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

In [2]: df = pd.read_csv("1_fiat500_VehicleSelection_Dataset.csv")[0:1500].dropna(axised from the selection_Dataset.csv")

L									
Out[2]:	ID mod		model	engine_power	age_in_days	km	previous_owners	lat	
	0	1.0	lounge	51.0	882.0	25000.0	1.0	44.907242	8.61155
	1	2.0	рор	51.0	1186.0	32500.0	1.0	45.666359	12.2418
	2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.4
	3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.6346
	4	5.0	рор	73.0	3074.0	106880.0	1.0	41.903221	12.4956
	1495	1496.0	pop	62.0	3347.0	0.00008	3.0	44.283878	11.8881
	1496	1497.0	pop	51.0	1461.0	91055.0	3.0	44.508839	11.4690
	1497	1498.0	lounge	51.0	397.0	15840.0	3.0	38.122070	13.3611
	1498	1499.0	sport	51.0	1400.0	60000.0	1.0	45.802021	9.18778
	1499	1500.0	pop	51.0	1066.0	53100.0	1.0	38.122070	13.3611

1500 rows × 9 columns

In [3]: df.head()

[3]:		ID	model	engine_power	age_in_days	km	previous_owners	lat	lon
	0	1.0	lounge	51.0	882.0	25000.0	1.0	44.907242	8.611559868
	1	2.0	рор	51.0	1186.0	32500.0	1.0	45.666359	12.24188995
	2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.41784
	3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.63460922
	4	5.0	рор	73.0	3074.0	106880.0	1.0	41.903221	12.49565029
	. =								

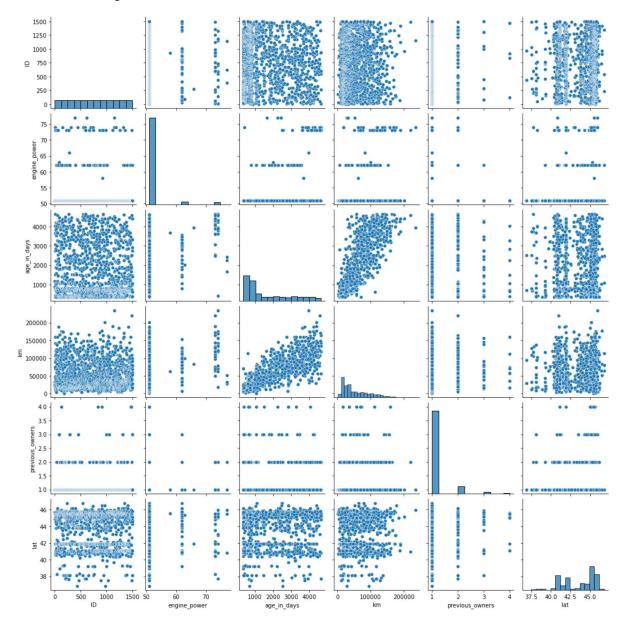
Data cleaning and pre processing

```
In [4]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 1500 entries, 0 to 1499
         Data columns (total 9 columns):
          #
               Column
                                  Non-Null Count
                                                   Dtype
               _ _ _ _ _
                                  -----
                                                    ----
          0
               ID
                                                   float64
                                  1500 non-null
          1
               model
                                  1500 non-null
                                                   object
               engine_power
                                  1500 non-null
                                                   float64
          2
                                                   float64
          3
               age_in_days
                                  1500 non-null
          4
                                  1500 non-null
                                                   float64
          5
                                 1500 non-null
                                                   float64
               previous owners
          6
               lat
                                  1500 non-null
                                                   float64
          7
               lon
                                  1500 non-null
                                                   object
          8
                                  1500 non-null
                                                    object
               price
         dtypes: float64(6), object(3)
         memory usage: 105.6+ KB
In [5]:
         df.describe()
Out[5]:
                         ID
                            engine_power
                                                                                            lat
                                          age_in_days
                                                                    previous_owners
          count 1500.000000
                              1500.000000
                                          1500.000000
                                                        1500.000000
                                                                        1500.000000 1500.000000
                 750.500000
                                51.875333
                                          1641.629333
                                                       53074.900000
                                                                           1.126667
                                                                                      43.545904
          mean
                                                       39955.013731
            std
                 433.157015
                                 3.911606
                                          1288.091104
                                                                           0.421197
                                                                                       2.112907
                                           366.000000
           min
                   1.000000
                                51.000000
                                                        1232.000000
                                                                           1.000000
                                                                                      36.855839
           25%
                 375.750000
                                51.000000
                                           670.000000
                                                       20000.000000
                                                                           1.000000
                                                                                      41.802990
           50%
                 750.500000
                                51.000000
                                          1035.000000
                                                       38720.000000
                                                                           1.000000
                                                                                      44.360376
           75%
                1125.250000
                                          2616.000000
                                                                           1.000000
                                                                                      45.467960
                                51.000000
                                                       78170.250000
           max 1500.000000
                                77.000000
                                          4658.000000 235000.000000
                                                                           4.000000
                                                                                      46.795612
In [6]: | df.columns
Out[6]: Index(['ID', 'model', 'engine_power', 'age_in_days', 'km', 'previous_owners',
                 'lat', 'lon', 'price'],
                dtype='object')
```

EDA and VISUALIZATION

In [7]: sns.pairplot(df)

Out[7]: <seaborn.axisgrid.PairGrid at 0x26b9813ad90>

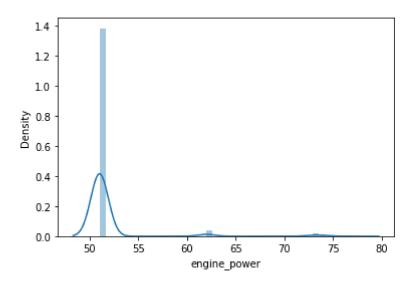


```
In [8]: | sns.distplot(df["engine_power"])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: Fut ureWarning: `distplot` is a deprecated function and will be removed in a futu re version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

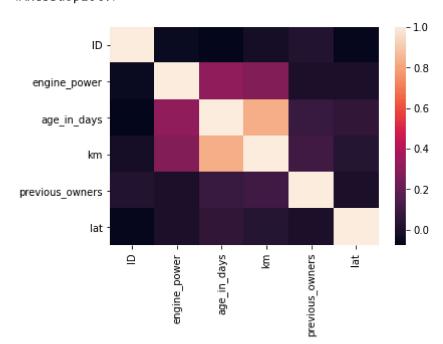
warnings.warn(msg, FutureWarning)

Out[8]: <AxesSubplot:xlabel='engine_power', ylabel='Density'>



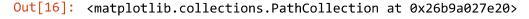
In [10]: | sns.heatmap(df1.corr())

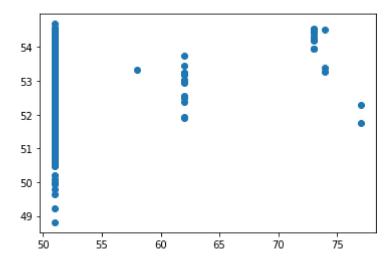
Out[10]: <AxesSubplot:>



split the data into training and test data

```
In [12]: x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.3)
In [13]: | lr = LinearRegression()
          lr.fit(x_train, y_train)
Out[13]: LinearRegression()
In [14]: |lr.intercept_
Out[14]: 52.76671328218245
         coeff = pd.DataFrame(lr.coef_, x.columns, columns =['Co-efficient'])
In [15]:
          coeff
Out[15]:
                          Co-efficient
                      ID
                           -0.000219
                            0.000745
              age_in_days
                            0.000005
                      km
          previous_owners
                           -0.581619
                           -0.036866
                      lat
         prediction = lr.predict(x_test)
In [16]:
          plt.scatter(y_test, prediction)
```





```
In [17]: lr.score(x_test,y_test)
```

Out[17]: 0.107544924049784