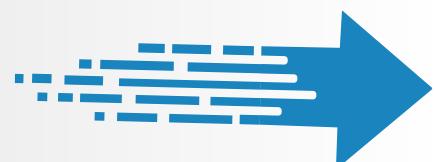


# ELECTRIC VEHICLES

## SALE PROJECT

Drive into the Future with Our  
Limited-Time Electric  
Vehicle

PROJECT



# GO GREEN SAVE GREEN

We're leading the charge towards a cleaner, greener future with our comprehensive electric vehicle (EV) and charging station services



Timmerman Industries  
123 Anywhere St., Any City

## Charging Station

Charging Location  
123 Anywhere St., Any City



# Electric vehicle market sales

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

ev_data = pd.read_csv('Electric_Vehicle_Population_Data.csv')

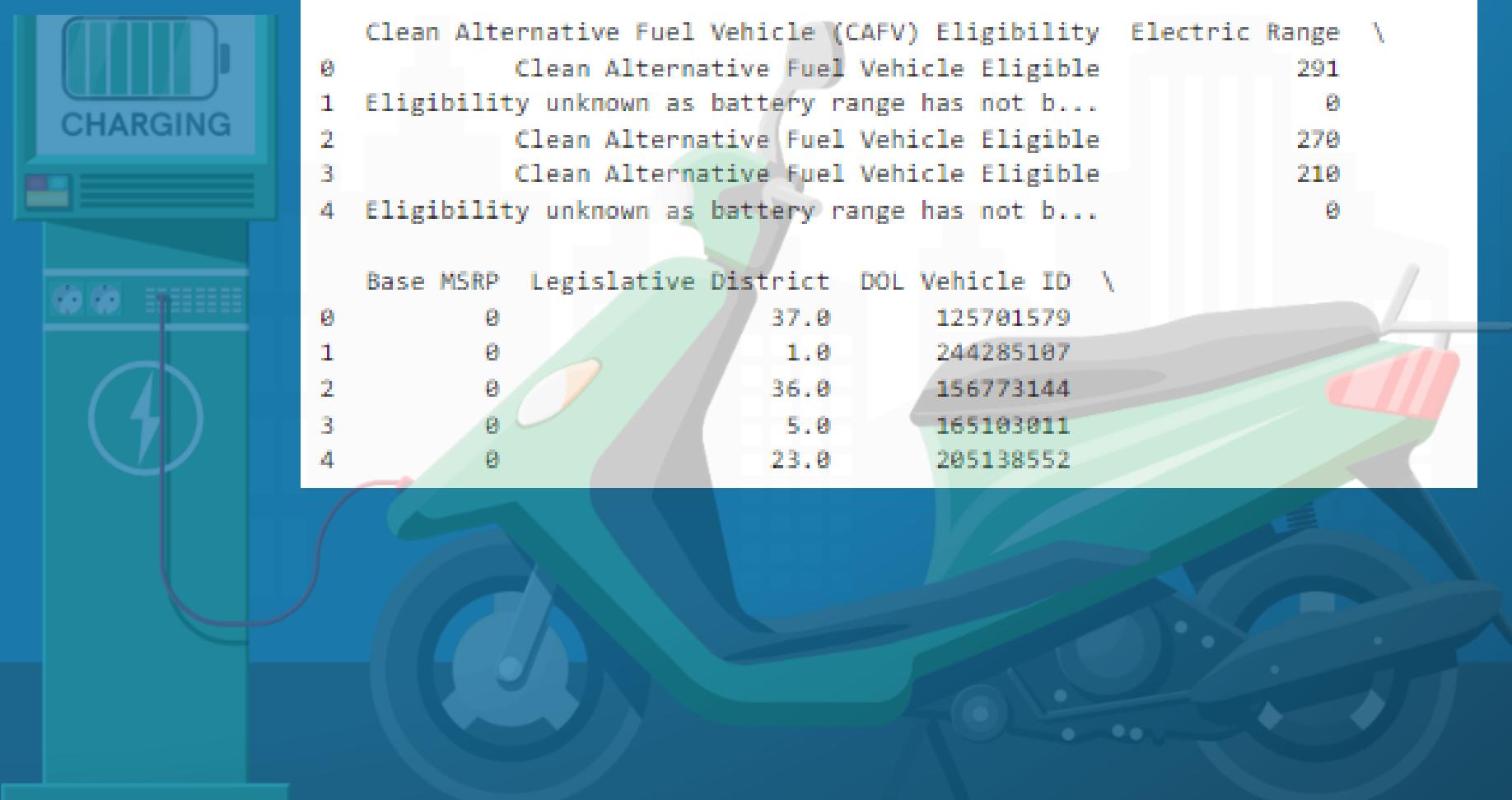
print(ev_data.head())
```

```
      VIN (1-10)    County     City State Postal Code Model Year Make \
0  5YJYGDEE1L      King   Seattle   WA  98122.0  2020  TESLA
1  7SAYGDEE9P  Snohomish  Bothell   WA  98021.0  2023  TESLA
2  5YJSA1E4XX      King   Seattle   WA  98109.0  2019  TESLA
3  5YJSA1E27G      King  Issaquah   WA  98027.0  2016  TESLA
4  5YJYGDEE5M    Kitsap  Suquamish   WA  98392.0  2021  TESLA
```

```
          Model           Electric Vehicle Type \
0  MODEL Y  Battery Electric Vehicle (BEV)
1  MODEL Y  Battery Electric Vehicle (BEV)
2  MODEL S  Battery Electric Vehicle (BEV)
3  MODEL S  Battery Electric Vehicle (BEV)
4  MODEL Y  Battery Electric Vehicle (BEV)
```

```
          Clean Alternative Fuel Vehicle (CAFV) Eligibility  Electric Range \
0                  Clean Alternative Fuel Vehicle Eligible                291
1  Eligibility unknown as battery range has not b...                   0
2                  Clean Alternative Fuel Vehicle Eligible                270
3                  Clean Alternative Fuel Vehicle Eligible                210
4  Eligibility unknown as battery range has not b...                   0
```

```
      Base MSRP  Legislative District  DOL Vehicle ID \
0        0.0            37.0       125701579
1        0.0            1.0       244285187
2        0.0            36.0      156773144
3        0.0            5.0       165103011
4        0.0            23.0      205138552
```



	VIN (1-10)	County	City	State	Postal Code	Model Year	Make	\
0	5YJYGDEE1L	King	Seattle	WA	98122-0	2020	TESLA	
1	7SAYGDEE9P	Snohomish	Bothell	WA	98021-0	2023	TESLA	
2	5YJSA1E4XK	King	Seattle	WA	98109-0	2019	TESLA	
3	5YJSA1E27G	King	Issaquah	WA	98027-0	2016	TESLA	
4	5YJYGDEE5M	Kitsap	Suquamish	WA	98392-0	2021	TESLA	
	Model	Electric Vehicle Type						\
0	MODEL Y	Battery Electric Vehicle (BEV)						
1	MODEL Y	Battery Electric Vehicle (BEV)						
2	MODEL S	Battery Electric Vehicle (BEV)						
3	MODEL S	Battery Electric Vehicle (BEV)						
4	MODEL Y	Battery Electric Vehicle (BEV)						
	Clean Alternative Fuel Vehicle (CAFV) Eligibility	Electric Range						\
0	Clean Alternative Fuel Vehicle Eligible	291						
1	Eligibility unknown as battery range has not b...	0						
2	Clean Alternative Fuel Vehicle Eligible	270						
3	Clean Alternative Fuel Vehicle Eligible	210						
4	Eligibility unknown as battery range has not b...	0						
	Base MSRP	Legislative District	DOL	Vehicle ID				\
0	0	37.0		125781579				
1	0	1.0		244285107				
2	0	36.0		156773144				
3	0	5.0		165103011				
4	0	23.0		205138552				
	Vehicle Location							\
0	POINT (-122.30839 47.610365)							
1	POINT (-122.179458 47.882589)							
2	POINT (-122.34848 47.632405)							
3	POINT (-122.03646 47.534065)							
4	POINT (-122.55717 47.733415)							
	Electric Utility	2020 Census Tract						\
0	CITY OF SEATTLE - (WA) CITY OF TACOMA - (WA)	5.303301e+10						
1	PUGET SOUND ENERGY INC	5.306105e+10						
2	CITY OF SEATTLE - (WA) CITY OF TACOMA - (WA)	5.303301e+10						
3	PUGET SOUND ENERGY INC  CITY OF TACOMA - (WA)	5.303303e+10						
4	PUGET SOUND ENERGY INC	5.303594e+10						

```
ev_data.info()
```

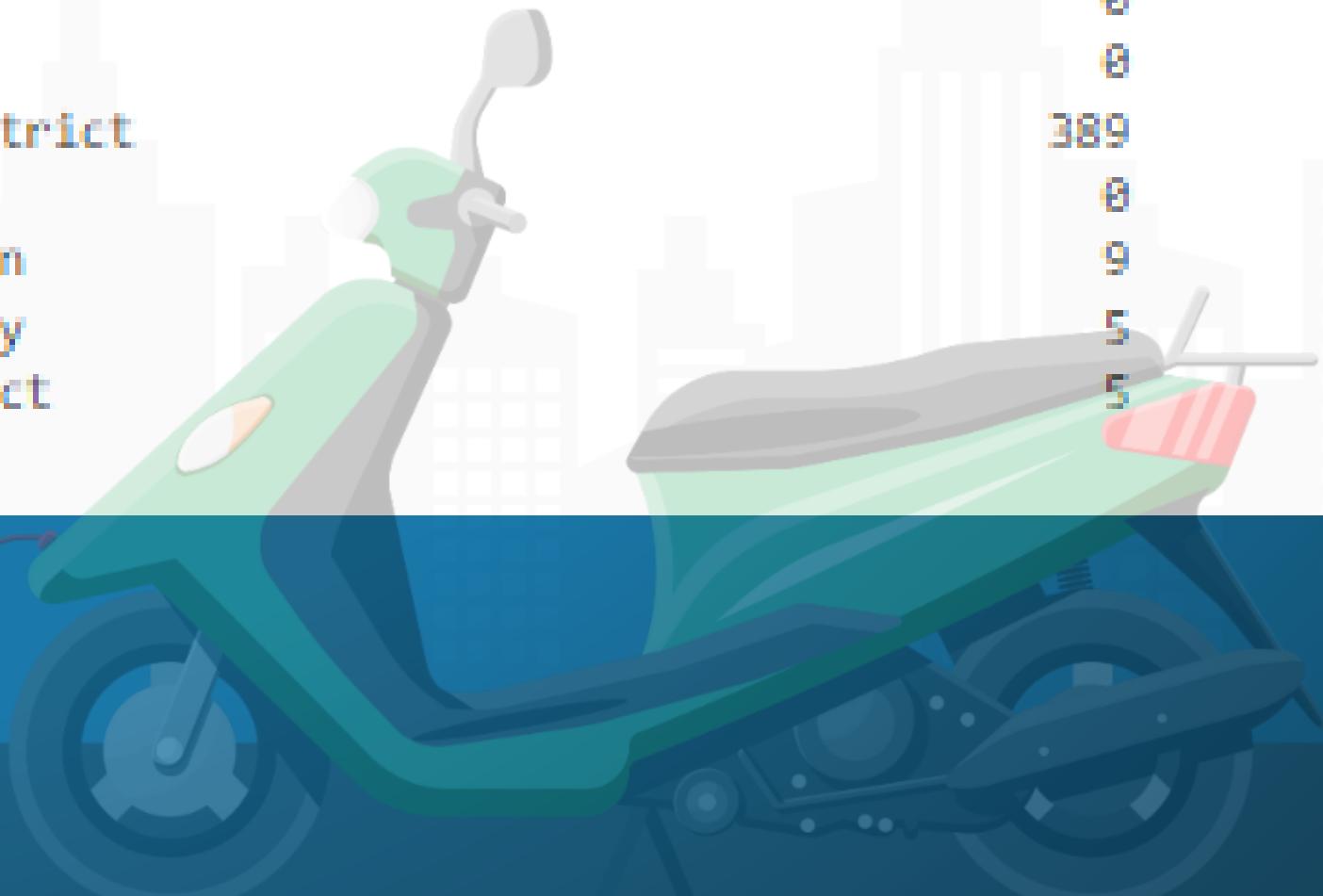
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 177866 entries, 0 to 177865
Data columns (total 17 columns):
```

#	Column	Non-Null Count	Dtype
0	VIN (1-10)	177866 non-null	object
1	County	177861 non-null	object
2	City	177861 non-null	object
3	State	177866 non-null	object
4	Postal Code	177861 non-null	float64
5	Model Year	177866 non-null	int64
6	Make	177866 non-null	object
7	Model	177866 non-null	object
8	Electric Vehicle Type	177866 non-null	object
9	Clean Alternative Fuel Vehicle (CAFV) Eligibility	177866 non-null	object
10	Electric Range	177866 non-null	int64
11	Base MSRP	177866 non-null	int64
12	Legislative District	177477 non-null	float64
13	DOL Vehicle ID	177866 non-null	int64
14	Vehicle Location	177857 non-null	object
15	Electric Utility	177861 non-null	object
16	2020 Census Tract	177861 non-null	float64

dtypes: float64(3), int64(4), object(10)  
memory usage: 23.1+ MB

```
ev_data.isnull().sum()
```

VIN (1-10)	0
County	5
City	5
State	0
Postal Code	5
Model Year	0
Make	0
Model	0
Electric Vehicle Type	0
Clean Alternative Fuel Vehicle (CAFV) Eligibility	0
Electric Range	0
Base MSRP	0
Legislative District	389
DOT Vehicle ID	0
Vehicle Location	9
Electric Utility	5
2020 Census Tract	5
dtype:	int64



```
ev_data = ev_data.dropna()
```

# For the task of market size of electric vehicles analysis, we can explore the following areas:

- # EV Adoption Over Time: Analyze the growth of the EV population by model year.
  - # Geographical Distribution: Understand where EVs are most commonly registered (e.g., by county).
  - # EV Types: Breakdown of the dataset by electric vehicle type (BEV, etc.).
  - # Make and Model Popularity: Identify the most popular makes and models among the registered EVs.
  - # Electric Range Analysis: Analyze the electric range of vehicles to see how EV technology is progressing.
  - # Estimated Growth in Market Size: Analyze and find the estimated growth in the market size of electric vehicles over time.
- # Let's start with analyzing the EV Adoption Over Time by visualizing the number of EVs registered over the years:
- # It will give us an insight into how the EV population has grown over the years:

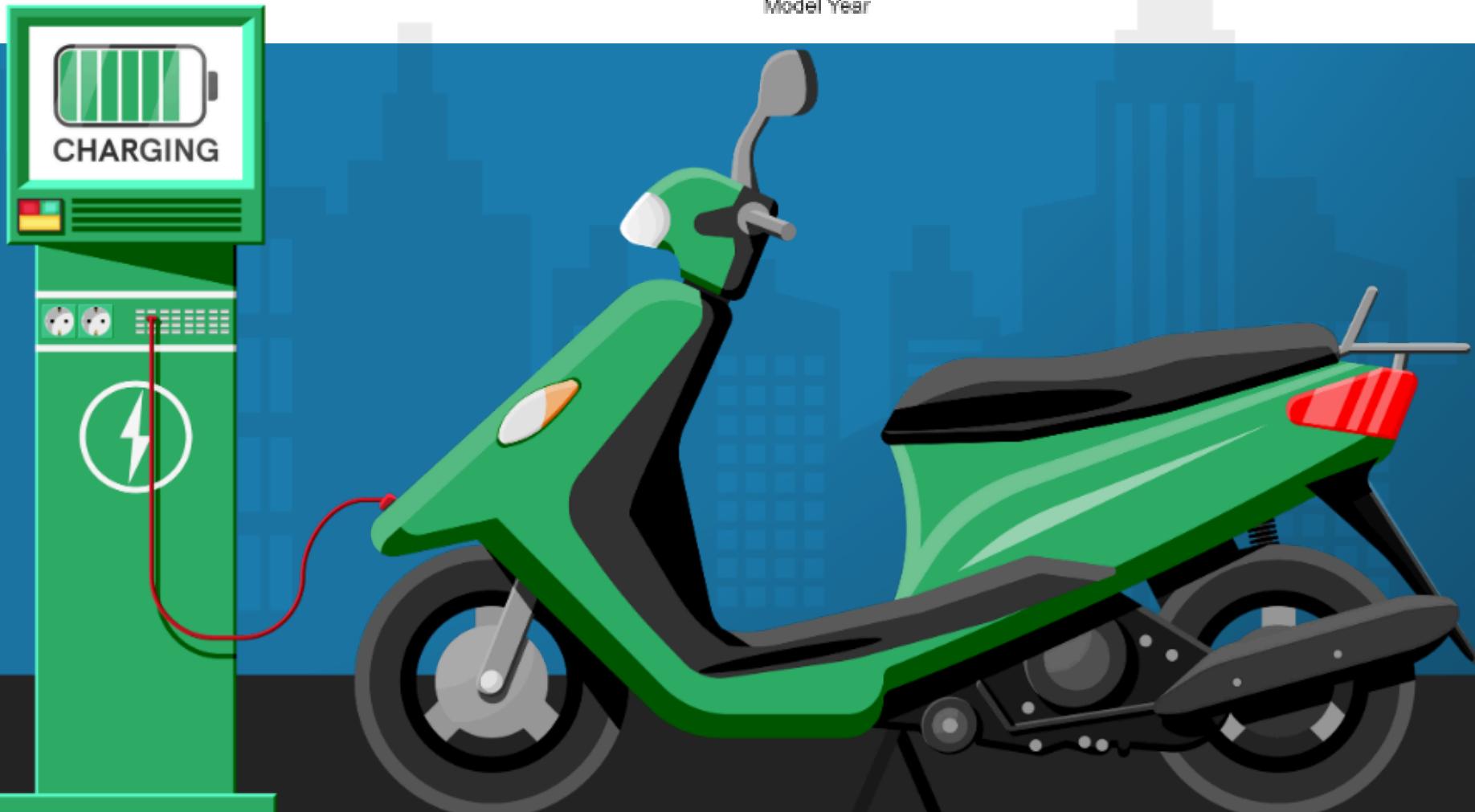
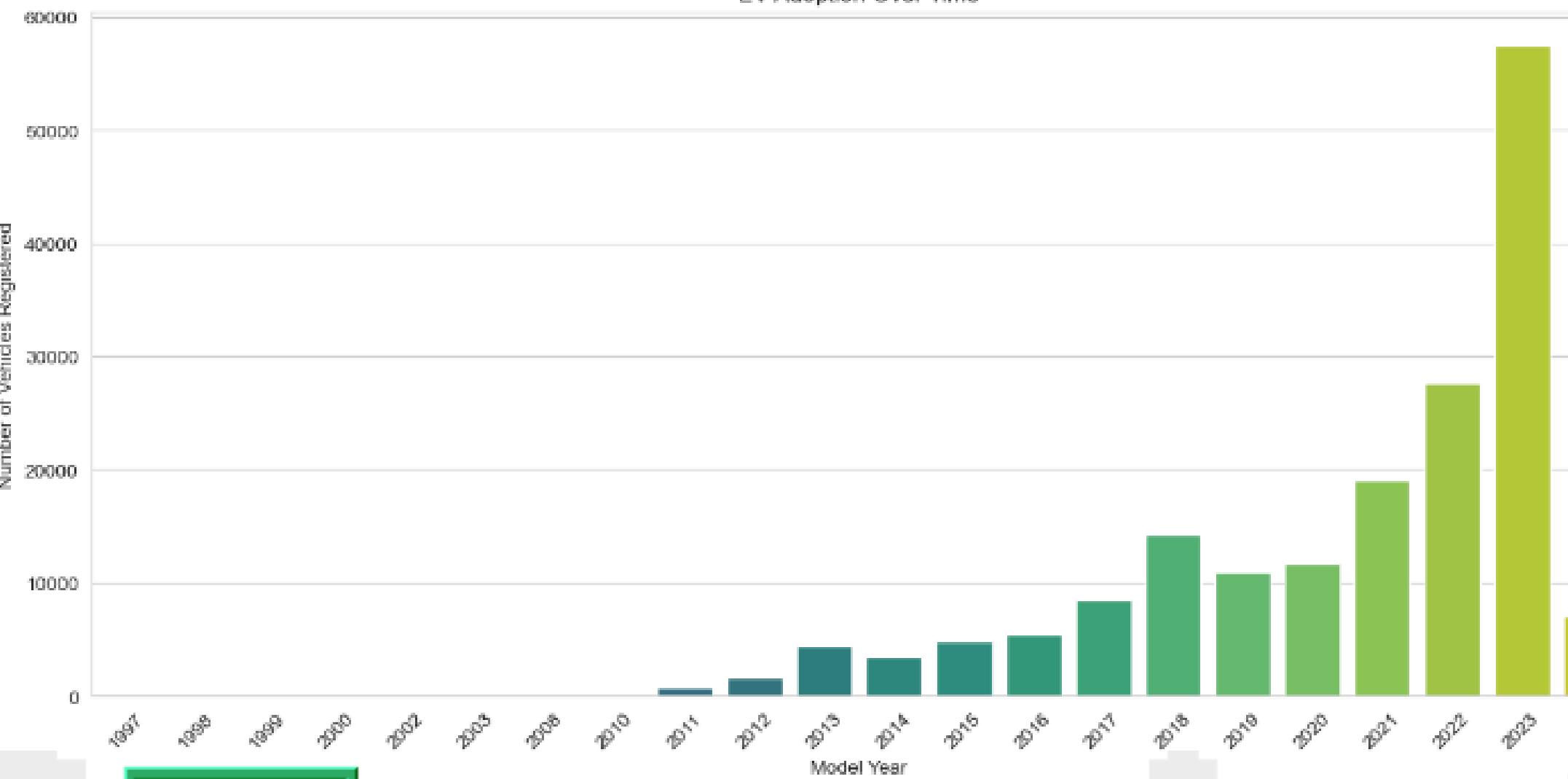


```
import seaborn as sns
sns.set_style("whitegrid")

# EV Adoption Over Time
plt.figure(figsize=(12, 6))
ev_adoption_by_year = ev_data["Model Year"].value_counts().sort_index()
sns.barplot(x=ev_adoption_by_year.index, y=ev_adoption_by_year.values)
plt.title('EV Adoption Over Time')
plt.xlabel('Model Year')
plt.ylabel('Number of Vehicles Registered')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



EV Adoption Over Time



