

electric-vehicle-task

October 12, 2024

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: df = pd.read_csv(r'C:\Users\user\Downloads\dataset.csv')
```

```
[3]: df
```

```
[3]:
```

	VIN (1-10)	County	City	State	Postal Code	Model Year	\
0	JTMEB3FV6N	Monroe	Key West	FL	33040	2022	
1	1G1RD6E45D	Clark	Laughlin	NV	89029	2013	
2	JN1AZ0CP8B	Yakima	Yakima	WA	98901	2011	
3	1G1FW6S08H	Skagit	Concrete	WA	98237	2017	
4	3FA6POSU1K	Snohomish	Everett	WA	98201	2019	
...	
112629	7SAYGDEF2N	King	Duvall	WA	98019	2022	
112630	1N4BZ1CP7K	San Juan	Friday Harbor	WA	98250	2019	
112631	1FMCU0KZ4N	King	Vashon	WA	98070	2022	
112632	KNDCD3LD4J	King	Covington	WA	98042	2018	
112633	YV4BR0CL8N	King	Covington	WA	98042	2022	

	Make	Model	Electric Vehicle Type	\
0	TOYOTA	RAV4 PRIME	Plug-in Hybrid Electric Vehicle (PHEV)	
1	CHEVROLET	VOLT	Plug-in Hybrid Electric Vehicle (PHEV)	
2	NISSAN	LEAF	Battery Electric Vehicle (BEV)	
3	CHEVROLET	BOLT EV	Battery Electric Vehicle (BEV)	
4	FORD	FUSION	Plug-in Hybrid Electric Vehicle (PHEV)	
...	
112629	TESLA	MODEL Y	Battery Electric Vehicle (BEV)	
112630	NISSAN	LEAF	Battery Electric Vehicle (BEV)	
112631	FORD	ESCAPE	Plug-in Hybrid Electric Vehicle (PHEV)	
112632	KIA	NIRO	Plug-in Hybrid Electric Vehicle (PHEV)	
112633	VOLVO	XC90	Plug-in Hybrid Electric Vehicle (PHEV)	

	Clean Alternative Fuel Vehicle (CAFV) Eligibility	Electric Range	\
0	Clean Alternative Fuel Vehicle Eligible	42	

1	Clean Alternative Fuel Vehicle Eligible	38
2	Clean Alternative Fuel Vehicle Eligible	73
3	Clean Alternative Fuel Vehicle Eligible	238
4	Not eligible due to low battery range	26
...
112629	Eligibility unknown as battery range has not b...	0
112630	Clean Alternative Fuel Vehicle Eligible	150
112631	Clean Alternative Fuel Vehicle Eligible	38
112632	Not eligible due to low battery range	26
112633	Not eligible due to low battery range	18

	Base MSRP	Legislative District	DOL Vehicle ID \
0	0	NaN	198968248
1	0	NaN	5204412
2	0	15.0	218972519
3	0	39.0	186750406
4	0	38.0	2006714
...
112629	0	45.0	217955265
112630	0	40.0	103663227
112631	0	34.0	193878387
112632	0	47.0	125039043
112633	0	47.0	194673692

	Vehicle Location \
0	POINT (-81.80023 24.5545)
1	POINT (-114.57245 35.16815)
2	POINT (-120.50721 46.60448)
3	POINT (-121.7515 48.53892)
4	POINT (-122.20596 47.97659)
...	...
112629	POINT (-121.98609 47.74068)
112630	POINT (-123.01648 48.53448)
112631	POINT (-122.4573 47.44929)
112632	POINT (-122.09124 47.33778)
112633	POINT (-122.09124 47.33778)

	Electric Utility	2020 Census Tract
0	NaN	12087972100
1	NaN	32003005702
2	PACIFICORP	53077001602
3	PUGET SOUND ENERGY INC	53057951101
4	PUGET SOUND ENERGY INC	53061041500
...
112629	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	53033032401
112630	BONNEVILLE POWER ADMINISTRATION ORCAS POWER &...	53055960301
112631	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	53033027702

```
112632    PUGET SOUND ENERGY INC||CITY OF TACOMA - (WA)    53033032007
112633    PUGET SOUND ENERGY INC||CITY OF TACOMA - (WA)    53033032005
```

```
[112634 rows x 17 columns]
```

```
[4]: df.head()
```

```
[4]:   VIN (1-10)   County   City State   Postal Code   Model Year   Make \
0  JTMEB3FV6N   Monroe   Key West   FL           33040       2022   TOYOTA
1  1G1RD6E45D    Clark   Laughlin   NV           89029       2013  CHEVROLET
2  JN1AZ0CP8B   Yakima   Yakima     WA           98901       2011   NISSAN
3  1G1FW6S08H   Skagit   Concrete   WA           98237       2017  CHEVROLET
4  3FA6P0SU1K  Snohomish   Everett    WA           98201       2019    FORD
```

```
   Model   Electric Vehicle Type \
0  RAV4 PRIME  Plug-in Hybrid Electric Vehicle (PHEV)
1      VOLT   Plug-in Hybrid Electric Vehicle (PHEV)
2      LEAF      Battery Electric Vehicle (BEV)
3    BOLT EV      Battery Electric Vehicle (BEV)
4    FUSION  Plug-in Hybrid Electric Vehicle (PHEV)
```

```
   Clean Alternative Fuel Vehicle (CAFV) Eligibility   Electric Range \
0      Clean Alternative Fuel Vehicle Eligible           42
1      Clean Alternative Fuel Vehicle Eligible           38
2      Clean Alternative Fuel Vehicle Eligible           73
3      Clean Alternative Fuel Vehicle Eligible          238
4      Not eligible due to low battery range           26
```

```
   Base MSRP   Legislative District   DOL Vehicle ID \
0           0                   NaN       198968248
1           0                   NaN       5204412
2           0                  15.0       218972519
3           0                  39.0       186750406
4           0                  38.0       2006714
```

```
   Vehicle Location   Electric Utility   2020 Census Tract
0  POINT (-81.80023 24.5545)           NaN       12087972100
1  POINT (-114.57245 35.16815)           NaN       32003005702
2  POINT (-120.50721 46.60448)    PACIFICORP       53077001602
3  POINT (-121.7515 48.53892)  PUGET SOUND ENERGY INC       53057951101
4  POINT (-122.20596 47.97659)  PUGET SOUND ENERGY INC       53061041500
```

```
[5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 112634 entries, 0 to 112633
Data columns (total 17 columns):
```

#	Column	Non-Null Count	Dtype
0	VIN (1-10)	112634 non-null	object
1	County	112634 non-null	object
2	City	112634 non-null	object
3	State	112634 non-null	object
4	Postal Code	112634 non-null	int64
5	Model Year	112634 non-null	int64
6	Make	112634 non-null	object
7	Model	112614 non-null	object
8	Electric Vehicle Type	112634 non-null	object
9	Clean Alternative Fuel Vehicle (CAFV) Eligibility	112634 non-null	object
10	Electric Range	112634 non-null	int64
11	Base MSRP	112634 non-null	int64
12	Legislative District	112348 non-null	float64
13	DOL Vehicle ID	112634 non-null	int64
14	Vehicle Location	112610 non-null	object
15	Electric Utility	112191 non-null	object
16	2020 Census Tract	112634 non-null	int64

dtypes: float64(1), int64(6), object(10)
memory usage: 14.6+ MB

```
[6]: df.isnull().sum()
```

```
[6]: VIN (1-10)          0
County                0
City                  0
State                 0
Postal Code           0
Model Year            0
Make                  0
Model                20
Electric Vehicle Type  0
Clean Alternative Fuel Vehicle (CAFV) Eligibility  0
Electric Range        0
Base MSRP             0
Legislative District  286
DOL Vehicle ID        0
Vehicle Location      24
Electric Utility      443
2020 Census Tract     0
dtype: int64
```

```
[7]: df.describe()
```

```
[7]:          Postal Code    Model Year  Electric Range    Base MSRP  \
count  112634.000000    112634.000000    112634.000000    112634.000000
```

mean	98156.226850	2019.003365	87.812987	1793.439681
std	2648.733064	2.892364	102.334216	10783.753486
min	1730.000000	1997.000000	0.000000	0.000000
25%	98052.000000	2017.000000	0.000000	0.000000
50%	98119.000000	2020.000000	32.000000	0.000000
75%	98370.000000	2022.000000	208.000000	0.000000
max	99701.000000	2023.000000	337.000000	845000.000000

	Legislative District	DOL Vehicle ID	2020 Census Tract
count	112348.000000	1.126340e+05	1.126340e+05
mean	29.805604	1.994567e+08	5.296650e+10
std	14.700545	9.398427e+07	1.699104e+09
min	1.000000	4.777000e+03	1.101001e+09
25%	18.000000	1.484142e+08	5.303301e+10
50%	34.000000	1.923896e+08	5.303303e+10
75%	43.000000	2.191899e+08	5.305307e+10
max	49.000000	4.792548e+08	5.603300e+10

```
[9]: df.shape
```

```
[9]: (112634, 17)
```

0.1 EDA_Exploratory Data Analysis

```
[10]: df.duplicated().sum()
```

```
[10]: 0
```

```
[11]: # df["Model"]=df["Model"].fillna(df["Model"].mode()[0])
```

```
[12]: # df["Legislative District"]=df["Legislative District"].fillna(df["Legislative_
↳District"].mean())
```

```
[13]: # df["Vehicle Location"]=df["Vehicle Location"].fillna(df["Vehicle Location"].
↳mode()[0])
```

```
[15]: # df["Electric Utility"]=df["Electric Utility"].fillna(df["Electric Utility"].
↳mode()[0])
```

```
[16]: # df.isna().sum()
```

```
[17]: # df.to_csv("Analysis on Electric Vehicles")
```

```
[18]: # df.shape
```

```
[19]: # df.info()
```

0.2 Univariate Analysis

0.2.1 Import required library - plotly.express

```
[20]: import plotly.express as px
```

```
[21]: fig = px.box(df, y='Electric Range',
                  title="Box Plot of Electric Range",
                  labels={'Electric Range': 'Electric Range'},
                  color_discrete_sequence=["#FF5733"]) # Custom color (e.g., orange)

