```
In [114]: import pandas as pd
In [115]: |data=pd.read_csv("fiat500.csv")
In [116]: data
Out[116]:
                        model engine_power age_in_days
                     ID
                                                             km previous_owners
                                                                                        lat
                                                                                                 lon price
                                                     882
                                                          25000
                0
                                         51
                                                                               1 44.907242
                                                                                            8.611560
                                                                                                      8900
                      1 lounge
                1
                      2
                           pop
                                         51
                                                    1186
                                                           32500
                                                                               1 45.666359 12.241890
                                                                                                      8800
                2
                          sport
                                         74
                                                    4658
                                                         142228
                                                                               1 45.503300 11.417840
                                                                                                      4200
                3
                                         51
                                                         160000
                                                                               1 40.633171 17.634609
                                                                                                      6000
                      4 lounge
                                                    2739
                                                         106880
                                                                              1 41.903221 12.495650
                                         73
                                                                                                      5700
                                                    3074
                           pop
             1533 1534
                                                    3712 115280
                                                                               1 45.069679
                                                                                            7.704920
                                                                                                      5200
                          sport
                                         51
                                                    3835 112000
             1534
                   1535
                        lounge
                                         74
                                                                               1 45.845692
                                                                                            8.666870
                                                                                                      4600
             1535 1536
                                                                                            9.413480
                                         51
                                                    2223
                                                          60457
                                                                               1 45.481541
                                                                                                      7500
                           pop
             1536 1537
                                                    2557
                                                          80750
                                                                               1 45.000702
                                                                                            7.682270
                                                                                                      5990
                        lounge
                                         51
             1537 1538
                                                          54276
                                                                              1 40.323410 17.568270
                                         51
                                                    1766
                                                                                                     7900
                           pop
            1538 rows × 9 columns
```

In [117]:

data=data.drop(['lat','lon','ID',],axis=1)

In [118]: data

Out[118]:

	model	engine_power	age_in_days	km	previous_owners	price
0	lounge	51	882	25000	1	8900
1	рор	51	1186	32500	1	8800
2	sport	74	4658	142228	1	4200
3	lounge	51	2739	160000	1	6000
4	рор	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

1538 rows × 6 columns

```
In [119]: data=pd.get_dummies(data)
In [120]: data.shape
    #data['model'] =data['model'].map({'lounge':1,'pop':2,'sport':3})
Out[120]: (1538, 8)
```

In [121]: data

Out[121]:

	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 [122]: y=data['price']
```

In [123]: x=data.drop('price',axis=1)

```
In [124]: y
Out[124]: 0
                  8900
                  8800
          2
                  4200
          3
                  6000
                  5700
          1533
                  5200
          1534
                  4600
          1535
                  7500
          1536
                  5990
          1537
                  7900
          Name: price, Length: 1538, dtype: int64
In [125]: !pip install scikit-learn
          Requirement already satisfied: scikit-learn in ./anaconda3/lib/python3.10/site-packages (1.2.1)
          Requirement already satisfied: joblib>=1.1.1 in ./anaconda3/lib/python3.10/site-packages (from scikit-lear
          n) (1.1.1)
          Requirement already satisfied: scipy>=1.3.2 in ./anaconda3/lib/python3.10/site-packages (from scikit-learn)
          (1.10.0)
          Requirement already satisfied: threadpoolctl>=2.0.0 in ./anaconda3/lib/python3.10/site-packages (from sciki
          t-learn) (2.2.0)
          Requirement already satisfied: numpy>=1.17.3 in ./anaconda3/lib/python3.10/site-packages (from scikit-lear
          n) (1.23.5)
In [126]: 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 [127]: x test.head(5)
Out[127]:
                                          km previous owners model lounge model pop model sport
                 engine power age in days
                                  3197 120000
                                                          2
                                                                      0
             481
                         51
                                                                                1
                                                                                           0
             76
                         62
                                                          1
                                  2101 103000
                                                                      0
                                                                                1
                                                                                           0
            1502
                         51
                                   670
                                        32473
                                                          1
                                                                      1
                                                                                0
                                                                                           0
            669
                         51
                                   913
                                        29000
                                                          1
                                                                      1
                                                                                0
                                                                                           0
                                                                                           0
            1409
                         51
                                   762
                                        18800
                                                          1
                                                                      1
                                                                                0
In [128]: x_train.shape
Out[128]: (1030, 7)
In [129]: y train
Out[129]: 527
                     9990
           129
                     9500
           602
                     7590
           331
                     8750
           323
                     9100
                    . . .
           1130
                   10990
                    9800
           1294
           860
                     5500
           1459
                     9990
           1126
                     8900
           Name: price, Length: 1030, dtype: int64
In [130]: from sklearn.linear model import LinearRegression
           reg = LinearRegression() #creating object of LinearRegression
           reg.fit(x_train,y_train) #training and fitting LR object using training data
Out[130]:
            ▼ LinearRegression
           LinearRegression()
```

```
In [131]: ypred=req.predict(x test)
In [132]: | ypred
                                  7948.63618724,
                                                  9704.82523573,
                                                                  7971.05970955,
                  9653.06224923,
                                                                  6698.19524313,
                 10399.51752022.
                                  9176.43567301.
                                                  5803.03205787,
                  8257.83550573, 10452.95284574,
                                                  9948.66454584,
                                                                  9789.65062843,
                 10582.50828537, 7568.91955482,
                                                  6804.97705225,
                                                                  8065.01292384,
                 10310.29143419, 8836.34894739,
                                                  8390.05091229,
                                                                  9582.13932508,
                                                                  7145.15315349,
                  9745.34784981, 10045.45021387, 10294.09872915,
                  9727.85493167,
                                  6281.78952194,
                                                  7901.36245623,
                                                                  9387.9203723 ,
                  5039.55649797,
                                  9351.49777725,
                                                  9980.70844784, 10094.79341516,
                                                                  5234.05388382.
                  6359.24321991,
                                  9856.10227211,
                                                  9099.07023804,
                  5534.45288323,
                                  4495.02309231, 10199.78432943, 10024.87037067,
                  5465.58034188,
                                  8520.72057674,
                                                  7034.71038647, 10054.65061446,
                                  6008.34860428,
                 10191.12067767,
                                                  9748.18097947,
                                                                  9669.4333196 ,
                  9145.3756075 ,
                                  9175.66562699, 10087.86753845,
                                                                  9825.02990067.
                  7340.29803785,
                                  5083.8487301 ,
                                                  9441.50914802, 10243.05490667,
                                                  6126.99295838,
                  5556.42300245, 10676.01945733,
                                                                  9845.16661356,
                  9850.77978959, 7840.83596305,
                                                  6552.05146566,
                                                                  9938.82104889,
                  8327.79232274, 9119.62204137,
                                                  6111.83787367, 10410.00504522,
                  6360.97695249,
                                  8601.59209793,
                                                  8377.80258216,
                                                                  9803.81343895,
                  8285.09831762. 10091.75635129. 10003.86694939. 10028.60283146.
In [133]: from sklearn .metrics import r2 score
          r2 score(y test,ypred)
Out[133]: 0.8415526986865394
In [134]: from sklearn.metrics import mean squared error
          mean squared error(ypred,y test)
Out[134]: 581887.727391353
In [135]: #from sklearn.metrics import accuracy score
          #accuracy score(y test)
```

In [136]: Results= pd.DataFrame(columns=['price', 'predicted']) Results['price']=y_test

Results['predicted']=ypred
Results=Results.reset_index()
Results['Id']=Results.index

Results.head(15)

Out[136]:

	index	price	predicted	ld
0	481	7900	5867.650338	0
1	76	7900	7133.701423	1
2	1502	9400	9866.357762	2
3	669	8500	9723.288745	3
4	1409	9700	10039.591012	4
5	1414	9900	9654.075826	5
6	1089	9900	9673.145630	6
7	1507	9950	10118.707281	7
8	970	10700	9903.859527	8
9	1198	8999	9351.558284	9
10	1088	9890	10434.349636	10
11	576	7990	7732.262557	11
12	965	7380	7698.672401	12
13	1488	6800	6565.952404	13
14	1432	8900	9662.901035	14

```
In [137]: from sklearn.model selection import GridSearchCV
          from sklearn.linear model import Ridge
          alpha = [1e-15, 1e-10, 1e-8, 1e-4, 1e-3,1e-2, 1, 5, 10, 20,30]
          ridge = Ridge()
          parameters = {'alpha': alpha}
          ridge regressor = GridSearchCV(ridge, parameters)
          ridge regressor.fit(x_train, y_train)
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=5.56109e-26): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.70876e-26): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=6.91585e-23): result may not be accurate.
            return linalg.solve(A, Xy, assume_a="pos", overwrite_a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.08003e-23): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.01022e-23): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.57959e-23): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.24161e-23): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=6.92759e-21): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.09091e-21): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlqWarning: I
```

```
ll-conditioned matrix (rcond=7.02112e-21): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.57414e-21): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.23284e-21): result may not be accurate.
            return linalq.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=6.9277e-17): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.09099e-17): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.02123e-17): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.57407e-17): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
          /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning: I
          ll-conditioned matrix (rcond=7.23274e-17): result may not be accurate.
            return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
Out[137]:
             GridSearchCV
           ▶ estimator: Ridge
                 ▶ Rildge
In [138]: ridge_regressor.best_params_
Out[138]: {'alpha': 30}
```

```
In [139]: ridge=Ridge(alpha=30)
    ridge.fit(x_train,y_train)
    y_pred_ridge=ridge.predict(x_test)
```

Out[145]:

	index	Price	Predicted	Id
0	481	7900	5869.741155	0
1	76	7900	7149.563327	1
2	1502	9400	9862.785355	2
3	669	8500	9719.283532	3
4	1409	9700	10035.895686	4
5	1414	9900	9650.311090	5
6	1089	9900	9669.183317	6
7	1507	9950	10115.128380	7
8	970	10700	9900.241944	8
9	1198	8999	9347.080772	9

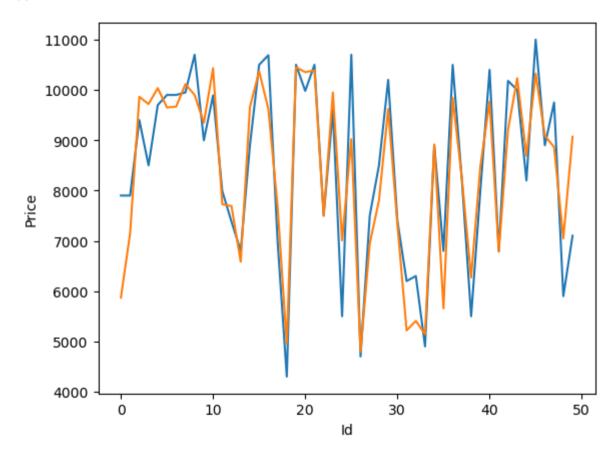
Dundinted Id

index Dries

```
In [146]: import seaborn as sns
import matplotlib.pyplot as plt

sns.lineplot(x='Id',y='Price',data=Results.head(50))
sns.lineplot(x='Id',y='Predicted',data=Results.head(50))
plt.plot()
```

Out[146]: []



In []: