A real estate agent want help to predict the house price for regions in Usa.he gave us the dataset to work on to use linear Regression model.Create a model that helps him to estimate

Data Collection

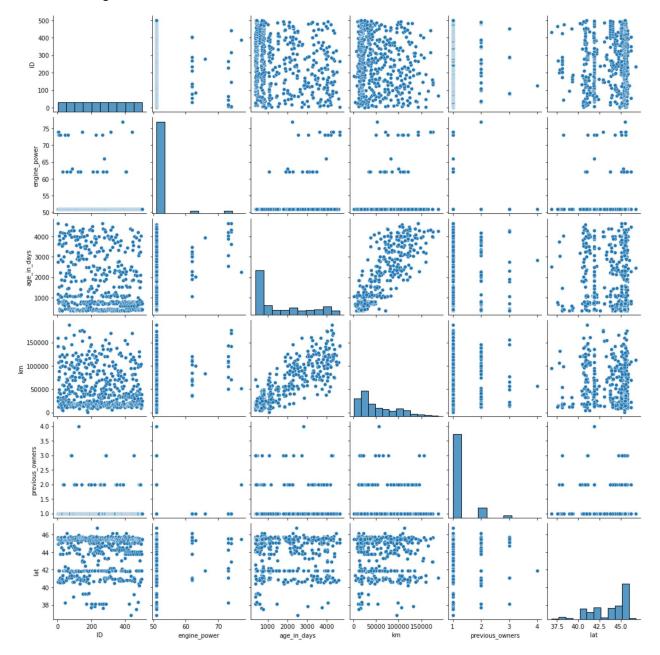
```
#import Libraries
In [1]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
In [22]:
         #import the dataset
          data=pd.read csv(r"C:\Users\user\Desktop\Vicky\1 fiat500 VehicleSelection Dataset.csv")[0:500
In [23]:
         #to display top 10 rows
          data.head()
Out[23]:
                                                                                                   Unnamed: L
                model engine_power age_in_days
                                                     km previous_owners
                                                                              lat
                                                                                         Ion price
          0 1.0
                 lounge
                                51.0
                                           882.0
                                                 25000.0
                                                                     1.0 44.907242 8.611559868
                                                                                              8900
                                                                                                        NaN
             2.0
                                          1186.0
                                                 32500.0
                                                                     1.0 45.666359 12.24188995
                                                                                              8800
          1
                                51.0
                                                                                                        NaN
                   pop
          2 3.0
                                74.0
                                          4658.0 142228.0
                                                                     1.0 45.503300
                                                                                     11.41784
                                                                                              4200
                                                                                                        NaN
                  sport
             4.0 lounge
                                51.0
                                          2739.0 160000.0
                                                                     1.0 40.633171 17.63460922
                                                                                              6000
                                                                                                        NaN
           4 5.0
                   pop
                                73.0
                                          3074.0 106880.0
                                                                     1.0 41.903221 12.49565029
                                                                                             5700
                                                                                                        NaN
In [24]:
         #to display null values
          data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 500 entries, 0 to 499
          Data columns (total 11 columns):
           #
               Column
                                 Non-Null Count Dtype
           0
               ID
                                 500 non-null
                                                  float64
           1
               model
                                 500 non-null
                                                  object
           2
               engine power
                                 500 non-null
                                                  float64
           3
               age_in_days
                                 500 non-null
                                                  float64
           4
                                 500 non-null
                                                  float64
               km
           5
               previous owners 500 non-null
                                                  float64
           6
               lat
                                 500 non-null
                                                  float64
           7
                                 500 non-null
                                                  object
               lon
           8
                                 500 non-null
                                                  object
               price
               Unnamed: 9
                                 0 non-null
                                                  float64
           10 Unnamed: 10
                                 0 non-null
                                                  object
          dtypes: float64(7), object(4)
          memory usage: 43.1+ KB
```

```
In [25]: #to display summary of statistics
          data.describe()
Out[25]:
                            engine power
                                          age in days
                                                                     previous owners
                                                                                             lat Unnamed: 9
           count 500.000000
                                 500.00000
                                            500.000000
                                                          500.000000
                                                                           500.00000
                                                                                     500.000000
                                                                                                        0.0
            mean
                 250.500000
                                 51.90800
                                           1677.516000
                                                        53279.784000
                                                                             1.16000
                                                                                      43.664013
                                                                                                       NaN
                 144.481833
                                  4.03337
                                           1339.277861
                                                                                       2.139034
                                                                                                       NaN
             std
                                                        41893.569817
                                                                             0.44135
             min
                    1.000000
                                 51.00000
                                            366.000000
                                                         1232,000000
                                                                             1.00000
                                                                                      36.855839
                                                                                                       NaN
             25%
                  125.750000
                                 51.00000
                                            578.000000
                                                        18199.500000
                                                                             1.00000
                                                                                      41.903221
                                                                                                       NaN
             50%
                  250.500000
                                 51.00000
                                           1066.000000
                                                        38000.000000
                                                                             1.00000
                                                                                      44.508839
                                                                                                       NaN
             75%
                  375.250000
                                 51.00000
                                           2769.000000
                                                        81900.000000
                                                                             1.00000
                                                                                       45.467960
                                                                                                       NaN
            max 500.000000
                                 77.00000
                                           4658.000000 188000.000000
                                                                             4.00000
                                                                                                       NaN
                                                                                      46.792019
In [26]:
          #to display columns name
          data.columns
Out[26]: Index(['ID', 'model', 'engine_power', 'age_in_days', 'km', 'previous_owners',
                   'lat', 'lon', 'price', 'Unnamed: 9', 'Unnamed: 10'],
                 dtype='object')
In [28]: data1=data[['ID', 'model', 'engine_power', 'age_in_days', 'km', 'previous_owners',
                   'lat', 'lon', 'price']]
```

EDA and Visualization

In [29]: sns.pairplot(data1)

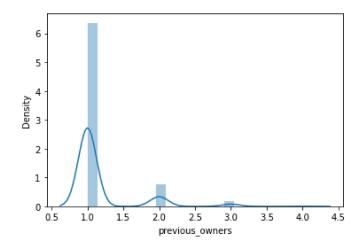
Out[29]: <seaborn.axisgrid.PairGrid at 0x1760ca0d250>



```
In [30]: sns.distplot(data['previous_owners'])
```

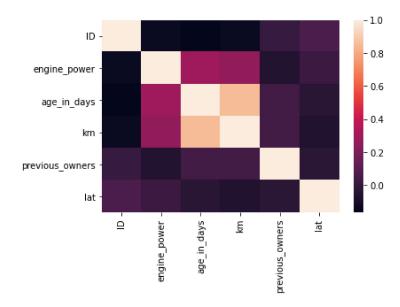
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `di
stplot` is a deprecated function and will be removed in a future version. Please adapt your
code to use either `displot` (a figure-level function with similar flexibility) or `histplot
` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

Out[30]: <AxesSubplot:xlabel='previous_owners', ylabel='Density'>





Out[31]: <AxesSubplot:>



To train the model

we are going to train the linear regression model ;We need to split the two variable x and y where x in independent variable (input) and y is dependent of x(output) so we could ignore address columns as it is not requires for our model

```
In [103]: x=data1[[ 'lat', 'price' ]]
           y=data1['km']
In [104]:
           #To split test and train data
           from sklearn.model_selection import train_test_split
           x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.6)
In [105]: from sklearn.linear_model import LinearRegression
           lr=LinearRegression()
           lr.fit(x_train,y_train)
Out[105]: LinearRegression()
In [106]: |lr.intercept_
Out[106]: 231360.35663553115
In [107]:
          coeff = pd.DataFrame(lr.coef_,x.columns,columns=["Co-efficient"])
           coeff
Out[107]:
                 Co-efficient
             lat -322.678413
                 -18.980317
           price
In [108]:
          prediction = lr.predict(x_train)
           plt.scatter(y_train,prediction)
Out[108]: <matplotlib.collections.PathCollection at 0x1760e62d4f0>
            140000
            120000
            100000
            80000
            60000
            40000
            20000
                0
                                   75000 100000 125000 150000 175000
                       25000
                             50000
 In [90]: |lr.score(x_test,y_test)
 Out[90]: 0.05380900324421822
  In [ ]:
  In [ ]:
```