

In [1]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

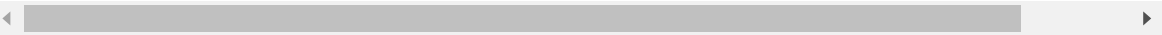
In [2]:

```
df=pd.read_csv(r'C:\Users\user\Downloads\fiat500_VehicleSelection_Dataset (2).csv')
df
```

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]:

```
df.head(10)
```

Out[3]:

|   | 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  | pop    | 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  | pop    | 73           | 3074        | 106880 | 1               | 41.903221 | 12.495650 |
| 5 | 6  | pop    | 74           | 3623        | 70225  | 1               | 45.000702 | 7.682270  |
| 6 | 7  | lounge | 51           | 731         | 11600  | 1               | 44.907242 | 8.611560  |
| 7 | 8  | lounge | 51           | 1521        | 49076  | 1               | 41.903221 | 12.495650 |
| 8 | 9  | sport  | 73           | 4049        | 76000  | 1               | 45.548000 | 11.549470 |
| 9 | 10 | sport  | 51           | 3653        | 89000  | 1               | 45.438301 | 10.991700 |

In [4]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1538 entries, 0 to 1537
Data columns (total 9 columns):
#   Column              Non-Null Count  Dtype
---  -
0   ID                   1538 non-null   int64
1   model                1538 non-null   object
2   engine_power         1538 non-null   int64
3   age_in_days          1538 non-null   int64
4   km                   1538 non-null   int64
5   previous_owners      1538 non-null   int64
6   lat                  1538 non-null   float64
7   lon                  1537 non-null   float64
8   price                1538 non-null   int64
dtypes: float64(2), int64(6), object(1)
memory usage: 108.3+ KB
```

In [5]:

```
df.describe()
```

Out[5]:

|       | ID          | engine_power | age_in_days | km            | previous_owners | lat         |
|-------|-------------|--------------|-------------|---------------|-----------------|-------------|
| count | 1538.000000 | 1538.000000  | 1538.000000 | 1538.000000   | 1538.000000     | 1538.000000 |
| mean  | 769.500000  | 51.904421    | 1650.980494 | 53396.011704  | 1.123537        | 43.541361   |
| std   | 444.126671  | 3.988023     | 1289.522278 | 40046.830723  | 0.416423        | 2.133511    |
| min   | 1.000000    | 51.000000    | 366.000000  | 1232.000000   | 1.000000        | 36.855831   |
| 25%   | 385.250000  | 51.000000    | 670.000000  | 20006.250000  | 1.000000        | 41.802990   |
| 50%   | 769.500000  | 51.000000    | 1035.000000 | 39031.000000  | 1.000000        | 44.394090   |
| 75%   | 1153.750000 | 51.000000    | 2616.000000 | 79667.750000  | 1.000000        | 45.467960   |
| max   | 1538.000000 | 77.000000    | 4658.000000 | 235000.000000 | 4.000000        | 46.795611   |

In [6]:

```
df.columns
```

Out[6]:

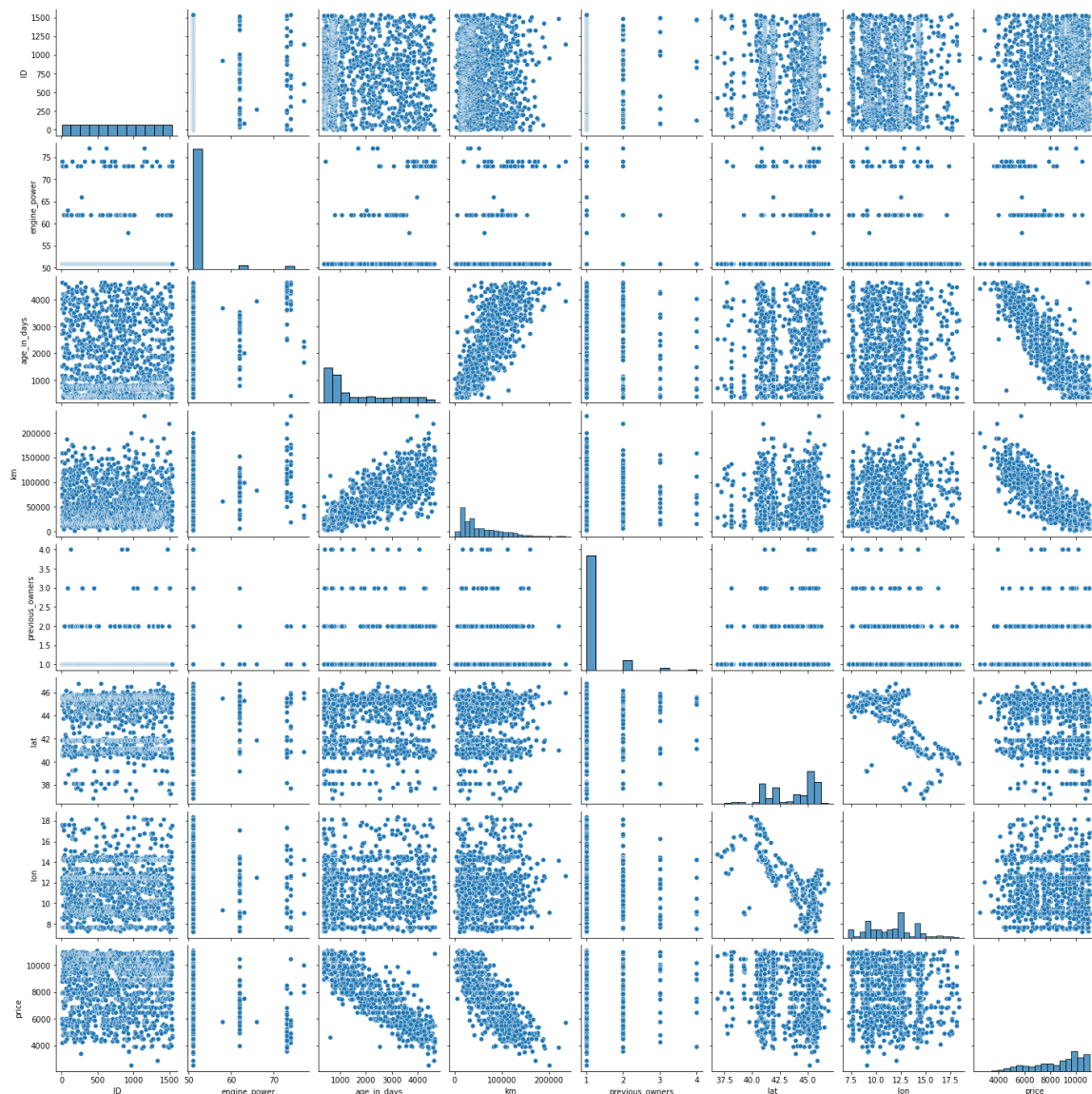
```
Index(['ID', 'model', 'engine_power', 'age_in_days', 'km', 'previous_owners',  
      'lat', 'lon', 'price'],  
      dtype='object')
```

In [7]:

```
sns.pairplot(df)
```

Out[7]:

&lt;seaborn.axisgrid.PairGrid at 0x26bf6619dc0&gt;



In [8]:

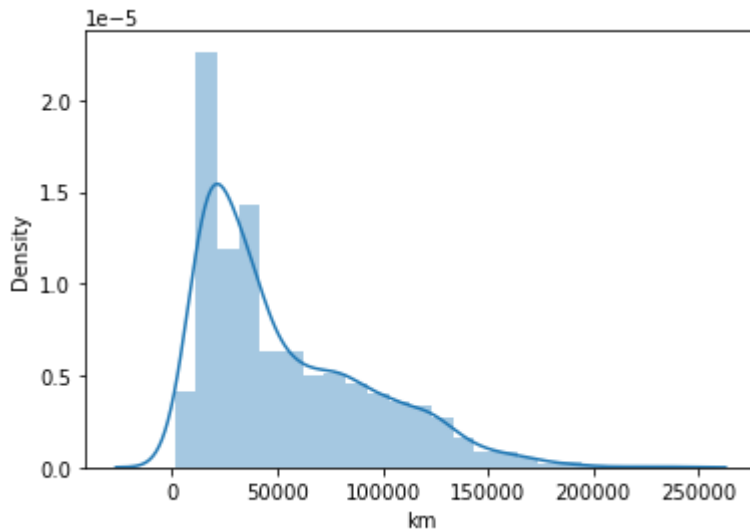
```
sns.distplot(df['km'])
```

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 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='km', ylabel='Density'>

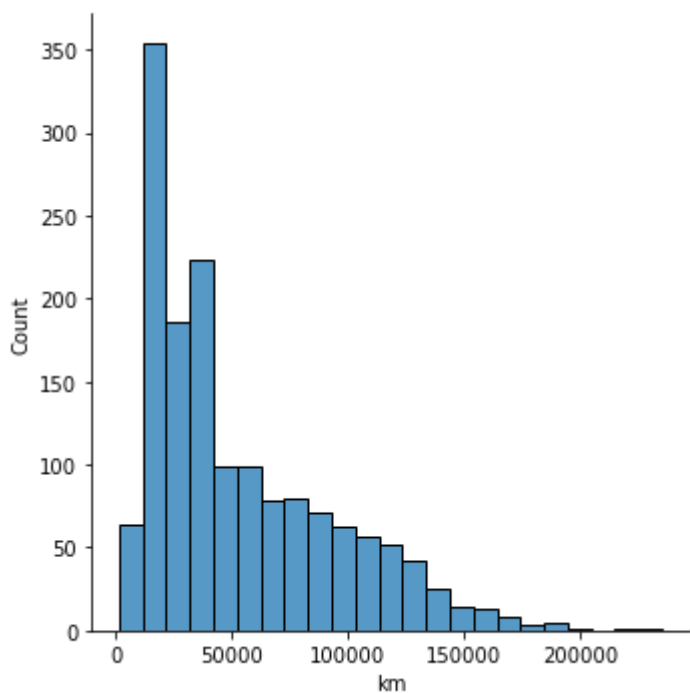


In [9]:

```
sns.displot(df["km"])
```

Out[9]:

<seaborn.axisgrid.FacetGrid at 0x26bf9415d30>



In [10]:

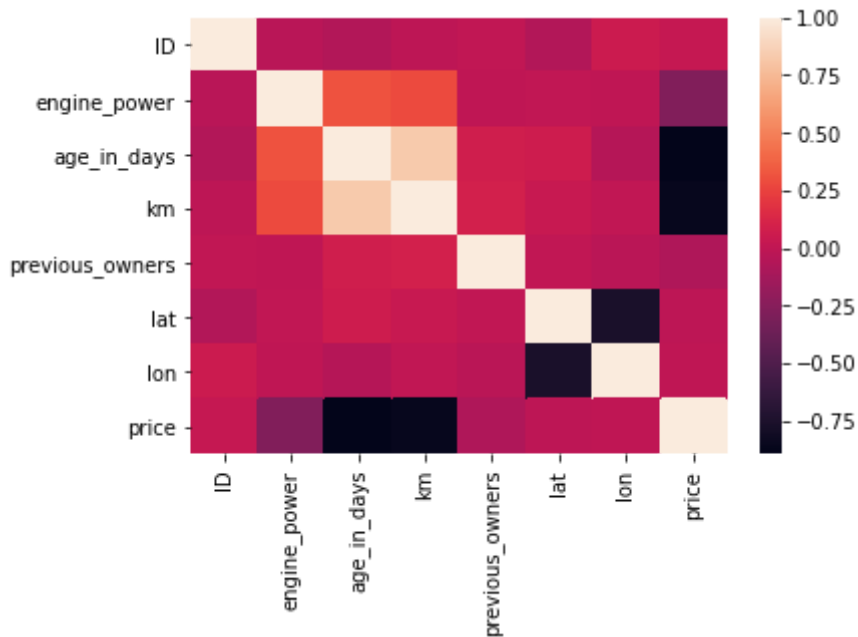
```
df1=df[['ID', 'model', 'engine_power', 'age_in_days', 'km', 'previous_owners',
        'lat', 'lon', 'price']]
```

In [11]:

```
sns.heatmap(df1.corr())
```

Out[11]:

&lt;AxesSubplot:&gt;



In [14]:

```
x=df1[['engine_power', 'age_in_days', 'km']]
y=df1[['price']]
```

In [15]:

```
from sklearn.model_selection import train_test_split
```

In [16]:

```
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

In [17]:

```
from sklearn.linear_model import LinearRegression
```

```
lr=LinearRegression()
lr.fit(x_train,y_train)#ValueError: Input contains NaN, infinity or a value too large for
```

Out[17]:

LinearRegression()

In [18]:

```
print(lr.intercept_)
```

[10559.14815828]

In [19]:

```
coef= pd.DataFrame(lr.coef_)  
coef
```

Out[19]:

|   | 0        | 1         | 2         |
|---|----------|-----------|-----------|
| 0 | 9.089696 | -0.885904 | -0.018897 |

In [20]:

```
print(lr.score(x_test,y_test))
```

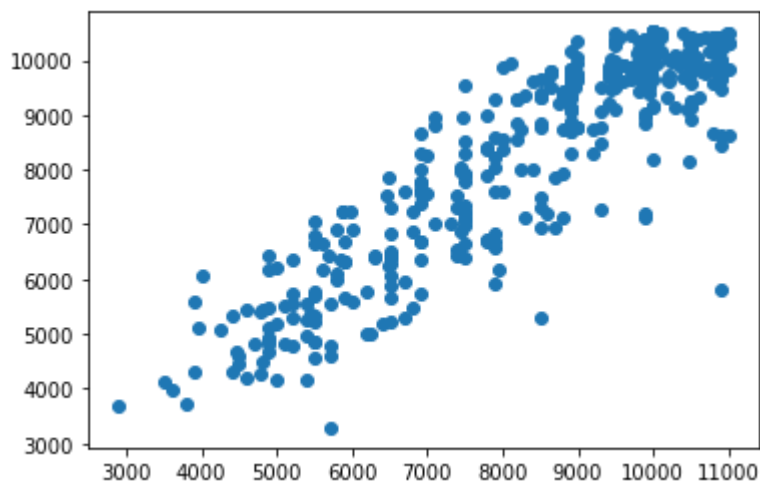
0.8126192264373202

In [21]:

```
prediction = lr.predict(x_test)  
plt.scatter(y_test,prediction)
```

Out[21]:

<matplotlib.collections.PathCollection at 0x26bfae4b100>



In [22]:

```
lr.score(x_test,y_test)
```

Out[22]:

0.8126192264373202

In [23]:

```
lr.score(x_train,y_train)
```

Out[23]:

0.853532785051327

In [24]:

```
from sklearn.linear_model import Ridge,Lasso
```

In [25]:

```
rr=Ridge(alpha=10)  
rr.fit(x_train,y_train)
```

Out[25]:

Ridge(alpha=10)

In [26]:

```
rr.score(x_test,y_test)
```

Out[26]:

0.8126193443083323

In [27]:

```
la=Lasso(alpha=10)  
la.fit(x_train,y_train)
```

Out[27]:

Lasso(alpha=10)

In [28]:

```
la.score(x_test,y_test)
```

Out[28]:

0.8126309693731364