Boston_Housing_Price_Prediction

March 1, 2019

```
1
   • OS: Ubuntu 18.04.2 LTS
   • Language: Python 3.7.2
   • Library:
   • numpy 1.16.1
   • pandas 0.24.1
   • sklearn 0.0.9
   • matplotlib 3.0.2
In [1]: from datetime import datetime
        import time
        import numpy as np
        import pandas as pd
        import seaborn as sns
        import matplotlib.pyplot as plt
        import tensorflow as tf
        from sklearn import linear_model
        from sklearn import datasets
        from sklearn.svm import l1_min_c
        from sklearn.linear_model import LassoCV
        import sklearn.model_selection
2
   • raw data : rawdata
   • : txt file
In [2]: ##
        rawdata = pd.read_csv("./housing_data.txt", sep="\s+", header=None)
        rawdata.columns = ["CRIM", "ZN", "INDUS", "CHAS", "NOX", "RM", "AGE", "DIS", "RAD", "TAX", "P
In [3]: #
```

rawdata.head(10)

```
Out[3]:
                                   CRIM
                                                         ZN
                                                                   INDUS CHAS
                                                                                                          NOX
                                                                                                                               RM
                                                                                                                                               AGE
                                                                                                                                                                    DIS RAD
                                                                                                                                                                                                   TAX \
                           0.00632
                                                                      2.31
                                                                                                    0.538
                                                                                                                                             65.2 4.0900
                                                                                                                                                                                              296.0
                    0
                                                 18.0
                                                                                             0
                                                                                                                       6.575
                                                                                                                                                                                      1
                          0.02731
                                                                                                                                                                                             242.0
                    1
                                                      0.0
                                                                      7.07
                                                                                             0 0.469
                                                                                                                       6.421
                                                                                                                                            78.9 4.9671
                                                                                                                                                                                      2
                    2 0.02729
                                                      0.0
                                                                      7.07
                                                                                             0 0.469
                                                                                                                      7.185
                                                                                                                                            61.1
                                                                                                                                                           4.9671
                                                                                                                                                                                             242.0
                    3 0.03237
                                                      0.0
                                                                                             0 0.458
                                                                                                                       6.998
                                                                                                                                            45.8 6.0622
                                                                                                                                                                                             222.0
                                                                      2.18
                          0.06905
                                                      0.0
                                                                      2.18
                                                                                             0 0.458
                                                                                                                     7.147
                                                                                                                                            54.2
                                                                                                                                                           6.0622
                                                                                                                                                                                             222.0
                          0.02985
                                                     0.0
                                                                      2.18
                                                                                             0 0.458 6.430
                                                                                                                                            58.7
                                                                                                                                                            6.0622
                                                                                                                                                                                      3 222.0
                                                                                             0 0.524 6.012
                    6 0.08829 12.5
                                                                      7.87
                                                                                                                                            66.6 5.5605
                                                                                                                                                                                      5 311.0
                    7 0.14455
                                                  12.5
                                                                      7.87
                                                                                             0 0.524 6.172
                                                                                                                                            96.1 5.9505
                                                                                                                                                                                      5 311.0
                    8 0.21124
                                                                                             0 0.524 5.631
                                                 12.5
                                                                      7.87
                                                                                                                                          100.0 6.0821
                                                                                                                                                                                      5 311.0
                    9 0.17004 12.5
                                                                      7.87
                                                                                             0 0.524 6.004
                                                                                                                                            85.9 6.5921
                                                                                                                                                                                      5 311.0
                                                                B LSTAT MEDV
                            PTRATIO
                                                   396.90
                                                                           4.98
                    0
                                    15.3
                                                                                          24.0
                                    17.8
                                                                           9.14
                    1
                                                   396.90
                                                                                          21.6
                                                                          4.03 34.7
                    2
                                    17.8 392.83
                     3
                                    18.7
                                                   394.63
                                                                           2.94 33.4
                                                                           5.33 36.2
                     4
                                    18.7 396.90
                    5
                                    18.7 394.12
                                                                           5.21 28.7
                     6
                                    15.2 395.60 12.43 22.9
                    7
                                    15.2 396.90
                                                                       19.15 27.1
                     8
                                    15.2 386.63
                                                                        29.93 16.5
                                   15.2 386.71 17.10 18.9
In [4]: #
                                 n = 506
                                 p = 14
                    n, p = rawdata.shape
2.1 ,
       • Training data: 404 row(80%)
       • Test data: 102 row(20%)
In [5]: # X, Y data
                    x_raw = rawdata.iloc[:, 0:13]
                    y_raw = rawdata.iloc[:, 13]
                    train_x, test_x, train_y, test_y = sklearn.model_selection.train_test_split(x_raw, y_rain_x, test_x, train_y, test_y = sklearn.model_selection.train_test_split(x_raw, y_rain_x, test_x, train_y, test_y = sklearn.model_selection.train_test_split(x_raw, y_rain_y, y_rain_y
2.1.1 null, null
In [6]: train_x.isnull().sum() # train_x
Out[6]: CRIM
                                                 0
                     ZN
                                                 0
```

INDUS

CHAS

NOX

RM

0

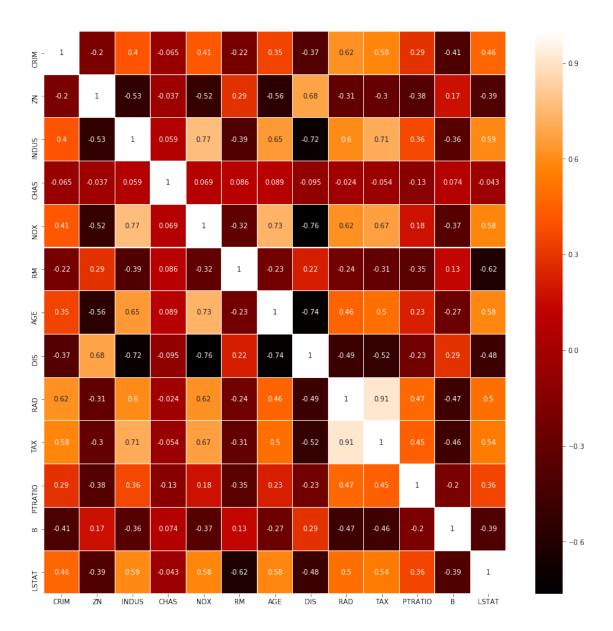
0

0

0

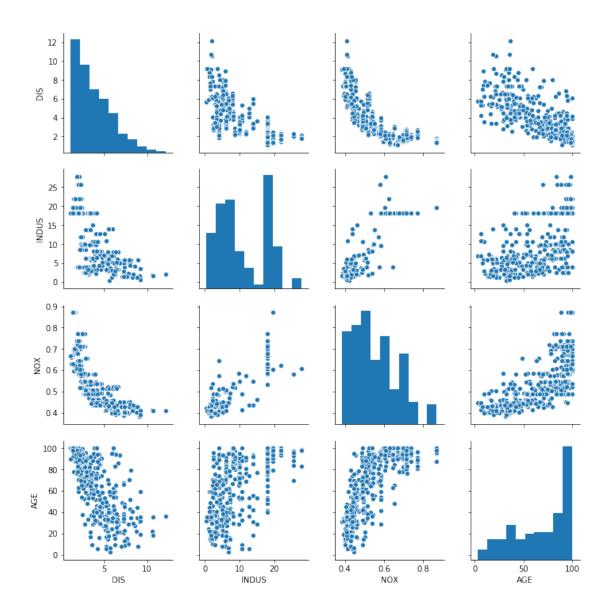
```
AGE
                    0
        DIS
                    0
        RAD
                    0
        TAX
                    0
        PTRATIO
                    0
                    0
        LSTAT
                    0
        dtype: int64
In [7]: train_y.isnull().sum() # train_y
Out[7]: 0
In [8]: test_x.isnull().sum() # text_x
Out[8]: CRIM
                    0
        ZN
                    0
        INDUS
                    0
        CHAS
                    0
        NOX
                    0
        RM
                    0
                    0
        AGE
        DIS
                    0
        RAD
        TAX
        PTRATIO
                    0
        В
                    0
        LSTAT
                    0
        dtype: int64
In [9]: test_y.isnull().sum() # test_y
Out[9]: 0
3
In [10]: train_x.describe() #
Out[10]:
                                      ZN
                       CRIM
                                                INDUS
                                                              CHAS
                                                                           NOX
                                                                                         RM
         count
                 404.000000
                             404.000000
                                          404.000000
                                                       404.000000
                                                                    404.000000
                                                                                 404.000000
                   3.262956
                               11.733911
                                           10.954356
                                                         0.076733
                                                                      0.552274
                                                                                   6.301433
         mean
         std
                   8.052195
                               23.710472
                                            6.878348
                                                         0.266497
                                                                      0.116741
                                                                                   0.726747
         min
                   0.006320
                               0.000000
                                            0.460000
                                                         0.000000
                                                                      0.385000
                                                                                   3.561000
         25%
                   0.071615
                                0.000000
                                            4.950000
                                                         0.000000
                                                                      0.448000
                                                                                   5.888000
         50%
                   0.227290
                                0.000000
                                            8.560000
                                                         0.000000
                                                                      0.524000
                                                                                   6.229500
         75%
                               12.500000
                   2.904685
                                           18.100000
                                                         0.000000
                                                                      0.624000
                                                                                   6.635000
                  88.976200
                             100.000000
                                           27.740000
                                                                                   8.780000
                                                         1.000000
                                                                      0.871000
         max
```

```
AGE
                           DIS
                                        RAD
                                                     TAX
                                                             PTRATIO
                                                                                 В
       404.000000
                    404.000000
                                 404.000000
                                             404.000000
                                                          404.000000
                                                                       404.000000
count
        67.677723
                      3.831537
                                   9.269802
                                             401.368812
                                                           18.383416
                                                                       359.642599
mean
std
        28.435612
                      2.099385
                                   8.636812
                                             167.482335
                                                            2.157794
                                                                        86.938206
         2.900000
                      1.129600
                                   1.000000
                                             187.000000
                                                           12.600000
                                                                         3.500000
min
25%
        42.050000
                      2.111750
                                   4.000000
                                             277.000000
                                                           17.225000
                                                                       376.745000
50%
        76.500000
                      3.298600
                                   5.000000
                                             329.000000
                                                           18.700000
                                                                       391.565000
75%
        93.950000
                      5.218725
                                  12.000000
                                             666.000000
                                                           20.200000
                                                                       396.307500
       100.000000
                     12.126500
                                  24.000000
                                             711.000000
                                                           22.000000
                                                                       396.900000
max
            LSTAT
       404.000000
count
        12.356881
mean
std
         7.275672
min
         1.730000
25%
         6.720000
50%
        10.530000
75%
        16.457500
        37.970000
max
```



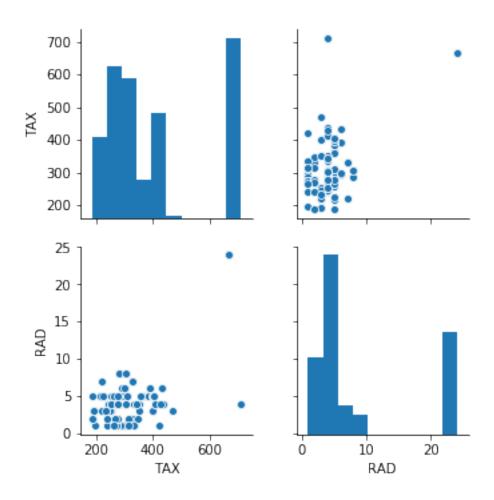
In [177]: sns.pairplot(train_x, vars=["DIS", "INDUS", "NOX", "AGE"])

Out[177]: <seaborn.axisgrid.PairGrid at 0x7f4f17bd78d0>

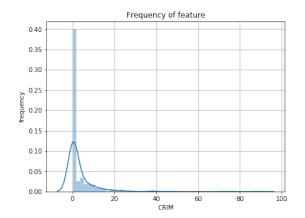


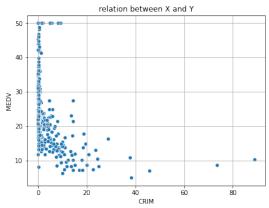
In [178]: sns.pairplot(train_x, vars=["TAX", "RAD"])

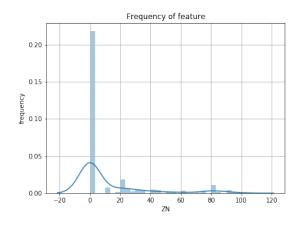
Out[178]: <seaborn.axisgrid.PairGrid at 0x7f4f17e83e10>

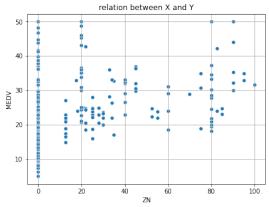


```
In [30]: # 1
    plt.figure(figsize=(15, 5))
    plt.subplot(121)
    plt.title("Frequency of feature")
    plt.grid()
    sns.distplot(train_x['CRIM'])
    plt.ylabel("frequency")
    plt.subplot(122)
    plt.title("relation between X and Y")
    plt.grid()
    sns.scatterplot(train_x['CRIM'], train_y)
    plt.show()
```

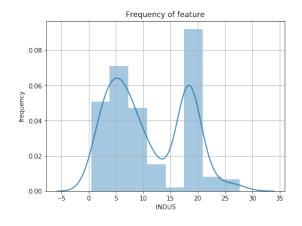


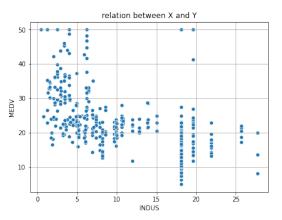




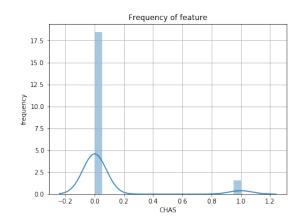


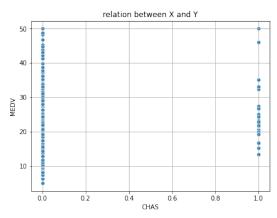
```
plt.grid()
sns.distplot(train_x['INDUS'])
plt.ylabel("frequency")
plt.subplot(122)
plt.title("relation between X and Y")
plt.grid()
sns.scatterplot(train_x['INDUS'], train_y)
plt.show()
```

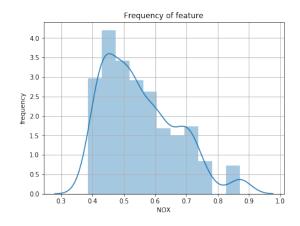


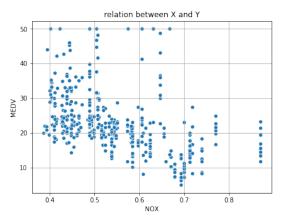


```
In [33]: # ( 1, 0)
    plt.figure(figsize=(15, 5))
    plt.subplot(121)
    plt.title("Frequency of feature")
    plt.grid()
    sns.distplot(train_x['CHAS'])
    plt.ylabel("frequency")
    plt.subplot(122)
    plt.title("relation between X and Y")
    plt.grid()
    sns.scatterplot(train_x['CHAS'], train_y)
    plt.show()
```

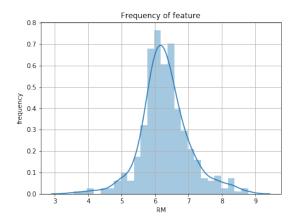


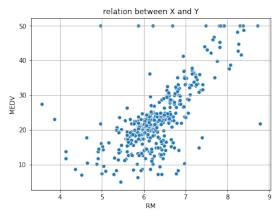


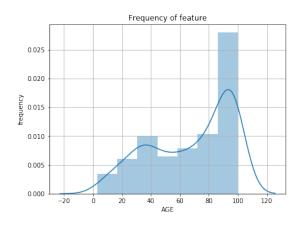




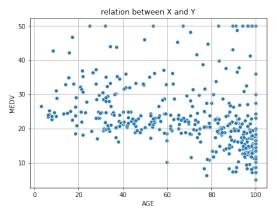
```
In [35]: # 1
    plt.figure(figsize=(15, 5))
    plt.subplot(121)
    plt.title("Frequency of feature")
    plt.grid()
    sns.distplot(train_x['RM'])
    plt.ylabel("frequency")
    plt.subplot(122)
    plt.title("relation between X and Y")
    plt.grid()
    sns.scatterplot(train_x['RM'], train_y)
    plt.show()
```





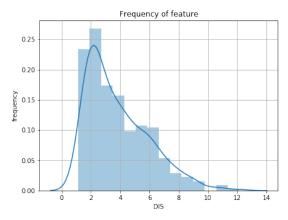


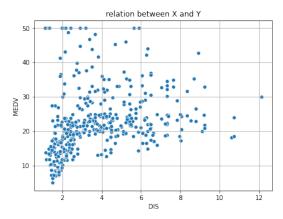
plt.show()



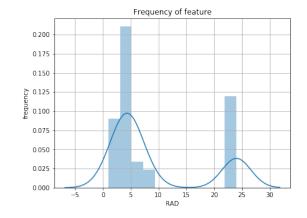
```
In [38]: # 5
          plt.figure(figsize=(15, 5))
          plt.subplot(121)
          plt.title("Frequency of feature")
```

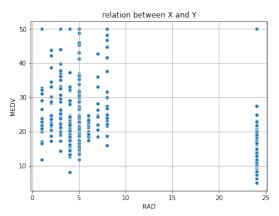
```
plt.grid()
sns.distplot(train_x['DIS'])
plt.ylabel("frequency")
plt.subplot(122)
plt.title("relation between X and Y")
plt.grid()
sns.scatterplot(train_x['DIS'], train_y)
plt.show()
```

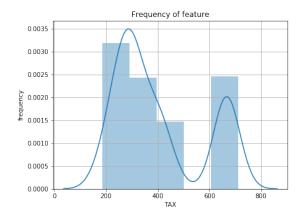


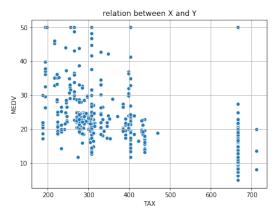


```
In [39]: #
    plt.figure(figsize=(15, 5))
    plt.subplot(121)
    plt.title("Frequency of feature")
    plt.grid()
    sns.distplot(train_x['RAD'])
    plt.ylabel("frequency")
    plt.subplot(122)
    plt.title("relation between X and Y")
    plt.grid()
    sns.scatterplot(train_x['RAD'], train_y)
    plt.show()
```

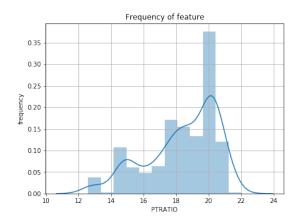


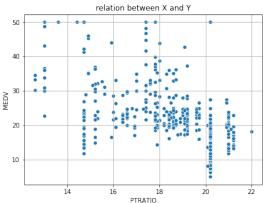


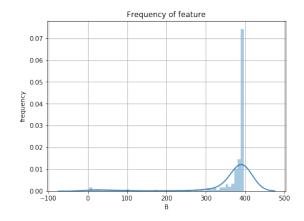




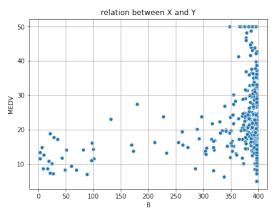
```
In [41]: # /
    plt.figure(figsize=(15, 5))
    plt.subplot(121)
    plt.title("Frequency of feature")
    plt.grid()
    sns.distplot(train_x['PTRATIO'])
    plt.ylabel("frequency")
    plt.subplot(122)
    plt.title("relation between X and Y")
    plt.grid()
    sns.scatterplot(train_x['PTRATIO'], train_y)
    plt.show()
```



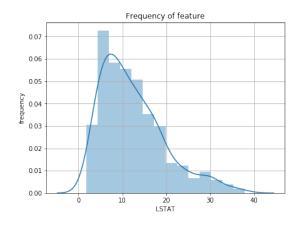


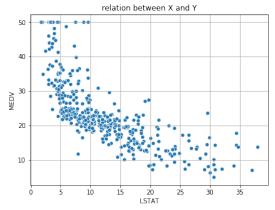


plt.show()



```
plt.grid()
sns.distplot(train_x['LSTAT'])
plt.ylabel("frequency")
plt.subplot(122)
plt.title("relation between X and Y")
plt.grid()
sns.scatterplot(train_x['LSTAT'], train_y)
plt.show()
```



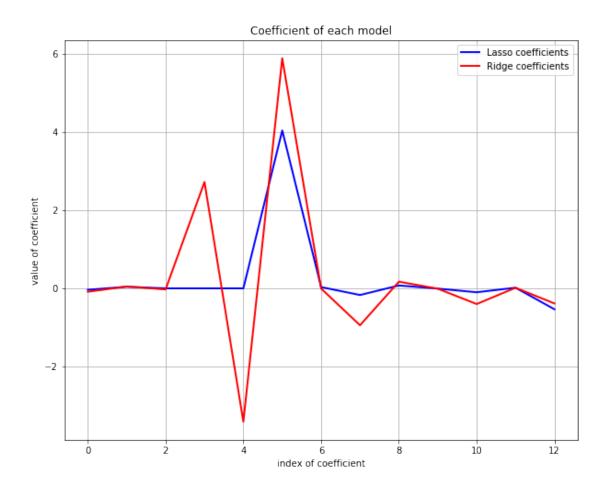


4

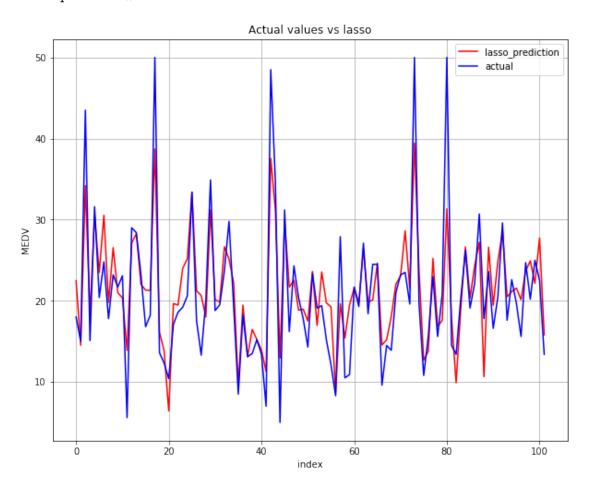
4.1 Model1 = Lasso regression(L1 regularization)

4.2 Model2 = Ridge regression(L2 regularization)

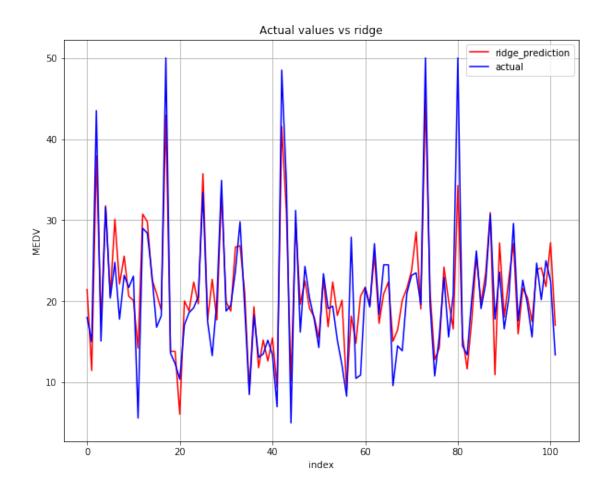
```
Out[166]: Ridge(alpha=0.0, copy_X=True, fit_intercept=False, max_iter=None,
             normalize=True, random_state=None, solver='auto', tol=1e-06)
In [167]: #
          coef_ridge = clf_ridge.coef_
          print(coef_ridge)
[-8.61158712e-02 4.64502631e-02 -2.65007636e-02 2.71886515e+00
-3.40834404e+00 5.88156430e+00 -5.06153133e-03 -9.42950569e-01
 1.70793544e-01 -8.68774357e-03 -4.00698002e-01 1.53968223e-02
-3.85297737e-01]
In [168]: # Lasso vs Ridge plot
          plt.figure(figsize=(10, 8))
          plt.title("Coefficient of each model")
          plt.grid()
          plt.plot(coefficient, color='blue', linewidth=2, label='Lasso coefficients')
          plt.plot(coefficient_ridge, color='red', linewidth=2, label='Ridge coefficients')
          plt.xlabel("index of coefficient")
          plt.ylabel("value of coefficient")
          plt.legend()
          plt.show
Out[168]: <function matplotlib.pyplot.show(*args, **kw)>
```



```
plt.title("Actual values vs lasso")
plt.grid()
plt.plot(compare_y['index'], compare_y['predict_lasso'], 'r-', label='lasso_prediction
plt.plot(compare_y['index'], compare_y['real_MEDV'], 'b-', label='actual')
plt.xlabel("index")
plt.ylabel("MEDV")
plt.legend()
plt.show()
```



```
In [173]: # () vs Ridge () plot
    plt.figure(figsize=(10, 8))
    plt.title("Actual values vs ridge")
    plt.grid()
    plt.plot(compare_y['index'], compare_y['predict_ridge'], 'r-', label='ridge_prediction
    plt.plot(compare_y['index'], compare_y['real_MEDV'], 'b-', label='actual')
    plt.xlabel("index")
    plt.ylabel("MEDV")
    plt.legend()
    plt.show()
```



In [174]: # Lasso R^2 value

clf_lasso.score(test_x, test_y)

Out[174]: 0.7486259621705256

In [175]: # Ridge R^2 value

clf_ridge.score(test_x, test_y)

Out[175]: 0.8145234131230182