```

# geographical distribution at county Level

ev_county_distribution = ev_data['County'].value_counts()
top_counties = ev_county_distribution.head(3).index

# filtering the dataset for these top counties
top_counties_data = ev_data[ev_data['County'].isin(top_counties)]

# analyzing the distribution of EVs within the cities of these top counties
ev_city_distribution_top_counties = top_counties_data.groupby(['County', 'City']).size().sort_values(ascending=False).reset_index(name='Number of Vehicles')

# visualize the top 10 cities across these counties
top_cities = ev_city_distribution_top_counties.head(10)

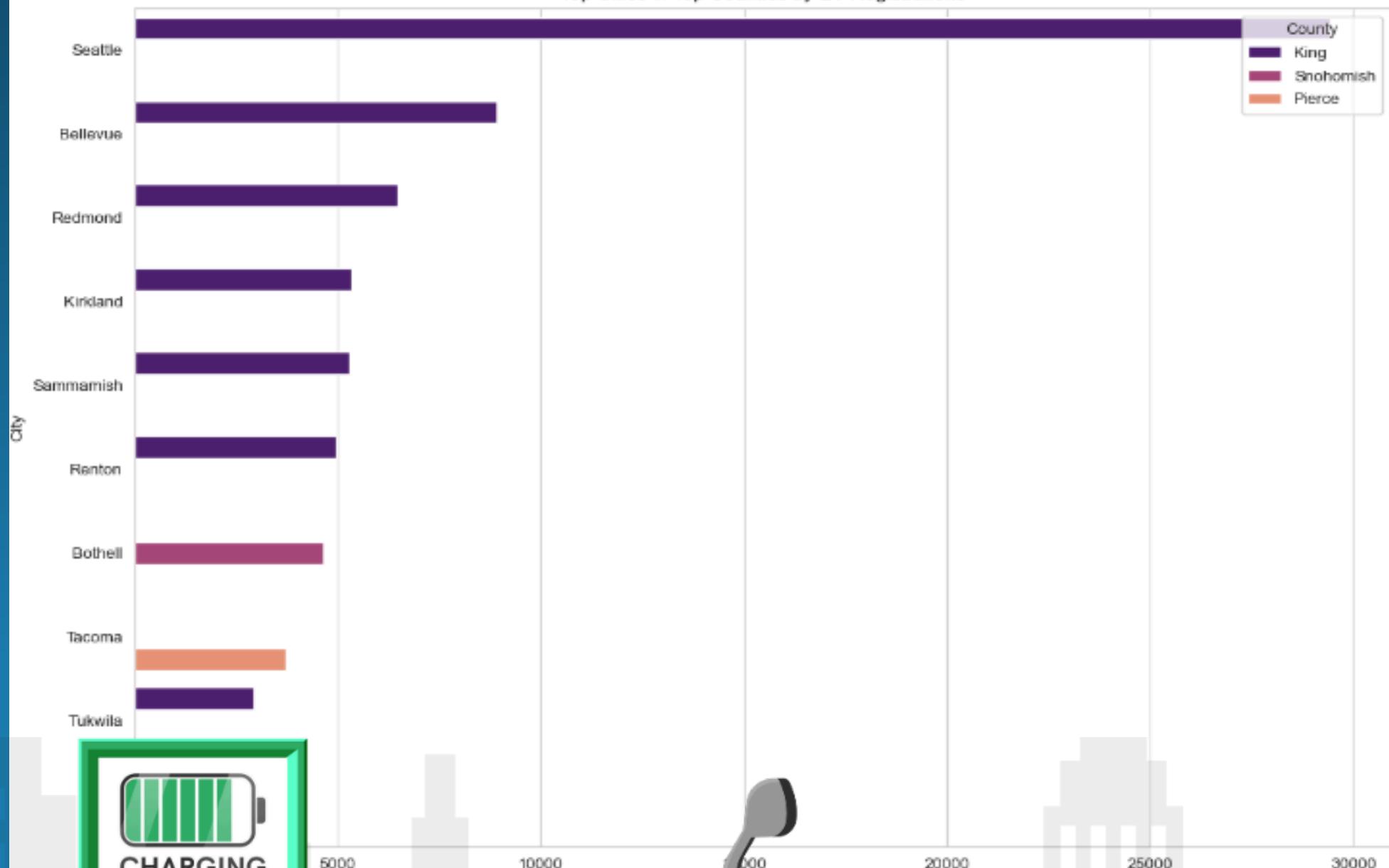
plt.figure(figsize=(12, 8))
sns.barplot(x='Number of Vehicles', y='City', hue='County', data=top_cities, palette="magma")
plt.title('Top Cities in Top Counties by EV Registrations')
plt.xlabel('Number of Vehicles Registered')
plt.ylabel('City')
plt.legend(title='County')
plt.tight_layout()

plt.show()

```



Top Cities in Top Counties by EV Registrations



```
# analyzing the distribution of electric vehicle Types
ev_type_distribution = ev_data['Electric Vehicle Type'].value_counts()

plt.figure(figsize=(10, 6))
sns.barplot(x=ev_type_distribution.values, y=ev_type_distribution.index, palette="rocket")
plt.title('Distribution of Electric Vehicle Types')
plt.xlabel('Number of Vehicles Registered')
plt.ylabel('Electric Vehicle Type')
plt.tight_layout()

plt.show()
```

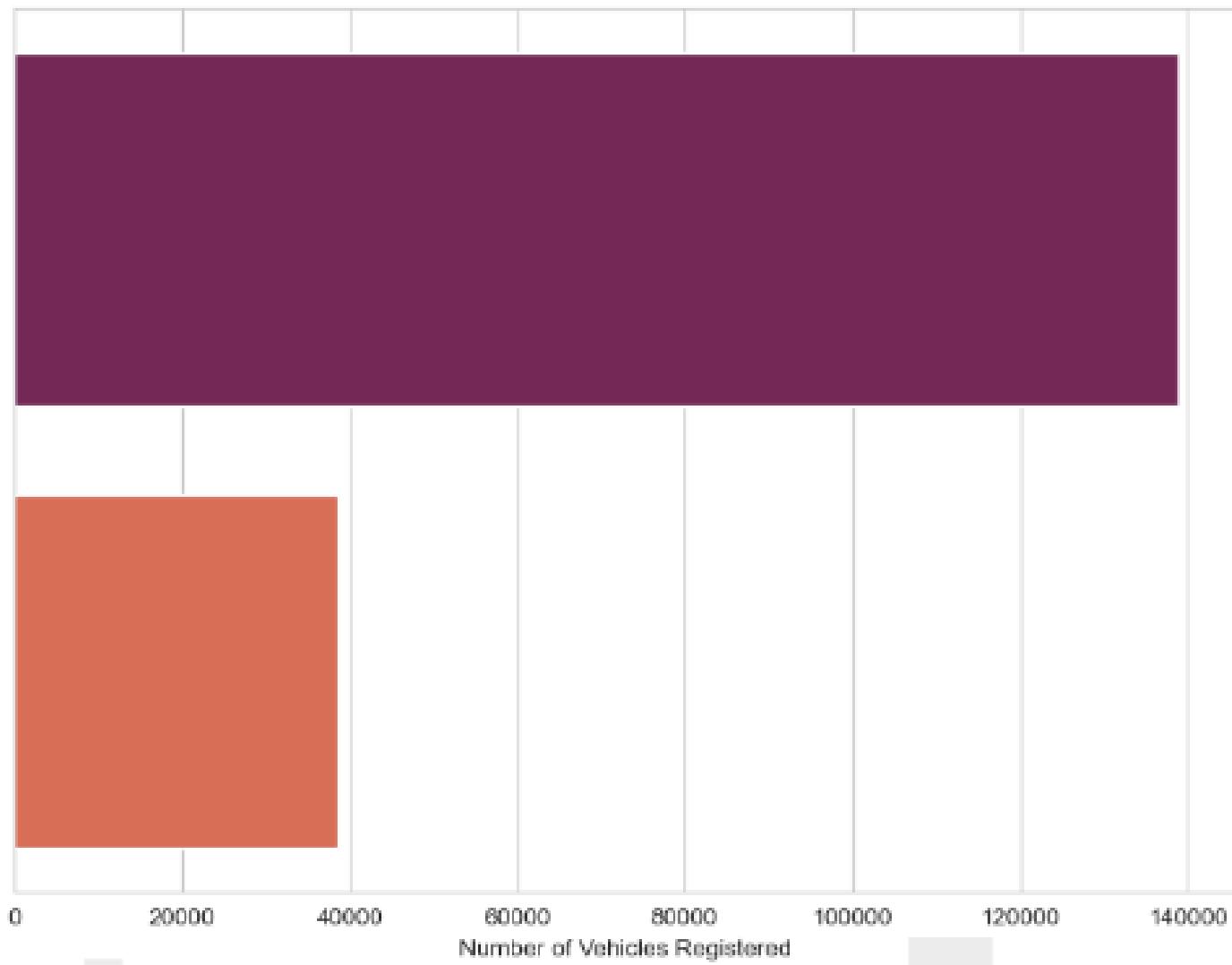


Distribution of Electric Vehicle Types

Electric Vehicle Type

Battery Electric Vehicle (BEV)

Plug-in Hybrid Electric Vehicle (PHEV)



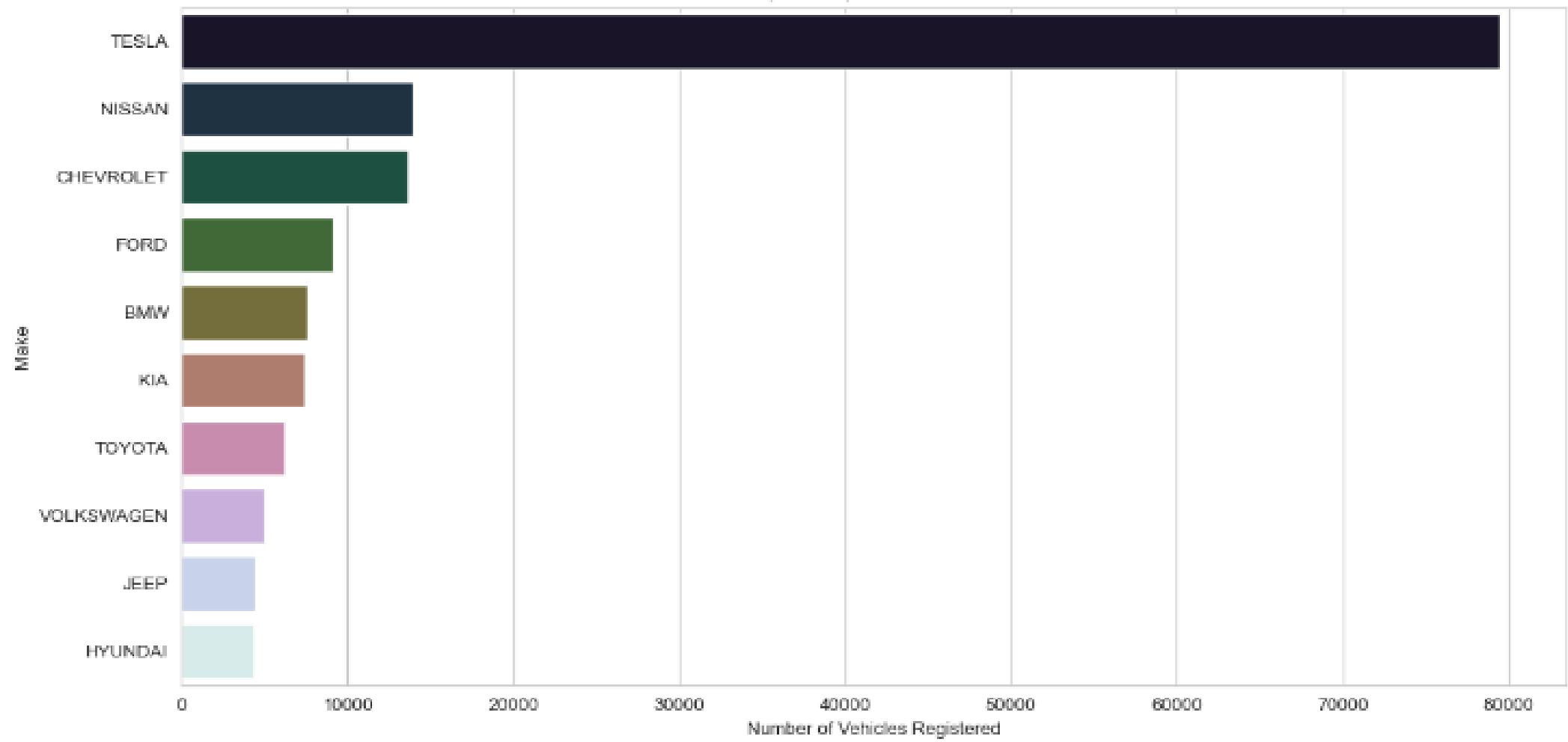
```
# analyzing the popularity of EV manufacturers
ev_make_distribution = ev_data[ 'Make' ].value_counts().head(10) # Limiting to top 10 for clarity

plt.figure(figsize=(12, 6))
sns.barplot(x=ev_make_distribution.values, y=ev_make_distribution.index, palette="cubehelix")
plt.title('Top 10 Popular EV Makes')
plt.xlabel('Number of Vehicles Registered')
plt.ylabel('Make')

plt.tight_layout()
plt.show()
```



Top 10 Popular EV Makes



```
# selecting the top 3 manufacturers based on the number of vehicles registered
top_3_makes = ev_make_distribution.head(3).index

# filtering the dataset for these top manufacturers
top_makes_data = ev_data[ev_data['Make'].isin(top_3_makes)]

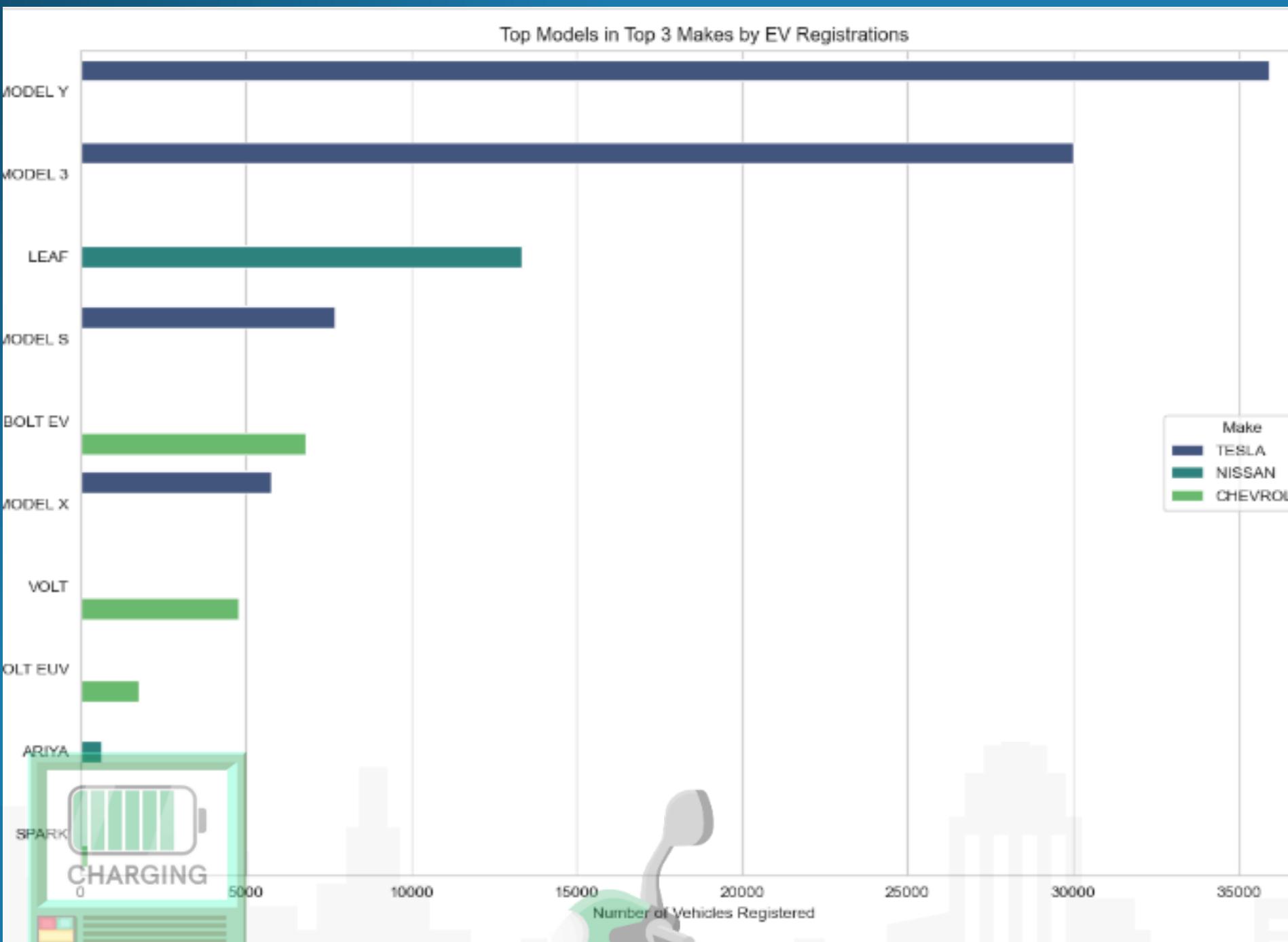
# analyzing the popularity of EV models within these top manufacturers
ev_model_distribution_top_makes = top_makes_data.groupby(['Make', 'Model']).size().sort_values(ascending=False).reset_index(name='Number of Vehicles')

# visualizing the top 10 models across these manufacturers for clarity
top_models = ev_model_distribution_top_makes.head(10)

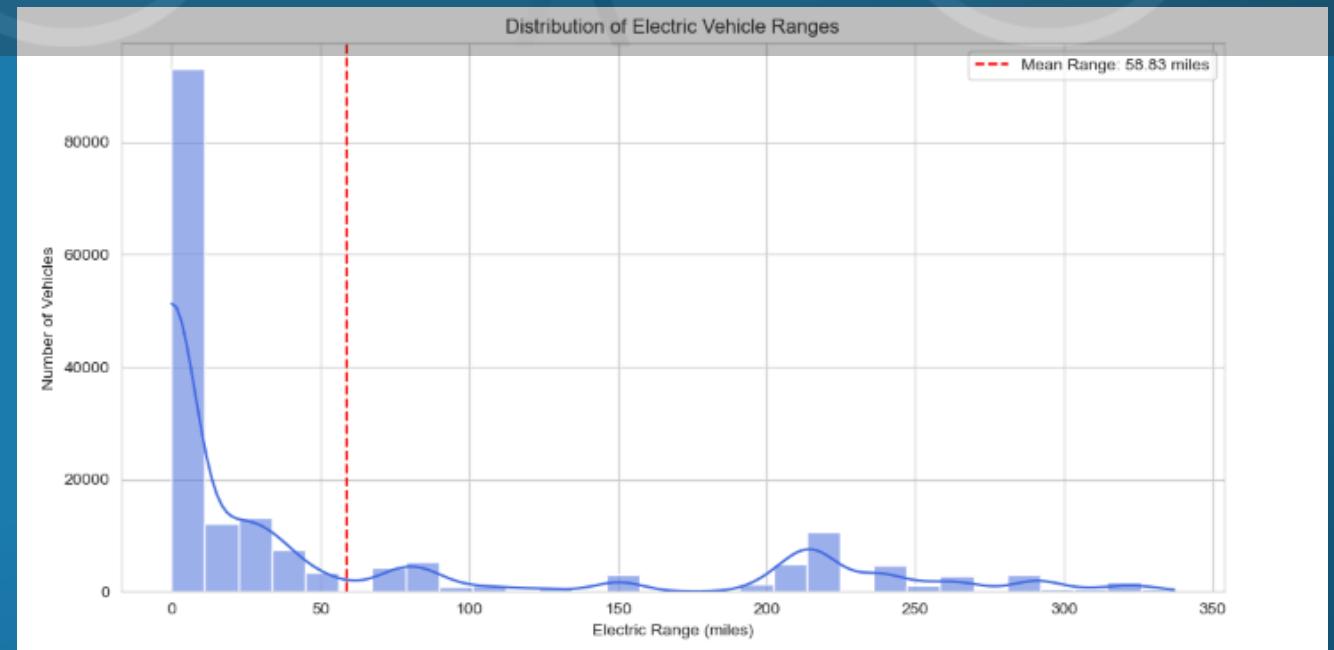
plt.figure(figsize=(12, 8))
sns.barplot(x='Number of Vehicles', y='Model', hue='Make', data=top_models, palette="viridis")
plt.title('Top Models in Top 3 Makes by EV Registrations')
plt.xlabel('Number of Vehicles Registered')
plt.ylabel('Model')
plt.legend(title='Make', loc='center right')

plt.tight_layout()
plt.show()
```





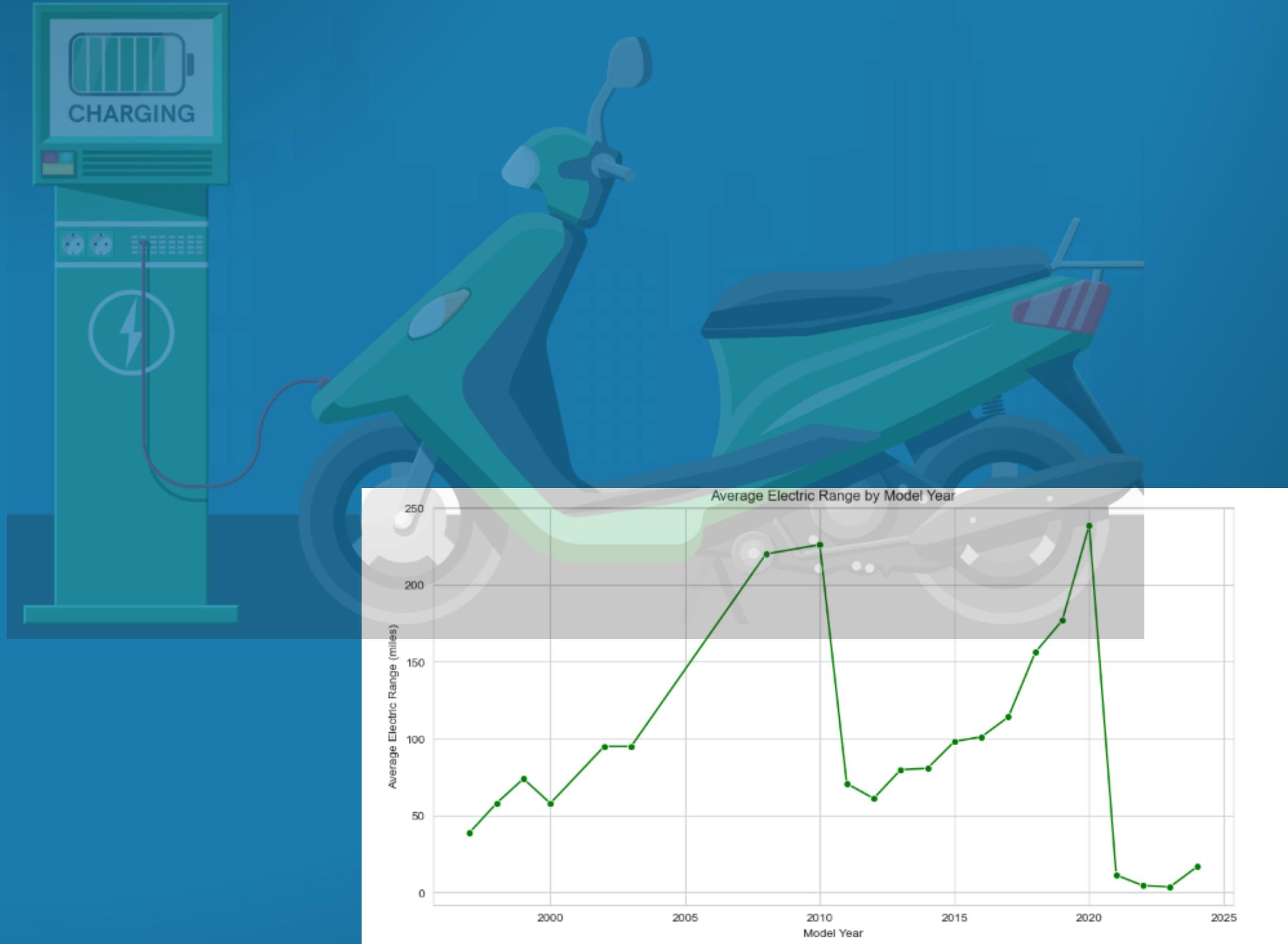
```
# analyzing the distribution of electric range
plt.figure(figsize=(12, 6))
sns.histplot(ev_data['Electric Range'], bins=30, kde=True, color='royalblue')
plt.title('Distribution of Electric Vehicle Ranges')
plt.xlabel('Electric Range (miles)')
plt.ylabel('Number of Vehicles')
plt.axvline(ev_data['Electric Range'].mean(), color='red', linestyle='--', label=f'Mean Range: {ev_data["Electric Range"].mean():.2f} miles')
plt.legend()
plt.show()
```



```
# calculating the average electric range by model year
average_range_by_year = ev_data.groupby('Model Year')[['Electric Range']].mean().reset_index()

plt.figure(figsize=(12, 6))
sns.lineplot(x='Model Year', y='Electric Range', data=average_range_by_year, marker='o', color='green')
plt.title('Average Electric Range by Model Year')
plt.xlabel('Model Year')
plt.ylabel('Average Electric Range (miles)')
plt.grid(True)

plt.show()
```



```

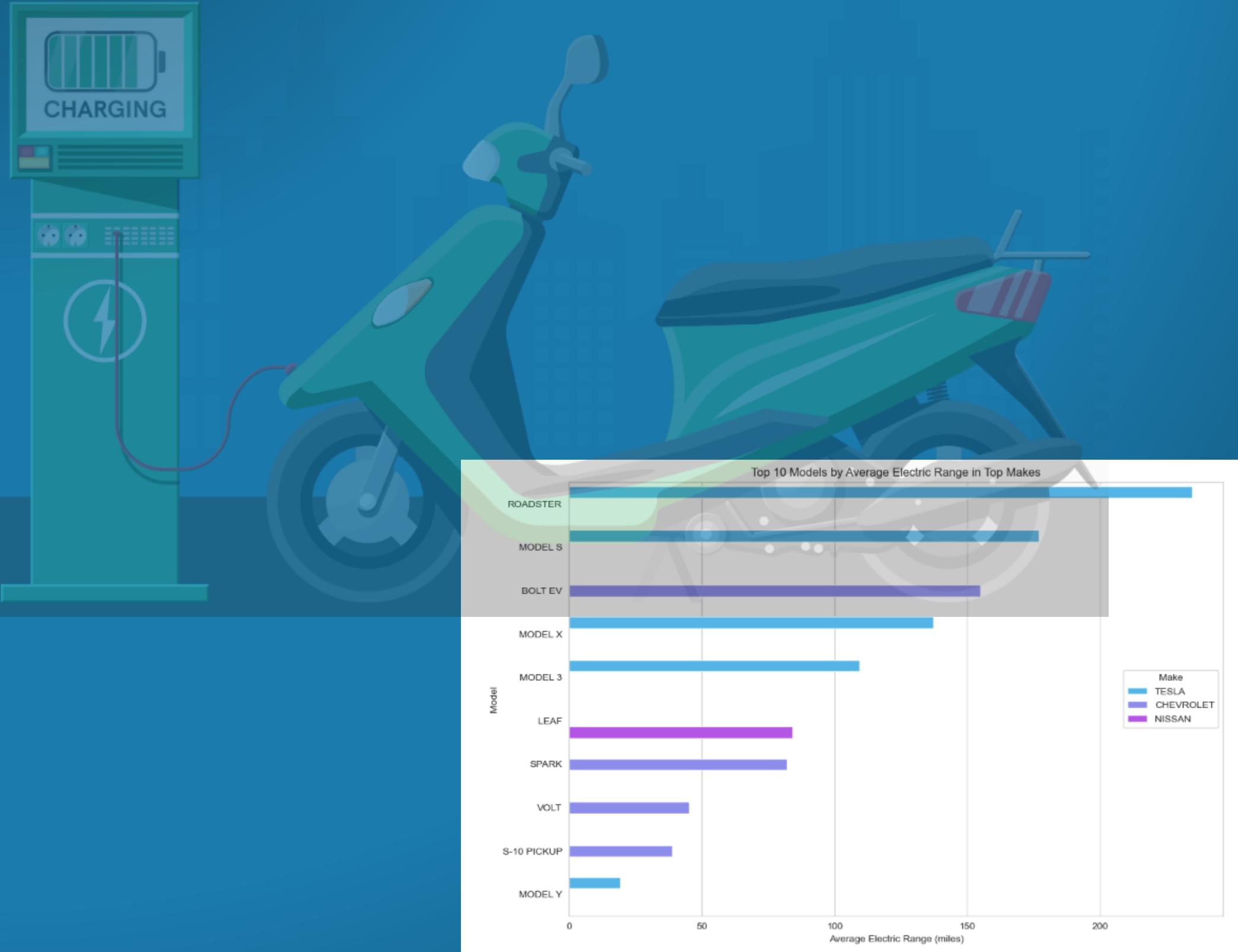
average_range_by_model = top_makes_data.groupby(['Make', 'Model'])['Electric Range'].mean().sort_values(ascending=False).reset_index()

# the top 10 models with the highest average electric range
top_range_models = average_range_by_model.head(10)

plt.figure(figsize=(12, 8))
barplot = sns.barplot(x='Electric Range', y='Model', hue='Make', data=top_range_models, palette="cool")
plt.title('Top 10 Models by Average Electric Range in Top Makes')
plt.xlabel('Average Electric Range (miles)')
plt.ylabel('Model')
plt.legend(title='Make', loc='center right')

plt.show()

```



## Estimated Market Size Analysis of Electric Vehicles in the United States

```
# Now, let's move forward towards finding the estimated market size of electric vehicles in the United States.
```

```
# I'll first count the number of EVs registered every year:
```

```
# calculate the number of EVs registered each year
```

```
ev_registration_counts = ev_data['Model Year'].value_counts().sort_index()
```

```
ev_registration_counts
```

```
: Model Year
```

```
1997      1
1998      1
1999      5
2000      7
2002      2
2003      1
2008     19
2010     23
2011    775
2012   1614
2013   4399
2014   3496
2015   4826
2016   5469
2017   8534
2018  14286
2019  10913
2020  11740
2021  19063
2022  27708
2023  57519
2024  7072
```

```
Name: count, dtype: int64
```



# THANK YOU

