

In [1]:

```

1 import numpy as np
2 import pandas as pd
3 import seaborn as sns
4 import matplotlib.pyplot as plt
5 from sklearn import preprocessing, svm
6 from sklearn.model_selection import train_test_split
7 from sklearn.linear_model import LinearRegression

```

In [2]:

```

1 dt=pd.read_csv(r"C:\Users\HP\Downloads\fiat500_VehicleSelection_Dataset (2).csv")
2 dt

```

Out[2]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon
0	1	lounge	51	882	25000	1	44.907242	8.6115
1	2	pop	51	1186	32500	1	45.666359	12.2418
2	3	sport	74	4658	142228	1	45.503300	11.4178
3	4	lounge	51	2739	160000	1	40.633171	17.6346
4	5	pop	73	3074	106880	1	41.903221	12.4956
...	...	...	...	...	...	...	...	...
1533	1534	sport	51	3712	115280	1	45.069679	7.7049
1534	1535	lounge	74	3835	112000	1	45.845692	8.6668
1535	1536	pop	51	2223	60457	1	45.481541	9.4134
1536	1537	lounge	51	2557	80750	1	45.000702	7.6822
1537	1538	pop	51	1766	54276	1	40.323410	17.5682

1538 rows × 9 columns

In [3]:

```

1 dt=dt[['engine_power','price']]
2 dt.columns=['Engine','Price']

```

In [4]:

```
1 dt.head(10)
```

Out[4]:

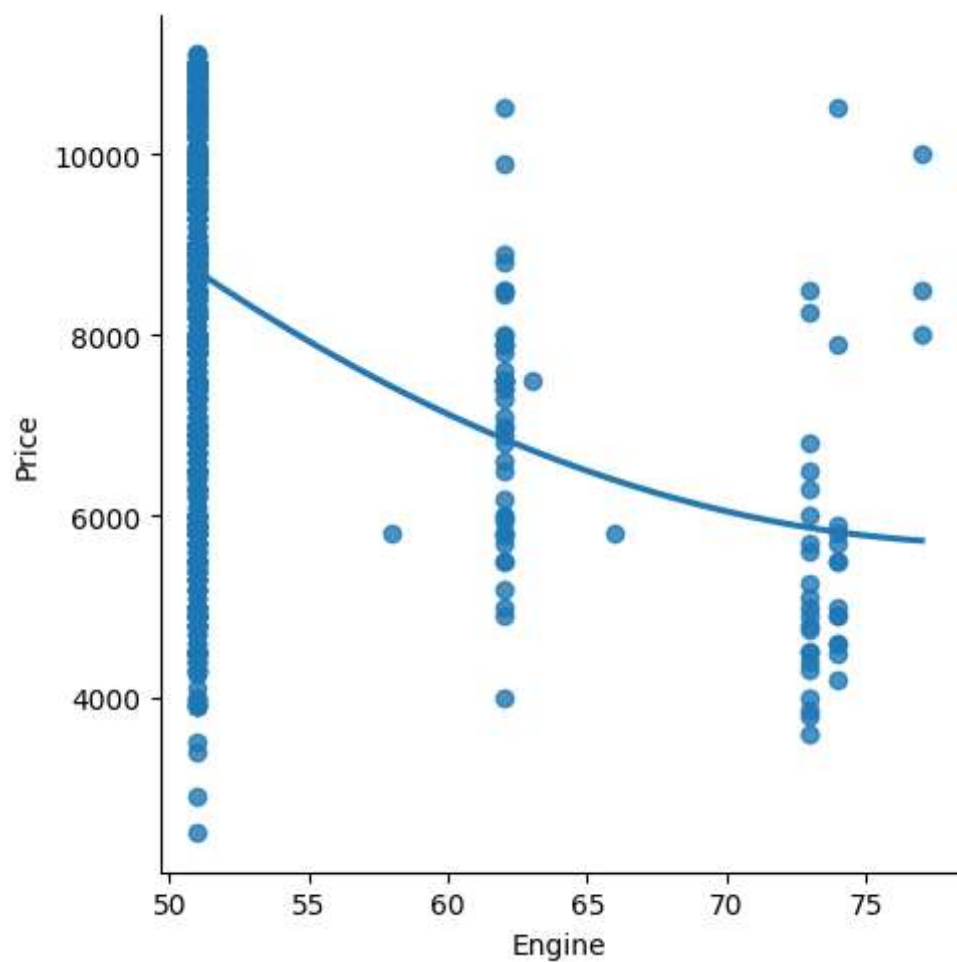
	Engine	Price
0	51	8900
1	51	8800
2	74	4200
3	51	6000
4	73	5700
5	74	7900
6	51	10750
7	51	9190
8	73	5600
9	51	6000

In [5]:

```
1 sns.lmplot(x='Engine',y='Price',data=dt,order=2,ci=None)
```

Out[5]:

&lt;seaborn.axisgrid.FacetGrid at 0x14ca0cb8dd0&gt;



In [6]:

1 dt.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1538 entries, 0 to 1537
Data columns (total 2 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0   Engine  1538 non-null    int64
 1   Price   1538 non-null    int64
dtypes: int64(2)
memory usage: 24.2 KB
```

In [7]:

1 dt.describe()

Out[7]:

	Engine	Price
<b>count</b>	1538.000000	1538.000000
<b>mean</b>	51.904421	8576.003901
<b>std</b>	3.988023	1939.958641
<b>min</b>	51.000000	2500.000000
<b>25%</b>	51.000000	7122.500000
<b>50%</b>	51.000000	9000.000000
<b>75%</b>	51.000000	10000.000000
<b>max</b>	77.000000	11100.000000

In [8]:

1 dt.fillna(method='ffill')

Out[8]:

	Engine	Price
<b>0</b>	51	8900
<b>1</b>	51	8800
<b>2</b>	74	4200
<b>3</b>	51	6000
<b>4</b>	73	5700
...	...	...
<b>1533</b>	51	5200
<b>1534</b>	74	4600
<b>1535</b>	51	7500
<b>1536</b>	51	5990
<b>1537</b>	51	7900

1538 rows × 2 columns

In [9]:

```
1 x=np.array(dt['Engine']).reshape(-1,1)
2 y=np.array(dt['Price']).reshape(-1,1)
```

In [10]:

```
1 dt.dropna(inplace=True)
```

C:\Users\HP\AppData\Local\Temp\ipykernel\_6568\735218168.py:1: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy) ([https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy))

```
dt.dropna(inplace=True)
```

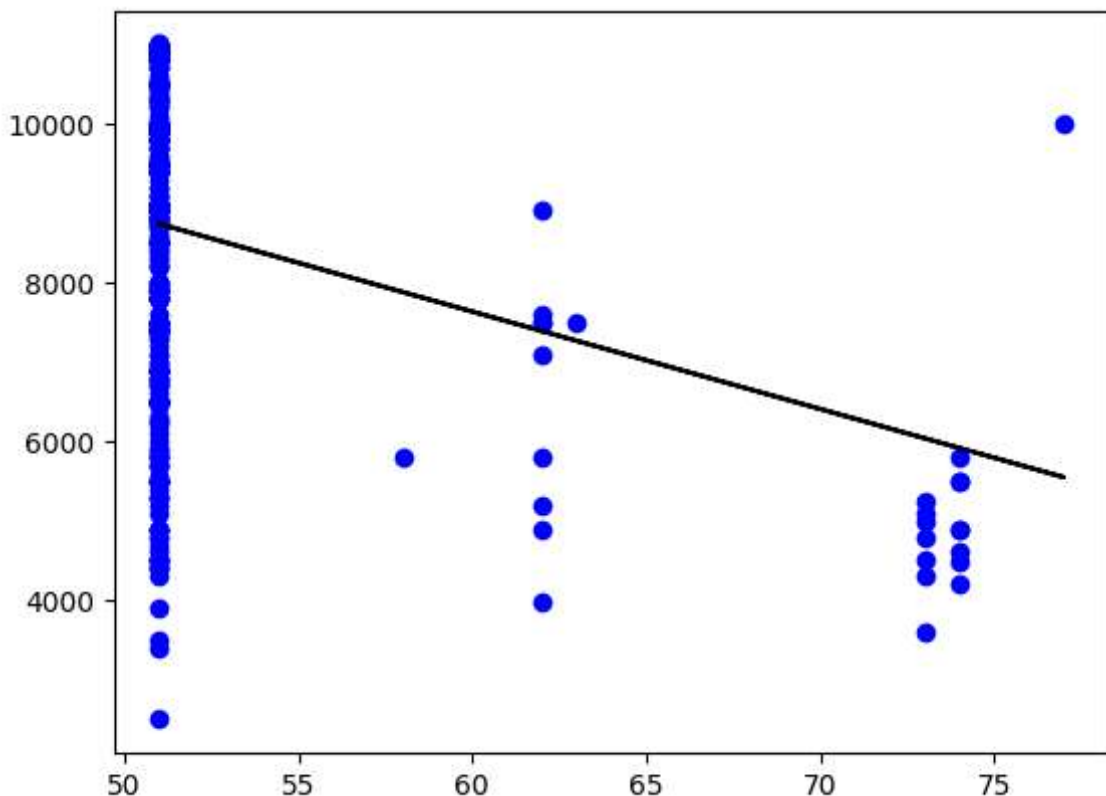
In [11]:

```
1 X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
2 reg=LinearRegression()
3 reg.fit(X_train,y_train)
4 print(reg.score(X_test,y_test))
```

0.11826973051509415

In [12]:

```
1 y_pred=reg.predict(X_test)
2 plt.scatter(X_test,y_test,color='b')
3 plt.plot(X_test,y_pred,color='k')
4 plt.show()
```

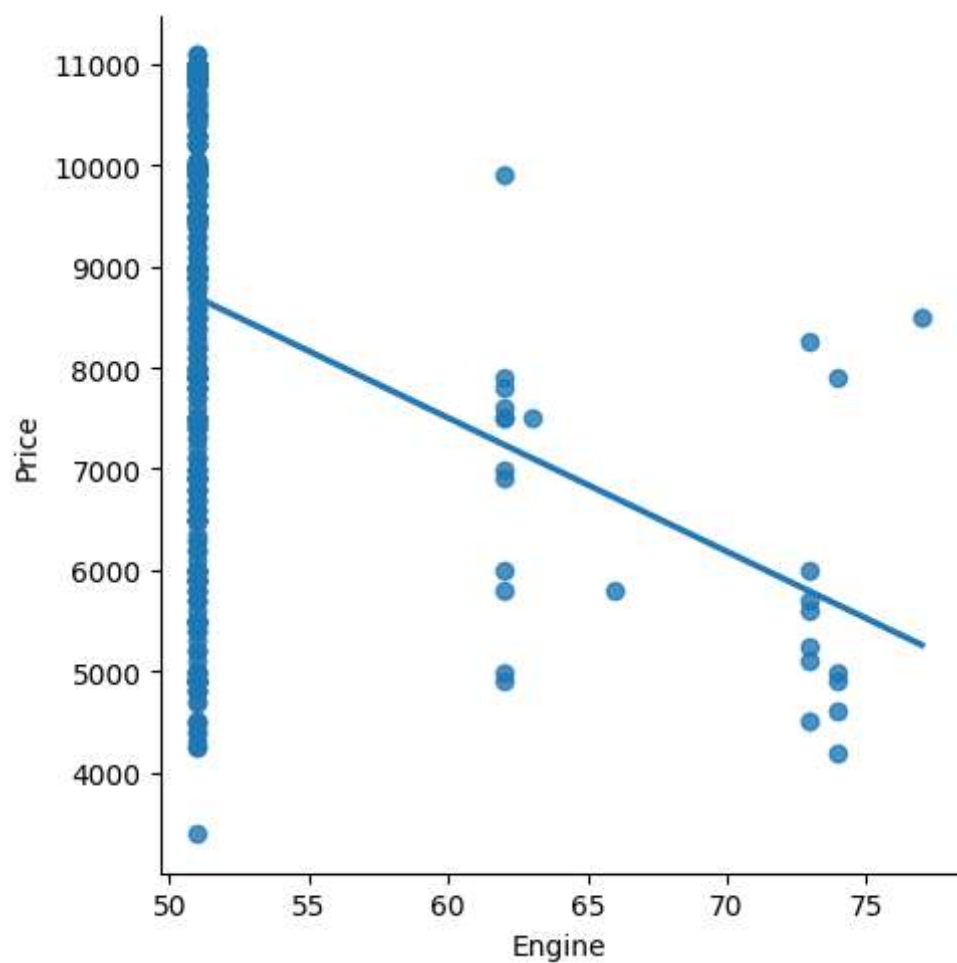


In [13]:

```
1 dt500=dt[:][:500]  
2 sns.lmplot(x="Engine",y="Price",data=dt500,order=1,ci=None)
```

Out[13]:

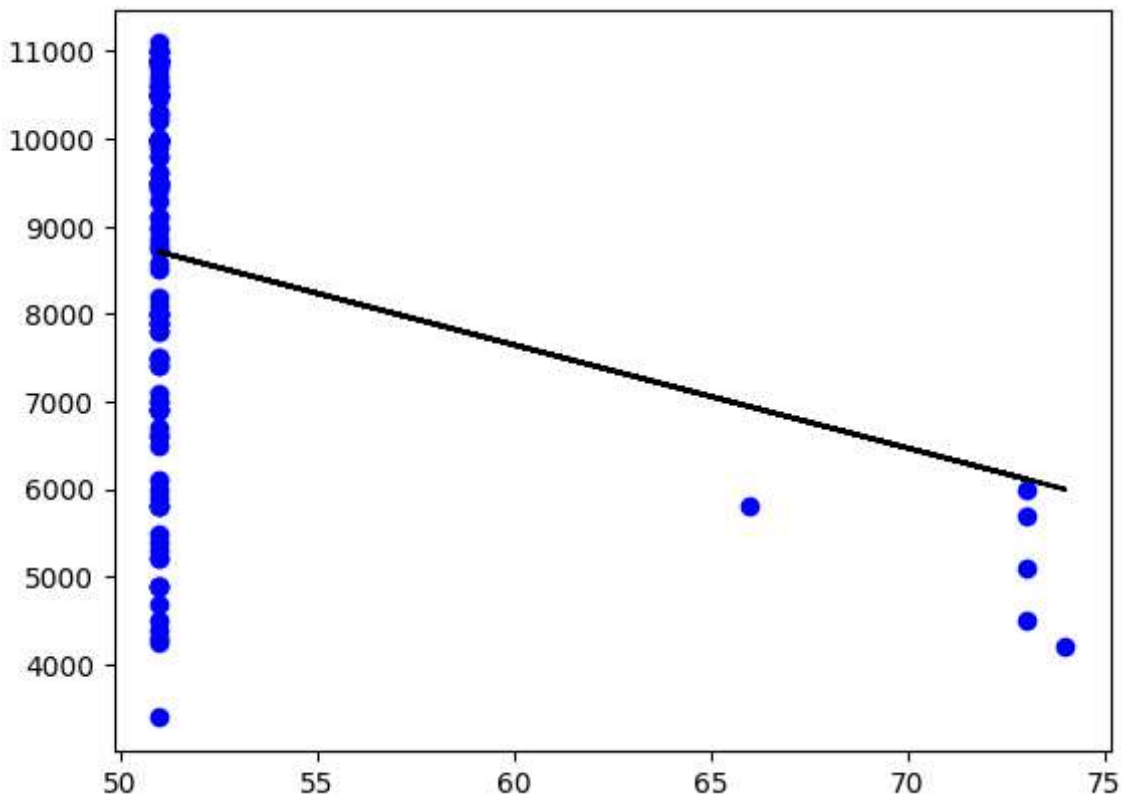
&lt;seaborn.axisgrid.FacetGrid at 0x14c94d42cd0&gt;



In [14]:

```
1 dt500.fillna(method='ffill',inplace=True)
2 X=np.array(dt500['Engine']).reshape(-1,1)
3 y=np.array(dt500['Price']).reshape(-1,1)
4 dt500.dropna(inplace=True)
5 X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.25)
6 reg=LinearRegression()
7 reg.fit(X_train,y_train)
8 print("Regression:",reg.score(X_test,y_test))
9 y_pred=reg.predict(X_test)
10 plt.scatter(X_test,y_test,color="b")
11 plt.plot(X_test,y_pred,color='k')
12 plt.show()
```

Regression: 0.10752350728507254



In [15]:

```
1 from sklearn.linear_model import LinearRegression
2 from sklearn.metrics import r2_score
3 model=LinearRegression()
4 model.fit(X_train,y_train)
5 y_pred=model.predict(X_test)
6 r2=r2_score(y_test,y_pred)
7 print("R2 score:",r2)
```

R2 score: 0.10752350728507254

In [16]:

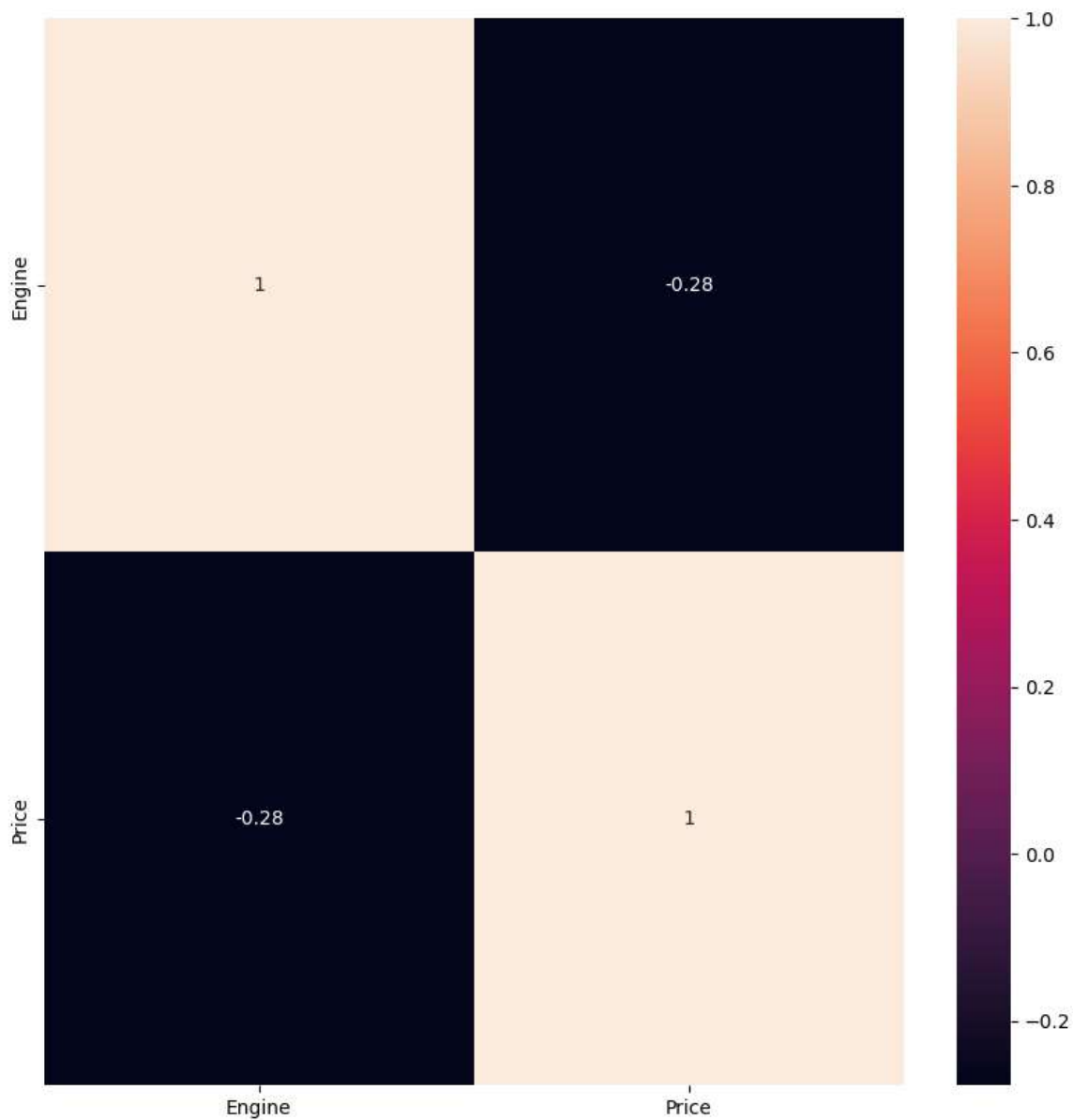
```
1 from sklearn.linear_model import Ridge, RidgeCV, Lasso
2 from sklearn.preprocessing import StandardScaler
```

In [17]:

```
1 plt.figure(figsize = (10, 10))
2 sns.heatmap(dt.corr(), annot = True)
3
```

Out[17]:

&lt;Axes: &gt;



In [18]:

```
1 features = dt.columns[0:2]
2 target = dt.columns[-1]
3 #X and y values
4 X = dt[features].values
5 y = dt[target].values
6 #splot
7 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
8 print("The dimension of X_train is {}".format(X_train.shape))
9 print("The dimension of X_test is {}".format(X_test.shape))
10 #Scale features
11 scaler = StandardScaler()
12 X_train = scaler.fit_transform(X_train)
13 X_test = scaler.transform(X_test)
```

The dimension of X\_train is (1076, 2)

The dimension of X\_test is (462, 2)

In [19]:

```
1 #Model
2 lr = LinearRegression()
3 #Fit model
4 lr.fit(X_train, y_train)
5 #predict
6 #prediction = lr.predict(X_test)
7 #actual
8 actual = y_test
9 train_score_lr = lr.score(X_train, y_train)
10 test_score_lr = lr.score(X_test, y_test)
11 print("\nLinear Regression Model:\n")
12 print("The train score for lr model is {}".format(train_score_lr))
13 print("The test score for lr model is {}".format(test_score_lr))
```

Linear Regression Model:

The train score for lr model is 1.0

The test score for lr model is 1.0



In [20]:

```
1 #Model
2 lr = LinearRegression()
3 #Fit model
4 lr.fit(X_train, y_train)
5 #predict
6 #prediction = lr.predict(X_test)
7 #actual
8 actual = y_test
9 train_score_lr = lr.score(X_train, y_train)
10 test_score_lr = lr.score(X_test, y_test)
11 print("\nLinear Regression Model:\n")
12 print("The train score for lr model is {}".format(train_score_lr))
13 print("The test score for lr model is {}".format(test_score_lr))
```

Linear Regression Model:

The train score for lr model is 1.0  
The test score for lr model is 1.0

In [21]:

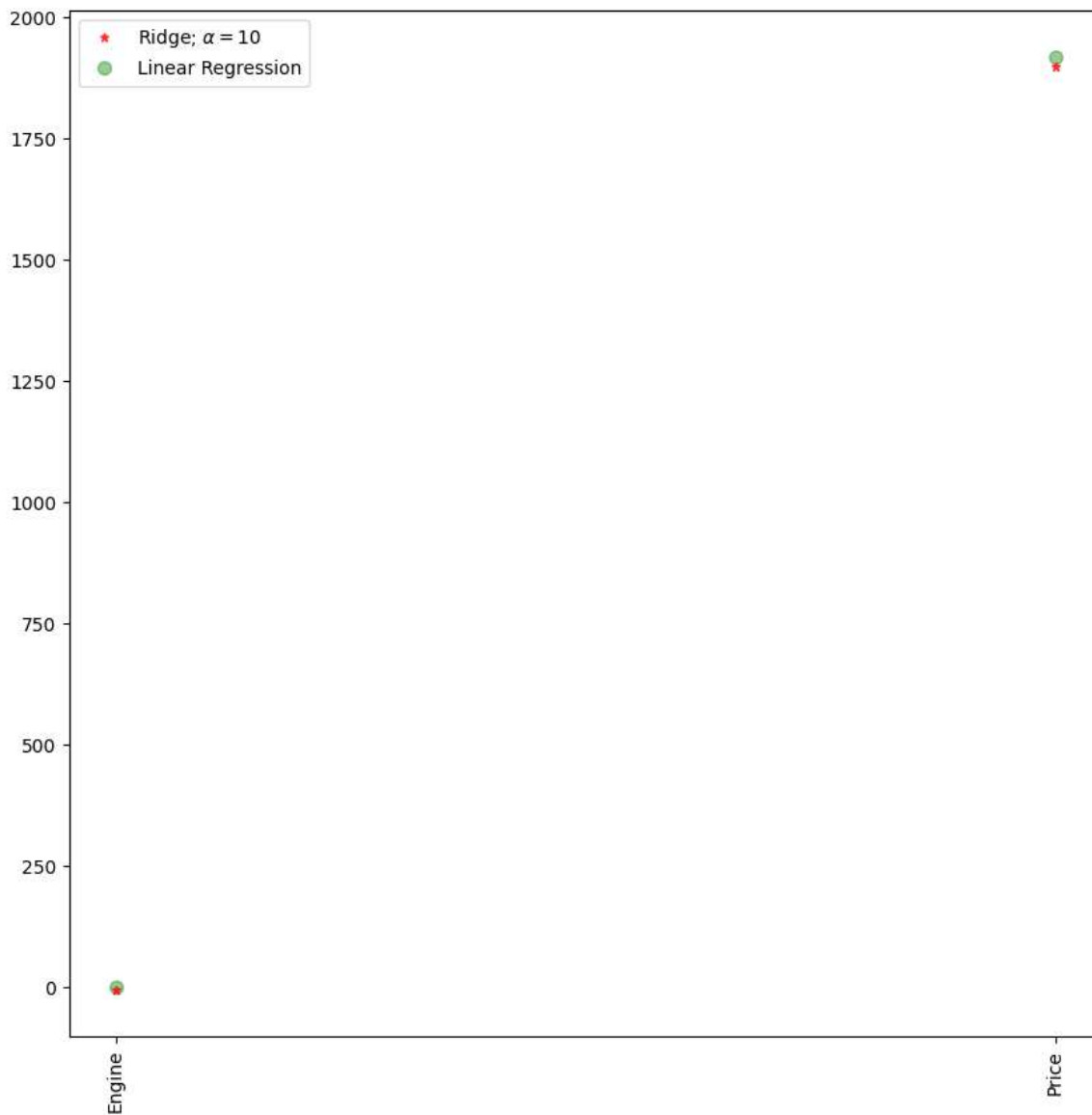
```
1 #Ridge Regression Model
2 ridgeReg = Ridge(alpha=10)
3 ridgeReg.fit(X_train,y_train)
4 #train and test scorefor ridge regression
5 train_score_ridge = ridgeReg.score(X_train, y_train)
6 test_score_ridge = ridgeReg.score(X_test, y_test)
7 print("\nRidge Model:\n")
8 print("The train score for ridge model is {}".format(train_score_ridge))
9 print("The test score for ridge model is {}".format(test_score_ridge))
10
```

Ridge Model:

The train score for ridge model is 0.9999088581979684  
The test score for ridge model is 0.9999100853681022

In [22]:

```
1 plt.figure(figsize = (10, 10))
2 plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,color='red')
3 #plt.plot(rr100.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6,color='blue')
4 plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='green')
5 plt.xticks(rotation = 90)
6 plt.legend()
7 plt.show()
```



In [23]:

```
1 #Lasso regression model
2 print("\nLasso Model: \n")
3 lasso = Lasso(alpha = 10)
4 lasso.fit(X_train,y_train)
5 train_score_ls =lasso.score(X_train,y_train)
6 test_score_ls =lasso.score(X_test,y_test)
7 print("The train score for ls model is {}".format(train_score_ls))
8 print("The test score for ls model is {}".format(test_score_ls))
9
```

Lasso Model:

The train score for ls model is 0.9999728562194999

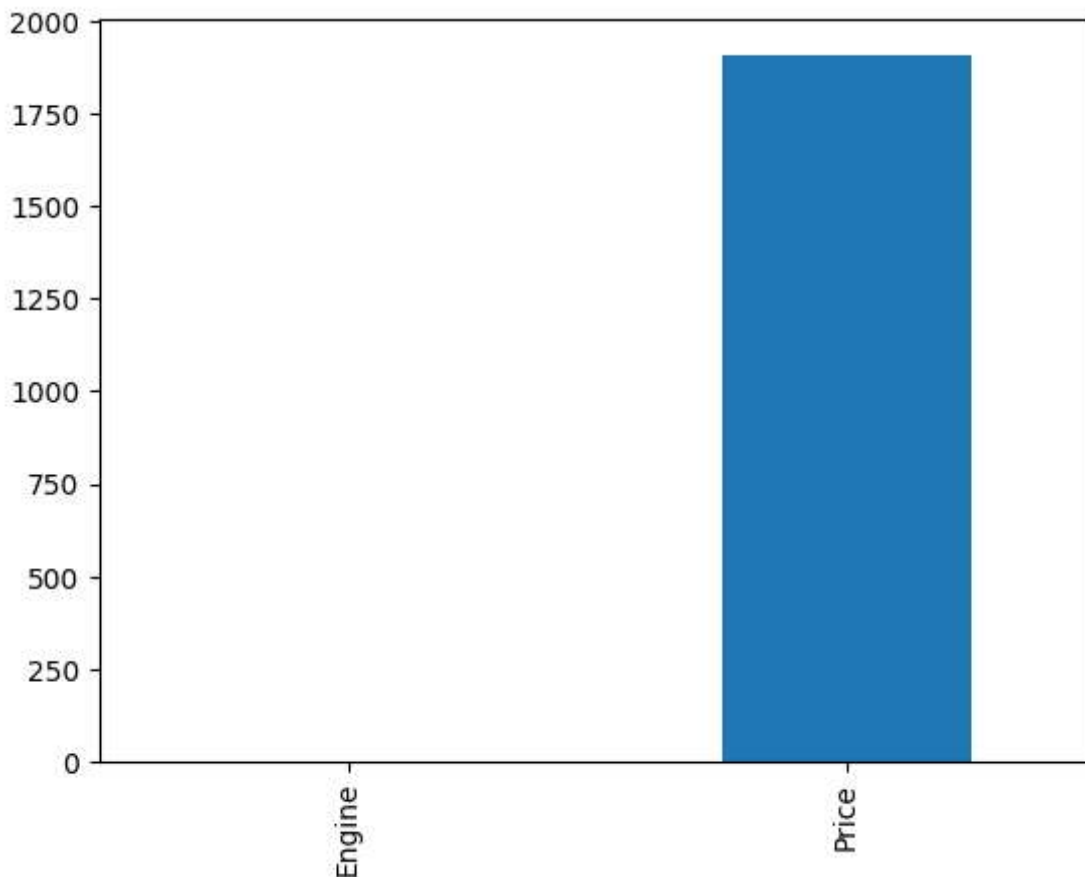
The test score for ls model is 0.9999728508562553

In [24]:

```
1 pd.Series(lasso.coef_, features).sort_values(ascending = True).plot(kind = "bar")
```

Out[24]:

&lt;Axes: &gt;



In [25]:

```
1 #Using the linear CV model
2 from sklearn.linear_model import LassoCV
3 #Lasso Cross validation
4 lasso_cv = LassoCV(alphas = [0.0001, 0.001,0.01, 0.1, 1, 10], random_state=0).fit(X_train, y_train)
5 #score
6 print(lasso_cv.score(X_train, y_train))
7 print(lasso_cv.score(X_test, y_test))
8
```

0.999999999501757

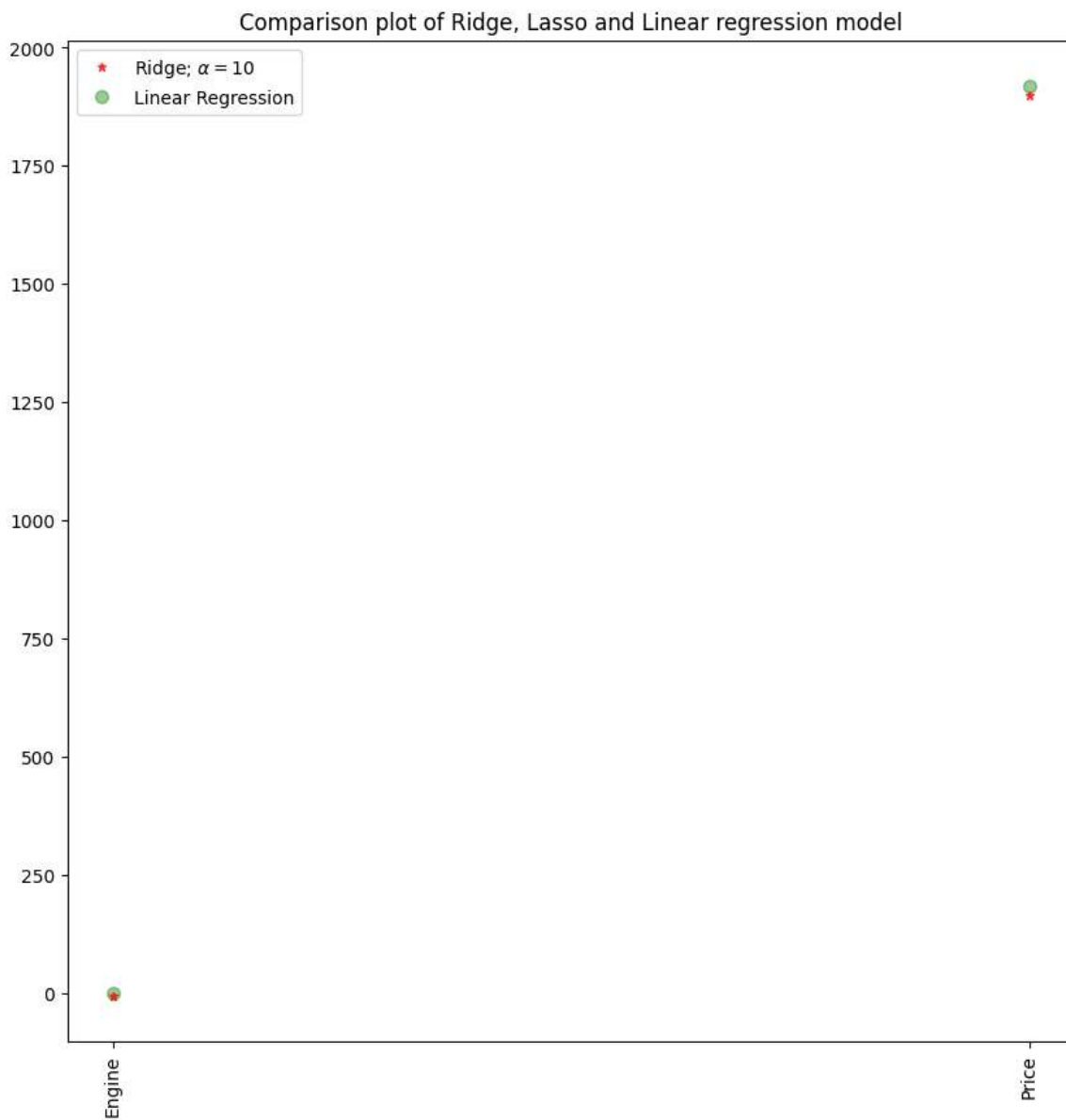
0.999999999638806

In [26]:

```

1  #plot size
2  plt.figure(figsize = (10, 10))
3  #add plot for ridge regression
4  plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,color='red')
5  #add plot for lasso regression
6  #plt.plot(lasso_cv.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6,color='blue')
7  #add plot for linear model
8  plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='green')
9  #rotate axis
10 plt.xticks(rotation = 90)
11 plt.legend()
12 plt.title("Comparison plot of Ridge, Lasso and Linear regression model")
13 plt.show()
14

```



In [27]:

```
1 #Using the linear CV model
2 from sklearn.linear_model import RidgeCV
3 #Ridge Cross validation
4 ridge_cv = RidgeCV(alphas = [0.0001, 0.001,0.01, 0.1, 1, 10]).fit(X_train, y_train)
5 #score
6 print("The train score for ridge model is {}".format(ridge_cv.score(X_train, y_train)))
7 print("The train score for ridge model is {}".format(ridge_cv.score(X_test, y_test)))
```

The train score for ridge model is 0.9999999999999899

The train score for ridge model is 0.9999999999999901

In [28]:

```
1 from sklearn.linear_model import ElasticNet
2 regr=ElasticNet()
3 regr.fit(X,y)
4 print(regr.coef_)
5 print(regr.intercept_)
```

[-0. 0.99999973]

0.0022802498624514556

In [29]:

```
1 y_pred_elastic=regr.predict(X_train)
2
```

In [30]:

```
1 mean_squared_error=np.mean((y_pred_elastic-y_train)**2)
2 print("Mean Squared Error on test set",mean_squared_error)
```

Mean Squared Error on test set 77371869.93693775

In [ ]:

```
1
```