```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn import preprocessing, svm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

| Out[2]: | | ID | model | engine_power | age_in_days | km | previous_owners | lat | lon |
|---------|------|------|--------|--------------|-------------|--------|-----------------|-----------|-----------|
| | 0 | 1 | lounge | 51 | 882 | 25000 | 1 | 44.907242 | 8.611560 |
| | 1 | 2 | рор | 51 | 1186 | 32500 | 1 | 45.666359 | 12.241890 |
| | 2 | 3 | sport | 74 | 4658 | 142228 | 1 | 45.503300 | 11.417840 |
| | 3 | 4 | lounge | 51 | 2739 | 160000 | 1 | 40.633171 | 17.634609 |
| | 4 | 5 | рор | 73 | 3074 | 106880 | 1 | 41.903221 | 12.495650 |
| | | | | | | | | | |
| | 1533 | 1534 | sport | 51 | 3712 | 115280 | 1 | 45.069679 | 7.704920 |
| | 1534 | 1535 | lounge | 74 | 3835 | 112000 | 1 | 45.845692 | 8.666870 |
| | 1535 | 1536 | pop | 51 | 2223 | 60457 | 1 | 45.481541 | 9.413480 |
| | 1536 | 1537 | lounge | 51 | 2557 | 80750 | 1 | 45.000702 | 7.682270 |
| | 1537 | 1538 | рор | 51 | 1766 | 54276 | 1 | 40.323410 | 17.568270 |
| | | | | | | | | | |

1538 rows × 9 columns

```
In [3]: dt=dt[['engine_power','price']]
    dt.columns=['Engine','Pric']
```

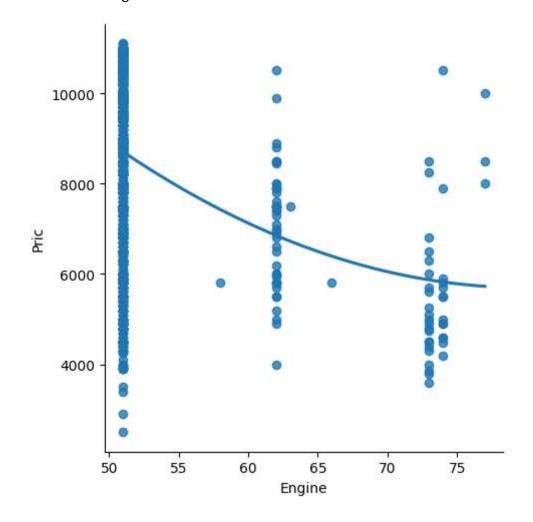
In [4]: dt.head(10)

| | | _ | |
|-------------|---|------|----|
| \triangle | + | Γ⁄Ι] | ١. |
| υu | L | 4 | ١. |

| | Engine | Pric |
|---|--------|-------|
| 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]: sns.lmplot(x='Engine',y='Pric',data=dt,order=2,ci=None)

Out[5]: <seaborn.axisgrid.FacetGrid at 0x2f647c4cfd0>



In [7]: dt.describe()

Out[7]:

| | Engine | Pric |
|-------|-------------|--------------|
| count | 1538.000000 | 1538.000000 |
| mean | 51.904421 | 8576.003901 |
| std | 3.988023 | 1939.958641 |
| min | 51.000000 | 2500.000000 |
| 25% | 51.000000 | 7122.500000 |
| 50% | 51.000000 | 9000.000000 |
| 75% | 51.000000 | 10000.000000 |
| max | 77.000000 | 11100.000000 |

In [8]: | dt.fillna(method='ffill')

Out[8]:

| | Engine | Pric |
|------|--------|------|
| 0 | 51 | 8900 |
| 1 | 51 | 8800 |
| 2 | 74 | 4200 |
| 3 | 51 | 6000 |
| 4 | 73 | 5700 |
| | | |
| 1533 | 51 | 5200 |
| 1534 | 74 | 4600 |
| 1535 | 51 | 7500 |
| 1536 | 51 | 5990 |
| 1537 | 51 | 7900 |
| | | |

1538 rows × 2 columns

```
In [9]: x=np.array(dt['Engine']).reshape(-1,1)
y=np.array(dt['Pric']).reshape(-1,1)
```

In [10]: dt.dropna(inplace=True)

C:\Users\91903\AppData\Local\Temp\ipykernel_5464\735218168.py:1: SettingWithC
opyWarning:

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

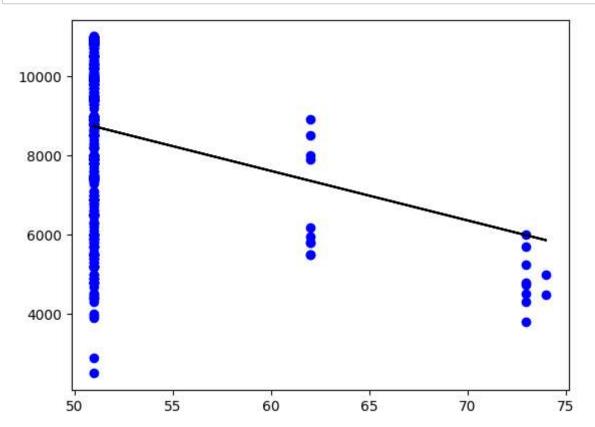
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

dt.dropna(inplace=True)

```
In [11]: X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
    reg=LinearRegression()
    reg.fit(X_train,y_train)
    print(reg.score(X_test,y_test))
```

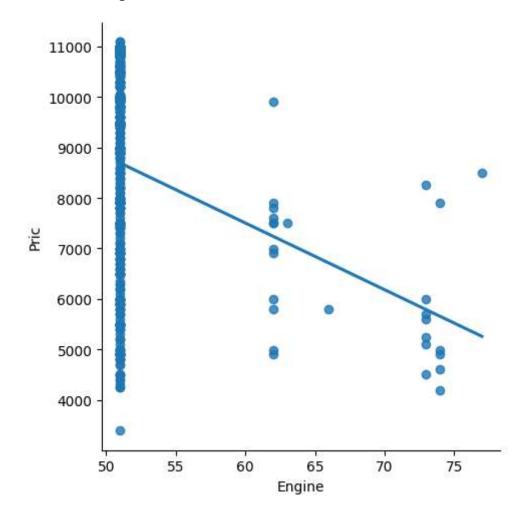
0.09048902894808675

```
In [12]: y_pred=reg.predict(X_test)
    plt.scatter(X_test,y_test,color='b')
    plt.plot(X_test,y_pred,color='k')
    plt.show()
```



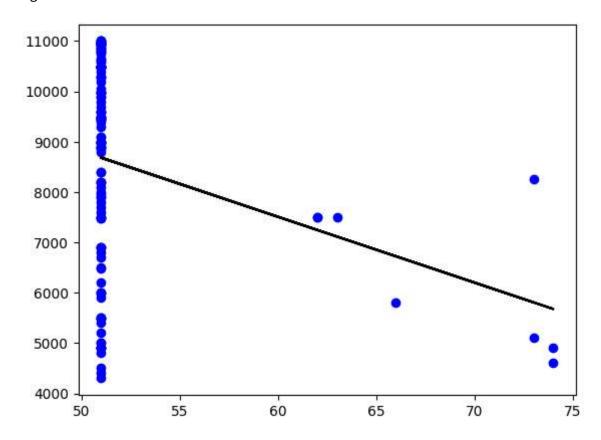
```
In [13]: dt500=dt[:][:500]
sns.lmplot(x="Engine",y="Pric",data=dt500,order=1,ci=None)
```

Out[13]: <seaborn.axisgrid.FacetGrid at 0x2f6359d0550>



```
In [14]: dt500.fillna(method='ffill',inplace=True)
    X=np.array(dt500['Engine']).reshape(-1,1)
    y=np.array(dt500['Pric']).reshape(-1,1)
    dt500.dropna(inplace=True)
    X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.25)
    reg=LinearRegression()
    reg.fit(X_train,y_train)
    print("Regression:",reg.score(X_test,y_test))
    y_pred=reg.predict(X_test)
    plt.scatter(X_test,y_test,color="b")
    plt.plot(X_test,y_pred,color='k')
    plt.show()
```

Regression: 0.08585204069415087



```
In [15]: from sklearn.linear_model import LinearRegression
    from sklearn.metrics import r2_score
    mode1=LinearRegression()
    mode1.fit(X_train,y_train)
    y_pred=mode1.predict(X_test)
    r2=r2_score(y_test,y_pred)
    print("R2 score:",r2)
```

R2 score: 0.08585204069415087

#conclusion: Linear regression is not fit for the model

Ridge and Lasso Regression

```
In [16]: from sklearn.preprocessing import StandardScaler
         from sklearn.linear_model import Ridge
         from sklearn.linear_model import RidgeCV
         from sklearn.linear_model import Lasso
In [17]: |plt.figure(figsize = (10, 10))
         sns.heatmap(dt.corr(), annot = True)
Out[17]: <Axes: >
                                                                                 - 1.0
                                                                                 - 0.8
                           1
                                                         -0.28
                                                                                 - 0.6
                                                                                  0.4
In [19]: features = dt.columns[0:2]
         target = dt.columns[-1]
         #X and y values
         X = dt[features].values
         y = dt[target].values
         #splot
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, rando
         print("The dimension of X train is {}".format(X train.shape))
         print("The dimension of X_test is {}".format(X_test.shape))
         #Scale features
         scaler = StandardScaler()
         X_train = scaler.fit_transform(X_train)
         X_test = scaler.transform(X_test)
         The dimension of X_train is (1076, 2)
```

The dimension of X_test is (462, 2)

```
In [20]: lr = LinearRegression()
#Fit model
lr.fit(X_train, y_train)
#predict
#prediction = lr.predict(X_test)
#actual
actual = y_test
train_score_lr = lr.score(X_train, y_train)
test_score_lr = lr.score(X_test, y_test)
print("\nLinear Regression Model:\n")
print("The train score for lr model is {}".format(train_score_lr))
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]: ridgeReg = Ridge(alpha=10)
    ridgeReg.fit(X_train,y_train)
    #train and test scorefor ridge regression
    train_score_ridge = ridgeReg.score(X_train, y_train)
    test_score_ridge = ridgeReg.score(X_test, y_test)
    print("\nRidge Model:\n")
    print("The train score for ridge model is {}".format(train_score_ridge))
    print("The test score for ridge model is {}".format(test_score_ridge))
```

Ridge Model:

The train score for ridge model is 0.9999088581979684 The test score for ridge model is 0.9999100853681023

```
plt.figure(figsize = (10, 10))
In [22]:
          plt.plot(features, ridgeReg.coef_, alpha=0.7, linestyle='none', marker='*', markers
          plt.plot(features, lr.coef_, alpha=0.4, linestyle='none', marker='o', markersize=7,
          plt.xticks(rotation = 90)
          plt.legend()
          plt.show()
            2000
                      Ridge; \alpha = 10
                                                                                              9
                     Linear Regression
            1750
            1500
            1250
            1000
            750
            500
            250
              0
                                                                                             Pric
```

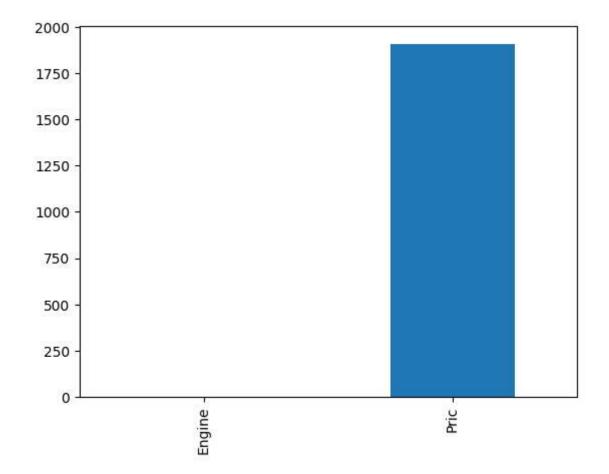
```
In [23]: print("\nLasso Model: \n")
  lasso = Lasso(alpha = 10)
  lasso.fit(X_train,y_train)
  train_score_ls =lasso.score(X_train,y_train)
  test_score_ls =lasso.score(X_test,y_test)
  print("The train score for ls model is {}".format(train_score_ls))
  print("The train score for ls model is {}".format(test_score_ls))
```

Lasso Model:

The train score for ls model is 0.9999728562194999 The train score for ls model is 0.9999728508562553

In [24]: pd.Series(lasso.coef_, features).sort_values(ascending = True).plot(kind = "ba

Out[24]: <Axes: >



```
In [25]: #Using the Linear CV model
    from sklearn.linear_model import LassoCV
    #Lasso Cross validation
    lasso_cv = LassoCV(alphas = [0.0001, 0.001, 0.1, 1, 10], random_state=0).
    #score
    print(lasso_cv.score(X_train, y_train))
    print(lasso_cv.score(X_test, y_test))
```

- 0.999999999501757
- 0.999999999638806

```
In [26]:
          plt.figure(figsize = (10, 10))
          plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markers
          plt.plot(features, ridgeReg.coef_,alpha=0.6,linestyle='none',marker='d',markers
          plt.plot(features, lr.coef_, alpha=0.4, linestyle='none', marker='o', markersize=7,
          plt.xticks(rotation = 90)
          plt.legend()
          plt.show()
           2000
                     Ridge; \alpha = 10
                     Ridge; \alpha = grid
                     Linear Regression
           1750
           1500
           1250
           1000
            750
            500
            250
              0
```

ElasticNet Regression