



Problem Statement/object

Market Size Analysis is the process of estimating the potential sales for a product or service within a particular market segment. In the context of electric vehicles (EVs), it involves assessing the total volume of EV registrations to understand the growth of the market, forecast future trends, and help stakeholders make informed decisions regarding production,infrastructure development, and policy-making.The provided dataset contains the following columns, each representing different aspects of the electric vehicle (EV) population in the United States:

- VIN (1-10): Partial Vehicle Identification Number.
- County: The county in which the vehicle is registered.
- City: The city in which the vehicle is registered.
- State: The state in which the vehicle is registered. It appears that this dataset may be focused on Washington (WA) state.
- Postal Code: The postal code where the vehicle is registered.
- Model Year: The year of the vehicle model
- Make: The manufacturer of the vehicle.
- Model:The model of the vehicle.
- Electric Vehicle Type: The type of electric vehicle, e.g., Battery Electric Vehicle (BEV).
- Clean Alternative Fuel Vehicle (CAFV) Eligibility: Eligibility status for clean alternative fuel vehicle programs.
- Electric range: The maximum range of the vehicle on a single charge (in miles).
- Base MSRP: The Manufacturer's Suggested Retail Price.
- Legislative District:The legislative district where the vehicle is registered.
- DOL Vehicle ID:Department of Licensing Vehicle Identification.
- Vehicle Location: Geographic coordinates of the vehicle location.
- Electric Utility: The electric utility service provider for the vehicle's location.
- 2020 Census Tract:The census tract for the vehicle's location.

```
In [3]: 1
C:\Program Files (x86)\Microsoft Visual Studio\Shared\Anaconda3_64\lib\site-packages\IPython\core\interactiveshell.py:2785: DtypeWarning: Columns (10,12) have mixed types. S
pecify dtype option on import or set low_memory=False.
  interactivity=interactivity, compiler=compiler, result=result)

In [4]: 1 df
(PHEV) range
24 JN1AZ0CP8C Thurston Tenino WA 98589.0 2012 NISSAN LEAF Battery Electric Vehicle (BEV) Clean Alternative Fuel Vehicle Eligible 73 0 20 131685669 POINT (-122.85403 46.856085) PUGET SOUND ENERGY INC 5.306701e-
25 1FADP5CU4G Thurston Olympia WA 98501.0 2016 FORD C-MAX Plug-in Hybrid Electric Vehicle (PHEV) Not eligible due to low battery range 19 0 22 101841806 POINT (-122.89692 47.043535) PUGET SOUND ENERGY INC 5.306701e-
26 1N4AZ1CP4J Thurston Olympia WA 98501.0 2018 NISSAN LEAF Battery Electric Vehicle (BEV) Clean Alternative Fuel Vehicle Eligible 151 0 22 475432268 POINT (-122.89692 47.043535) PUGET SOUND ENERGY INC 5.306701e-
27 WA1AAAGE4M King Seattle WA 98108.0 2021 AUDI E-TRON Battery Electric Vehicle (BEV) Clean Alternative Fuel 222 0 37 156817314 POINT (-122.3268963 CITY OF SEATTLE - (WA)CITY OF TACOMA - 5.303301e-
```

1. Display the top 5 rows

```
In [5]: 1
```

Top 5 rows of the dataset:

	VIN (1-10)	County	City	State	Postal Code	Model Year	Make	Model	Electric Vehicle Type	Clean Alternative Fuel Vehicle (CAFV) Eligibility	Electric Range	Base MSRP	Legislative District	DOL Vehicle ID	Vehicle Location	Electric Utility	2020 Census Tract
0	5YJYGDEE1L	King	Seattle	WA	98122.0	2020	TESLA	MODEL Y	Battery Electric Vehicle (BEV)	Clean Alternative Fuel Vehicle Eligible	291	0	37	125701579	POINT (-122.30839 47.610365)	CITY OF SEATTLE - (WA) CITY OF TACOMA - (WA)	5.303301e+10
1	7SAYGDEE9P	Snohomish	Bothell	WA	98021.0	2023	TESLA	MODEL Y	Battery Electric Vehicle (BEV)	Eligibility unknown as battery range has not b...	0	0	1	244285107	POINT (-122.179458 47.802589)	PUGET SOUND ENERGY INC	5.306105e+10
2	5YJSA1E4XK	King	Seattle	WA	98109.0	2019	TESLA	MODEL S	Battery Electric Vehicle (BEV)	NaN	270	0	36	156773144	POINT (-122.34848 47.632405)	CITY OF SEATTLE - (WA) CITY OF TACOMA - (WA)	5.303301e+10
3	5YJSA1E27G	King	Issaquah	WA	98027.0	2016	TESLA	MODEL S	Battery Electric Vehicle (BEV)	Clean Alternative Fuel Vehicle Eligible	210	0	5	165103011	POINT (-122.03646 47.534065)	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	5.303303e+10
4	5YJYGDEE5M	Kitsap	Suquamish	WA	98392.0	2021	NaN	MODEL Y	Battery Electric Vehicle (BEV)	Eligibility unknown as battery range has not b...	0	0	23	205138552	POINT (-122.55717 47.733415)	PUGET SOUND ENERGY INC	5.303594e+10

2. Display the last 5 rows

```
In [6]: 1
```

Last 5 rows of the dataset:

	VIN (1-10)	County	City	State	Postal Code	Model Year	Make	Model	Electric Vehicle Type	Clean Alternative Fuel Vehicle (CAFV) Eligibility	Electric Range	Base MSRP	Legislative District	DOL Vehicle ID	Vehicle Location	Electric Utility	2020 Census Tract
177861	7SAYGDEE3N	Pierce	Bonney Lake	WA	98391.0	2022	TESLA	MODEL Y	Battery Electric Vehicle (BEV)	Eligibility unknown as battery range has not b...	0	0	31	195224452	POINT (-122.183805 47.18062)	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	5.305307e+10
177862	KM8K23AG1P	Mason	Shelton	WA	98584.0	2023	HYUNDAI	KONA ELECTRIC	Battery Electric Vehicle (BEV)	Eligibility unknown as battery range has not b...	0	0	35	228454180	POINT (-123.105305 47.211085)	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	5.304596e+10
177863	5YJYGDEE6M	Grant	Quincy	WA	98848.0	2021	TESLA	MODEL Y	Battery Electric Vehicle (BEV)	Eligibility unknown as battery range has not b...	0	0	13	168797219	POINT (-119.8493873 47.2339933)	PUD NO 2 OF GRANT COUNTY	5.302501e+10
177864	WVGKMPE27M	King	Black Diamond	WA	98010.0	2021	VOLKSWAGEN	ID.4	Battery Electric Vehicle (BEV)	Eligibility unknown as battery range has not b...	0	0	5	182448801	POINT (-122.00451 47.312185)	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	5.303303e+10
177865	5YJ3E1EA8M	Pierce	Tacoma	WA	98422.0	2021	TESLA	MODEL 3	Battery Electric Vehicle (BEV)	Eligibility unknown as battery range has not b...	0	0	27	211464683	POINT (-122.38578 47.28971)	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	5.305394e+10

3. Check the shape of the dataset

```
In [7]: 1
```

Shape of the dataset:

(177866, 17)

4. Check the data types of each feature

```
In [8]: 1
```

Data types of each feature:

```
VIN (1-10)          object
County             object
City               object
State              object
Postal Code        float64
Model Year         int64
Make               object
Model              object
Electric Vehicle Type  object
Clean Alternative Fuel Vehicle (CAFV) Eligibility  object
Electric Range      object
Base MSRP           int64
Legislative District object
DOL Vehicle ID      int64
Vehicle Location    object
Electric Utility     object
2020 Census Tract   float64
dtype: object
```

5. Check the statistical summary

```
In [9]: 1
```

Statistical summary of the dataset:

	Postal Code	Model Year	Base MSRP	DOL Vehicle ID	2020 Census Tract
count	177861.000000	177866.000000	177866.000000	1.778660e+05	1.778610e+05
mean	98172.453506	2020.515512	1073.109363	2.202313e+08	5.297672e+10
std	2442.450668	2.989384	8358.624956	7.584987e+07	1.578047e+09
min	1545.000000	1997.000000	0.000000	4.385000e+03	1.001020e+09
25%	98052.000000	2019.000000	0.000000	1.814743e+08	5.303301e+10
50%	98122.000000	2022.000000	0.000000	2.282522e+08	5.303303e+10
75%	98370.000000	2023.000000	0.000000	2.548445e+08	5.305307e+10
max	99577.000000	2024.000000	845000.000000	4.792548e+08	5.603300e+10

6. Check the null values

```
In [10]: 1
```

Null values in the dataset:

VIN (1-10)	0
County	5
City	5
State	0
Postal Code	5
Model Year	0
Make	7
Model	4
Electric Vehicle Type	6
Clean Alternative Fuel Vehicle (CAFV) Eligibility	2
Electric Range	3
Base MSRP	0
Legislative District	389
DOL Vehicle ID	0
Vehicle Location	9
Electric Utility	5
2020 Census Tract	5
dtype: int64	

7. Check the duplicate values

```
In [11]: 1
```

Duplicate values in the dataset:

0

8. Check the anomalies or wrong entries

```
In [12]: 1
```

Anomalies or wrong entries (Example: Checking Model Year for future years):

VIN (1-10)	County	City	State	Postal Code	Model Year	Make	Model	Electric Vehicle Type	Clean Alternative Fuel Vehicle (CAFV) Eligibility	Electric Range	Base MSRP	Legislative District	DOL Vehicle ID	Vehicle Location	Electric Utility	2020 Census Tract
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1. Descriptive Statistics:

- What are the mean, median, and standard deviation of the base

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In [13]: 1
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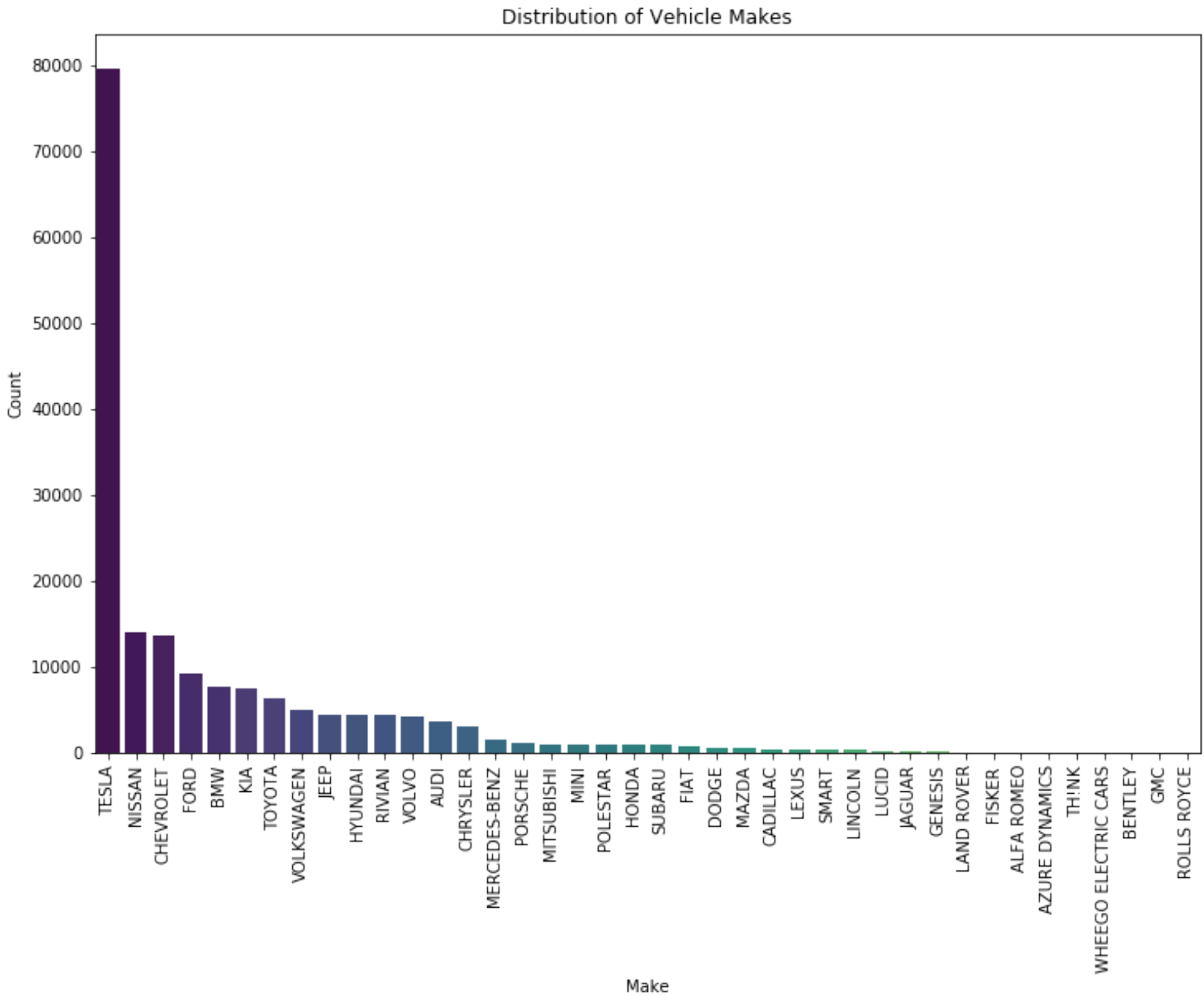
Mean MSRP: 1073.1093632284978
Median MSRP: 0.0
Standard Deviation MSRP: 8358.624956079144

2. Data Distribution:

- What is the distribution of vehicle makes in the dataset? Represent it using a bar chart.

In [14]:

1



3. Model Year Analysis:

- What are the most common model years in the dataset? Provide a frequency table and histogram.

In [15]:

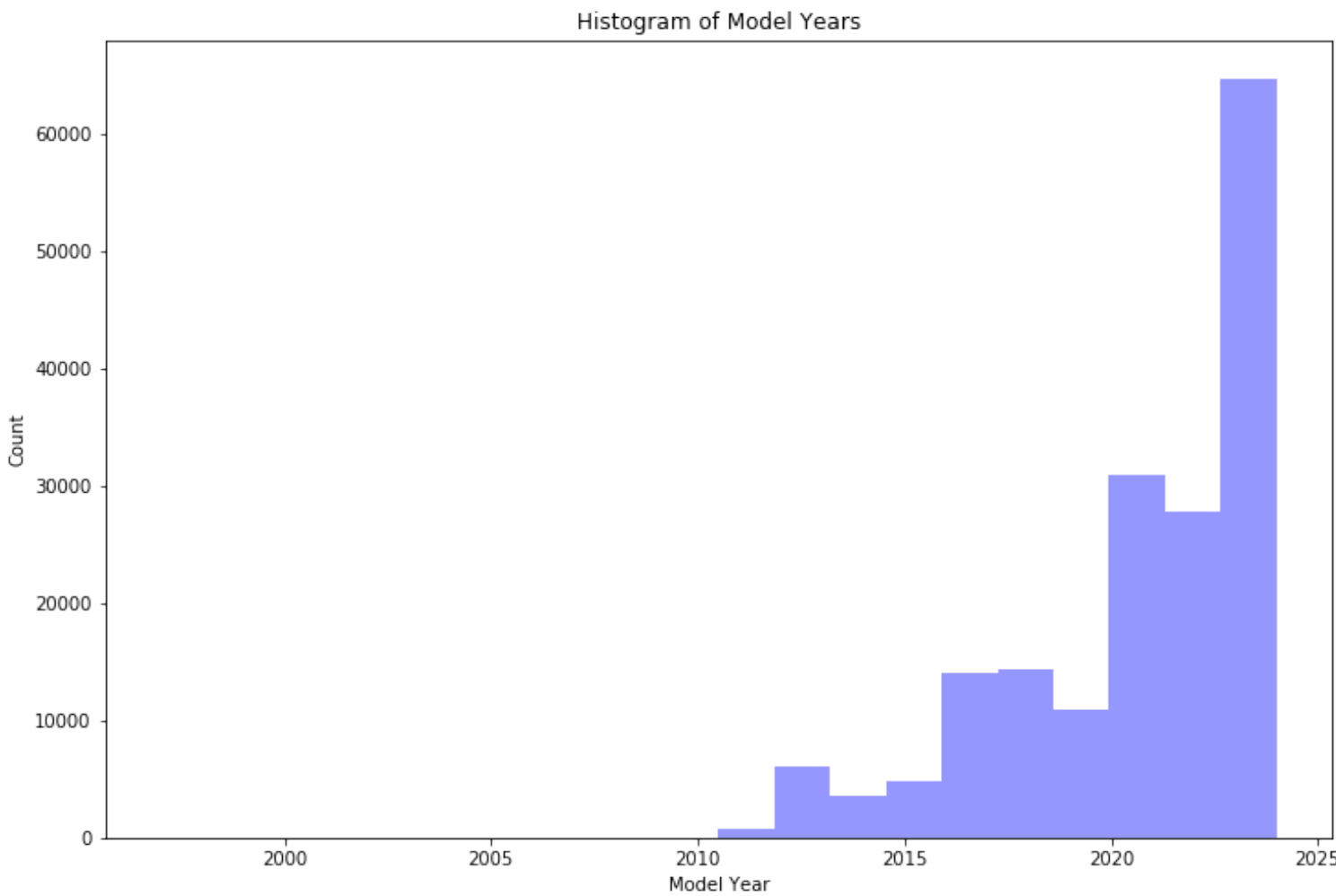
1

Frequency Table of Model Years:

1997	1
1998	1
1999	5
2000	7
2002	2
2003	1
2008	20
2010	23
2011	775
2012	1618
2013	4409
2014	3509
2015	4844
2016	5483
2017	8562
2018	14323
2019	10940
2020	11768
2021	19132
2022	27776
2023	57587
2024	7080

Name: Model Year, dtype: int64

C:\Program Files (x86)\Microsoft Visual Studio\Shared\Anaconda3_64\lib\site-packages\matplotlib\axes_axes.py:6462: UserWarning: The 'normed' kwarg is deprecated, and has been replaced by the 'density' kwarg.
warnings.warn("The 'normed' kwarg is deprecated, and has been "



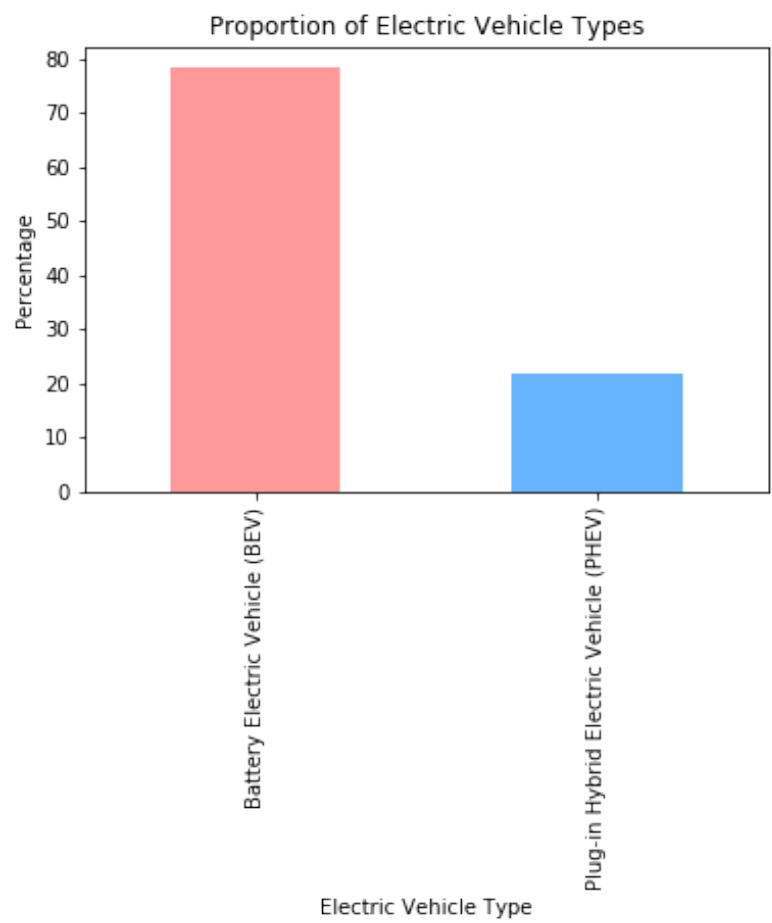
4. Electric Vehicle Type:

- What is the proportion of Battery Electric Vehicles (BEV) versus other types of electric vehicles?

In [16]:

1

```
Proportion of Electric Vehicle Types:
Battery Electric Vehicle (BEV)      78.266052
Plug-in Hybrid Electric Vehicle (PHEV)  21.733948
Name: Electric Vehicle Type, dtype: float64
```



5. Electric Range Analysis:

- What is the average electric range for vehicles of different makes? Provide a summary table.

In [17]:

1

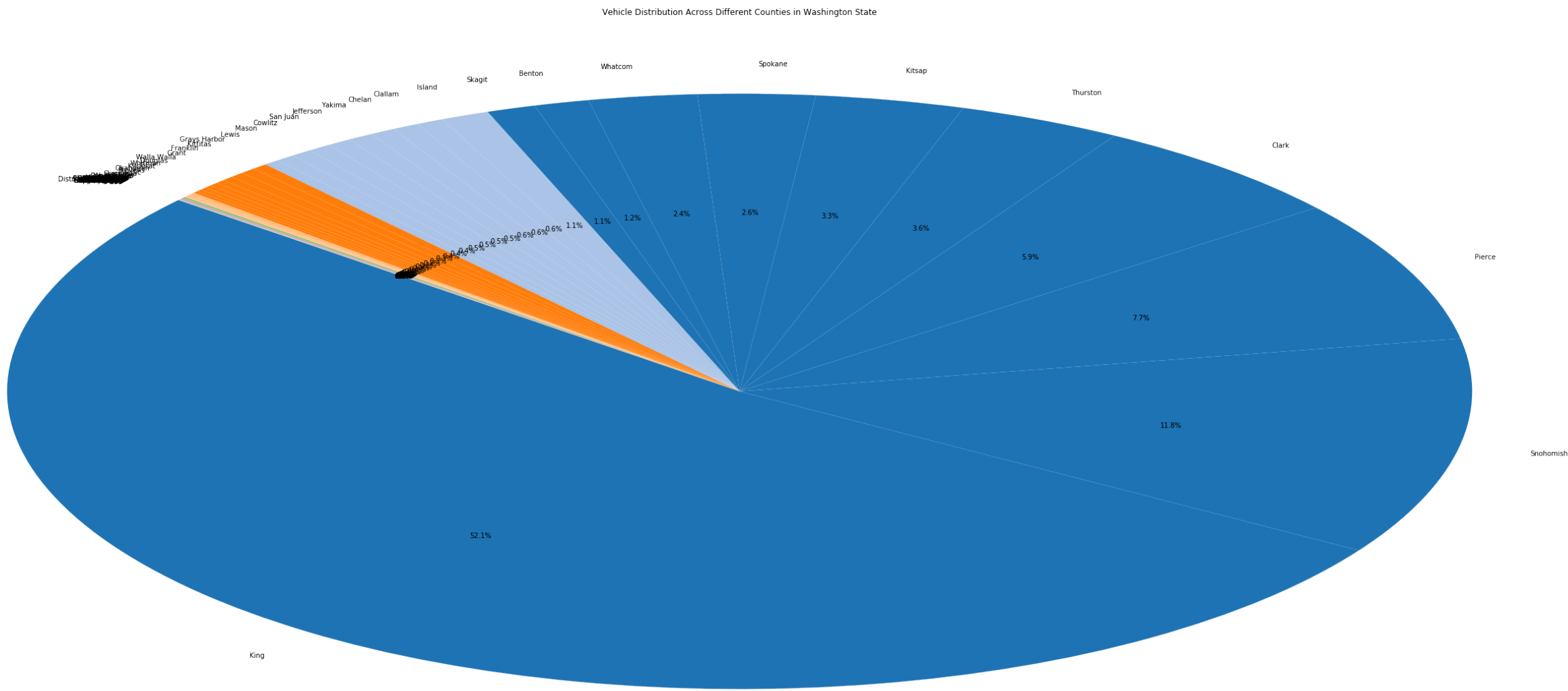
```
Average Electric Range for Vehicles of Different Makes:
Make
JAGUAR      203.741379
BMW         201.159730
WHEEGO ELECTRIC CARS  100.000000
THINK       100.000000
CHEVROLET   94.552785
FIAT        85.645408
NISSAN      80.348053
TESLA       77.341268
SMART       62.325926
AZURE DYNAMICS  56.000000
AUDI        50.984641
HONDA       46.600240
PORSCHE     44.812116
KIA         44.650161
POLESTAR    37.776644
ALFA ROMEO  33.000000
CHRYSLER    32.212162
DODGE       32.000000
MITSUBISHI  30.646138
TOYOTA      28.095102
MAZDA       25.781513
LAND ROVER  25.000000
LINCOLN     23.543071
VOLKSWAGEN  22.927058
JEEP        22.365402
BENTLEY     19.666667
HYUNDAI     19.362914
LEXUS       18.800000
MINI        18.036789
VOLVO       16.135737
FORD        10.812914
MERCEDES-BENZ  9.346130
CADILLAC    8.798429
FISKER      8.755102
SUBARU      1.350181
RIVIAN      0.000000
LUCID       0.000000
ROLLS ROYCE 0.000000
GENESIS     0.000000
GMC         0.000000
Name: Electric Range, dtype: float64
```

6. County Distribution:

- How are vehicles distributed across different counties in Washington state? Represent the distribution using a pie chart.

In [18]:

1



7. Price Analysis:

- Compare the average base MSRP of vehicles eligible for the Clean Alternative Fuel Vehicle (CAFV) program versus those that are not.

In [19]:

1

```
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interactivity=interactivity, compiler=compiler, result=result)

Average Base MSRP for CAFV Eligible Vehicles: $2028.13
Average Base MSRP for Non-Eligible CAFV Vehicles: $505.16
```

8. Geographical Analysis:

- How does the base MSRP vary across different cities in Washington state?

In [20]:

1

102	Dayton	4368.750000
333	Raymond	4248.437500
268	Newman Lake	4205.813953
221	Lyle	4192.307692
13	Anderson Island	4126.315789
417	Touchet	3993.750000
433	Valleyford	3883.333333
292	Orondo	3744.642857
241	Medina	3698.636364
293	Oroville	3678.947368
14	Ariel	3678.947368
59	Cathlamet	3595.900000
..
193	Kittitas	0.000000
190	Keyport	0.000000
189	Kettle Falls	0.000000
184	Keller	0.000000
183	Kapowsin	0.000000
182	Kalama	0.000000
181	Joint Base Lewis-McChord	0.000000

9. Legislative Districts:

- Which legislative districts have the highest number of registered electric vehicles? Provide a ranked list.

In [21]:

1

	Legislative District	Number of Registered EVs
0	41	8441
1	45	7425
2	5	6810
3	48	6631
4	1	6265
5	36	5922
6	43	5049
7	46	5033
8	11	4871
9	34	4449
10	37	4326
11	44	4263
12	21	4037
13	22	3857
14	40	3796
15	45	3312
16	18	3302
17	26	3253
18	48	3186
19	41	3112
20	31	3002
21	32	2957
22	10	2892
23	23	2804
24	11	2636
25	27	2522
26	42	2444
27	39	2396
28	28	2277
29	33	2248
...
62	7	810
63	5	804
64	34	706
65	22	542
66	49	510
67	44	414
68	21	373
69	15	352
70	30	306
71	24	291
72	26	285
73	20	266
74	35	264
75	39	197
76	2	175
77	19	152
78	14	141
79	10	126
80	31	98
81	15	91
82	38	34
83	7	29
84	13	9
85	12	5
86	9	5
87	16	4
88	40	2
89	3	2
90	6	1
91	?	1

[92 rows x 2 columns]

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interactivity=interactivity, compiler=compiler, result=result)

10. Electric Utility Providers:

- What is the distribution of electric utility service providers for the vehicles in the dataset?

In [22]:

1

	Electric Utility	Number of Vehicles
0	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	65990
1	PUGET SOUND ENERGY INC	35882
2	CITY OF SEATTLE - (WA) CITY OF TACOMA - (WA)	31381
3	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF C...	10173
4	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	7828
5	PUGET SOUND ENERGY INC PUD NO 1 OF WHATCOM CO...	4008
6	BONNEVILLE POWER ADMINISTRATION AVISTA CORP ...	2797
7	BONNEVILLE POWER ADMINISTRATION PUD 1 OF SNOH...	1537
8	PACIFICORP	1306
9	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF B...	1244
10	MODERN ELECTRIC WATER COMPANY	1127
11	PUD NO 1 OF CHELAN COUNTY	1047
12	BONNEVILLE POWER ADMINISTRATION PUGET SOUND E...	951
13	BONNEVILLE POWER ADMINISTRATION ORCAS POWER &...	938
14	BONNEVILLE POWER ADMINISTRATION CITY OF RICHL...	903
15	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF C...	897
16	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF C...	838
17	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	708
18	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF G...	609
19	PUD NO 2 OF GRANT COUNTY	561
20	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF F...	549
21	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	471
22	AVISTA CORP	400
23	BONNEVILLE POWER ADMINISTRATION VERA IRRIGATI...	399
24	NON WASHINGTON STATE ELECTRIC UTILITY	384
25	PUD NO 1 OF DOUGLAS COUNTY	362
26	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	335
27	BONNEVILLE POWER ADMINISTRATION INLAND POWER ...	335
28	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF K...	284
29	CITY OF TACOMA - (WA) TANNER ELECTRIC COOP	251
..
46	BONNEVILLE POWER ADMINISTRATION CITY OF MILTO...	93
47	BONNEVILLE POWER ADMINISTRATION AVISTA CORP ...	71
48	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF W...	60
49	PUD NO 1 OF PEND OREILLE COUNTY	55
50	BONNEVILLE POWER ADMINISTRATION AVISTA CORP ...	44
51	CITY OF CHENEY - (WA)	43
52	PUD NO 1 OF WHATCOM COUNTY	42
53	BONNEVILLE POWER ADMINISTRATION TOWN OF RUSTO...	40
54	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	39
55	BONNEVILLE POWER ADMINISTRATION BIG BEND ELEC...	34
56	BONNEVILLE POWER ADMINISTRATION TOWN OF EATON...	29
57	CITY OF TACOMA - (WA)	24
58	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF M...	21
59	BONNEVILLE POWER ADMINISTRATION COLUMBIA RURA...	20
60	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF F...	18
61	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF K...	16
62	BONNEVILLE POWER ADMINISTRATION CITY OF MCCLE...	15
63	BONNEVILLE POWER ADMINISTRATION PACIFICORP C...	14
64	CITY OF SUMAS - (WA) PUD NO 1 OF WHATCOM COUNTY	8
65	CITY OF CHEWELAH	5
66	PORTLAND GENERAL ELECTRIC CO	4
67	BONNEVILLE POWER ADMINISTRATION CITY OF COULE...	4
68	BONNEVILLE POWER ADMINISTRATION BENTON RURAL ...	4
69	BONNEVILLE POWER ADMINISTRATION NESPELEM VALL...	3
70	BONNEVILLE POWER ADMINISTRATION CITY OF TACOM...	2
71	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF A...	2
72	CITY OF SEATTLE - (WA)	1
73	BONNEVILLE POWER ADMINISTRATION PENINSULA LIG...	1
74	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF C...	1
75	BONNEVILLE POWER ADMINISTRATION PUD NO 1 OF J...	1

[76 rows x 2 columns]

```
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  interactivity=interactivity, compiler=compiler, result=result)
```

11. Census Tract Analysis:

- How are vehicles distributed across different 2020 Census Tracts? Provide insights based on vehicle counts per tract.

In [23]:

1

	2020 Census Tract	Number of Vehicles
0	5.303303e+10	2479
1	5.303303e+10	983
2	5.303303e+10	820
3	5.303303e+10	801
4	5.306701e+10	672
5	5.303301e+10	651
6	5.303303e+10	601
7	5.306105e+10	581
8	5.303303e+10	577
9	5.303302e+10	558

```
C:\Program Files (x86)\Microsoft Visual Studio\Shared\Anaconda3_64\lib\site-packages\IPython\core\interactiveshell.py:2785: DtypeWarning: Columns (10,12) have mixed types. Specify dtype option on import or set low_memory=False.
  interactivity=interactivity, compiler=compiler, result=result)
```

12. Electric Range Correlation:

- Is there a correlation between the electric range and the base MSRP of the vehicles? Provide the correlation coefficient and interpret the result.

In [24]:

1

Correlation coefficient between Electric Range and Base MSRP: 0.00
There is little to no correlation between electric range and base MSRP.

```
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  interactivity=interactivity, compiler=compiler, result=result)
```

13. VIN Analysis:

- Identify any patterns or commonalities in the VIN (1-10) for the vehicles. Are there any frequent prefixes or suffixes?


```
In [25]: 1

Most Common VIN Prefixes:
5YJ      50385
7SA      29274
1G1      13389
1N4      12120
KND      7325
1C4      4480
KM8      3781
YV4      3763
JTD      3761
3FM      3322
Name: VIN (1-10), dtype: int64

Most Common VIN Suffixes:
E6P      1606
E7P      1573
E2P      1545
E5P      1525
E8P      1515
EXP      1503
E9P      1497
E3P      1486
E0P      1467
E4P      1455
Name: VIN (1-10), dtype: int64
```

14. Eligibility Status:

- What percentage of vehicles are eligible for the Clean Alternative Fuel Vehicle (CAFV) program?

```
In [26]: 1

Percentage of vehicles eligible for the CAFV program: 0.00%
```

15. Model Popularity:

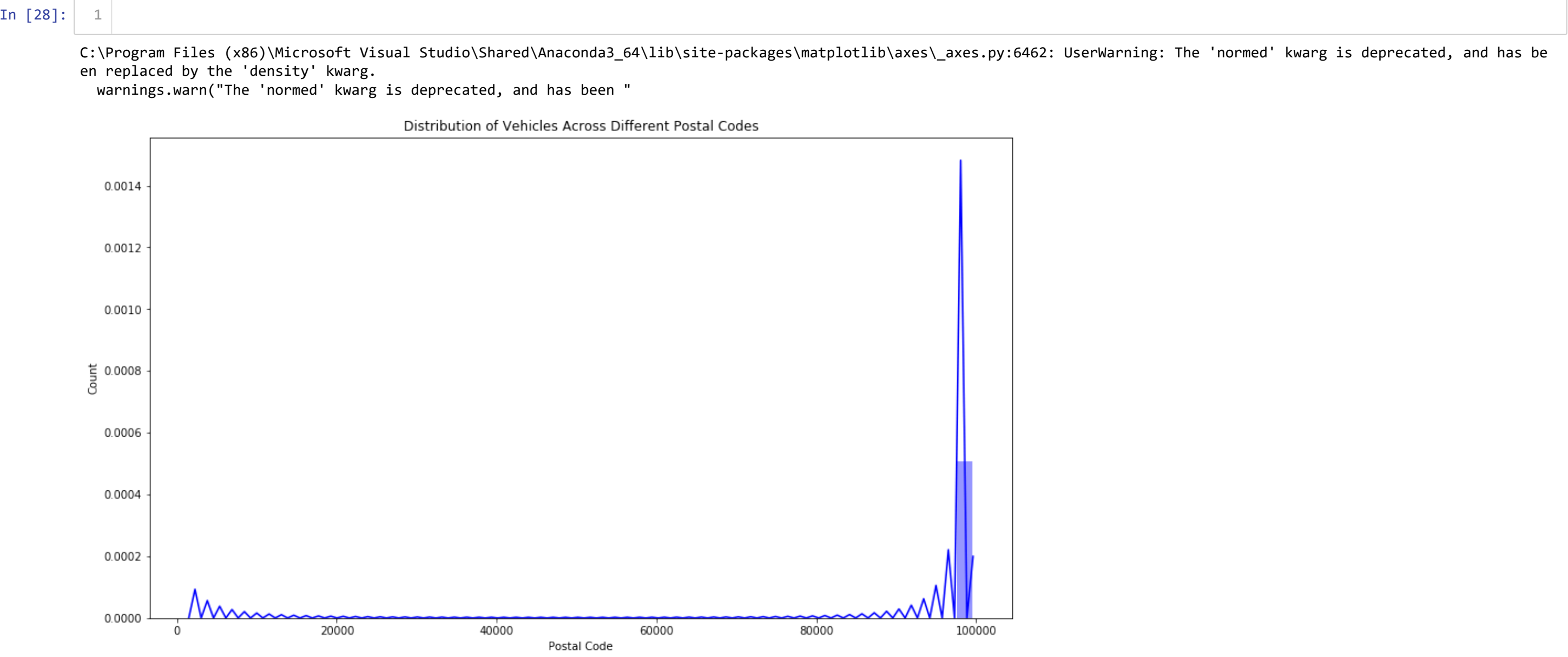
- Which vehicle models are the most popular in the dataset? Provide a frequency table of the top 10 models.

```
In [27]: 1

Top 10 Most Popular Vehicle Models:
MODEL Y      35989
MODEL 3      30091
LEAF         13365
MODEL S       7734
BOLT EV      6821
MODEL X       5796
VOLT         4796
ID.4         3937
WRANGLER      3392
MUSTANG MACH-E 3322
Name: Model, dtype: int64
```

16. Postal Code Distribution:

- How are vehicles distributed across different postal codes? Provide a heatmap or density plot.

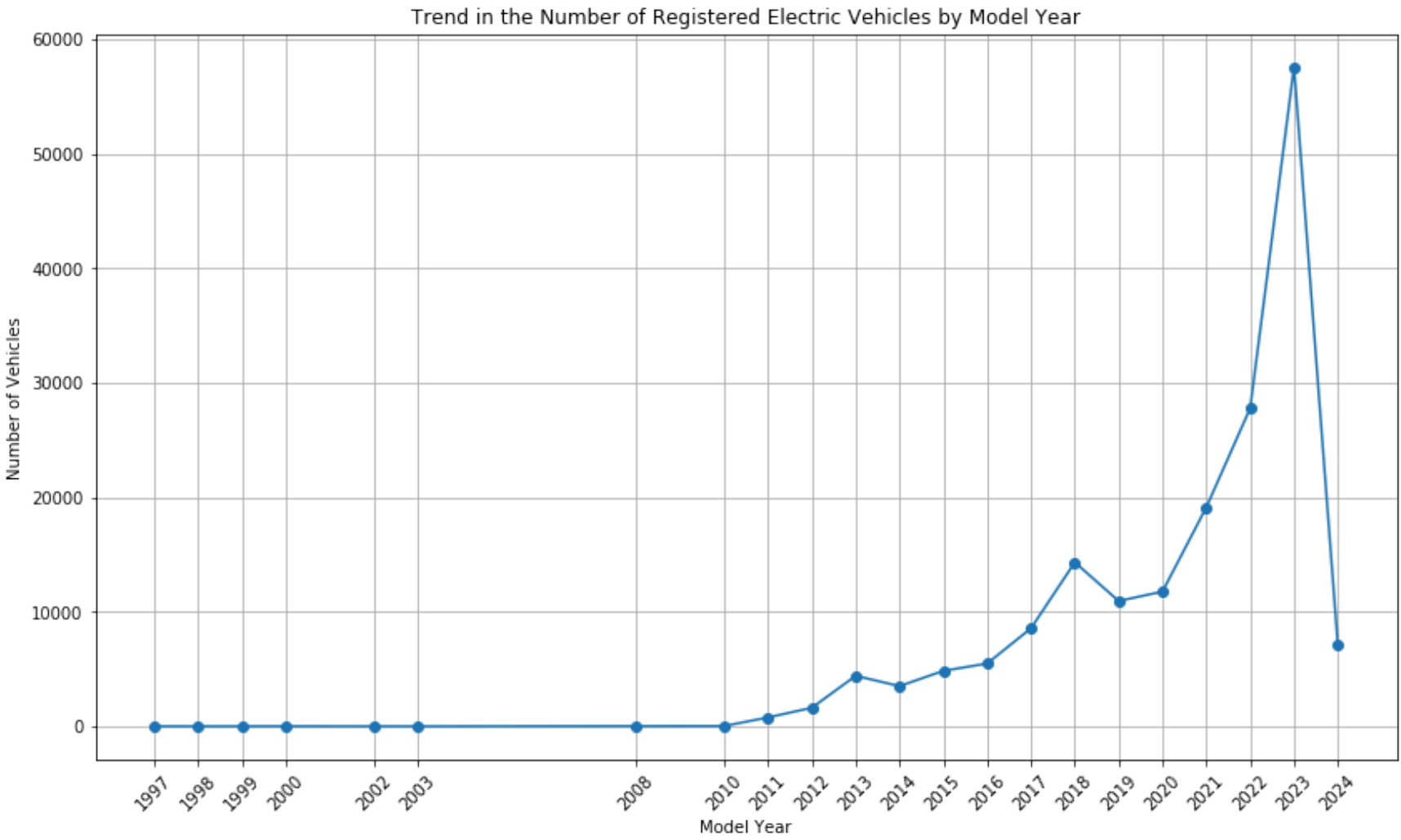


18. Model Year Trend:

- Analyze the trend in the number of registered electric vehicles by model year. Provide a line chart to show any increase or decrease over the years.

In [29]:

1



19. Range vs. Year:

- Is there a trend between the model year and the electric range of the vehicles? Provide a scatter plot and analyze the trend.

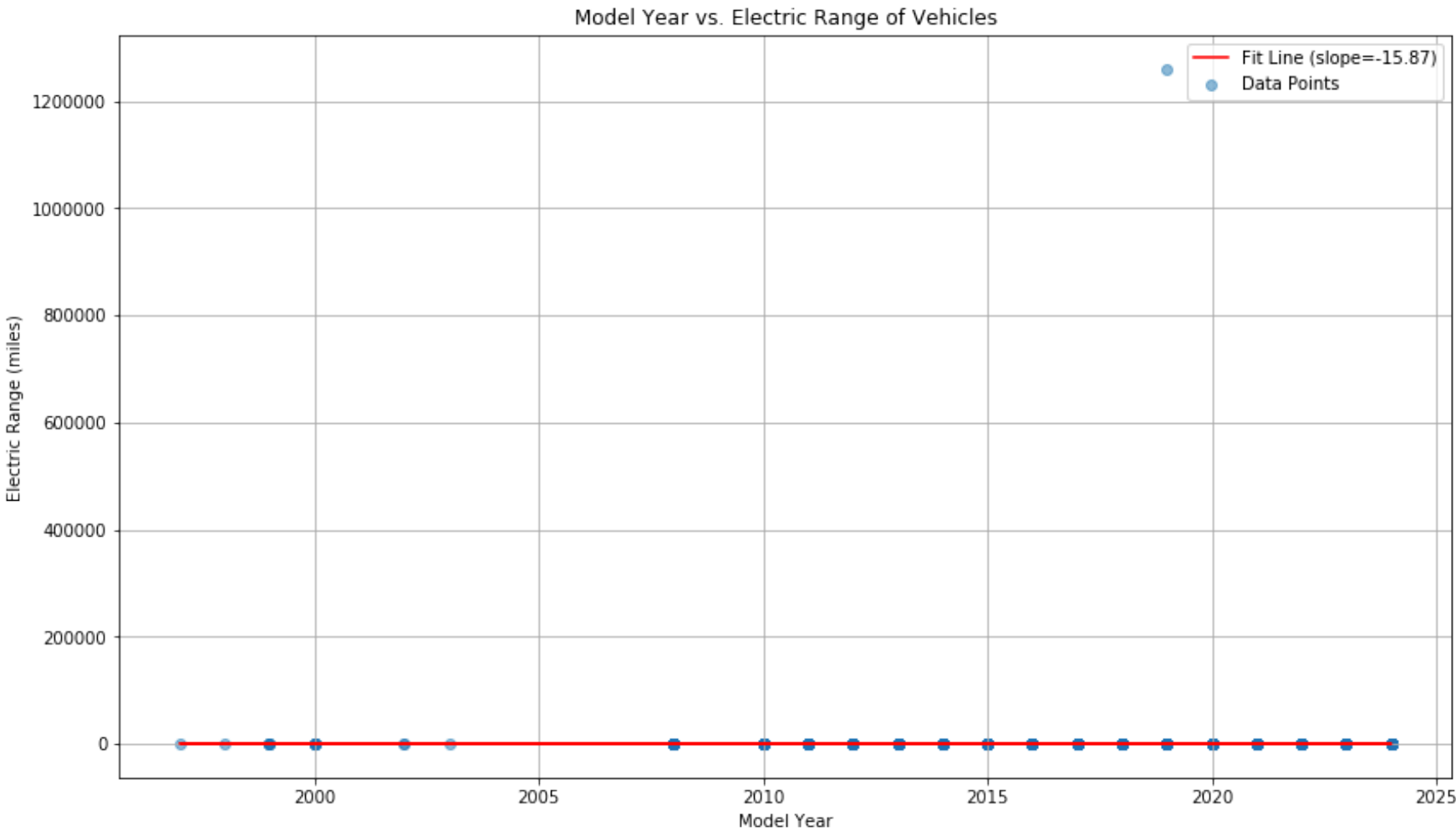
In [32]:

1

```
C:\Program Files (x86)\Microsoft Visual Studio\Shared\Anaconda3_64\lib\site-packages\ipykernel_launcher.py:9: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy (http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy)
if __name__ == '__main__':
C:\Program Files (x86)\Microsoft Visual Studio\Shared\Anaconda3_64\lib\site-packages\ipykernel_launcher.py:10: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy (http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy)
# Remove the CWD from sys.path while we load stuff.
```



Slope: -15.874249257702395
Intercept: 32140.093746317798
R-squared: 0.000252064540932783

20. Legislative District and MSRP:

- How does the average base MSRP vary across different legislative districts?

In [33]: 1

C:\Program Files (x86)\Microsoft Visual Studio\Shared\Anaconda3_64\lib\site-packages\ipykernel_launcher.py:8: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy> (<http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy>)

Average Base MSRP by Legislative District

Legislative District	Average Base MSRP (\$)
1	300
2	350
3	400
4	450
5	500
6	550
7	600
8	650
9	680
10	700
11	720
12	750
13	780
14	800
15	820
16	850
17	880
18	900
19	920
20	950
21	980
22	1000
23	1020
24	1050
25	1080
26	1100
27	1120
28	1150
29	1180
30	1200
31	1220
32	1250
33	1280
34	1300
35	1320
36	1350
37	1380
38	1400
39	1420
40	1450
41	1480
42	1500
43	1520
44	1550
45	1580
46	1600
47	1650
48	1700
49	1800
50	3100

In []: 1