

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [2]: df=pd.read_csv(r"C:\Users\user\Downloads\1_fiat500_VehicleSelection_Dataset.csv")
df.fillna(0,inplace=True)
df
```

Out[2]:

	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	8.61155986
1	2.0	pop	51.0	1186.0	32500.0	1.0	45.666359	12.2418899
2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.4178
3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.6346092
4	5.0	pop	73.0	3074.0	106880.0	1.0	41.903221	12.4956502
...
1544	0.0	0	0.0	0.0	0.0	0.0	0.000000	length
1545	0.0	0	0.0	0.0	0.0	0.0	0.000000	conca
1546	0.0	0	0.0	0.0	0.0	0.0	0.000000	Null value
1547	0.0	0	0.0	0.0	0.0	0.0	0.000000	fin
1548	0.0	0	0.0	0.0	0.0	0.0	0.000000	search

1549 rows × 11 columns

```
In [3]: df.head()
```

Out[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	8.61155986
1	2.0	pop	51.0	1186.0	32500.0	1.0	45.666359	12.2418899
2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.4178
3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.6346092
4	5.0	pop	73.0	3074.0	106880.0	1.0	41.903221	12.4956502

In [4]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1549 entries, 0 to 1548
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    1549 non-null   float64
1   model                 1549 non-null   object
2   engine_power          1549 non-null   float64
3   age_in_days           1549 non-null   float64
4   km                    1549 non-null   float64
5   previous_owners       1549 non-null   float64
6   lat                   1549 non-null   float64
7   lon                   1549 non-null   object
8   price                 1549 non-null   object
9   Unnamed: 9            1549 non-null   float64
10  Unnamed: 10           1549 non-null   object
dtypes: float64(7), object(4)
memory usage: 133.2+ KB
```

In [5]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1549 entries, 0 to 1548
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    1549 non-null   float64
1   model                 1549 non-null   object
2   engine_power          1549 non-null   float64
3   age_in_days           1549 non-null   float64
4   km                    1549 non-null   float64
5   previous_owners       1549 non-null   float64
6   lat                   1549 non-null   float64
7   lon                   1549 non-null   object
8   price                 1549 non-null   object
9   Unnamed: 9            1549 non-null   float64
10  Unnamed: 10           1549 non-null   object
dtypes: float64(7), object(4)
memory usage: 133.2+ KB
```

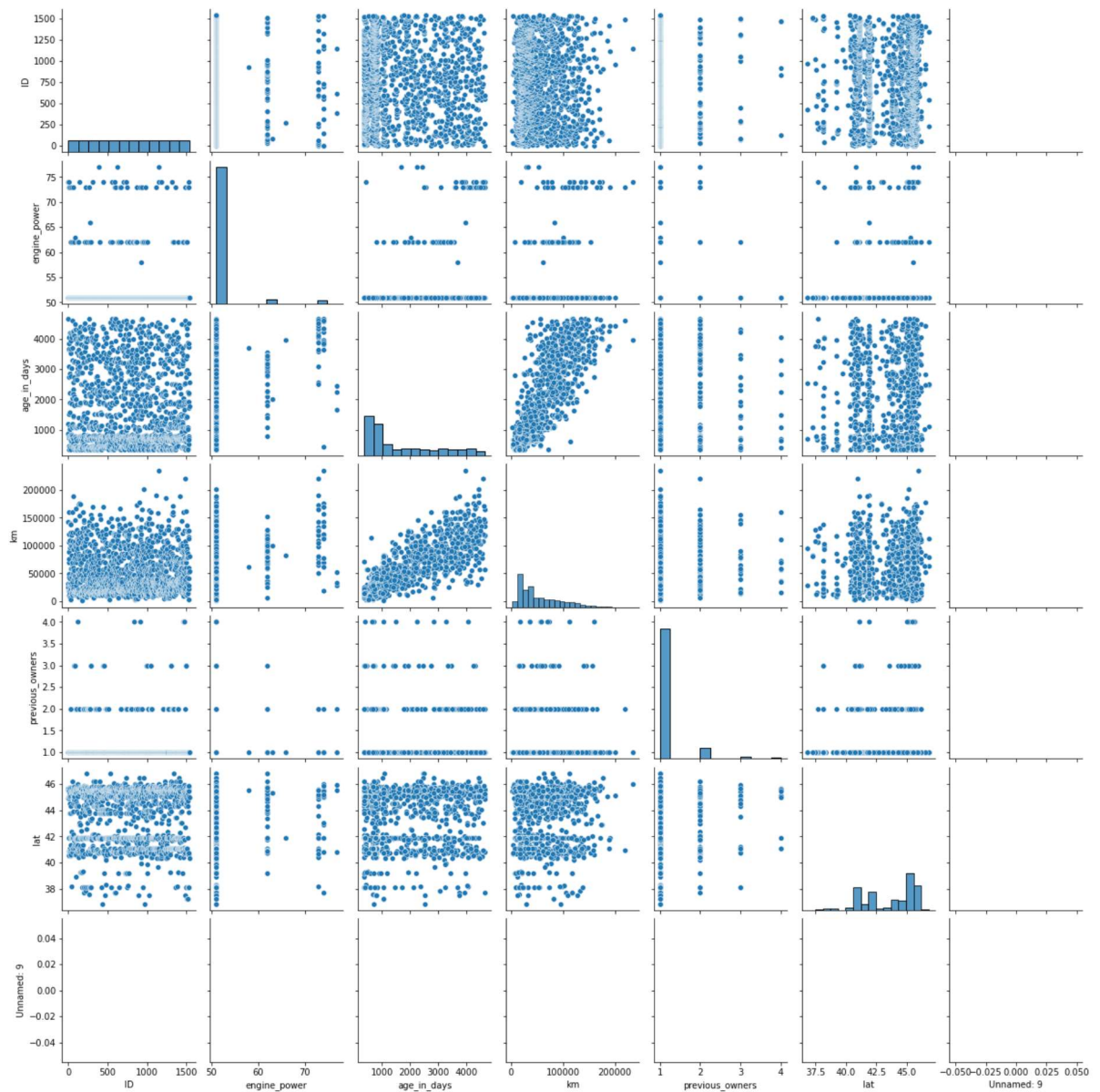
In [6]: df=pd.read_csv("1_fiat500_VehicleSelection_Dataset.csv")

In [8]:
import seaborn **as** sns

In [9]:

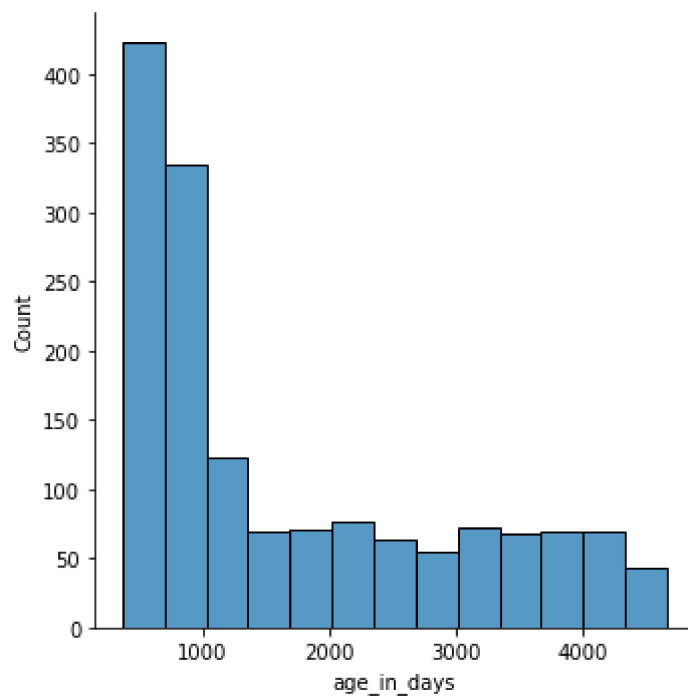
```
sns.pairplot(df)
```

Out[9]: <seaborn.axisgrid.PairGrid at 0x16d214fa460>



```
In [10]: sns.displot(df['age_in_days'])
```

```
Out[10]: <seaborn.axisgrid.FacetGrid at 0x16d23d940a0>
```

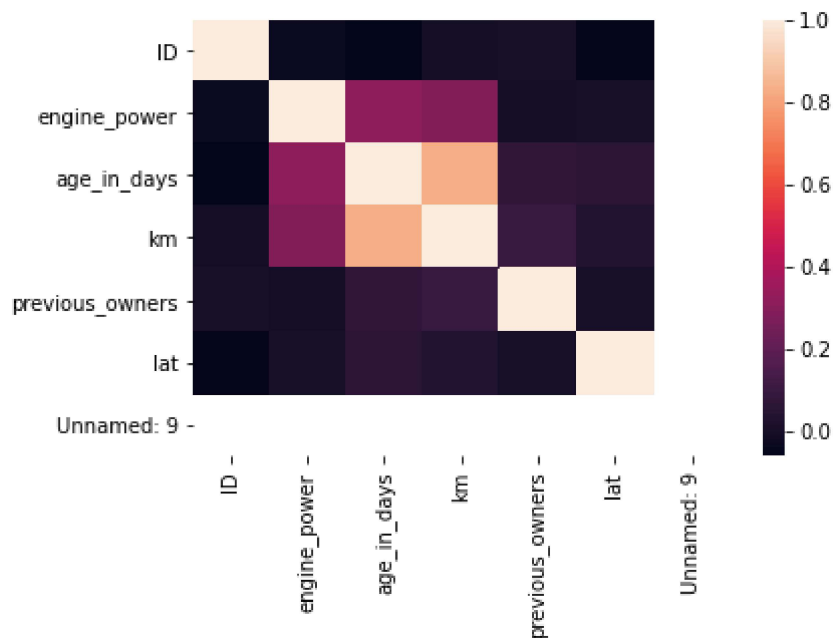


