#### SUMESH R -20104169

```
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(axis="columns")
                                                                     previous_owners
                               engine_power age_in_days
                                                                                            lat
Out[2]:
                   ID
                        model
                                                                                                         lon
                                                                                                              p
             0
                                                            25000.0
                                                                                 1.0 44.907242 8.611559868
                                                    882.0
                                                                                                              8
                   1.0
                       lounge
                                         51.0
             1
                   2.0
                                         51.0
                                                   1186.0
                                                            32500.0
                                                                                 1.0 45.666359
                                                                                                12.24188995
                                                                                                              8
                          pop
             2
                   3.0
                                         74.0
                                                                                 1.0 45.503300
                                                   4658.0
                                                           142228.0
                                                                                                    11.41784
                         sport
                                                                                                              4
             3
                                         51.0
                                                   2739.0
                                                           160000.0
                                                                                     40.633171
                                                                                                17.63460922
                                                                                                              6
                   4.0
                       lounge
             4
                   5.0
                                        73.0
                                                   3074.0
                                                           106880.0
                                                                                 1.0
                                                                                     41.903221
                                                                                                12.49565029
                                                                                                              5
                          pop
          1495 1496.0
                                         62.0
                                                            80000.0
                                                                                     44.283878 11.88813972
                                                                                                              7
                                                   3347.0
                                                                                 3.0
                          pop
          1496 1497.0
                                         51.0
                                                            91055.0
                                                                                 3.0 44.508839
                                                                                                              7
                          pop
                                                   1461.0
                                                                                                11.46907997
          1497 1498.0
                                         51.0
                                                    397.0
                                                            15840.0
                                                                                 3.0 38.122070 13.36112022 10
                       lounge
          1498 1499.0
                         sport
                                         51.0
                                                   1400.0
                                                            60000.0
                                                                                 1.0 45.802021
                                                                                                9.187789917 10
          1499 1500.0
                                         51.0
                                                   1066.0
                                                            53100.0
                                                                                 1.0 38.122070 13.36112022
                                                                                                             3
                          pop
         1500 rows × 9 columns
In [3]:
          df.head()
Out[3]:
                 model engine power age in days
                                                         km
                                                              previous_owners
                                                                                      lat
                                                                                                       price
            1.0
                 lounge
                                  51.0
                                              882.0
                                                     25000.0
                                                                               44.907242 8.611559868
                                                                                                       8900
            2.0
                                  51.0
                                             1186.0
                                                     32500.0
                                                                               45.666359 12.24188995
                                                                                                       8800
                    pop
             3.0
                  sport
                                  74.0
                                             4658.0
                                                    142228.0
                                                                               45.503300
                                                                                             11.41784
                                                                                                       4200
            4.0
                 lounge
                                  51.0
                                             2739.0
                                                    160000.0
                                                                               40.633171
                                                                                          17.63460922
                                                                                                       6000
            5.0
                                  73.0
                                             3074.0 106880.0
                                                                           1.0 41.903221 12.49565029
                    pop
```

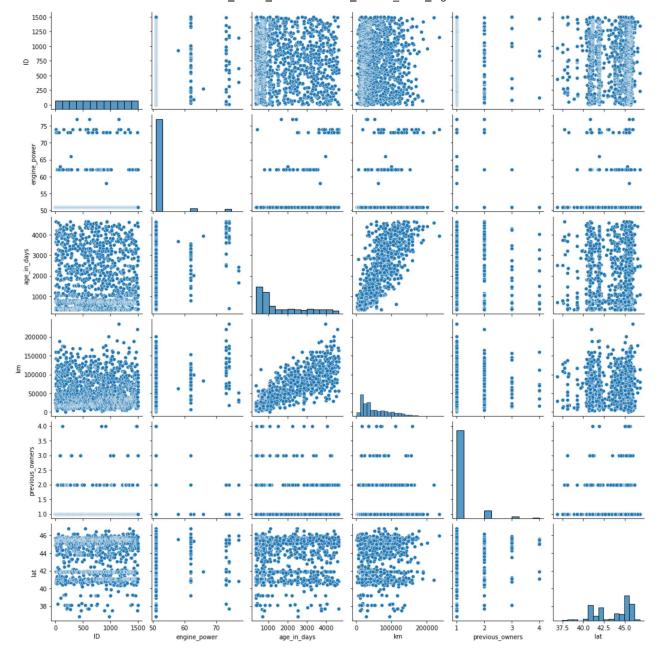
# Data cleaning and pre processing

```
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
         #
             ID
         0
                               1500 non-null
                                               float64
         1
             model
                               1500 non-null
                                               object
                               1500 non-null
                                               float64
         2
             engine power
                                               float64
         3
                               1500 non-null
             age in days
         4
                               1500 non-null
                                               float64
         5
                                               float64
             previous owners
                               1500 non-null
         6
                               1500 non-null
                                               float64
         7
                               1500 non-null
                                               object
             lon
         8
             price
                               1500 non-null
                                               object
        dtypes: float64(6), object(3)
        memory usage: 105.6+ KB
In [5]:
         df.describe()
Out[5]:
                          engine_power age_in_days
                                                           km
                                                                previous_owners
                                                                                       lat
         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
                433.157015
                               3.911606
                                      1288.091104
                                                    39955.013731
                                                                      0.421197
                                                                                  2.112907
          std
                 1.000000
                              51.000000
                                        366.000000
                                                    1232.000000
                                                                      1.000000
                                                                                 36.855839
          min
          25%
                375.750000
                              51.000000
                                        670.000000
                                                    20000.000000
                                                                      1.000000
                                                                                 41.802990
                750.500000
                                                    38720.000000
          50%
                              51.000000
                                      1035.000000
                                                                      1.000000
                                                                                 44.360376
              1125.250000
                              51.000000
                                       2616.000000
                                                    78170.250000
                                                                      1.000000
                                                                                 45.467960
          max 1500.000000
                              77.000000 4658.000000 235000.000000
                                                                      4.000000
                                                                                 46.795612
In [6]:
         df.columns
dtype='object')
```

## **EDA and VISUALIZATION**

```
In [7]: sns.pairplot(df)
```

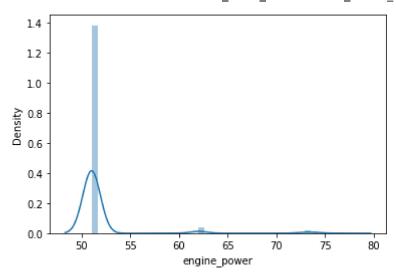
Out[7]: <seaborn.axisgrid.PairGrid at 0x26b9813ad90>



In [8]: sns.distplot(df["engine\_power"])

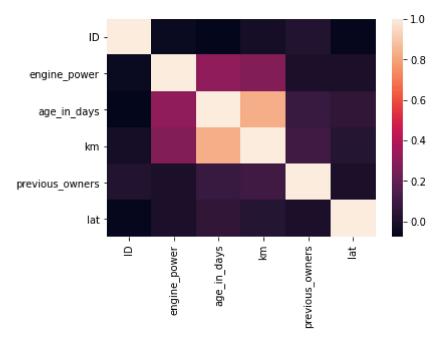
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning:
 `distplot` is a deprecated function and will be removed in a future version. Please adap
 t your code to use either `displot` (a figure-level function with similar flexibility) o
 r `histplot` (an axes-level function 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_
          52.76671328218245
Out[14]:
In [15]:
           coeff = pd.DataFrame(lr.coef_, x.columns, columns =['Co-efficient'])
                          Co-efficient
Out[15]:
                      ID
                            -0.000219
              age_in_days
                             0.000745
                      km
                             0.000005
          previous_owners
                            -0.581619
                      lat
                            -0.036866
In [16]:
           prediction = lr.predict(x_test)
           plt.scatter(y_test, prediction)
          <matplotlib.collections.PathCollection at 0x26b9a027e20>
Out[16]:
          54
          53
          52
          51
          49
             50
                      55
                               60
                                       65
                                                70
                                                         75
```

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

0.107544924049784 Out[17]: