```
In [119]: import pandas as pd
In [120]: data=pd.read csv("/home/placement/Desktop/data analysis(413)/fiat500.csv")
In [121]: data
Out[121]:
                        model engine_power age_in_days
                                                             km previous owners
                                                                                       lat
                                                                                                lon price
                                                          25000
                0
                      1 lounge
                                         51
                                                    882
                                                                              1 44.907242
                                                                                            8.611560
                                                                                                     8900
                                         51
                                                   1186
                                                          32500
                                                                              1 45.666359 12.241890
                                                                                                     8800
                1
                           pop
                                                         142228
                                                                              1 45.503300 11.417840
                2
                                                                                                     4200
                          sport
                                         74
                                                   4658
                                                         160000
                3
                      4 lounge
                                         51
                                                   2739
                                                                              1 40.633171 17.634609
                                                                                                     6000
                                         73
                                                                              1 41.903221 12.495650
                                                   3074 106880
                                                                                                     5700
                           pop
                                                                                            7.704920
             1533
                  1534
                          sport
                                         51
                                                   3712 115280
                                                                              1 45.069679
                                                                                                     5200
                  1535
                                                   3835
                                                         112000
                                                                              1 45.845692
                                                                                           8.666870
                                                                                                     4600
             1534
                        lounge
                                         74
                                                          60457
                                                                                           9.413480
             1535
                  1536
                           pop
                                         51
                                                   2223
                                                                              1 45.481541
                                                                                                     7500
             1536
                  1537
                        lounge
                                         51
                                                   2557
                                                          80750
                                                                              1 45.000702
                                                                                           7.682270
                                                                                                     5990
             1537 1538
                                         51
                                                   1766
                                                          54276
                                                                              1 40.323410 17.568270
                                                                                                     7900
                           pop
            1538 rows × 9 columns
```

In [122]: z=data1.loc[(data1.previous_owners==1)]

In [123]: z

Out[123]:

	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 [124]: data1=data.drop(["lat","lon","ID"],axis=1)
```

In [125]: data1=pd.get_dummies(data1)

In [126]: data1

Out[126]:

	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

1538 rows × 8 columns

```
In [130]: 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)
          e+05
            model = cd fast.enet coordinate descent(
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ coordinate descent.py:631: C
          onvergenceWarning: Objective did not converge. You might want to increase the number of iterations, check
          the scale of the features or consider increasing regularisation. Duality gap: 2.703e+08, tolerance: 3.517
          e+05
            model = cd fast.enet coordinate descent(
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ coordinate descent.py:631: C
          onvergenceWarning: Objective did not converge. You might want to increase the number of iterations, check
          the scale of the features or consider increasing regularisation. Duality gap: 2.854e+08, tolerance: 3.711
          e+05
            model = cd fast.enet coordinate descent(
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ coordinate descent.py:631: C
          onvergenceWarning: Objective did not converge. You might want to increase the number of iterations, check
          the scale of the features or consider increasing regularisation. Duality gap: 2.909e+08, tolerance: 3.576
          e+05
            model = cd fast.enet coordinate descent(
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ coordinate descent.py:631: C
          onvergenceWarning: Objective did not converge. You might want to increase the number of iterations, check
          the coals of the features or consider increasing regularisation. Duality can. 2 701a:00 telegranes. 2 E10
In [131]: elastic=ElasticNet(alpha=.01)
          elastic.fit(x train,y train)
          y pred elastic=elastic.predict(x test)
```

In [132]: from sklearn.metrics import r2_score
r2_score(y_test,y_pred_elastic)

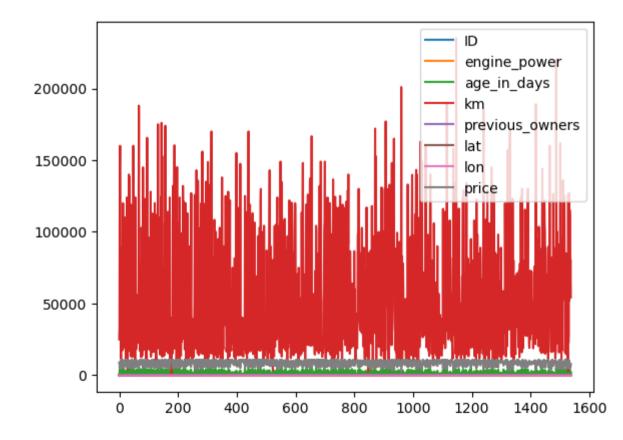
Out[132]: 0.8488682857174344

In [134]: from sklearn.metrics import mean_squared_error
 elastic_Error=mean_squared_error(y_pred_elastic,y_test)
 elastic_Error

Out[134]: 603966.023413073

In [135]: data.plot()

Out[135]: <Axes: >



```
In [136]: elastic_regressor.best_params_
Out[136]: {'alpha': 0.01}
In [ ]:
In [ ]:
```