```
In [11]: df1=df.drop(['Unnamed: 10'],axis=1)
df1
df1=df1.drop(df1.index[1537:])
df1.isna().sum()
```

```
Out[11]: ID          0
model            0
engine_power     0
age_in_days      0
km              0
previous_owners  0
lat             0
lon            0
price           0
Unnamed: 9      1537
dtype: int64
```

```
In [12]: sns.heatmap(df1.corr())
```

```
Out[12]: <AxesSubplot:>
```



```
In [13]: from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
In [14]: df1.isna().sum()
```

```
Out[14]: ID          0
model          0
engine_power   0
age_in_days    0
km             0
previous_owners 0
lat           0
lon           0
price         0
Unnamed: 9     1537
dtype: int64
```

```
In [15]: y=df1['price']
x=df1.drop(['price','Unnamed: 9','model'],axis=1)
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
print(x_train)
```

	ID	engine_power	age_in_days	km	previous_owners	lat
\						
662	663.0	51.0	1492.0	23016.0	1.0	45.131672
252	253.0	51.0	790.0	27595.0	2.0	45.688259
1335	1336.0	51.0	366.0	12950.0	1.0	44.512428
151	152.0	51.0	2892.0	67000.0	1.0	45.674839
1477	1478.0	51.0	1917.0	126426.0	1.0	41.769051
...
456	457.0	51.0	456.0	14800.0	1.0	44.063129
730	731.0	51.0	762.0	12337.0	1.0	43.782372
1412	1413.0	51.0	1431.0	38000.0	1.0	45.215408
1361	1362.0	51.0	3227.0	73000.0	1.0	38.122070
1392	1393.0	51.0	1766.0	29400.0	2.0	41.741779

	lon
662	8.449170113
252	8.731450081
1335	11.12837982
151	9.687859535
1477	12.66281033
...	...
456	12.44701004
730	11.25498962
1412	7.633669853
1361	13.36112022
1392	12.6441803

[1075 rows x 7 columns]

In [16]:

```
model=LinearRegression()
model.fit(x_train,y_train)
model.intercept_
```

Out[16]: 9390.660847702526

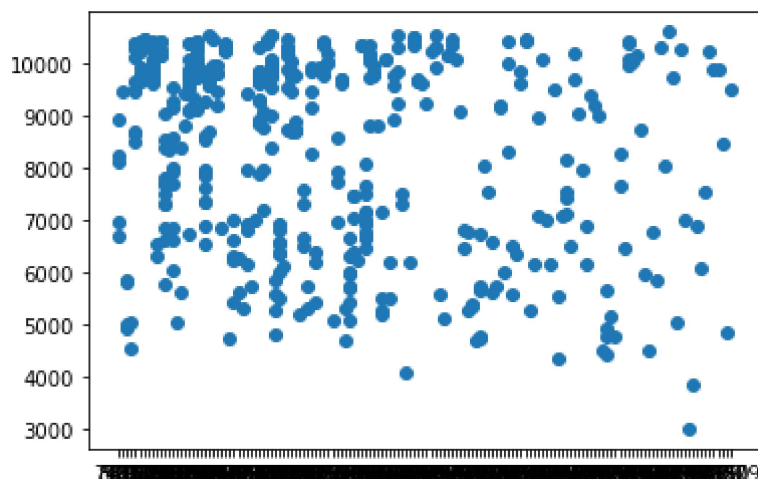
In [17]: model.coef_

Out[17]: array([-5.11101287e-02, 6.39840119e+00, -9.22719101e-01, -1.70984147e-02,
4.75965035e+00, 3.20330098e+01, -5.90726736e+00])

In [18]:

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

Out[18]: <matplotlib.collections.PathCollection at 0x16d25c92e50>



In [19]:

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

Out[19]: 0.8552727844290008

In []: