```
In [1]: import pandas as pd
In [2]: data=pd.read csv('/home/palcement/Downloads/fiat500.csv')
In [3]: data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 1538 entries, 0 to 1537
         Data columns (total 9 columns):
              Column
                                 Non-Null Count
          #
                                                  Dtype
              -----
          0
              ID
                                 1538 non-null
                                                  int64
                                1538 non-null
          1
              model
                                                  object
              engine power
                                1538 non-null
                                                  int64
          3
                                                  int64
              age in days
                                 1538 non-null
          4
                                1538 non-null
                                                  int64
              km
              previous owners 1538 non-null
                                                  int64
                                1538 non-null
                                                  float64
              lat
          7
              lon
                                1538 non-null
                                                  float64
              price
                                1538 non-null
                                                  int64
         dtypes: float64(2), int64(6), object(1)
         memory usage: 108.3+ KB
In [4]:
        data.head(5)
Out[4]:
            ID model engine_power age_in_days
                                               km previous owners
                                                                       lat
                                                                               lon price
                                        882
                                             25000
                                                               1 44.907242
                                                                                   8900
            1 lounge
                              51
                                                                           8.611560
                                             32500
                              51
                                        1186
                                                               1 45.666359 12.241890
                                                                                   8800
                 pop
            3
                sport
                              74
                                       4658
                                            142228
                                                               1 45.503300 11.417840
                                                                                   4200
               lounge
                              51
                                            160000
                                                               1 40.633171 17.634609
                                                                                   6000
                              73
                                       3074 106880
                                                               1 41.903221 12.495650 5700
            5
                 pop
In [5]: data1=data.drop(['lat','lon','ID'],axis=1)
```

In [6]: data

Out[6]:

	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	рор	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 [7]: data1

Out[7]:

	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 [8]: datal=pd.get_dummies(datal)
```

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

In [10]: X

Out[10]:

	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 [11]: y
Out[11]: 0
                 8900
                 8800
         2
                 4200
         3
                 6000
         4
                 5700
                 . . .
                 5200
         1533
         1534
                 4600
         1535
                 7500
         1536
                 5990
         1537
                 7900
         Name: price, Length: 1538, dtype: int64
```

In [12]: data1

Out[12]:

	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 [13]: 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 [14]: X_train

Out[14]:

	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
1130	51	1127	24000	1	1	0	0
1294	51	852	30000	1	1	0	0
860	51	3409	118000	1	0	1	0
1459	51	762	16700	1	1	0	0
1126	51	701	39207	1	1	0	0

1030 rows × 7 columns

Ridge Regression

```
In [15]: | 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/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=5.56109e-26): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.70876e-26): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=6.91585e-23): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.08003e-23): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.01022e-23): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.57959e-23): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.24161e-23): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=6.92759e-21): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.09091e-21): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
```

```
Ill-conditioned matrix (rcond=7.02112e-21): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.57414e-21): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.23284e-21): result may not be accurate.
           return linalq.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=6.9277e-17): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.09099e-17): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.02123e-17): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.57407e-17): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
         /home/palcement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ ridge.py:216: LinAlgWarning:
         Ill-conditioned matrix (rcond=7.23274e-17): result may not be accurate.
           return linalg.solve(A, Xy, assume a="pos", overwrite a=True).T
Out[15]:
          ▶ GridSearchCV
          ▶ estimator: Ridge
                ▶ Ridge
In [16]: ridge_regressor.best_params_
Out[16]: {'alpha': 30}
```

In [17]: ridge=Ridge(alpha=30)

ridge.fit(X train,y train)

y pred ridge=ridge.predict(X test)