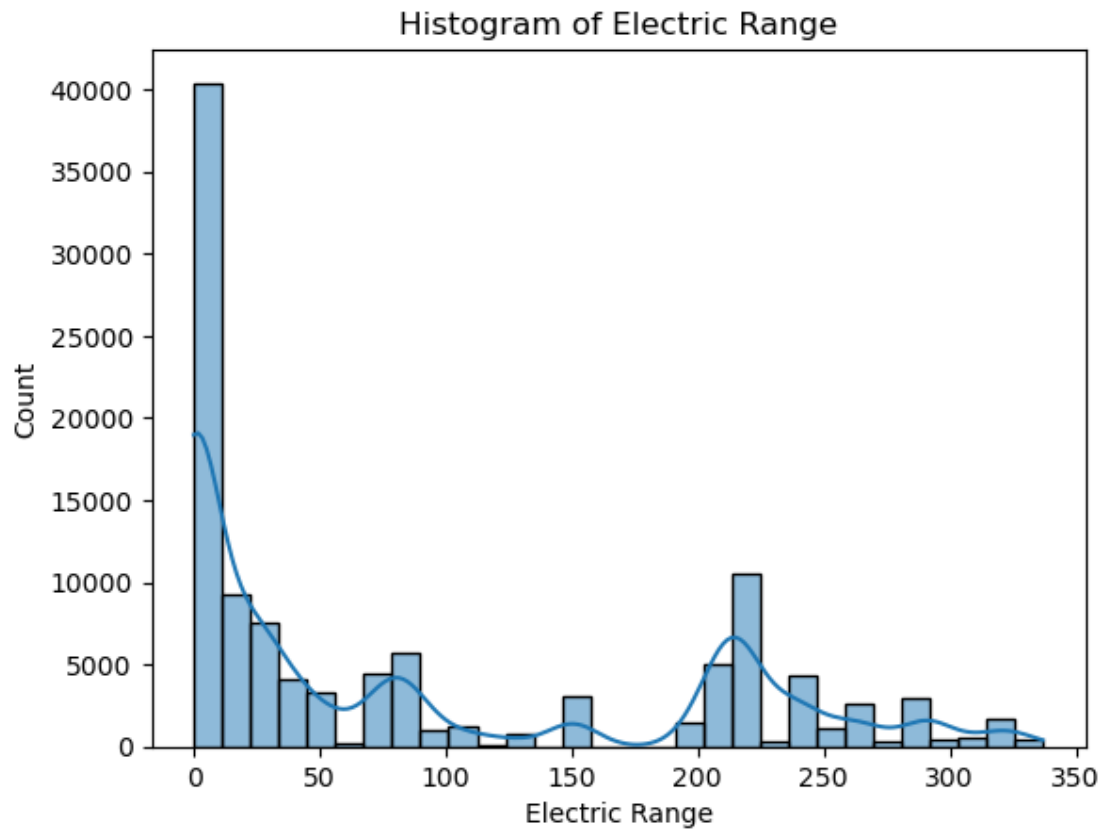
# Customize layout
fig.update_layout(yaxis_title="Electric Range", width=800, height=600)

# Show plot
fig.show()
```

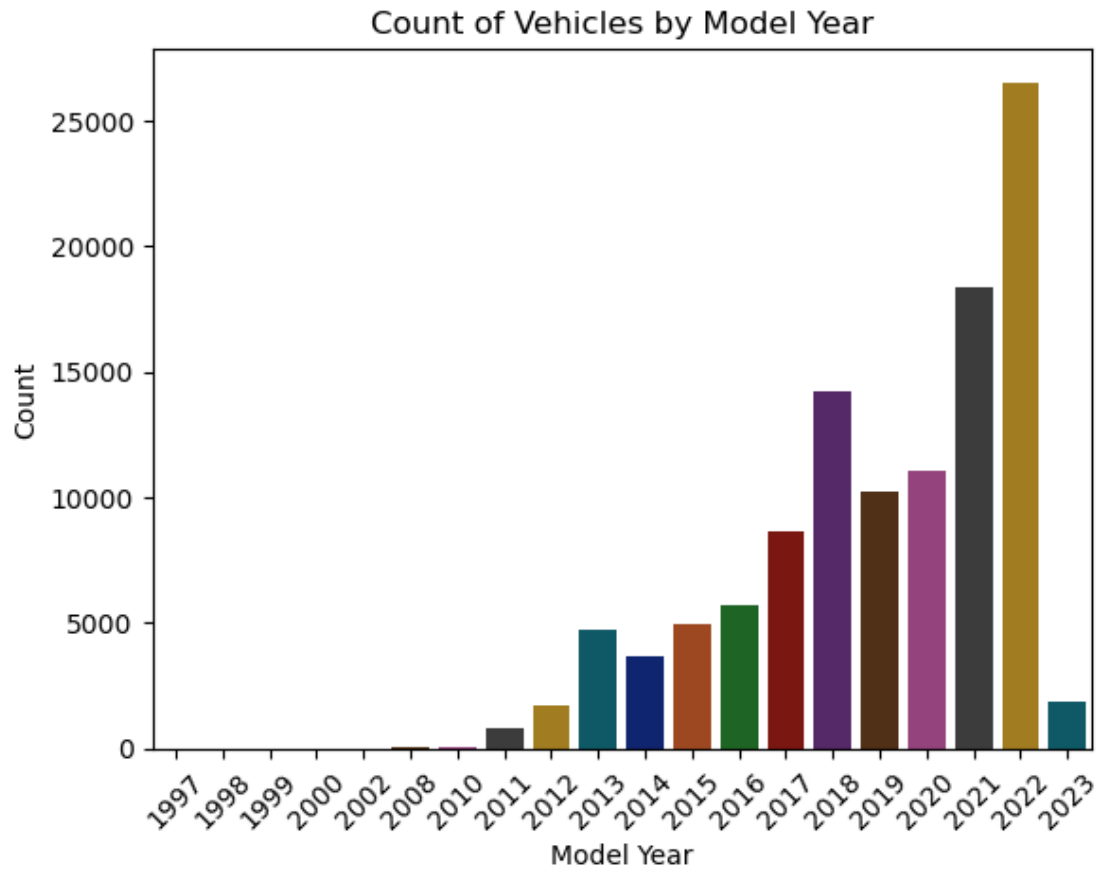
```
[22]: px.histogram(df, x='Base MSRP', title="Histogram of Base MSRP", nbins=30,
                  color_discrete_sequence=['#EF553B']).update_layout(
    xaxis_title="Base MSRP", yaxis_title="Count", width=800, height=600).show()
```

0.2.2 Histograms for numerical features

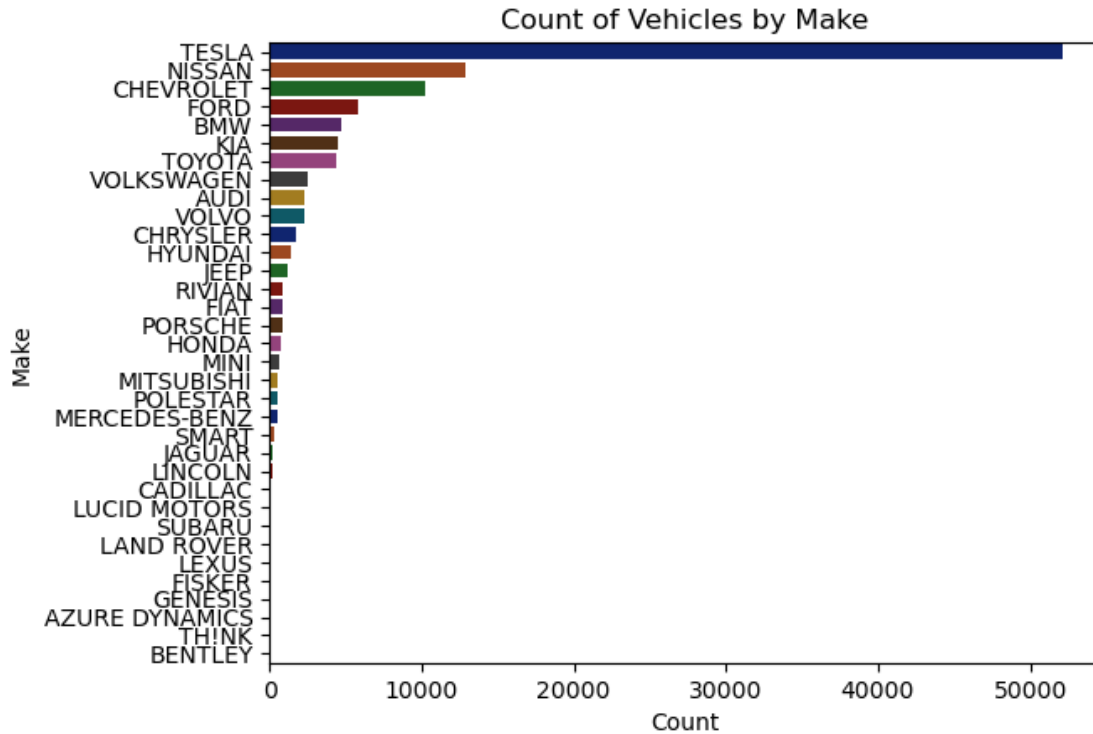
```
[23]: # Replace 'Electric Range' with the actual column name if it contains spaces.
sns.histplot(df['Electric Range'],
             bins=30, kde=True).set_title('Histogram of Electric Range')
plt.show()
```



```
[24]: # Use a darker color palette
sns.countplot(x='Model Year', data=df, palette='dark')
plt.title('Count of Vehicles by Model Year')
plt.xticks(rotation=45)
plt.xlabel('Model Year')
plt.ylabel('Count')
plt.show()
```



```
[25]: # Use a dark color palette
sns.countplot(y='Make', data=df, order=df['Make'].value_counts().index,
             palette='dark')
plt.title('Count of Vehicles by Make')
plt.ylabel('Make')
plt.xlabel('Count')
plt.show()
```

0.3 Bivariate Analysis

0.3.1 Scatter plot using plotly.express Numerical vs Numerical A(Electric Range vs Make)

0.3.2 Task- This is an open ended problem.apply exploratory data analysis (Univariate and Bivariate) on the dataset available above.

```
[26]: px.scatter(df,x = "Make",y ="Electric Range")
```

0.4 Box plot using plotly.Express

```
[27]: px.box(df, x = "Electric Vehicle Type", y = "Electric Range")
```

0.4.1 pie chart plot using plotly.Express

```
[28]: px.pie(df,names = "Make", values = "2020 Census Tract")
```

```
[29]: df["State"].unique()
```

```
[29]: array(['FL', 'NV', 'WA', 'IL', 'NY', 'VA', 'OK', 'KS', 'CA', 'NE', 'MD',
        'CO', 'DC', 'TN', 'SC', 'CT', 'OR', 'TX', 'SD', 'HI', 'GA', 'MS',
        'AR', 'NC', 'MO', 'UT', 'PA', 'DE', 'OH', 'WY', 'AL', 'ID', 'AZ',
```

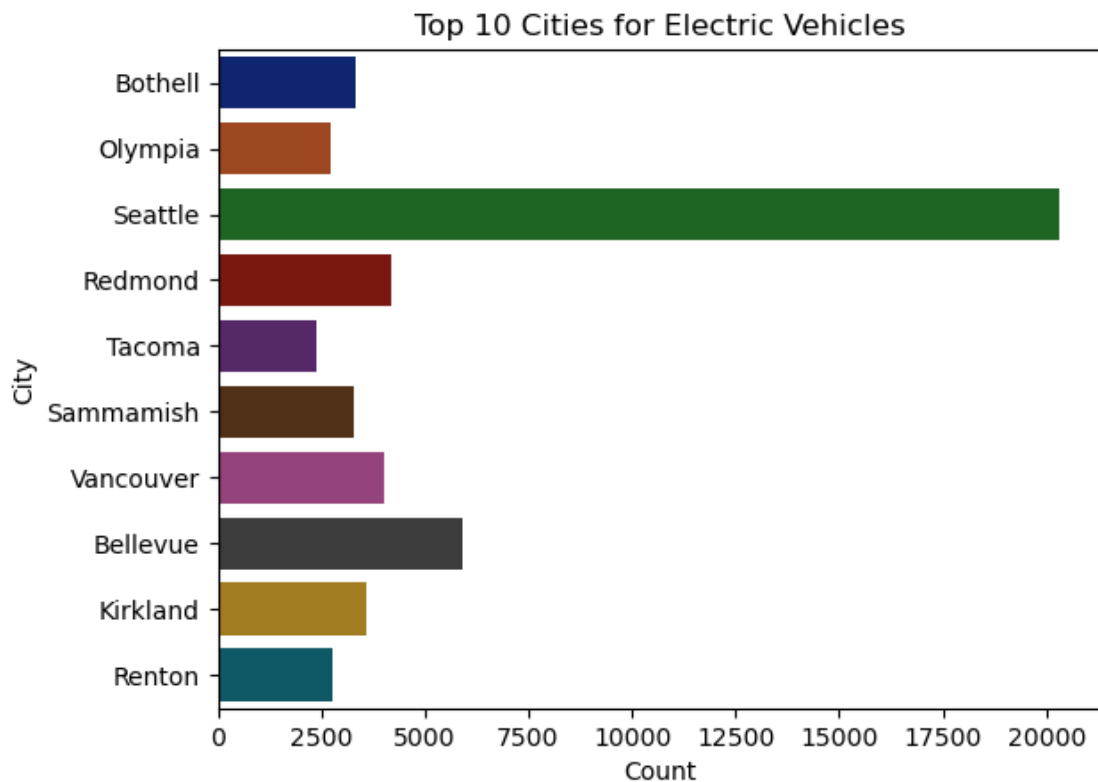
```
'AK', 'LA', 'NM', 'WI', 'KY', 'NJ', 'MN', 'MA', 'ME', 'RI', 'NH',  
'ND'], dtype=object)
```

```
[30]: grouped_df = df.groupby("State").agg({"Electric Range": "mean"})
```

```
[31]: df.shape
```

```
[31]: (112634, 17)
```

```
[32]: # Use a dark color palette  
sns.set_palette("dark")  
  
top_cities = df['City'].value_counts().nlargest(10).index # Top 10 cities  
sns.countplot(y='City', data=df[df['City'].isin(top_cities)])  
plt.title('Top 10 Cities for Electric Vehicles')  
plt.xlabel('Count')  
plt.ylabel('City')  
plt.show()
```



0.4.2 Heatmap of Correlation (for numeric variables)

```
[40]: # Select only numeric columns
df_numeric = df.select_dtypes(include=[np.number])

# Calculate the correlation matrix
correlation_matrix = df_numeric.corr()

# Create heatmap for correlation matrix
import plotly.express as px
fig = px.imshow(correlation_matrix, title="Correlation Heatmap", text_auto=True)
fig.show()
```

```
[34]: # Pie chart for Electric Vehicle Type
fig = px.pie(df, names='Electric Vehicle Type', title="Distribution of Electric_
↳Vehicle Types")
fig.show()
```

```
[36]: import pandas as pd, plotly.express as px

df = pd.read_csv(r"C:\Users\user\Downloads\dataset.csv")
state_nyc = df.groupby(['Postal Code', 'Model Year']).size().
↳reset_index(name='Number_of_Vehicles')

fig = px.choropleth_mapbox(state_nyc, geojson='https://raw.githubusercontent.com/python-visualization/folium/master/examples/data/us-states.json',
↳locations='Postal Code', color='Number_of_Vehicles',
↳featureidkey="properties.ZCTA5CE10", mapbox_style="carto-positron",
zoom=5, center={"lat": 47.7511, "lon": -120.7401}, animation_frame='Model_
↳Year')

fig.update_layout(margin={"r": 0, "t": 0, "l": 0, "b": 0}).show()
```

```
[38]: # pip install bar_chart_race
```

```
[39]: import pandas as pd, plotly.express as px

df = pd.read_csv(r"C:\Users\user\Downloads\dataset.csv")
d = df.groupby(['Make', 'Model Year']).size().
↳reset_index(name='Number_of_Vehicles')

px.bar(d, x='Number_of_Vehicles', y='Make', color='Make',
↳animation_frame='Model Year', orientation='h',
title='EV Makes and Their Count Over the Years', range_x=[0, 3000]).
↳update_traces(
texttemplate='%{x}', textposition='outside').update_layout(
```

```
yaxis_title='EV Makes', xaxis_title='Number of EV Vehicles', title_x=0.5,   
↪width=800, height=600).show()
```

```
[ ]:
```