ML6

May 23, 2022

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[2]: # Apply Linear, Ridge, Lasso Regression technique of machine learning
     # to analyze and build the model of the Diabetesdataset.
     # Display and compare the accuracy(Cross-Validation, R2 Score)
     # of all the models
[2]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     from sklearn.linear_model import LinearRegression, Ridge, RidgeCV, Lasso,
     from sklearn.datasets import load_diabetes
     from sklearn.model_selection import cross_val_score
     from sklearn.metrics import r2_score
[3]: diabetes = load diabetes()
     X = diabetes.data
     y = diabetes.target
[9]: X.shape
[9]: (442, 10)
[4]: print(X.shape)
    print(y.shape)
    (442, 10)
    (442,)
[5]: | lr = LinearRegression()
     lr.fit(X,y)
     y_predict = lr.predict(X)
     print(lr.score(X,y))
     lr_scores = cross_val_score(lr,X,y,cv=5)
     print(lr_scores.mean())
     print(lr.coef_)
     print(lr.intercept_)
     print(r2_score(y,y_predict))
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0.5177494254132934
     0.4823181221114939
     476.74583782 101.04457032 177.06417623 751.27932109
                                                            67.62538639]
     152.1334841628965
     0.5177494254132934
[30]: # Lasso
     lr = Lasso(alpha=0.05)
     lr.fit(X,y)
     y_predict = lr.predict(X)
     print(lr.score(X,y))
     lr_scores = cross_val_score(lr,X,y,cv=5)
     print(lr_scores.mean())
     # print(lr.coef_)
     # print(lr.intercept_)
     0.5131486010621044
     0.48204217044282477
     Γ -0.
                  -194.04749923 521.81917521 295.22384739 -99.44970769
                  -222.72076612 0.
                                              512.04940782
                                                            52.92594445]
     152.13348416289645
[31]: # LassoCV
     lr = LassoCV(alphas=[0.01,0.02,0.04,0.05,0.1,0.2,0.5])
     lr.fit(X,y)
     y_predict = lr.predict(X)
     print(lr.alpha_)
     print(lr.score(X,y))
     lr_scores = cross_val_score(lr,X,y,cv=5)
     print(lr_scores.mean())
     # print(lr.coef_)
     # print(lr.intercept_)
     0.04
     0.5138646937185254
     0.47995535949978924
     [ -0.
                  -202.27296915 522.96840122 299.88446204 -118.13019877
                  -212.94729492
                                              517.11486339
       -0.
                                16.5039356
                                                           56.3117281 ]
     152.13348416289645
[32]: # Ridge
     lr = Ridge(alpha=0.05)
     lr.fit(X,y)
     y_predict = lr.predict(X)
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print(lr.score(X,y))
                          lr_scores = cross_val_score(lr,X,y,cv=5)
                          print(lr_scores.mean())
                          # print(lr.coef_)
                          # print(lr.intercept_)
                       0.5144431310030684
                       0.48081634735342826
                        \begin{bmatrix} -2.69045402 & -221.28733251 & 507.59872561 & 311.48379569 & -137.96353066 \end{bmatrix} 
                                -36.98110839 -174.62635298 113.63280903 483.13024134
                                                                                                                                                                                                                                                                                      79.20916664]
                       152.13348416289642
[33]: # LassoCV
                          lr = RidgeCV(alphas=[0.01,0.02,0.04,0.05,0.1,0.2,0.5])
                          lr.fit(X,y)
                          y_predict = lr.predict(X)
                          print(lr.alpha_)
                          print(lr.score(X,y))
                          lr_scores = cross_val_score(lr,X,y,cv=5)
                          print(lr_scores.mean())
                          # print(lr.coef_)
                          # print(lr.intercept_)
                       0.01
                       0.5166287840315842
                       0.4802027531860105
                       [ \quad -7.19945679 \quad -234.55293001 \quad 520.58313622 \quad 320.52335582 \quad -380.60706569 \quad -380.6070659 \quad -380.6070659 \quad -380.6070659 \quad -380.6070659 \quad -380.6070659 \quad -380.6070659 \quad -380.607069 \quad -380.6070000000000000000
                                150.48375154 -78.59123221 130.31305868 592.34958662
                                                                                                                                                                                                                                                                                       71.1337681 ]
                       152.13348416289645
    []:
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