

```
In [1]: import pandas as pd
import warnings
warnings.filterwarnings("ignore")
```

```
In [2]: data=pd.read_csv("/home/palcement/Downloads/fiat500.csv")
data
```

Out[2]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	pop	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	pop	73	3074	106880	1	41.903221	12.495650	5700
...
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1538 rows × 9 columns

```
In [3]: data1=data.loc[(data.previous_owners==1)]  
data1
```

Out[3]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	pop	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	pop	73	3074	106880	1	41.903221	12.495650	5700
...
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1389 rows × 9 columns

```
In [4]: data1=data1.drop(['ID','lat','lon'],axis=1)  
data1
```

Out[4]:

	model	engine_power	age_in_days	km	previous_owners	price
0	lounge	51	882	25000	1	8900
1	pop	51	1186	32500	1	8800
2	sport	74	4658	142228	1	4200
3	lounge	51	2739	160000	1	6000
4	pop	73	3074	106880	1	5700
...
1533	sport	51	3712	115280	1	5200
1534	lounge	74	3835	112000	1	4600
1535	pop	51	2223	60457	1	7500
1536	lounge	51	2557	80750	1	5990
1537	pop	51	1766	54276	1	7900

1389 rows × 6 columns

```
In [5]: data1=pd.get_dummies(data1)
data1
```

Out[5]:

	engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
0	51	882	25000	1	8900	1	0	0
1	51	1186	32500	1	8800	0	1	0
2	74	4658	142228	1	4200	0	0	1
3	51	2739	160000	1	6000	1	0	0
4	73	3074	106880	1	5700	0	1	0
...
1533	51	3712	115280	1	5200	0	0	1
1534	74	3835	112000	1	4600	1	0	0
1535	51	2223	60457	1	7500	0	1	0
1536	51	2557	80750	1	5990	1	0	0
1537	51	1766	54276	1	7900	0	1	0

1389 rows × 8 columns

```
In [6]: y=data1['price']
X=data1.drop(['price'],axis=1)
```

In [7]:

y

Out[7]:

0	8900
1	8800
2	4200
3	6000
4	5700

...

1533	5200
1534	4600
1535	7500
1536	5990
1537	7900

Name: price, Length: 1389, dtype: int64

In [8]:

X

Out[8]:

	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
0	51	882	25000	1	1	0	0
1	51	1186	32500	1	0	1	0
2	74	4658	142228	1	0	0	1
3	51	2739	160000	1	1	0	0
4	73	3074	106880	1	0	1	0
...
1533	51	3712	115280	1	0	0	1
1534	74	3835	112000	1	1	0	0
1535	51	2223	60457	1	0	1	0
1536	51	2557	80750	1	1	0	0
1537	51	1766	54276	1	0	1	0

1389 rows × 7 columns

```
In [46]: from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.33,random_state=42)
```

```
In [47]: X_train
```

Out[47]:

	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
915	51	397	17081	1	1	0	0
12	51	456	18450	1	1	0	0
638	51	397	21276	1	1	0	0
190	51	821	19000	1	1	0	0
701	51	701	27100	1	1	0	0
...
1201	51	790	50740	1	0	1	0
1239	51	4383	107600	1	0	1	0
1432	51	701	42095	1	1	0	0
951	51	3684	78000	1	1	0	0
1235	51	1613	45000	1	1	0	0

930 rows × 7 columns

```
In [48]: y_train
```

```
Out[48]: 915      10900
          12       9700
          638     10850
          190     9990
          701     10300
          ...
          1201     8300
          1239     3950
          1432     8900
          951      6500
          1235     8800
          Name: price, Length: 930, dtype: int64
```

```
In [49]: y_test
```

```
Out[49]: 625      5400
          187      5399
          279      4900
          734     10500
          315      9300
          ...
          115     10650
          370      9900
          1179     5900
          93      10050
          147      9900
          Name: price, Length: 459, dtype: int64
```

```
In [50]: from sklearn.linear_model import ElasticNet
from sklearn.model_selection import GridSearchCV

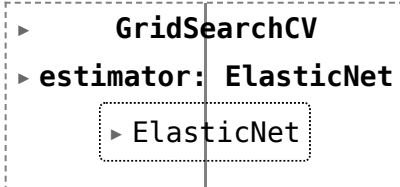
elastic = ElasticNet()

parameters = {'alpha': [1e-15, 1e-10, 1e-8, 1e-4, 1e-3, 1e-2, 1, 5, 10, 20]}

elastic_regressor = GridSearchCV(elastic, parameters)

elastic_regressor.fit(X_train, y_train)
```

```
Out[50]:
```



```
  ▶ GridSearchCV
  ▶ estimator: ElasticNet
    ▶ ElasticNet
```

```
In [51]: elastic_regressor.best_params_
```

```
Out[51]: {'alpha': 0.01}
```

```
In [52]: elastic=ElasticNet(alpha=0.1)
elastic.fit(X_train,y_train)
y_pred_elastic=elastic.predict(X_test)
```

```
In [53]: from sklearn.metrics import r2_score
r2_score(y_test,y_pred_elastic)
```

```
Out[53]: 0.8601270407940889
```

```
In [54]: from sklearn.metrics import mean_squared_error
elastic_Error=mean_squared_error(y_pred_elastic,y_test)
elastic_Error
```

```
Out[54]: 515678.8171884504
```



```
In [55]: Results=pd.DataFrame(columns=['Price','Predicted']) #price and predicted names are our wish
Results['Price']=y_test
Results['Predicted']=y_pred_elastic
Results=Results.reset_index()
Results['Id']=Results.index
```

```
In [56]: Results
```

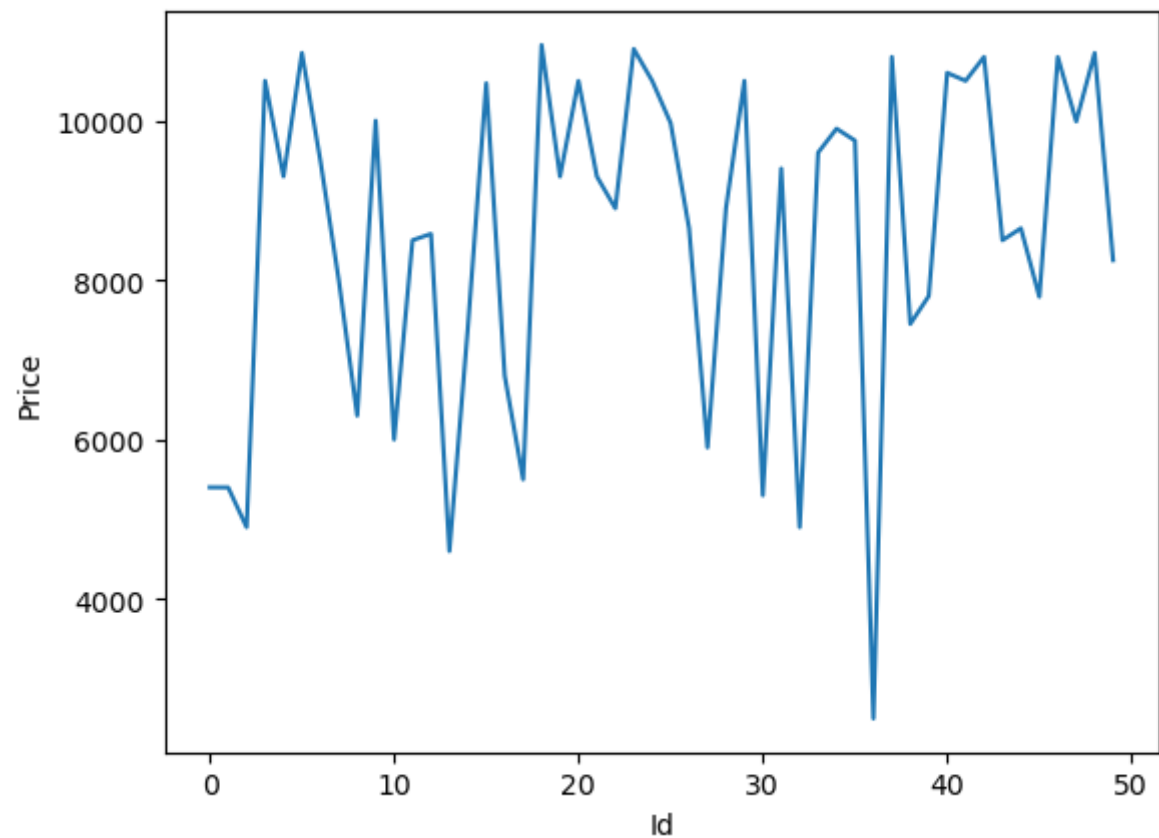
```
Out[56]:
```

	index	Price	Predicted	Id
0	625	5400	5478.361166	0
1	187	5399	5124.950418	1
2	279	4900	4833.208393	2
3	734	10500	9688.909121	3
4	315	9300	9402.252771	4
...
454	115	10650	10389.152700	454
455	370	9900	10260.455864	455
456	1179	5900	6773.307749	456
457	93	10050	10370.171613	457
458	147	9900	10063.002616	458

459 rows × 4 columns

```
In [58]: import seaborn as sns
import matplotlib.pyplot as plt
sns.lineplot(x='Id',y='Price',data=Results.head(50)) #Orange color
sns.lineplot(x='Id',y='Predicted',data=Results.head(50)) #Blue color
plt.plot()
```

Out[58]: []



In []: