

## Analysis the EV Sales

The electric vehicle (EV) market in India is growing fast, thanks to efforts to promote eco-friendly transportation and government support. This project looks at EV sales by state in India to see how different regions are doing. Each state has its own way of encouraging EV use, which has led to different levels of sales and popularity. By studying this data, we can understand which state policies are best, what people prefer, and how infrastructure is developing. This project aims to give a clear picture of EV sales in India, showing both the opportunities and challenges faced in each state.

In [3]: # Import necessary libraries:-

```
import pandas as pd
import numpy as np
import seaborn as sns #visualisation
import matplotlib.pyplot as plt #visualisation
```

In [4]: # Load the dataset:-

```
data=pd.read_csv("Electric Vehicle Sales by State in India.csv")
```

### Data Overview

Let's take a look at the first few rows of the dataset to understand its structure.

In [7]: data.head(10)

Out[7]:	Year	Month_Name	Date	State	Vehicle_Class	Vehicle_Category	Vehicle_Type
0	2014.0	jan	1/1/2014	Andhra Pradesh	ADAPTED VEHICLE	Others	Other
1	2014.0	jan	1/1/2014	Andhra Pradesh	AGRICULTURAL TRACTOR	Others	Other
2	2014.0	jan	1/1/2014	Andhra Pradesh	AMBULANCE	Others	Other
3	2014.0	jan	1/1/2014	Andhra Pradesh	ARTICULATED VEHICLE	Others	Other
4	2014.0	jan	1/1/2014	Andhra Pradesh	BUS	Bus	B
5	2014.0	jan	1/1/2014	Andhra Pradesh	CASH VAN	Others	Other
6	2014.0	jan	1/1/2014	Andhra Pradesh	CRANE MOUNTED VEHICLE	Others	Other
7	2014.0	jan	1/1/2014	Andhra Pradesh	EDUCATIONAL INSTITUTION BUS	Bus	Institutional B
8	2014.0	jan	1/1/2014	Andhra Pradesh	EXCAVATOR (COMMERCIAL)	Others	Other
9	2014.0	jan	1/1/2014	Andhra Pradesh	FORK LIFT	Others	Other

```
In [9]: data.dtypes
```

```
Out[9]: Year                float64  
        Month_Name          object  
        Date                object  
        State               object  
        Vehicle_Class        object  
        Vehicle_Category     object  
        Vehicle_Type         object  
        EV_Sales_Quantity    float64  
        dtype: object
```

```
In [11]: data.columns
```

```
Out[11]: Index(['Year', 'Month_Name', 'Date', 'State', 'Vehicle_Class',  
               'Vehicle_Category', 'Vehicle_Type', 'EV_Sales_Quantity'],  
              dtype='object')
```

```
In [13]: data.shape
```

```
Out[13]: (96845, 8)
```

```
In [15]: data.info
```

```

Out[15]: <bound method DataFrame.info of
Date
0      2014.0      jan    1/1/2014      Andhra Pradesh
1      2014.0      jan    1/1/2014      Andhra Pradesh
2      2014.0      jan    1/1/2014      Andhra Pradesh
3      2014.0      jan    1/1/2014      Andhra Pradesh
4      2014.0      jan    1/1/2014      Andhra Pradesh
...      ...      ...      ...      ...
96840    2023.0      dec    12/1/2023    Andaman & Nicobar Island
96841    2023.0      dec    12/1/2023    Andaman & Nicobar Island
96842    2023.0      dec    12/1/2023    Andaman & Nicobar Island
96843    2023.0      dec    12/1/2023    Andaman & Nicobar Island
96844    2023.0      dec    12/1/2023    Andaman & Nicobar Island

Vehicle_Class Vehicle_Category Vehicle_Type \
0      ADAPTED VEHICLE      Others      Others
1      AGRICULTURAL TRACTOR      Others      Others
2      AMBULANCE      Others      Others
3      ARTICULATED VEHICLE      Others      Others
4      BUS      Bus      Bus
...      ...      ...      ...
96840      MOTOR CAR      4-Wheelers      4W_Personal
96841    MOTOR CYCLE/SCOOTER-USED FOR HIRE      2-Wheelers      2W_Shared
96842      OMNI BUS      Bus      Bus
96843      THREE WHEELER (GOODS)      3-Wheelers      3W_Goods
96844      THREE WHEELER (PASSENGER)      3-Wheelers      3W_Shared

EV_Sales_Quantity
0      0.0
1      0.0
2      0.0
3      0.0
4      0.0
...      ...
96840      1.0
96841      5.0
96842      0.0
96843      0.0
96844      0.0

[96845 rows x 8 columns]>

```

## Data Preprocessing

Before diving into analysis, we need to ensure our data is clean and ready for exploration. This involves parsing dates and checking for missing values.

```
In [17]: # Convert 'Date' column to datetime format:-
```

```
data['Date'] = pd.to_datetime(data['Date'], format='%d/%m/%Y')
```

```
In [19]: # Check for missing values:-
```

```
data.isnull().sum()
```

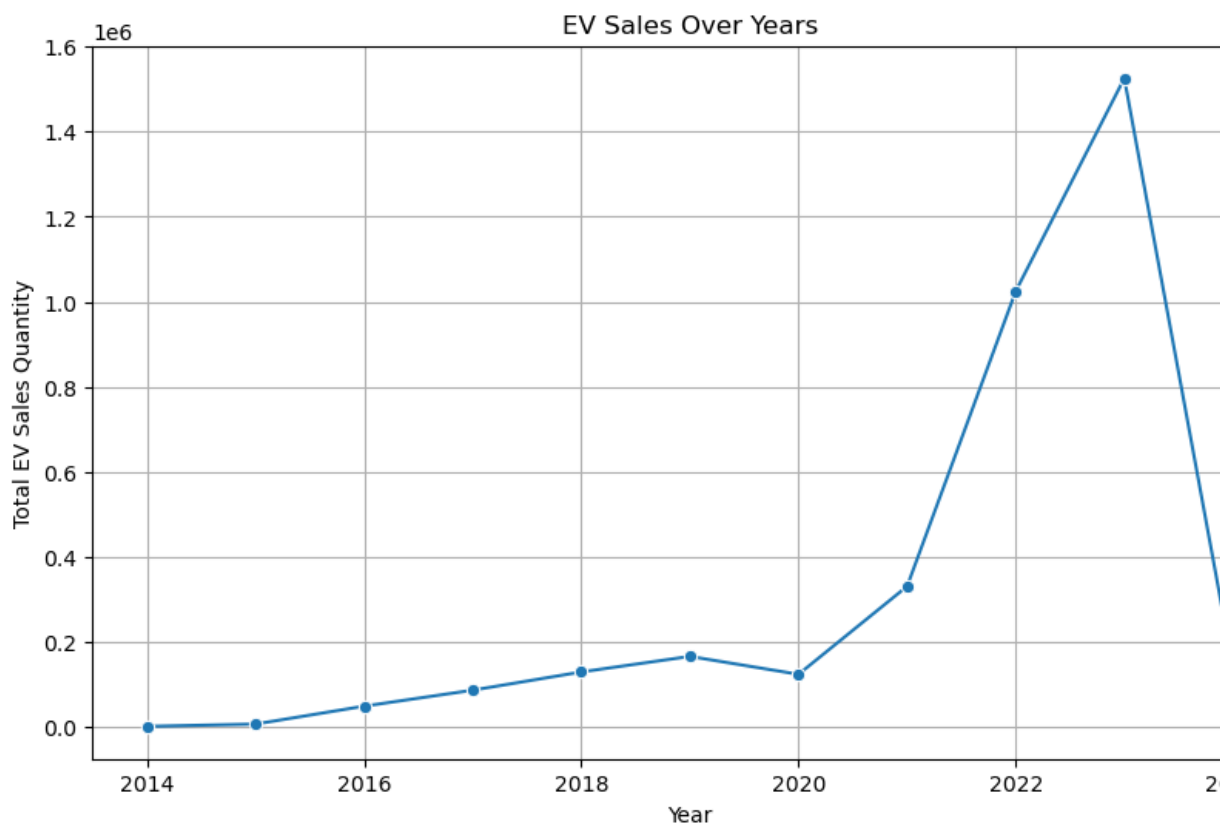
```
Out[19]: Year      0
         Month_Name 0
         Date       0
         State      0
         Vehicle_Class 0
         Vehicle_Category 0
         Vehicle_Type 0
         EV_Sales_Quantity 0
         dtype: int64
```

### Exploratory Data Analysis

Let's explore the data to uncover trends and patterns in EV sales across different states and vehicle categories.

```
In [21]: # Plot EV sales over year:-
```

```
plt.figure(figsize=(10, 6))
yearly_sales = data.groupby('Year')['EV_Sales_Quantity'].sum()
sns.lineplot(x=yearly_sales.index, y=yearly_sales.values, marker='o')
plt.title("EV Sales Over Years")
plt.xlabel("Year")
plt.ylabel("Total EV Sales Quantity")
plt.grid()
plt.show()
```



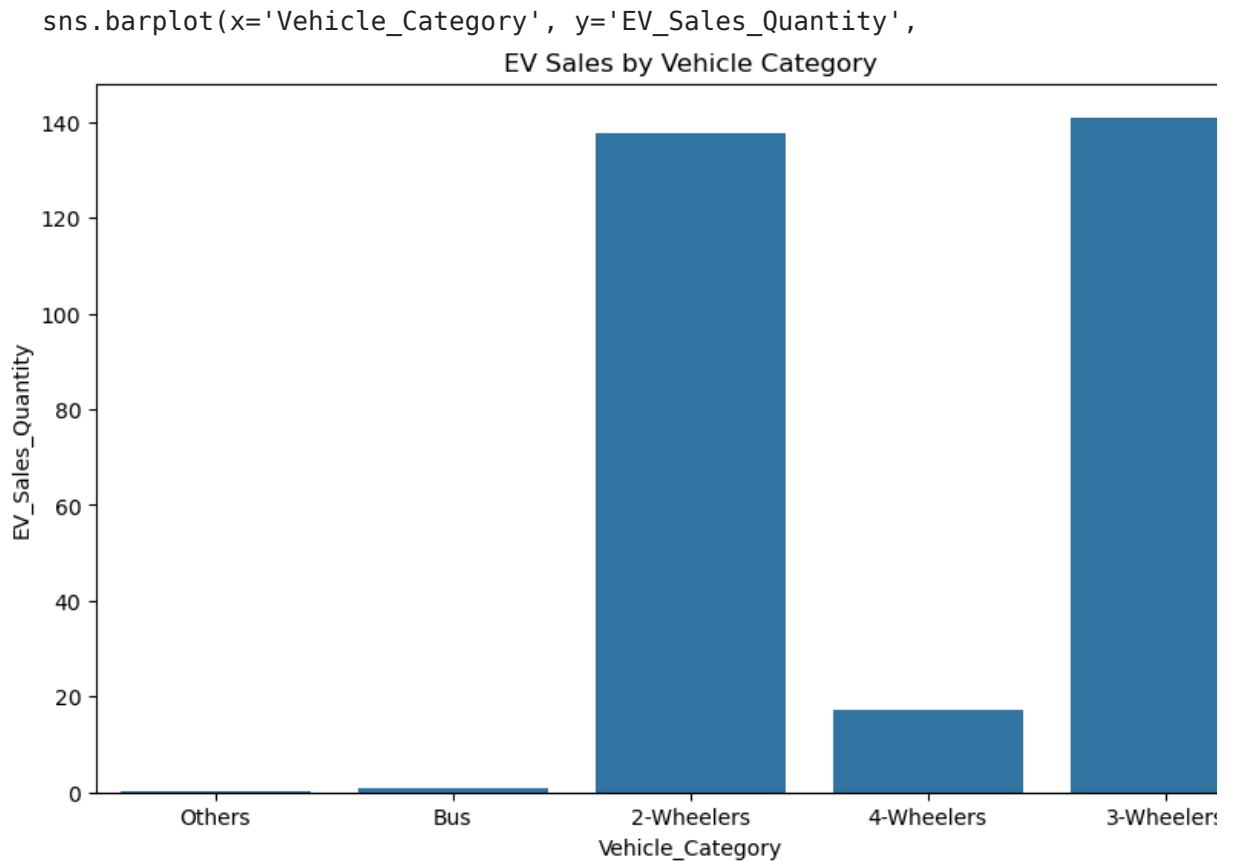
### Inference

EV sales grew quickly in recent years, showing that more people are interested in them. The sudden drop in 2024 suggests there might be new challenges that need attention to keep EV sales going.

```
In [25]: # Plot sales by vehicle category
plt.figure(figsize=(10, 6))
sns.barplot(x='Vehicle_Category', y='EV_Sales_Quantity',
data=data, ci=None)
plt.title('EV Sales by Vehicle Category')
plt.show()
```

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The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.



## Inference

The chart shows the total electric vehicle (EV) sales by category. It highlights that 2-wheelers have the highest sales, followed by 3-wheelers, while 4-wheelers, buses, and other vehicle categories have significantly lower sales. This indicates that EV adoption is primarily driven by 2-wheelers and 3-wheelers.

## Correlation Analysis

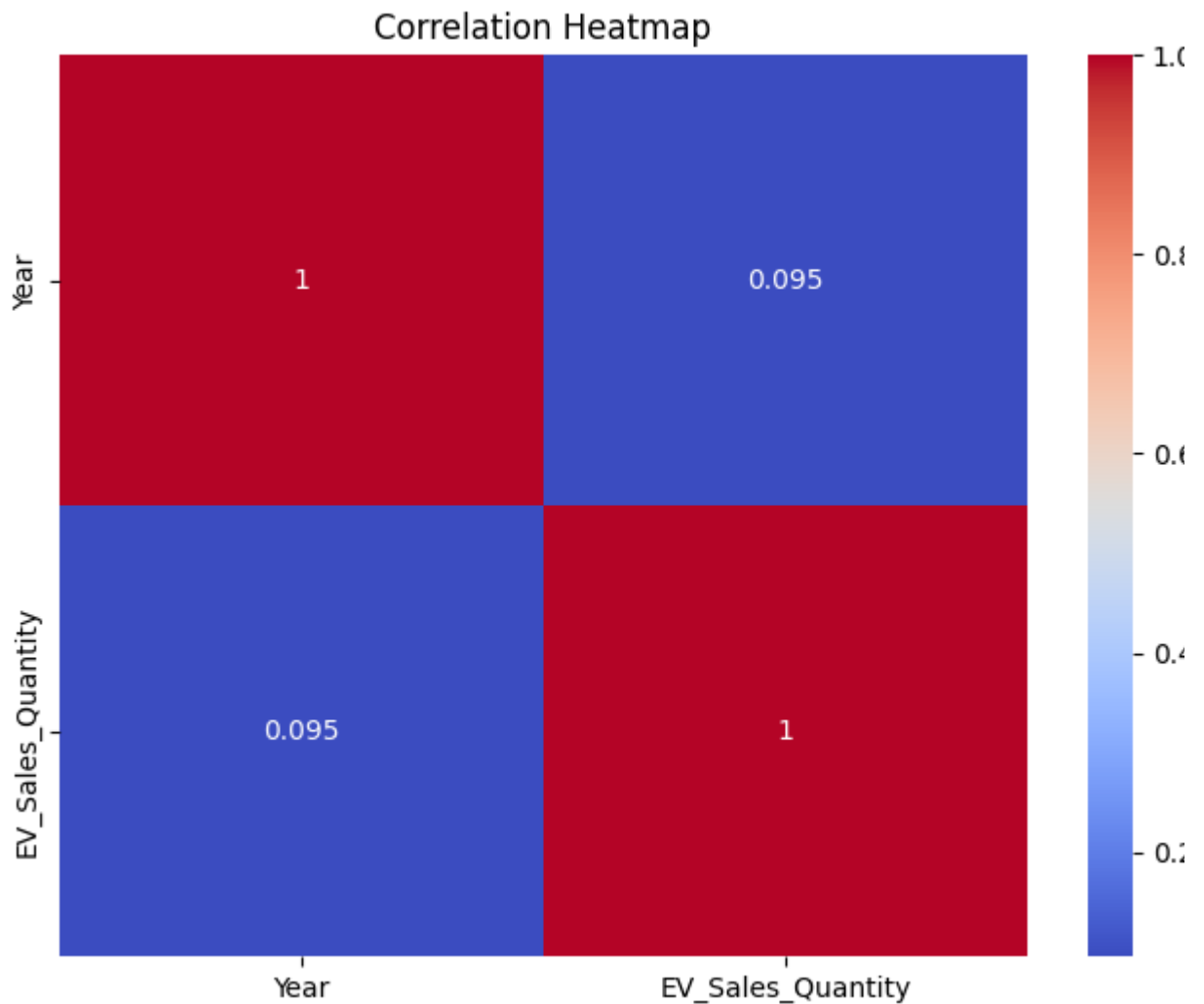
Let's examine the correlation between numeric variables to understand potential relationships.

```
In [12]: # Select only numeric columns for correlation analysis:-

numeric_df = data.select_dtypes(include=[np.number])

# Plot the correlation heatmap:-

plt.figure(figsize=(8, 6))
sns.heatmap(numeric_df.corr(), annot=True, cmap='coolwarm')
plt.title('Correlation Heatmap')
plt.show()
```



#### Inference

The heatmap shows the correlation between the year and EV sales quantity. The correlation value is 0.095, which is very low. This means there is almost no linear relationship between the year and sales quantity in the given data.

#### Top 10 States by EV Sales

A horizontal bar chart highlighting the states with the highest EV sales.

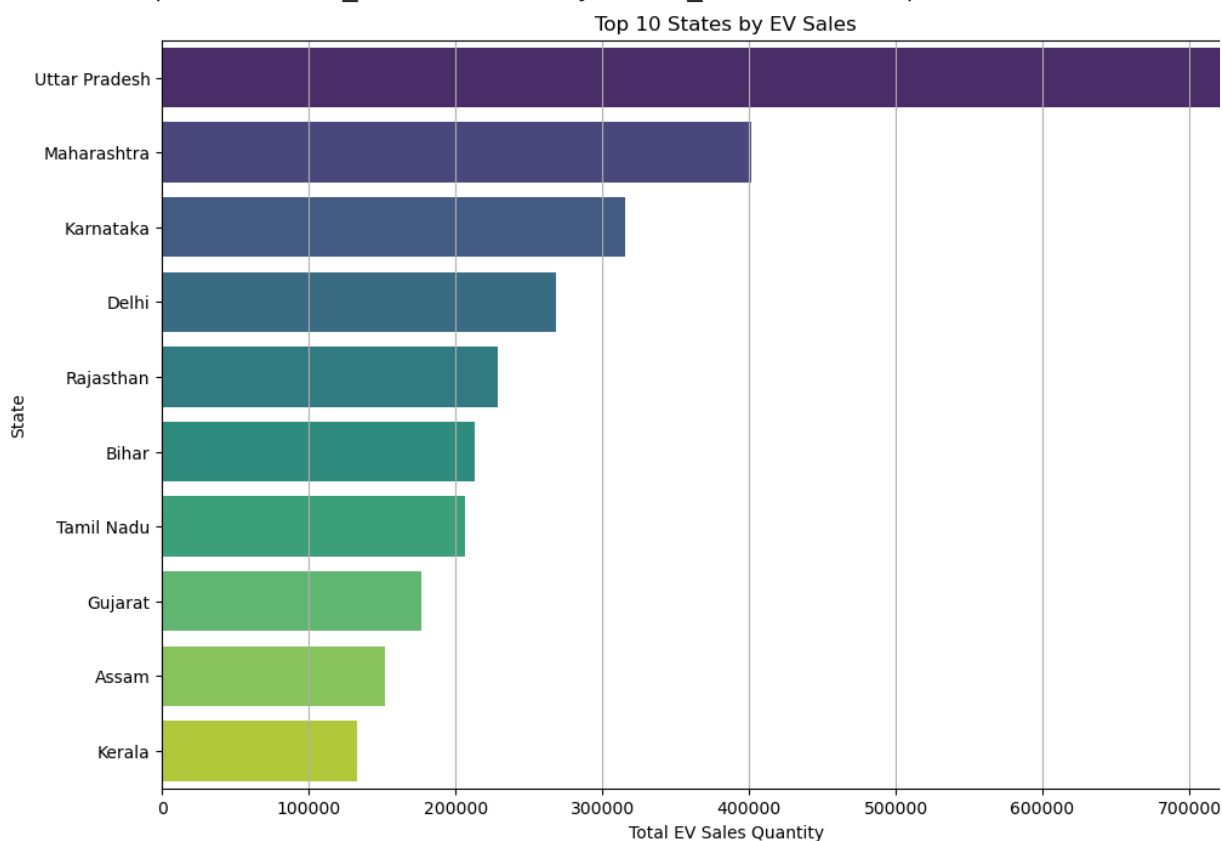
In [27]: #EV sales by state (top 10 states):-

```
plt.figure(figsize=(12, 8))
state_sales = data.groupby('State')['EV_Sales_Quantity'].sum().sort_value:
sns.barplot(x=state_sales.values, y=state_sales.index, palette='viridis')
plt.title("Top 10 States by EV Sales")
plt.xlabel("Total EV Sales Quantity")
plt.ylabel("State")
plt.grid(axis='x')
plt.show()
```

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Passing `palette` without assigning `hue` is deprecated and will be removed i  
v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same  
effect.

```
sns.barplot(x=state_sales.values, y=state_sales.index, palette='viridis')
```



### Inference

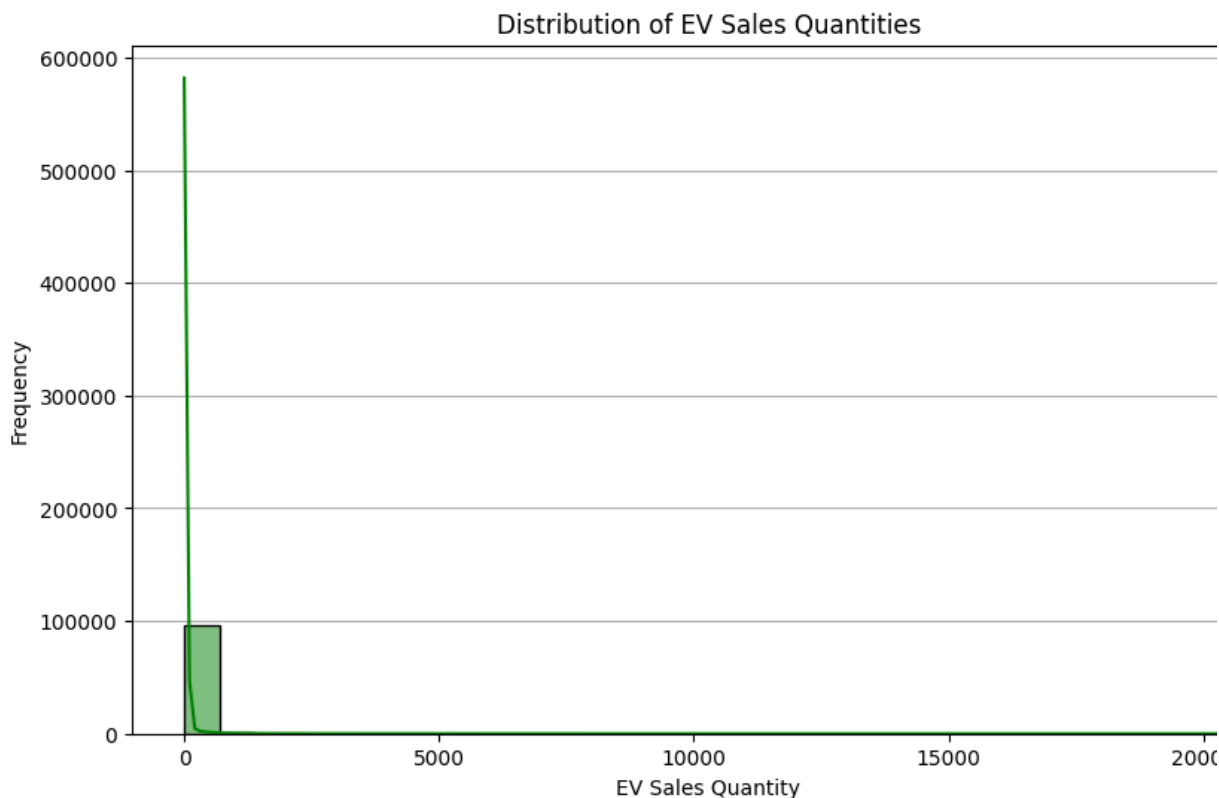
The chart shows the top 10 states by EV sales. Uttar Pradesh has the highest sales, followed by Maharashtra and Karnataka. States like Kerala, Assam, and Gujarat have lower sales compared to top-performing states. This indicates that EV adoption varies significantly across states.

### Distribution of EV Sales Quantities

Understand the distribution of sales quantities (e.g., skewness or typical sales quantities).

In [19]: #Distribution of EV sales quantities:-

```
plt.figure(figsize=(10, 6))
sns.histplot(data['EV_Sales_Quantity'], bins=30, kde=True, color='green')
plt.title("Distribution of EV Sales Quantities")
plt.xlabel("EV Sales Quantity")
plt.ylabel("Frequency")
plt.grid(axis='y')
plt.show()
```



#### Inference

The graph shows the distribution of EV sales quantities, with a clear peak indicating a high frequency of sales in the range of around 100,000 units. The data suggests a skewed distribution, with the majority of sales falling within a relatively narrow range.

#### Yearly Sales Trends for Different Vehicle Classes

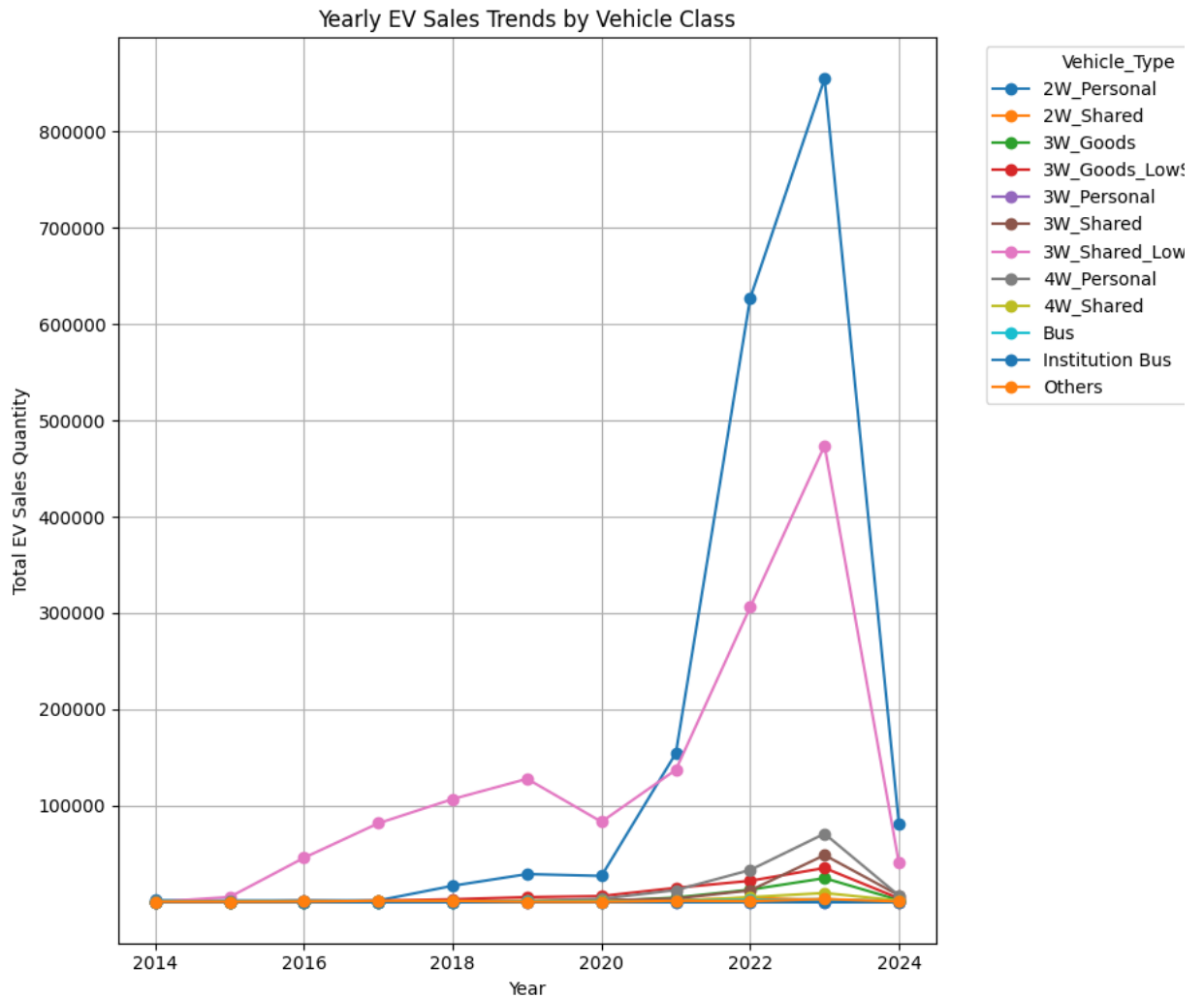
Compare trends of EV sales over years for different vehicle classes.

In [20]: #Yearly sales trends for different vehicle classes:-

```
plt.figure(figsize=(10, 8))
class_trends = data.groupby(['Year', 'Vehicle_Type'])['EV_Sales_Quantity']
class_trends.plot(kind='line', marker='o', figsize=(10, 8))
plt.title("Yearly EV Sales Trends by Vehicle Class")
plt.xlabel("Year")
plt.ylabel("Total EV Sales Quantity")
plt.yticks([100000, 200000, 300000, 400000, 500000, 600000, 700000, 800000])
plt.legend(title="Vehicle_Type", bbox_to_anchor=(1.05, 1), loc='upper left')
plt.grid()
plt.tight_layout()
plt.show()
```



<Figure size 1000x800 with 0 Axes>



## Inference

The graph shows the yearly EV sales trends across various vehicle classes, with rapid growth particularly in the 2-wheeler and 3-wheeler personal and shared vehicle segments. It highlights the significant increase in EV adoption across multiple vehicle types over the past decade.

## EV Sales by Vehicle Category

Visualize the distribution of sales across different vehicle categories.

