

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

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

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

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

```
In [4]: data1=data.loc[(data.previous_owners==1)]
```

In [5]: data1

Out[5]:

	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 [6]: data1=data.drop(['lat', 'lon', 'lon'],axis=1)

In [7]: data1=pd.get\_dummies(data1)

In [8]: data1

Out[8]:

	ID	engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
0	1	51	882	25000	1	8900	1	0	0
1	2	51	1186	32500	1	8800	0	1	0
2	3	74	4658	142228	1	4200	0	0	1
3	4	51	2739	160000	1	6000	1	0	0
4	5	73	3074	106880	1	5700	0	1	0
...	...	...	...	...	...	...	...	...	...
1533	1534	51	3712	115280	1	5200	0	0	1
1534	1535	74	3835	112000	1	4600	1	0	0
1535	1536	51	2223	60457	1	7500	0	1	0
1536	1537	51	2557	80750	1	5990	1	0	0
1537	1538	51	1766	54276	1	7900	0	1	0

1538 rows × 9 columns

```
In [9]: y=data1['price']
x=data1.drop('price',axis=1)
```

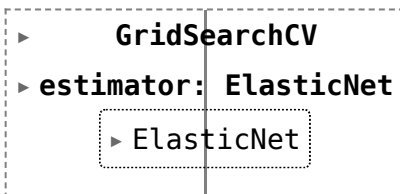
```
In [10]: from sklearn.model_selection import train_test_split
X_train,X_test,Y_train,Y_test=train_test_split(x,y,test_size=0.33,random_state=42)
```

In [ ]:

In [ ]:

In [12]: ccccc

Out[12]:



```
GridSearchCV
└── estimator: ElasticNet
    └── ElasticNet
```

In [13]: elastic\_regressor.best\_params\_

Out[13]: {'alpha': 0.01}

In [16]: elastic=ElasticNet(alpha=0.1)  
elastic.fit(X\_train,Y\_train)  
y\_pred\_elastic=elastic.predict(X\_test)

In [18]: from sklearn.metrics import r2\_score  
r2\_score(Y\_test,y\_pred\_elastic)

Out[18]: 0.8414565299012147

In [20]: from sklearn.metrics import mean\_squared\_error  
elastic\_Error=mean\_squared\_error(y\_pred\_elastic,Y\_test)  
elastic\_Error

Out[20]: 582240.9011940917

```
In [22]: Results=pd.DataFrame(columns=['price', 'predicted'])
Results['price']=Y_test
Results['predicted']=y_pred_elastic
Results=Results.reset_index()
Results['ID']=Results.index
Results
```

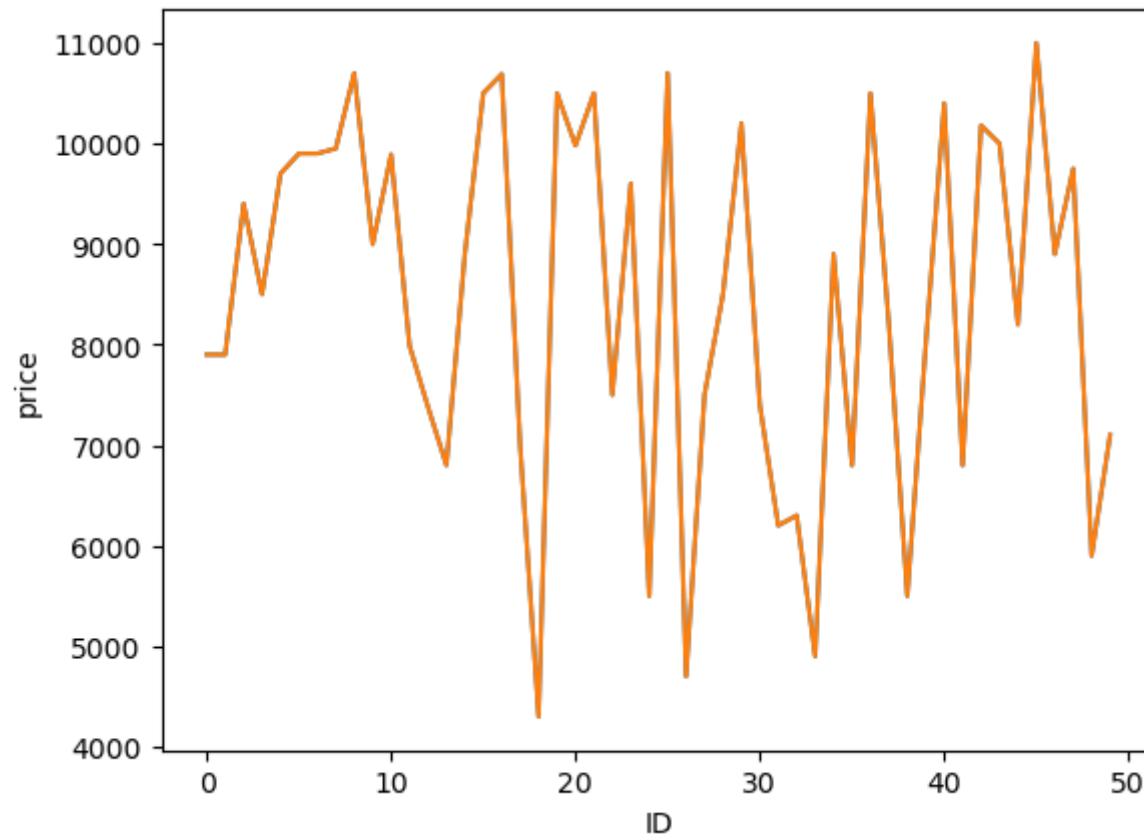
Out[22]:

	index	price	predicted	ID
0	481	7900	5903.123297	0
1	76	7900	7223.403259	1
2	1502	9400	9794.808007	2
3	669	8500	9724.783858	3
4	1409	9700	9974.200594	4
...	...	...	...	...
503	291	10900	10069.090270	503
504	596	5699	6320.490299	504
505	1489	9500	9959.608673	505
506	1436	6990	8310.780259	506
507	575	10900	10384.079146	507

508 rows × 4 columns

```
In [23]: import seaborn as sns
import matplotlib.pyplot as plt
sns.lineplot(x='ID',y='price',data=Results.head(50))
sns.lineplot(x='ID',y='price',data=Results.head(50))
plt.plot()
```

Out[23]: []



In [ ]: