In [41]:

2

```
In [35]:
             1
                import pandas as pd
             2
             3
                from sklearn.datasets import fetch_california_housing
                boston = fetch_california_housing()
                boston_df=pd.DataFrame(boston.data , columns= boston.feature_names)
             5
                boston_df['Price_House'] = boston.target
                boston_df
Out[35]:
                   MedInc HouseAge AveRooms AveBedrms Population AveOccup Latitude Longituc
                   8.3252
                                        6.984127
                                                                  322.0
                                                                          2.555556
                                                                                               -122.2
                0
                                41.0
                                                    1.023810
                                                                                      37.88
                1
                   8.3014
                                 21.0
                                        6.238137
                                                    0.971880
                                                                 2401.0
                                                                          2.109842
                                                                                      37.86
                                                                                               -122.2
                   7.2574
                                 52.0
                                        8.288136
                                                    1.073446
                                                                  496.0
                                                                          2.802260
                                                                                      37.85
                                                                                               -122.2
                   5.6431
                                 52.0
                                        5.817352
                                                    1.073059
                                                                  558.0
                                                                          2.547945
                                                                                      37.85
                                                                                               -122.2
                   3.8462
                                 52.0
                                        6.281853
                                                    1.081081
                                                                  565.0
                                                                          2.181467
                                                                                      37.85
                                                                                               -122.2
               ...
                       ...
                                  ...
                                                          ...
                                                                     ...
                                                                                         ...
            20635
                   1.5603
                                 25.0
                                        5.045455
                                                    1.133333
                                                                  845.0
                                                                          2.560606
                                                                                      39.48
                                                                                               -121.0
            20636
                   2.5568
                                 18.0
                                        6.114035
                                                    1.315789
                                                                  356.0
                                                                          3.122807
                                                                                      39.49
                                                                                               -121.2
                   1.7000
            20637
                                 17.0
                                        5.205543
                                                    1.120092
                                                                  1007.0
                                                                          2.325635
                                                                                      39.43
                                                                                               -121.2
            20638
                   1.8672
                                 18.0
                                        5.329513
                                                                  741.0
                                                                          2.123209
                                                                                      39.43
                                                                                               -121.3
                                                    1.171920
            20639
                   2.3886
                                 16.0
                                        5.254717
                                                    1.162264
                                                                  1387.0
                                                                          2.616981
                                                                                      39.37
                                                                                               -121.2
           20640 rows × 9 columns
                x=boston_df.iloc[:,:-1]
In [36]:
In [37]:
               x.shape
Out[37]:
           (20640, 8)
In [38]:
                y = boston_df.iloc[:,-1]
In [39]:
                y.shape
Out[39]: (20640,)
In [40]:
             1
                #scaling
             2
                from sklearn.preprocessing import StandardScaler
                sc = StandardScaler()
                x_sc = sc.fit_transform(x)
```

from sklearn.model_selection import train_test_split

xtrain,xtest,ytrain,ytest = train_test_split(x_sc, y ,test_size=0.3 , r

```
In [42]: 1 #applying linear regression
2 from sklearn.metrics import r2_score,mean_absolute_error
3 from sklearn.linear_model import LinearRegression
4 lr = LinearRegression()
5 lr.fit(xtrain,ytrain)
```

Out[42]: LinearRegression()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [56]: 1 predict = lr.predict(xtest)
2 print(r2_score(ytest,predict))
```

0.592807983357122

```
In [57]: 1 print(mean_absolute_error(ytest,predict))
```

0.5355113615407027

```
In [58]: 1 #applying linear regression
2 from sklearn.linear_model import Ridge
3 rr = Ridge()
4 rr.fit(xtrain,ytrain)
```

Out[58]: Ridge()

In [60]:

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

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print(mean_absolute_error(ytest,predict1))

```
0.5350945553017961
```

```
In [61]:
           1 #now applying repeated k fold
           2 from sklearn.model_selection import RepeatedKFold
           3 cv=RepeatedKFold(n_splits=10,n_repeats=3, random_state=1)
           4 from sklearn.metrics import r2 score
           5 ypred=rr.predict(xtest)
           6 r2_score(ytest,ypred)
           7
           8 from sklearn.preprocessing import StandardScaler
           9 sc= StandardScaler()
          10 x sc = sc.fit transform(x)
          11 xtrain,xtest,ytrain,ytest = train_test_split(x_sc,y, test_size=0.25 , r
          12
             model1=Ridge()
          13
             params={'alpha':[0.00001,0.0001,0.001,0.01,0.1,1,5,10]}
          14
          15 | from sklearn.model_selection import GridSearchCV
          16 | search = GridSearchCV(model1,params,cv=cv)
          17 result=search.fit(x_sc,y)
          18 result.best_params_
          19
Out[61]: {'alpha': 10}
In [54]:
           2 model2 = Ridge(alpha=10)
           3 model2.fit(xtrain,ytrain)
           4 ypred2=model2.predict(xtest)
           5 r2_score(ytest, ypred2)
Out[54]: 0.5930434772040074
In [62]:
           1 #applying lasso
           2 from sklearn.linear_model import Lasso
           3 | lr = Lasso()
           4 | lr.fit(xtrain,ytrain)
```

Out[62]: Lasso()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

0.9099307895899686

```
In [65]:
           1 #now applying repeated k fold
           2 from sklearn.model_selection import RepeatedKFold
           3 cv=RepeatedKFold(n_splits=10,n_repeats=3, random_state=1)
           4 from sklearn.metrics import r2_score
           5 ypred=lr.predict(xtest)
           6 r2_score(ytest,ypred)
          8 from sklearn.preprocessing import StandardScaler
          9 sc= StandardScaler()
          10 x sc = sc.fit transform(x)
          11 xtrain,xtest,ytrain,ytest = train_test_split(x_sc,y, test_size=0.25 , r
          12
             model1=Ridge()
          13
             params={'alpha':[0.00001,0.0001,0.001,0.01,0.1,1,5,10]}
          14
          15 from sklearn.model_selection import GridSearchCV
          16 | search = GridSearchCV(model1,params,cv=cv)
          17 result=search.fit(x_sc,y)
          18 result.best_params_
          19
Out[65]: {'alpha': 10}
In [66]:
          1 model2 = Ridge(alpha=10)
           2 model2.fit(xtrain,ytrain)
           3 ypred2=model2.predict(xtest)
           4 r2_score(ytest, ypred2)
Out[66]: 0.5930434772040074
 In [ ]:
```