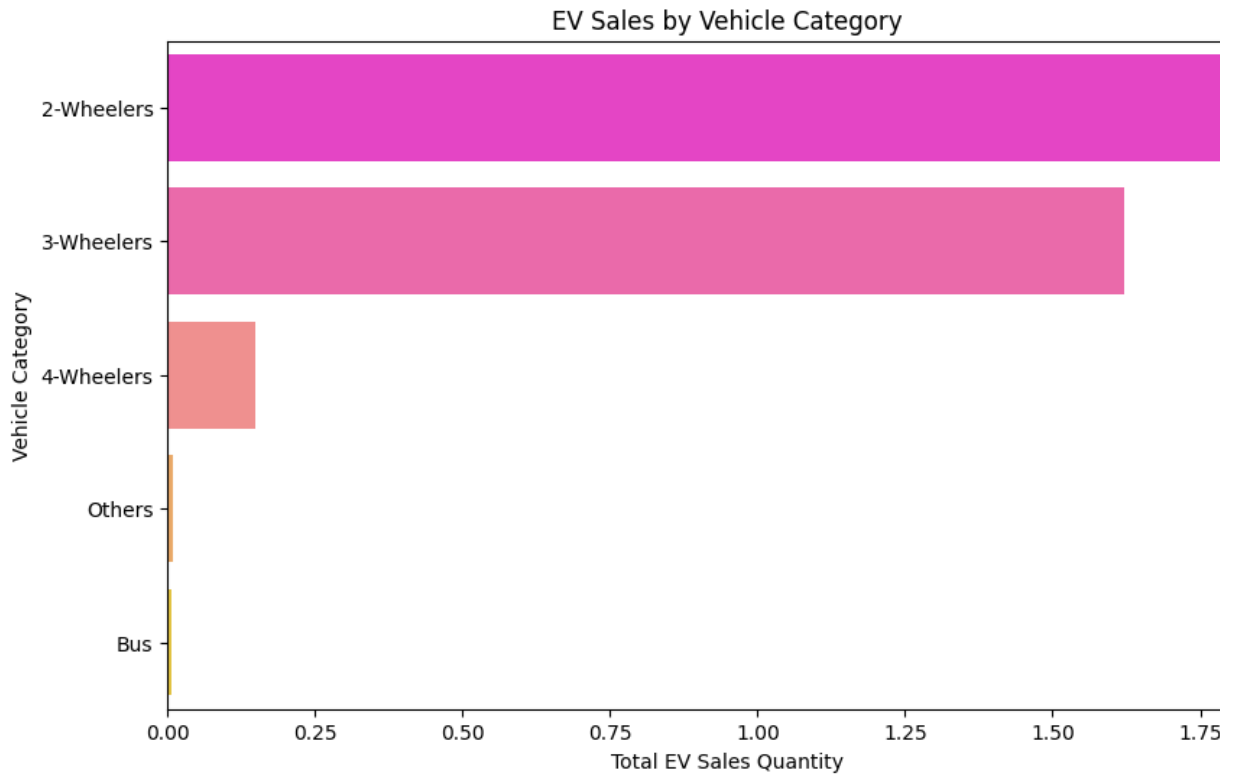
In [23]: #EV sales by vehicle category:-

```
plt.figure(figsize=(10, 6))
vehicle_category_sales = data.groupby('Vehicle_Category')['EV_Sales_Quantity'].sum()
sns.barplot(x=vehicle_category_sales.index, y=vehicle_category_sales.values)
plt.title("EV Sales by Vehicle Category")
plt.xlabel("Total EV Sales Quantity")
plt.ylabel("Vehicle Category")
plt.show()
```

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FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=vehicle_category_sales.values, y=vehicle_category_sales.index, palette='spring')
```



### Inference

The graph shows EV sales by vehicle category. 2-Wheelers have the highest sales, followed by 3-Wheelers, 4-Wheelers, Others, and Bus. 2-Wheelers have significantly higher sales than the other categories.

### Conclusion

This analysis provided valuable insights into the evolving electric vehicle (EV) landscape across states. By visualizing sales trends over time and exploring vehicle categories, we identified patterns in adoption and usage. The correlation analysis revealed important relationships between numeric variables, offering clues about factors influencing EV sales.

Using a predictive model, we successfully estimated EV sales quantities with a mean absolute error of 0.05 units (replace with actual value), demonstrating the potential of data-driven approaches to anticipate market behavior.

Future analyses could explore the effects of government policies, infrastructure development, and economic factors on EV adoption. Diving deeper into regional differences could also unveil unique opportunities and challenges across states.