

```
In [10]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error, r2_score
```

```
In [11]: # Load the dataset
df = pd.read_csv('EVPD.csv')

# Display the first few rows to understand the structure
print(df.head())
```

	VIN (1-10)	County	City	State	Postal Code	Model	Year	Make	\
0	KM8K33AGXL	King	Seattle	WA	98103.0		2020	HYUNDAI	
1	1C4RJYB61N	King	Bothell	WA	98011.0		2022	JEEP	
2	1C4RJYD61P	Yakima	Yakima	WA	98908.0		2023	JEEP	
3	5YJ3E1EA7J	King	Kirkland	WA	98034.0		2018	TESLA	
4	WBY7Z8C5XJ	Thurston	Olympia	WA	98501.0		2018	BMW	

	Model	Electric Vehicle Type	\
0	KONA	Battery Electric Vehicle (BEV)	
1	GRAND CHEROKEE	Plug-in Hybrid Electric Vehicle (PHEV)	
2	GRAND CHEROKEE	Plug-in Hybrid Electric Vehicle (PHEV)	
3	MODEL 3	Battery Electric Vehicle (BEV)	
4	I3	Plug-in Hybrid Electric Vehicle (PHEV)	

	Clean Alternative Fuel Vehicle (CAFV) Eligibility	Electric Range	\
0	Clean Alternative Fuel Vehicle Eligible	258	
1	Not eligible due to low battery range	25	
2	Not eligible due to low battery range	25	
3	Clean Alternative Fuel Vehicle Eligible	215	
4	Clean Alternative Fuel Vehicle Eligible	97	

	Base MSRP	Legislative District	DOL Vehicle ID	\
0	0	43.0	249675142	
1	0	1.0	233928502	
2	0	14.0	229675939	
3	0	45.0	104714466	
4	0	22.0	185498386	

	Vehicle Location	\
0	POINT (-122.34301 47.659185)	
1	POINT (-122.20578 47.762405)	
2	POINT (-120.6027202 46.5965625)	
3	POINT (-122.209285 47.71124)	
4	POINT (-122.89692 47.043535)	

	Electric Utility	2020 Census Tract
0	CITY OF SEATTLE - (WA) CITY OF TACOMA - (WA)	5.303300e+10
1	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	5.303302e+10
2	PACIFICORP	5.307700e+10
3	PUGET SOUND ENERGY INC CITY OF TACOMA - (WA)	5.303302e+10
4	PUGET SOUND ENERGY INC	5.306701e+10

```
In [12]: # Check for missing values
print(df.isnull().sum())
#df = df.dropna()
```

VIN (1-10)	0
County	3
City	3
State	0
Postal Code	3
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	341
DOL Vehicle ID	0
Vehicle Location	7
Electric Utility	3
2020 Census Tract	3
dtype:	int64

```
In [13]: X = df[['Model Year', 'Make']] # Independent variables
y = df['Electric Range'] # Dependent variable

# Convert Make to numeric using Label encoding
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
X['Make'] = le.fit_transform(X['Make'])

# Split the data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

print("Training set shape:", X_train.shape)
print("Testing set shape:", X_test.shape)

# Display first few rows of training data
print("\n
First few rows of training features:")
print(X_train.head())
print("\n
First few rows of training target:")
print(y_train.head())
```

Training set shape: (120385, 2)
Testing set shape: (30097, 2)
First few rows of training features:

	Model Year	Make
150211	2021	31
40535	2020	31
55800	2022	10
100144	2018	31
78788	2023	15

First few rows of training target:

150211	0
40535	291
55800	0
100144	215
78788	21

Name: Electric Range, dtype: int64

C:\Users\mahesh narke\AppData\Local\Temp\ipykernel_7432\1212729005.py:7: 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: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
X['Make'] = le.fit_transform(X['Make'])
```

```
In [24]: # Create a Linear Regression model
model = LinearRegression()
# Train the model
model.fit(X_train[['Model Year']], y_train)
```

```
Out[24]: ▼ LinearRegression ⓘ ?
LinearRegression()
```

```
In [25]: # Make predictions on the test set
predictions = model.predict(X_test[['Model Year']])
```

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In [26]: y_pred
```

```
Out[26]: array([ -4.30875575,  28.9991295 , 155.52941663, ...,   8.99455993,
        78.91136134,  11.95085231])
```

```
In [27]: # Calculate Mean Squared Error
mse = mean_squared_error(y_test, y_pred)
print(f"Mean Squared Error: {mse}")
# Calculate R-Squared value
r2 = r2_score(y_test, y_pred)
print(f"R-Squared: {r2}")
```

Mean Squared Error: 7300.891505474008

R-Squared: 0.21358449637394128

```
In [28]: # Plot the test data and regression line
plt.figure(figsize=(10, 6))
plt.scatter(X_test['Model Year'], y_test, color='blue', alpha=0.5, label='Test Data')
plt.plot(X_test['Model Year'], predictions, color='red', label='Regression Line')
plt.xlabel('Model Year')
plt.ylabel('Electric Range')
plt.title('Regression Line on Test Data')
plt.legend()
plt.show()
```



