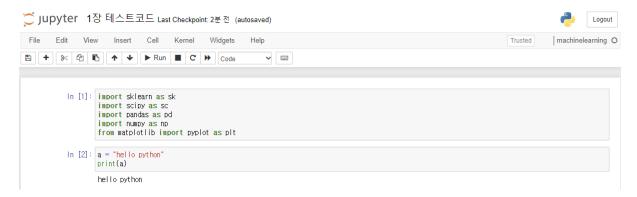
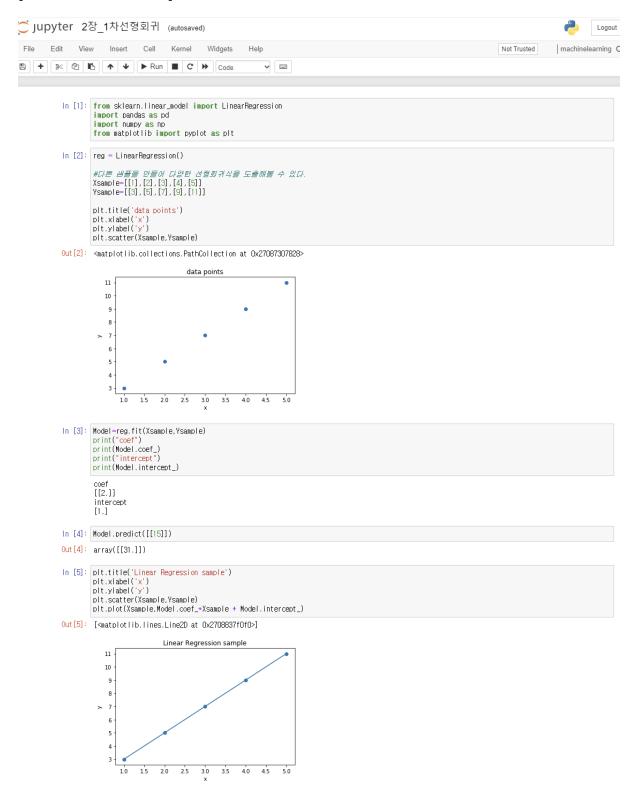
# 1주차 실습과제

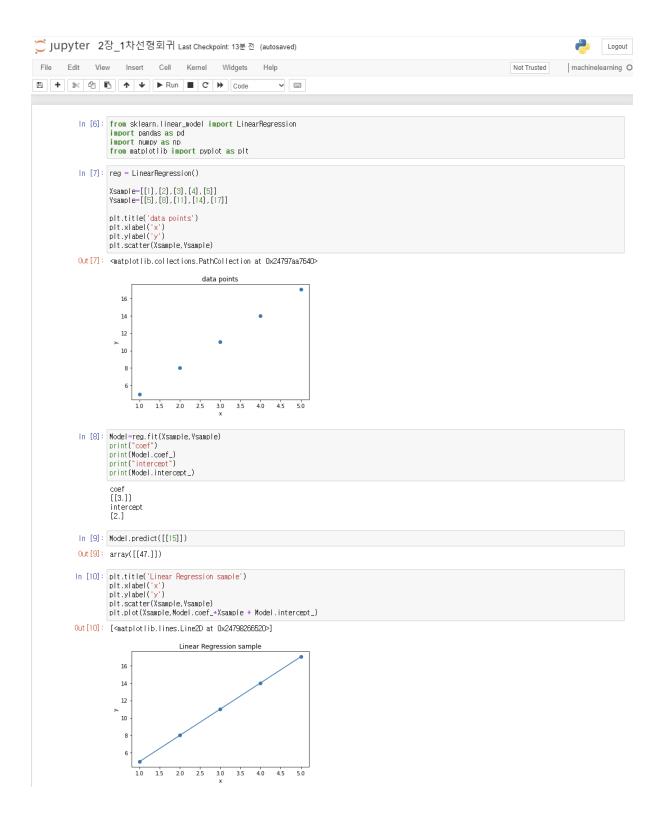
2016314786 김호진

# [1장 실습 - 테스트코드]

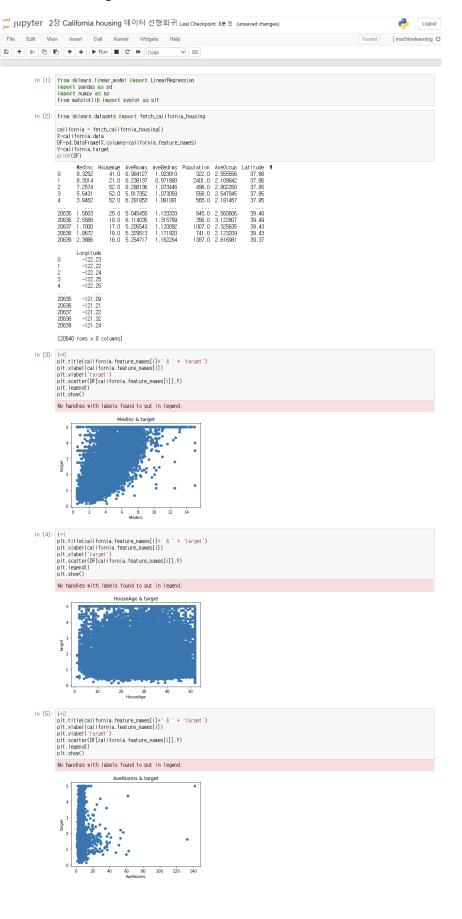


# [2장 실습 - 1차선형회귀]





## [2장 실습 - California housing 데이터 선형회귀]



```
In [6]: i=3
    plt.title(california.feature_names[i]*' & ' + 'target')
    plt.xlabel(california.feature_names[i])
    plt.ylabel('target')
    plt.scatter(Dr[california.feature_names[i]), y)
    plt.lepend()
    plt.show()
                    No handles with labels found to put in legend.
                                                                 AveBedrms & target
                       target
  In [7]: |=4 plt.title(california.feature_names[i]+' & ' + 'target') plt.xiabel(california.feature_names[i]) plt.yiabel('target') plt.scatter([Flcalifornia.feature_names[i]], Y) plt.legnd() plt.show()
                    No handles with labels found to put in legend.
                                                                  Population & target
                                                                 00 15000 20000 25000 30000
  In (8): i=5
    plt.title(california.feature_names[i]*' & ' + 'target')
    plt.xlabel(california.feature_names[i])
    plt.ylabel('target')
    plt.scatter(DF[california.feature_names[i]],Y)
    plt.lepend()
    plt.show()
                    No handles with labels found to put in legend.
                       target
  In [9]: i=6
    plt.title(california.feature_names[i]*' & ' + 'target')
    plt.xlabel(california.feature_names[i])
    plt.ylabel('target')
    plt.scatter(DF[california.feature_names[i]],Y)
    plt.lepend()
    plt.show()
                    No handles with labels found to put in legend.
                       target
In [10]: i=7
    plt.title(california.feature_names[i]*'& ' + 'target')
    plt.xlabe(california.feature_names[i])
    plt.ylabe('target')
    plt.scatter(DF[california.feature_names[i]],Y)
    plt.legend()
    plt.show()
                    No handles with labels found to put in legend.
                                                                       -120 -118
Longitude
                                                                                                             -116
                                     -124
                                                     -122
```

```
In [11]: reg = LinearRegression()
Model=reg.fit(X,Y)
print("coef")
print(Model.coef_)
print("intercept")
print(Model.intercept_)
                                             coef
[ 4.36693293e-01 9.43677803e-03 -1.07322041e-01 6.45065694e-01 -3.97638942e-06 -3.78654265e-03 -4.21314378e-01 -4.34513755e-01]
                                               intercept
-36.941920207184396
In [12]:

| Dit. title(california.feature_names[i]*' & ' * 'target')
| Dit. xlabel(california.feature_names[i])
| Dit. ylabel('target')
| Dit. scatter(DF[california.feature_names[i]], Y)
| Dit. plot(DF[california.feature_names[i]], W)
| Dit. plot(DF[california.feature_names[i]], Nodel.coef_[i]*DF[california.feature_names[i]], 'r-')
| Dit. show()
| Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Dit. show() | Di
                                                                                                                                              MedInc & target
                                                  target
w
  In [13]: i=1
    plt.title(california.feature_names[i])* & ' * 'target')
    plt.xlabel(california.feature_names[i])
    plt.ylabel('target')
    plt.scatter(DF[california.feature_names[i]], Y)
    plt.polt(DF[california.feature_names[i]], Model.coef_[i]*DF[california.feature_names[i]], 'r-')
    plt.legend()
    plt.sbow()
                                              No handles with labels found to put in legend.
                                                                                                                                       HouseAge & target
                                                  target
                                                                                                                                                 20 30
HouseAge
                                                                                                                                                                                                                           40
 In [14]: i=2
plt.title(california.feature_names[i]+' & ' + 'target')
plt.xlabel(california.feature_names[i])
plt.ylabel('target')
plt.scatter(DF[california.feature_names[i]], Y)
plt.plot(DF[california.feature_names[i]], Model.coef_[i]+DF[california.feature_names[i]], 'r-')
plt.legend()
plt.show()
                                              No handles with labels found to put in legend.
                                                                                                                                                       AveRooms & target
                                                                    2.5 -
                                                                    0.0
                                                                 -2.5
                                                 = −5.0
= −7.5
                                                            -10.0
                                                              -12.5
                                                              -15.0
                                                                                                                                                                                                                           100
                                                                                                                                                                                                                                                      120
  In [15]: i=3
plt.title(california.feature_names[i]+' & ' + 'target')
plt.xiabel(california.feature_names[i])
plt.ylabel('target')
plt.scatter(DF[california.feature_names[i]], Y)
plt.plot(DF[california.feature_names[i]], Model.coef_[i]*DF[california.feature_names[i]], 'r-')
plt.legend()
plt.show()
                                              No handles with labels found to put in legend.
                                                                                                                                           AveBedrms & target
                                                            20
                                                  target
10
```

```
In (16): i=4
    plt.title(california.feature_names[i]* % % '* 'target')
    plt.xlabel(california.feature_names[i])
    plt.ylabel('target')
    plt.scatter(DFcalifornia.feature_names[i]], y)
    plt.plot(DF[california.feature_names[i]], Wodel.coef_[i]*DF[california.feature_names[i]], 'r-')
    plt.legend()
    plt.show()
                       No handles with labels found to put in legend.
                                                                     Population & target
                                                 5000 10000 15000 20000 25000 30000 35000
Population
In [17]: i=5
plt.title(california.feature_names[]+' & ' + 'target')
plt.xlabel(california.feature_names[i])
plt.ylabel('target')
plt.scatter(DF(california.feature_names[i]], y)
plt.plot(DF(california.feature_names[i]], Model.coef_[i]*DF(california.feature_names[i]], 'r-')
plt.legend()
plt.show()
                                                                         AveOccup & target
                         target
                                                                                    600
AveOccup
In [18] i=6
ptt.title(california.feature_names[i]*' & ' * 'target')
ptt.xiabel(california.feature_names[i])
ptt.ylabel('target')
ptt.scatter(DF(california.feature_names[i]], Y)
ptt.plot(DF(california.feature_names[i]], Wodel.coef_[i]*DF(california.feature_names[i]], 'r-')
ptt.legend()
ptt.show()
                                                                            Latitude & target
                        target
                              -10
                              -15
 In [19]: i=7
    plt.title(california.feature_names[])+' & ' + 'target')
    plt.xiabel(california.feature_names[])
    plt.ylabel('target')
    plt.scatter(DF[california.feature_names[]], Y)
    plt.plot(DF[california.feature_names[]], Model.coef_[] + DF[california.feature_names[]], 'r-')
    plt.legend()
    plt.schow()
                       No handles with labels found to put in legend.
                                                                       Longitude & target
                       target
06
                                                                                                                     -116
                                                            -122
                                                                                -120 -118
Longitude
  In [20]: DF.mean()
Out [20]: Medinc
HouseAge
AveBooms
AveBedrims
Population
AveCocup
Latitude
Longitude
dtype: float64
                                                     3.870671
28.639486
5.429000
1.096675
1425.476744
3.070655
35.631861
-119.569704
  In [21]: Model.predict([DF.mean()])
  Out [21]: array([2.06855817])
```

### [3장 실습 - Lasso, Ridge 정규화]

```
♡ Jupyter 3장 Lasso, Ridge 정규화 Last Checkpoint: 8분전 (unsaved changes)
File Edit View Insert Cell Kernel Wildgets Help

□ + 3< □ □ + ► Run ■ □ → □ code ✓ □
                                                                                                                                                                                                                                                                                                                                                                                            Trusted | machinelearning O
                                                    In [1]: from sklearn.linear_model import LinearRepression import pands as sod import number as so from satiotitib import solet as pit from sklearn.linear_model import Ridge, Lasso
                                                              In [2]: from sklearn.datasets import fetch_california_housing
                                                                                california = fetch_california_housing()
X-california_data
IF-pd_DataFrame(X_columns-california_feature_names)
P-california_target
print(IF)
                                                                                           | Medical Househope Averboars Averboars Population Averboars Latitude # | 10 | 0.2552 | 41.0 | 0.394127 | 1.025500 | 0.252.0 | 2.055005 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 | 0.376.0 

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        25.0
        5.04545
        1.133333
        945.0
        2.58628
        19.40

        20536
        2.5568
        18.0
        6.114055
        1.337679
        56.0
        3.122007
        9.40

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        7.20554
        1.120362
        1007.0
        2.20505
        9.43

        20539
        1.8972
        18.0
        5.32543
        1.179302
        74.0
        2.12303
        9.43

        20539
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                                                                                                  Longitude
0 -122.23
1 -122.22
2 -122.24
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4 -122.25
                                                              In [3]: reg = LinearRegression()
Model = reg.flt(X,Y)
print("coef")
print(Model.coef_)
print(Model.coef_)
print(Model.intercept_)
                                                                                           coef
[ 4.36632536-01 9.435778036-03 -1.073223416-01 6.4500993640-01
-3.57638426-05 -3.789542656-03 -4.233145786-01 -4.345137656-01]
Intercent
-3.5.34152007184396
                                                                                     rid-Ridoe(alpha-ALPHA)
Model-rid-fit(X,Y)
print("coef")
print(Model.coef_)
print("intercept")
print(Model.intercept_)
                                                                                           coef
[ 4.3668397e-01 9.43993990e-03 -1.07303096e-01 6.44965230e-01 -5.97576456e-06 -3.78652421e-03 -4.21312879e-01 -4.34510869e-01]
                                                                                             intercept
-36.9415871633609
                                                           In [5]: ALPHA=0.5
                                                                                           coef
[ 4.35643795e-01 - 3.43559573e-03 -1.07227325e-01 - 6.44550954e-01 -
-3.97359500e-05 -3.7855054e-03 -4.27350954e-01 -4.3449554e-01]
Intercoek
-35.34023578950354
                                                              In [6]: ALPHA-1
                                                                                        conf [ 4, 9854280a-01 9, 43738519a-03 -1, 07132781a-01 6, 44082856-01 -3, 97034256-06 -3, 78035959a-03 -4, 21295005a-01 -4, 34494717a-01 ] interiors; -36, 936565232337
                                                           In [7]: ALPHA=2
                                                                                           coef
[ 4 .56458000+01 9.43901105+03 -1.09944000+01 6.400024234+01
-3.56450115+06 -3.780175778+33 -4.21284056+01 -4.34455500+01
intercept
-35.5550201400057
                                                                                | las-Lasso(a|oha-ALPHA)
| Model = las.fit(X,Y)
| print("coef")
| print(Model.coef_)
| print("intercept")
| print(Model.intercept_)
                                                                                     coef
[3.9962657e-01 1.59621512e-02 -0.0000000e-00 0.0000000e+00
1.7901960e-05 -3.3225135e-03 -1.14214450e-01 -9.92250999-02
intercent
-7.59459919173792
                                                              In [9]: ALPHA=0.5
                                                                                             cost | 1.2.8854841e-01 | 1.3331451e-02 | 0.0000000e+00 | -0.0000000e+00 | 1.7810540e-05 | -0.00000000e+00 | -0.0000000e+00 | -0.0000000e+00 | -0.0000000e+00 | -0.0000000e+00 | -0.00000000e+00 | -0.0000000e+00 | -0.000000e+00 | -0.0000000e+00 | -0.000000e+00 | -0.000000e+00 | -0.000000e+00 | -0.0000000e+00 | -0.000000e+00 | -0.0000000e+00 | -0.0000000e+00 | -0.0000000e+00 | -0.000
                                                         In [10]: ALPHA=1
                                                                                             coef
[1.45469232e-01 5.81496884e-03 0.00000000e+00 -0.00000000e+00
-6.37282607e-05 -0.00000000e+00 -0.00000000e+00 -0.00000000e+00
```

#### [3장 실습 - 선형회귀 변수 선택]

```
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                                                                                                                                                                                                Trusted machinelearning (
B + % 2 B ↑ ↓ Run ■ C > Code
                In [1]: import warnings warnings.filterwarnings(action='ignore')
                 In [2]: from sklearn.linear_model import LinearRegression
                             import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
                             from sklearn.linear_model import Ridge, Lasso
                In [3]: from sklearn.datasets import fetch_california_housing
                             california = fetch_california_housing()
                             X=california.data
DF=pd.DataFrame(X,columns=california.feature_names)
                            Y=california.target
print(DF)

        MedInc
        HouseAge
        AveRooms
        AveBedrins
        Population
        AveOccup
        Latitude
        W

        8.3252
        41.0
        6.984127
        1.023810
        322.0
        2.555556
        37.88

        8.3014
        21.0
        6.238137
        0.971880
        2401.0
        2.109842
        37.85

        7.2574
        52.0
        8.288136
        1.073446
        495.0
        2.902260
        37.85

        5.6431
        52.0
        5.817352
        1.073059
        558.0
        2.547945
        37.85

                             4
                                       3.8462
                                                         52.0 6.281853
                                                                                 1.081081
                                                                                                       565.0 2.181467
                             20635 1.5603
                                                         25.0 5.045455
                                                                                  1.133333
                                                                                                       845.0 2.560606
                                                                                                                                     39.48
                                                                                                     356.0 2.500006
356.0 3.122807
1007.0 2.325635
741.0 2.123209
1387.0 2.616981
                                       2.5568
                                                         18.0 6.114035
                                                                                  1.315789
                                                                                                                                     39.43
39.43
                                                        17.0 5.205543
18.0 5.329513
                             20637
                                       1.7000
                                                                                  1.120092
                                       1.8672
                                                                                    .171920
                             20639 2.3886
                                                         16.0 5.254717
                                                                                 1.162264
                                                                                                                                     39.37
                                       Longitude
                                          -122.23
-122.22
                             0
                                          -122.24
                             3
                                          -122.25
                             20635
                                          -121.09
                             20636
20637
                                         -121.21
-121.22
                             20638
                                          -121.32
                                          -121.24
                             [20640 rows x 8 columns]
                 In [4]: import statsmodels.api as sm
                             def forward_selection(data,target,cutoff = 0.05):
    initial_features = data.columns.tolist()
    best_features = []
    while(len(initial_features)>0):
                                        remaining_features = list(set(initial_features)-set(best_features))
new_pval = pd.Series(index=remaining_features)
                                         for new_column in remaining_features:
   model = sm.OLS(target, sm.add_constant(data[best_features+[new_column]])).fit()
   new_pval[new_column] = model.pvalues[new_column]
                                        min_p_value = new_pval.min()
if(min_p_value < cutoff):
                                              best_features.append(new_pval.idxmin())
                                              break
                                   return best_features
                 In [5]: forwarddata=forward_selection(DF,Y,0.000000000000001)
                             print(forwarddata)
                             ['MedInc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms']
                In [6]: forwarddata=forward_selection(DF,Y,0.01)
print(forwarddata)
                             ['MedInc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms', 'AveOccup']
                 In [7]: forwarddata=forward_selection(DF,Y,O.1)
                             print(forwarddata)
                             ['Medinc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms', 'AveOccup']
                 In [8]: forwarddata=forward_selection(DF,Y,1)
                             ['MedInc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms', 'AveOccup', 'Population']
```

```
In [9]: def backward_elimination(data, target, cutoff= 0.05):
    features = data.columns.tolist()
    while(len(features) > 0):
        features.with.constant = sm.add_constant(data[features])
        p.values = sm.0.S(farget, features_with_constant).fit().pvalues[i:]
        max_p.value = p.values.max()
        if(max_p.value >= cutoff):
        excluded_feature-p.values.idmax()
        features.remove(excluded_feature)
        else:
            break
    return features
 In [10]: backwarddata=backward_elimination(DF,Y,0.000000000000001) print(backwarddata)
               ['Medinc', 'HouseAge', 'AveRooms', 'AveBedrms', 'Latitude', 'Longitude']
 In [11]: backwarddata-backward_elimination(DF,Y,0.01) print(backwarddata)
               ['Medinc', 'HouseAge', 'AveRooms', 'AveBedrms', 'AveOccup', 'Latitude', 'Longitude']
 In [12]: backwarddata=backward_elimination(DF,Y,0.1) print(backwarddata)
               ['Medinc', 'HouseAge', 'AveRooms', 'AveBedrms', 'AveOccup', 'Latitude', 'Longitude']
 In [13]: backwarddata=backward_elimination(DF,Y,1) print(backwarddata)
               ['Medinc', 'HouseAge', 'AveRooms', 'AveBedrms', 'Population', 'AveOccup', 'Latitude', 'Longitude']
else:
break
                   else:
break
return best_features
 In [15]: stepdata-stepwise_selection(DF,Y,0.000000000000000) print(stepdata)
               ['Medinc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms']
 In [16]: stepdata-stepwise_selection(DF,Y,0.01) print(stepdata)
               ['Medinc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms', 'AveOccup']
 In [17]: stepdata=stepwise_selection(DF,Y,0.1) print(stepdata)
               ['MedInc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms', 'AveOccup']
 In [18]: stepdata=stepwise_selection(DF,Y,1) print(stepdata)
               ['MedInc', 'HouseAge', 'Latitude', 'Longitude', 'AveBedrms', 'AveRooms', 'AveOccup', 'Population']
In [19]: i-4
plt.title(california feature_names[i]*' & ' * 'target')
plt.xiabel(california.feature_names[i])
plt.yiabel('target')
plt.scatter(OF[california.feature_names[i]),V)
plt.legend()
plt.show()
               No handles with labels found to put in legend.
                                            Population & target
                               5000 10000 15000 20000 25000 30000 3500
Population
 In [20]: i=5
    plt.title(california.feature_names[i]*' & ' + 'target')
    plt.xiabel(california.feature_names[i])
    plt.yabel('target')
    plt.scatter(DF[california.feature_names[i]],Y)
    plt.legend()
    plt.show()
               No handles with labels found to put in legend.
                target
```