: int64
```

```
# Fill missing values (if any) using median for numerical columns or mode for categorical columns
df['EV_Sales_Quantity'].fillna(df['EV_Sales_Quantity'].median(), inplace=True)
df.fillna(df.mode().iloc[0], inplace=True)
```

```
<ipython-input-10-c57d6aa50147>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through inplace method. The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we
```

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[

```
df['EV_Sales_Quantity'].fillna(df['EV_Sales_Quantity'].median(), inplace=True)
```

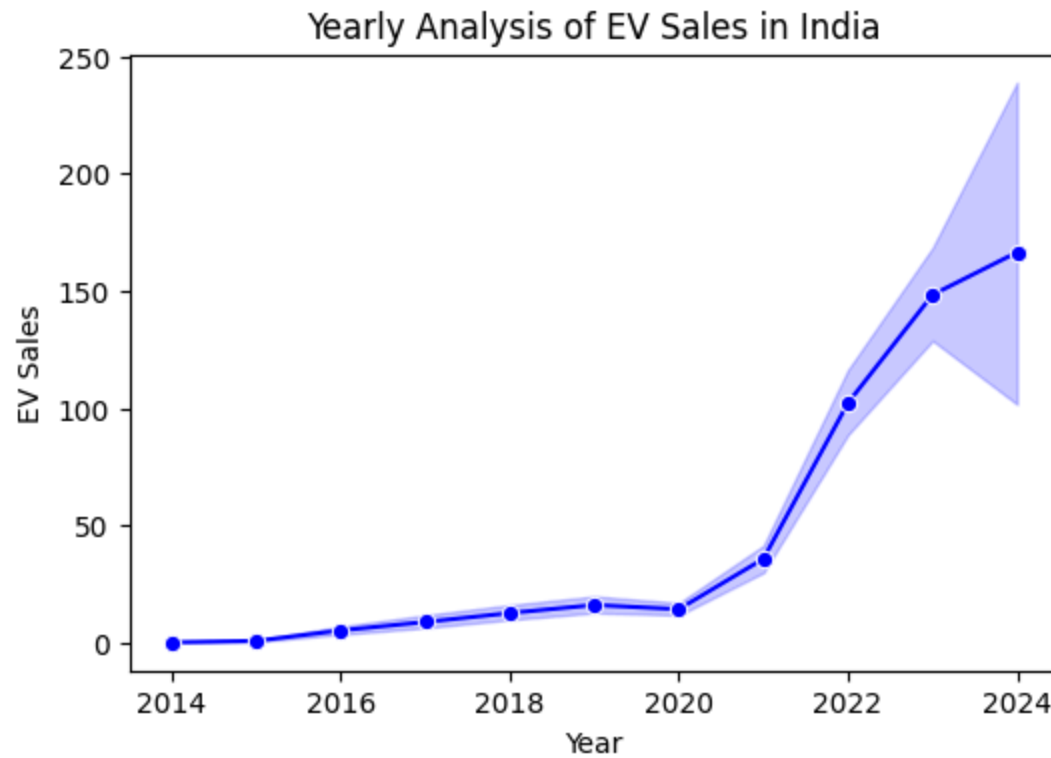
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✓ Exploratory Data Analysis (EDA)

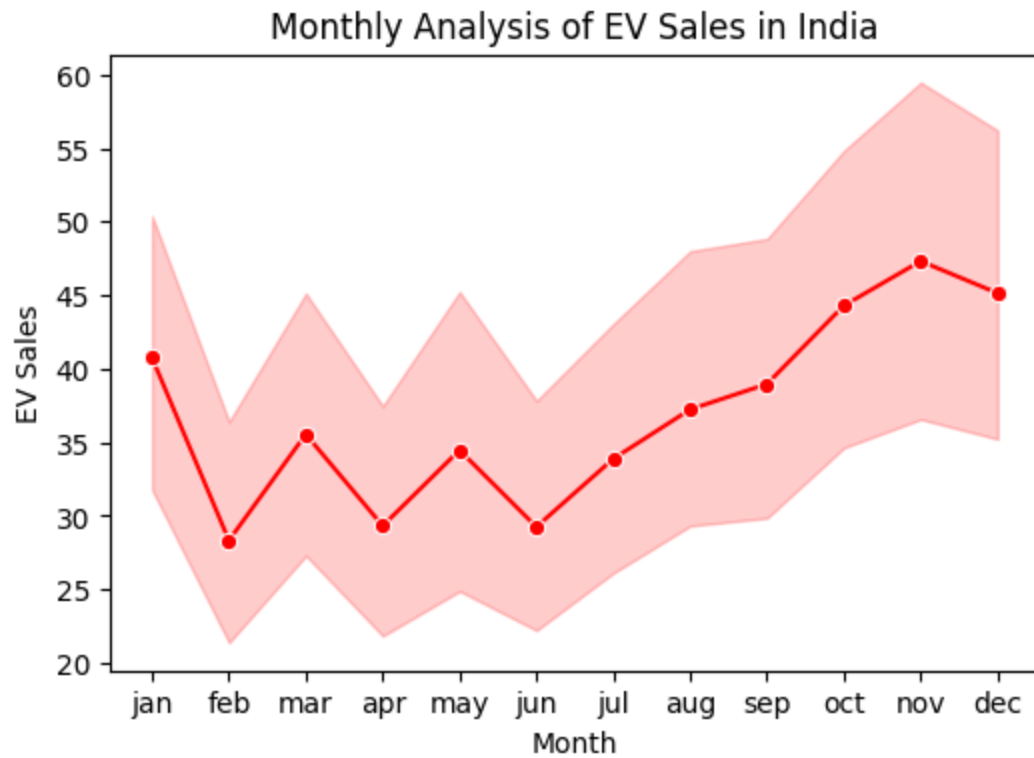
Visualize trends in EV sales over time, across states, and vehicle categories.

```
import matplotlib.pyplot as plt
import seaborn as sns
```

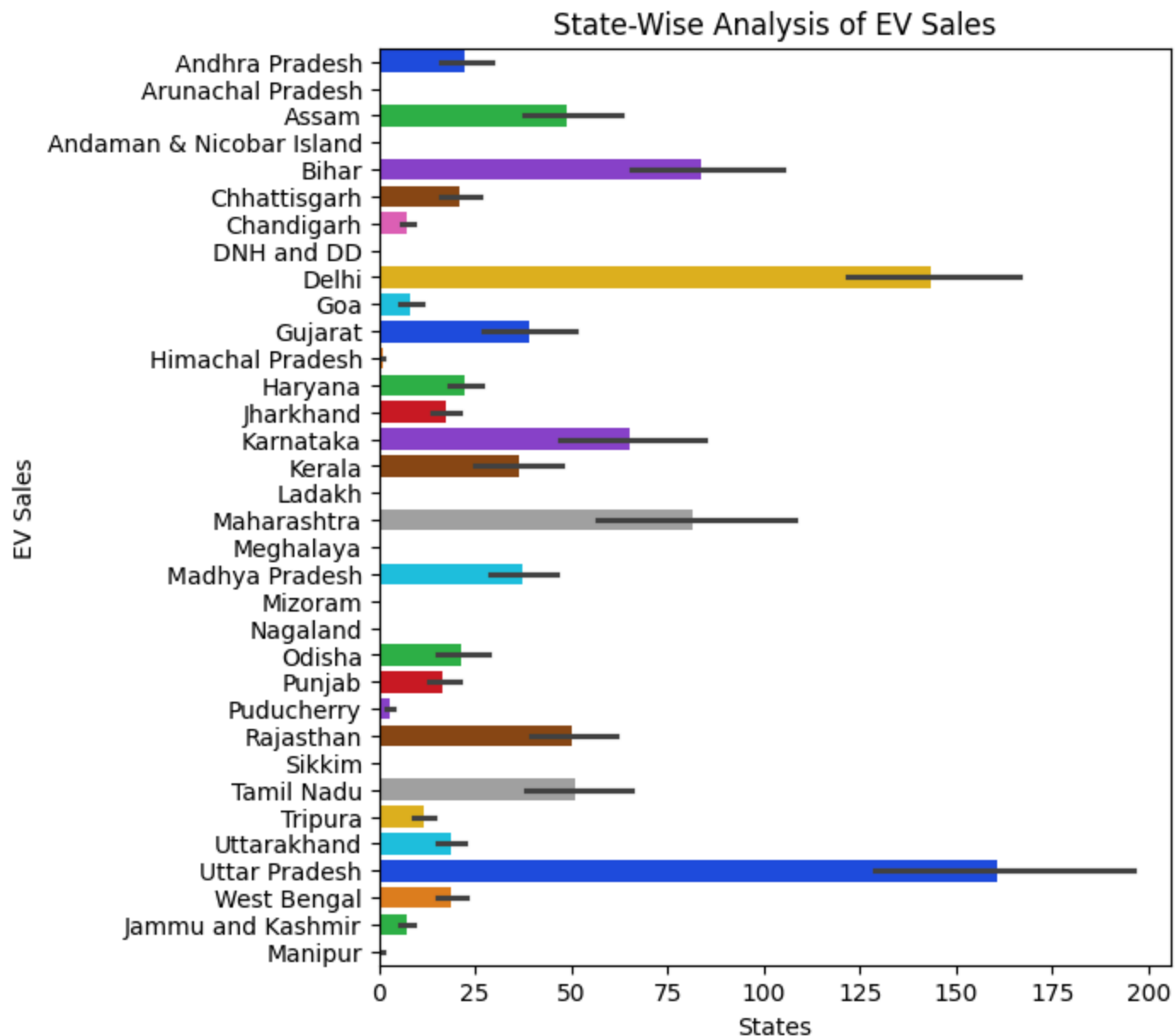
```
# Plot EV sales over the years
plt.figure(figsize=(6,4))
plt.title('Yearly Analysis of EV Sales in India')
sns.lineplot(x='Year', y='EV_Sales_Quantity', data=df,
marker='o', color='b')
plt.xlabel('Year')
plt.ylabel('EV Sales');
```



```
# Monthly Analysis of EV Sales in India
plt.figure(figsize=(6,4))
plt.title('Monthly Analysis of EV Sales in India')
sns.lineplot(x='Month_Name', y='EV_Sales_Quantity', data=df,
marker='o', color='r')
plt.xlabel('Month')
plt.ylabel('EV Sales');
```

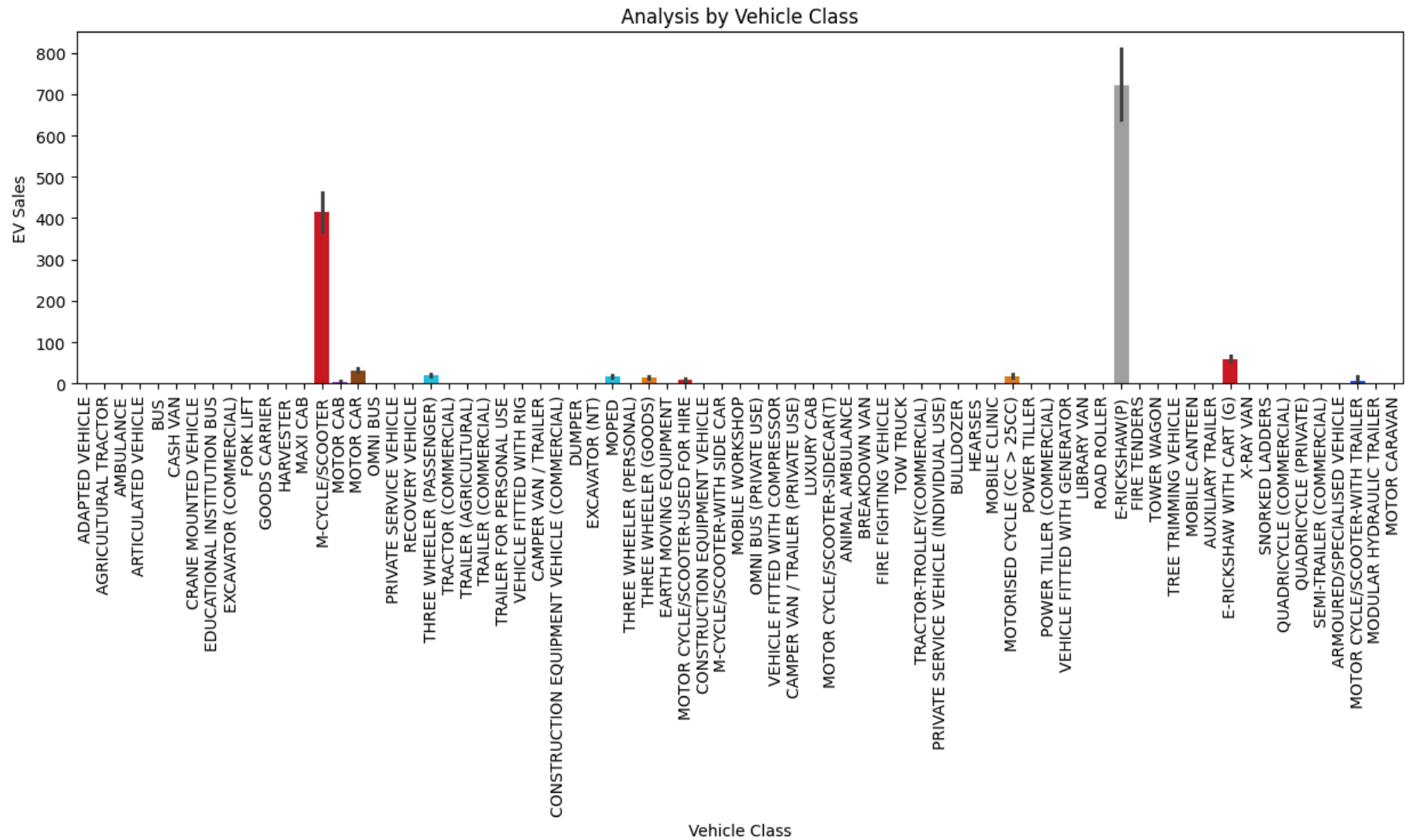


```
plt.figure(figsize=(6,7))
plt.title('State-Wise Analysis of EV Sales')
sns.barplot(y='State', x='EV_Sales_Quantity', data=df,
hue='State', palette='bright')
plt.xlabel('States')
plt.ylabel('EV Sales');
```

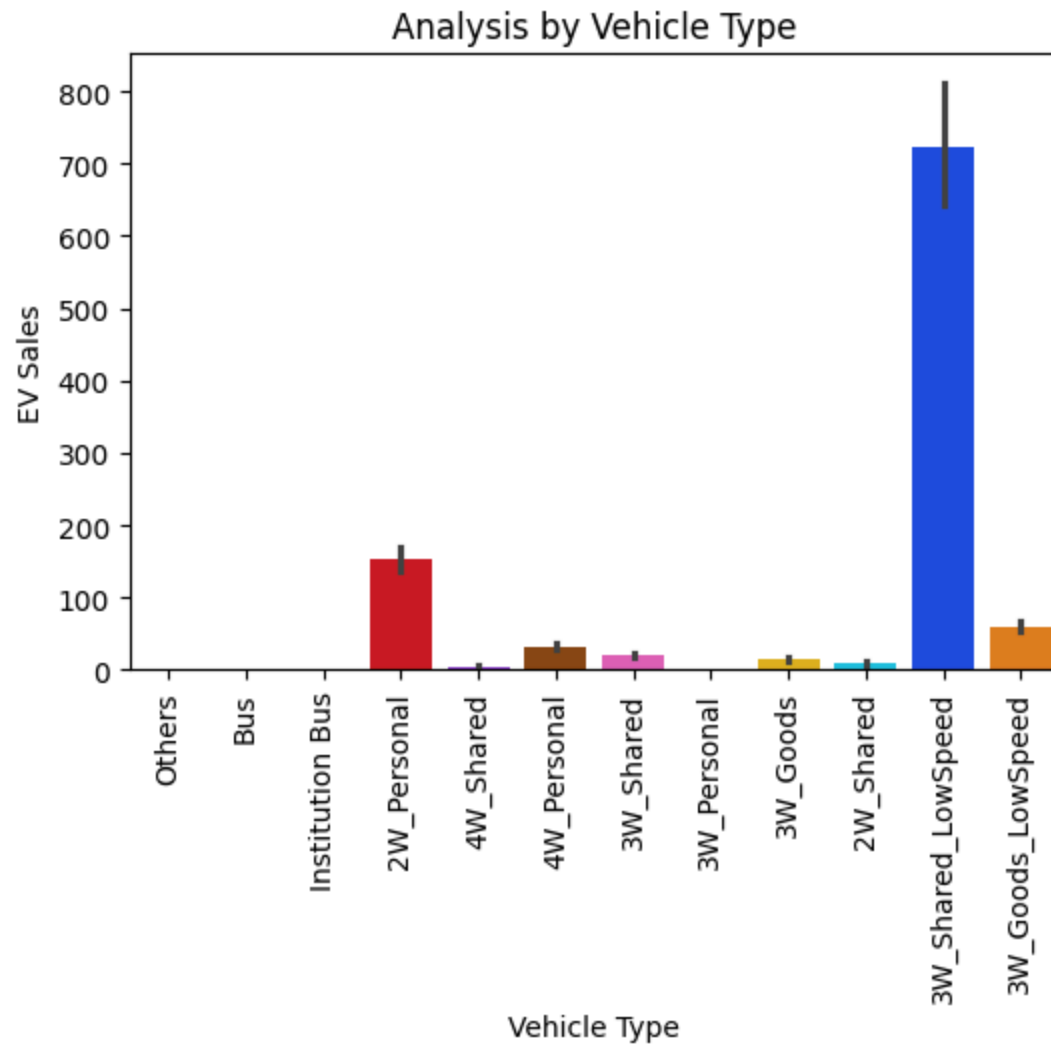


```
plt.figure(figsize=(15,4))
sns.barplot(x='Vehicle_Class', y='EV_Sales_Quantity', data=df,
hue='Vehicle_Class', palette='bright')
plt.title('Analysis by Vehicle Class')
```

```
plt.xlabel('Vehicle Class')  
plt.ylabel('EV Sales')  
plt.xticks(rotation=90);
```




```
plt.figure(figsize=(6,4))
sns.barplot(x='Vehicle_Type', y='EV_Sales_Quantity',data=df,
hue='Vehicle_Type', palette='bright')
plt.title('Analysis by Vehicle Type')
plt.xlabel('Vehicle Type')
plt.ylabel('EV Sales')
plt.xticks(rotation=90);
```



✓ Feature Engineering"

Create new features such as month and day from the Date column and encode categorical variables.

```
# Extract Month and Day from the Date column
```

```
df['Month'] = df['Date'].dt.month
```

```
df['Day'] = df['Date'].dt.day
```

```
# Encode categorical variables using one-hot encoding
```

```
df_encoded = pd.get_dummies(df, columns=['State','Vehicle_Class', 'Vehicle_Category', 'Vehicle_Type'],drop_first=True)
```

```
# Drop unnecessary columns like Date, Month_Name (if already extracted into numerical values)
```

```
df_encoded.drop(['Date', 'Month_Name'], axis=1, inplace=True)
```

```
df.head()
```



	Year	Month_Name	Date	State	Vehicle_Class	Vehicle_Category	Vehicle_Type	EV_Sales_Quantity	Month	Day
0	2014.0	jan	2014-01-01	Andhra Pradesh	ADAPTED VEHICLE	Others	Others	0.0	1	1
1	2014.0	jan	2014-01-01	Andhra Pradesh	AGRICULTURAL TRACTOR	Others	Others	0.0	1	1
2	2014.0	jan	2014-01-01	Andhra Pradesh	AMBULANCE	Others	Others	0.0	1	1
3	2014.0	jan	2014-01-01	Andhra Pradesh	ARTICULATED	Others	Others	0.0	1	1

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✓ Modeling

Use a regression model (e.g., Random Forest Regressor) to predict EV sales.

```
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error

# Split the data into features and target variable
X = df_encoded.drop('EV_Sales_Quantity', axis=1)
y = df_encoded['EV_Sales_Quantity']

# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Instantiate the model
model = RandomForestRegressor(n_estimators=100, random_state=42)

# Train the model
model.fit(X_train, y_train)

# Make predictions
y_pred = model.predict(X_test)
# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
print(f'Root Mean Squared Error: {rmse}')
```

➡ Root Mean Squared Error: 130.58175227721011

✓ Model Evaluation

```
# Plot actual vs predicted sales
plt.figure(figsize=(10, 6))
plt.scatter(y_test, y_pred)
plt.title('Actual vs Predicted EV Sales')
plt.xlabel('Actual EV Sales')
plt.ylabel('Predicted EV Sales')
plt.show()
```

