

```
In [123]: import pandas as pd
```

```
In [124]: data=pd.read_csv("/home/placement/Downloads/fiat500.csv")
```

```
In [125]: data.describe()
```

```
Out[125]:
```

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	price
count	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000
mean	769.500000	51.904421	1650.980494	53396.011704	1.123537	43.541361	11.563428	8576.003901
std	444.126671	3.988023	1289.522278	40046.830723	0.416423	2.133518	2.328190	1939.958641
min	1.000000	51.000000	366.000000	1232.000000	1.000000	36.855839	7.245400	2500.000000
25%	385.250000	51.000000	670.000000	20006.250000	1.000000	41.802990	9.505090	7122.500000
50%	769.500000	51.000000	1035.000000	39031.000000	1.000000	44.394096	11.869260	9000.000000
75%	1153.750000	51.000000	2616.000000	79667.750000	1.000000	45.467960	12.769040	10000.000000
max	1538.000000	77.000000	4658.000000	235000.000000	4.000000	46.795612	18.365520	11100.000000

```
In [126]: data.head(5)
```

```
Out[126]:
```

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

```
In [128]: data1
```

```
Out[128]:
```

	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

1538 rows × 6 columns

```
In [129]: data2=pd.get_dummies(data1)
```

```
In [130]: data2
```

```
Out[130]:
```

	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 [131]: data2.shape
```

```
Out[131]: (1538, 8)
```

```
In [132]: y=data2['price']  
x=data2.drop('price',axis=1)
```

In [133]:

x

Out[133]:

	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

1538 rows × 7 columns

In [134]:

y

Out[134]:

0	8900
1	8800
2	4200
3	6000
4	5700
...	...
1533	5200
1534	4600
1535	7500
1536	5990
1537	7900

Name: price, Length: 1538, dtype: int64

```
In [135]: 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 [136]: x_test.head(5)
```

Out[136]:

	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
481	51	3197	120000	2	0	1	0
76	62	2101	103000	1	0	1	0
1502	51	670	32473	1	1	0	0
669	51	913	29000	1	1	0	0
1409	51	762	18800	1	1	0	0

```
In [137]: y_test.head(5)
```

Out[137]:

481	7900
76	7900
1502	9400
669	8500
1409	9700

Name: price, dtype: int64

```
In [138]: x_train.head(5)
```

Out[138]:

	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
527	51	425	13111	1	1	0	0
129	51	1127	21400	1	1	0	0
602	51	2039	57039	1	0	1	0
331	51	1155	40700	1	1	0	0
323	51	425	16783	1	1	0	0

```
In [139]: y_train.head(5)
```

```
Out[139]: 527    9990
          129    9500
          602    7590
          331    8750
          323    9100
          Name: price, dtype: int64
```

```
In [140]: import warnings
          warnings.filterwarnings("ignore")
```

```
In [141]: 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)
```

```
Out[141]: ▸ GridSearchCV
          ▸ estimator: Ridge
              ▸ Ridge
```

```
In [142]: ridge_regressor.best_params_
```

```
Out[142]: {'alpha': 30}
```

```
In [143]: ridge=Ridge(alpha=30)
```

```
In [144]: ridge.fit(x_train,y_train)
          y_pred_ridge=ridge.predict(x_test)
```

```
In [145]: from sklearn.metrics import mean_squared_error
          Ridge_Error=mean_squared_error(y_pred_ridge,y_test)
          Ridge_Error
```

```
Out[145]: 579521.7970897449
```

```
In [146]: from sklearn.metrics import r2_score
          r2_score(y_test,y_pred_ridge)
```

```
Out[146]: 0.8421969385523054
```

```
In [147]: #only for lounge model
```

```
In [148]: data2=data.loc[(data.model=='lounge')]
data2
```

Out[148]:

	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
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
6	7	lounge	51	731	11600	1	44.907242	8.611560	10750
7	8	lounge	51	1521	49076	1	41.903221	12.495650	9190
11	12	lounge	51	366	17500	1	45.069679	7.704920	10990
...
1528	1529	lounge	51	2861	126000	1	43.841980	10.515310	5500
1529	1530	lounge	51	731	22551	1	38.122070	13.361120	9900
1530	1531	lounge	51	670	29000	1	45.764648	8.994500	10800
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990

1094 rows × 9 columns

```
In [149]: data3=pd.get_dummies(data2)
```


In [150]: data3

Out[150]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	price	model_lounge
0	1	51	882	25000	1	44.907242	8.611560	8900	1
3	4	51	2739	160000	1	40.633171	17.634609	6000	1
6	7	51	731	11600	1	44.907242	8.611560	10750	1
7	8	51	1521	49076	1	41.903221	12.495650	9190	1
11	12	51	366	17500	1	45.069679	7.704920	10990	1
...
1528	1529	51	2861	126000	1	43.841980	10.515310	5500	1
1529	1530	51	731	22551	1	38.122070	13.361120	9900	1
1530	1531	51	670	29000	1	45.764648	8.994500	10800	1
1534	1535	74	3835	112000	1	45.845692	8.666870	4600	1
1536	1537	51	2557	80750	1	45.000702	7.682270	5990	1

1094 rows × 9 columns

In [151]: data4=data3.drop(['ID','lat','lon'],axis=1)

```
In [152]: data4
```

```
Out[152]:
```

	engine_power	age_in_days	km	previous_owners	price	model_lounge
0	51	882	25000	1	8900	1
3	51	2739	160000	1	6000	1
6	51	731	11600	1	10750	1
7	51	1521	49076	1	9190	1
11	51	366	17500	1	10990	1
...
1528	51	2861	126000	1	5500	1
1529	51	731	22551	1	9900	1
1530	51	670	29000	1	10800	1
1534	74	3835	112000	1	4600	1
1536	51	2557	80750	1	5990	1

1094 rows × 6 columns

```
In [153]: data4.shape
```

```
Out[153]: (1094, 6)
```

```
In [154]: y=data4['price']  
x=data4.drop('price',axis=1)
```

In [155]: x

Out[155]:

	engine_power	age_in_days	km	previous_owners	model_lounge
0	51	882	25000	1	1
3	51	2739	160000	1	1
6	51	731	11600	1	1
7	51	1521	49076	1	1
11	51	366	17500	1	1
...
1528	51	2861	126000	1	1
1529	51	731	22551	1	1
1530	51	670	29000	1	1
1534	74	3835	112000	1	1
1536	51	2557	80750	1	1

1094 rows × 5 columns

In [156]:

y

Out[156]:

0	8900
3	6000
6	10750
7	9190
11	10990

...

1528	5500
1529	9900
1530	10800
1534	4600
1536	5990

Name: price, Length: 1094, dtype: int64

In [157]:

```
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 [158]:

x_test.head(5)

Out[158]:

	engine_power	age_in_days	km	previous_owners	model_lounge
676	51	762	18609	1	1
215	51	701	25000	1	1
146	51	4018	152900	1	1
1319	51	731	20025	1	1
1041	51	640	38231	1	1

In [159]:

y_test.head(5)

Out[159]:

676	10250
215	9790
146	5500
1319	9900
1041	8900

Name: price, dtype: int64

In [160]: `x_train.head(5)`

Out[160]:

	engine_power	age_in_days	km	previous_owners	model_lounge
441	51	762	36448	1	1
701	51	701	27100	1	1
695	51	3197	51083	1	1
1415	51	670	33000	1	1
404	51	456	14000	1	1

In [161]: `y_train.head(5)`

Out[161]:

```
441      8980
701     10300
695      5880
1415     10490
404      9499
Name: price, dtype: int64
```

In [162]: `import warnings`
`warnings.filterwarnings("ignore")`

In [163]: `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)`

Out[163]:

```

> GridSearchCV
  > estimator: Ridge
    > Ridge

```

```
In [164]: ridge_regressor.best_params_    #get alpha value or constant
```

```
Out[164]: {'alpha': 30}
```

```
In [165]: ridge=Ridge(alpha=30)
```

```
In [166]: ridge.fit(x_train,y_train)  
y_pred_ridge=ridge.predict(x_test) #predicted value
```

```
In [167]: from sklearn.metrics import mean_squared_error #rms value  
Ridge_Error=mean_squared_error(y_pred_ridge,y_test)  
Ridge_Error
```

```
Out[167]: 519771.8129989745
```

```
In [168]: from sklearn.metrics import r2_score  
r2_score(y_test,y_pred_ridge) #efficiency
```

```
Out[168]: 0.8373030813683994
```

```
In [169]: results=pd.DataFrame(columns=['actual','Predicted'])
results['actual']=y_test
results['Predicted']=y_pred_ridge
results=results.reset_index()
results['Id']=results.index
results.head(10)
```

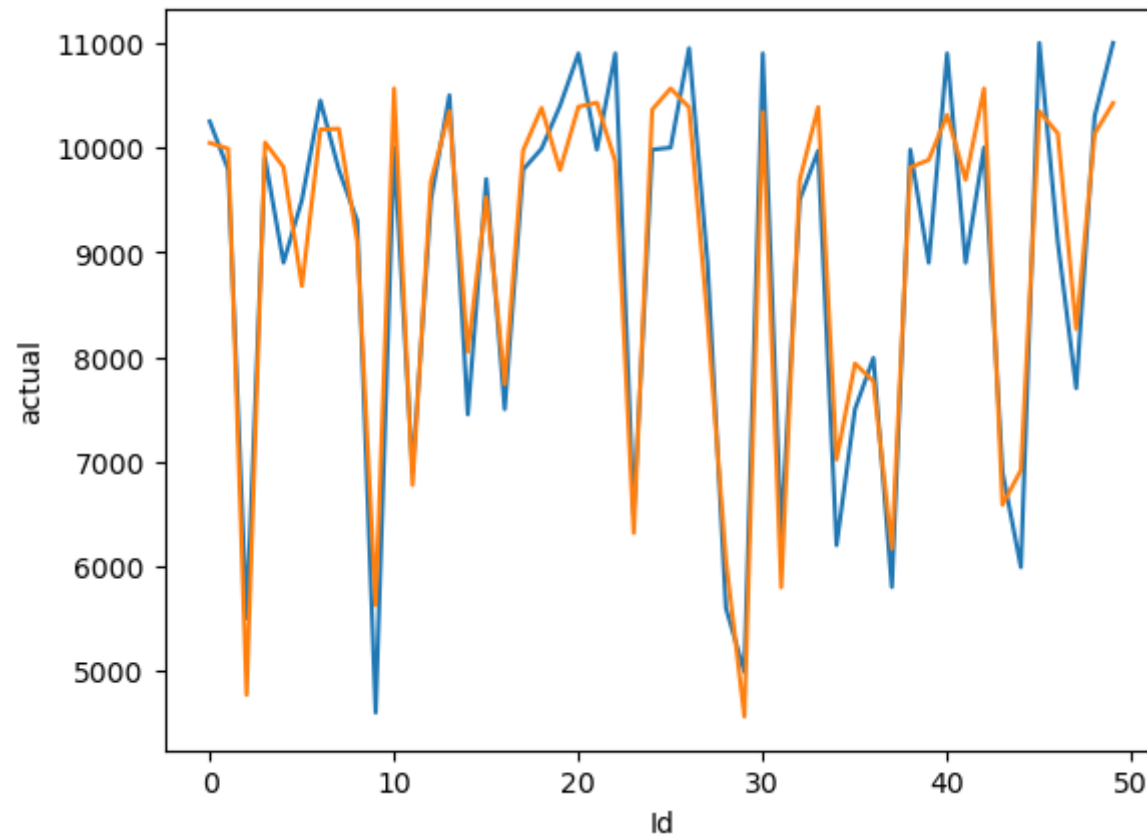
Out[169]:

	index	actual	Predicted	Id
0	676	10250	10045.347779	0
1	215	9790	9989.171535	1
2	146	5500	4769.099603	2
3	1319	9900	10048.683238	3
4	1041	8900	9813.944798	4
5	1425	9500	8678.143561	5
6	409	10450	10173.797921	6
7	617	9790	10180.627008	7
8	1526	9300	9107.315259	8
9	1010	4600	5625.007407	9

```
In [175]: import seaborn as hh
import matplotlib.pyplot as plt
```

```
In [183]: hh.lineplot(x='Id',y='actual',data=results.head(50))  
          hh.lineplot(x='Id',y='Predicted',data=results.head(50))
```

```
Out[183]: <Axes: xlabel='Id', ylabel='actual'>
```



In []: