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Project: Linear Regression analysis on "advertising dataset" I have downloaded the "Advertising.csv" file from "An Introduction to Statistical Learning with Applications in R. This project aims to show my skills in data analysis and linear regression modeling using Python and scikit-learn.

My work is for educational purposes.

In [218... # Python_Machine_Learning_Linear_Regression_project_02

Cross Validation

dtype: float64

-26.55120437 -13.73958657]

 $[-3.54007101 \ -27.85277703 \ -36.30204922 \ -10.95084931 \ -20.36331675$

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Make Training and Testing datasets from the "Advertising.csv" file.
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In [221... # Read the .csv file
         import pandas as pd
         df = pd.read_csv("Advertising.csv")
         # Choose X featurers
         X = df.drop('sales', axis = 1)
         # Choose y response * sales column)
         y = df['sales']
         # Split the X and y into Train and Test parts
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25, random_state = 101)
         # Scale the X dataset.
         # load StandardScaler
         from sklearn.preprocessing import StandardScaler
         scaler= StandardScaler()
         # Scalar fit on the training and testing X Data.
         scaler.fit(X_train)
         X_train = scaler.transform(X_train)
         X_test = scaler.transform(X_test)
In [223... # Create a Model
         # Use Ridge Model
         # Load model
         from sklearn.linear_model import Ridge
         # Make a model
         model = Ridge(alpha = 10) # Choose alpha
         # Fit the above Model on X_Train, and y_train dataset
         model.fit(X_train, y_train)
         # Predict y values on the X_test data using the fitted model.
         y_predicted = model.predict(X_test)
         # find mean squared error from y_test, and y_pred values
         from sklearn.metrics import mean_squared_error
         mean_squared_error(y_test, y_predicted)
Out[223... 2.3043155882835635
        # Adjust parameters as Necessary and repeat just last two steps.
         # Now repeat the above steps using different alpha values.
         model_10 = Ridge(alpha = 0.1)
         model_10.fit(X_train, y_train)
         y_predicted_10 = model_10.predict(X_test)
         mean_squared_error(y_test, y_predicted_10)
Out[225... 2.1096939229756075
In [227... # To minimize the data leakage, dataset is divided into three paarts (train, validation, and test). test data is used for final model test.
         # First split dataset
         X_train, X_test1, y_train, y_test1 = train_test_split(X, y, test_size = 0.25, random_state = 101)
         # second split dataset.
         X_train2, X_test, y_test2, y_test = train_test_split(X_test1, y_test1, test_size = 0.4, random_state = 101)
         # Use StandardScaler
         from sklearn.preprocessing import StandardScaler
         scalar = StandardScaler()
         # fit scalar on X_train data
         scaler.fit(X_train)
         # transform for three datasets
         X_train = scaler.transform(X_train)
         X_train2 = scaler.transform(X_train2)
         X_test = scaler.transform(X_test)
         # Make a linear model
         from sklearn.linear_model import Ridge
         model1 = Ridge(alpha = 100)
         # fit train datasets.
         model1.fit(X_train, y_train)
         # predict on X_train2
         y_test2_predict = model.predict(X_train2)
         # check error
         from sklearn.metrics import mean_squared_error
         mean_squared_error(y_test2, y_test2_predict)
         # Choose alpha = 0.1
         model2 = Ridge(alpha = 0.1)
         # # fit train datasets.
         model2.fit(X_train, y_train)
         # prediction
         second_predict = model2.predict(X_train2)
         # check error
         mean_squared_error(y_test2, second_predict)
         print (mean_squared_error(y_test2, second_predict))
         # # Final performance if you satisfy at alpha = 0.1
         y_test_prediction = model2.predict(X_test)
         print(y_test_prediction)
         mean_squared_error(y_test, y_test_prediction)
        1.9391609593785695
        [ 8.13368275 18.42064041 18.21958416 9.93548647 24.7130477 19.55515613
        21.42209051 8.89562661 10.11448833 16.19293195 15.44979271 7.83548211
        17.23369351 9.97825676 7.42449941 17.00033491 9.64243739 10.72810643
         9.88551126 7.67363797]
Out[227... 2.3654933683711636
In [229... | # Use K-Fold cross calidation for error analysis in the ("Advertising.csv").
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.75, random_state = 101)
         from sklearn.preprocessing import StandardScaler
         scalar = StandardScaler()
         scaler.fit(X_train)
         X_train = scaler.transform(X_train)
         X_test = scaler.transform(X_test)
         # Make a linear model(Ridge)
         from sklearn.linear_model import Ridge
         model_1 = Ridge(alpha = 100)
         from sklearn.model_selection import cross_val_score
         cv_scors = cross_val_score(model_1, X_train,y_train, scoring ='neg_mean_squared_error', cv = 7) # k-fole (cv) value = 7.
         # Absolute mean of cv_scors
         print(abs(scores.mean()))
         # Make another model
         model_2 = Ridge(alpha = 1000)
         scor_2 = cross_val_score(model_2, X_train, y_train, scoring ='neg_mean_squared_error', cv = 7) # k-fole (cv) value is 7.
         print(scor_2)
         print(abs(scor_2.mean()))
         # Use model to fit all train data
         model_final= Ridge(alpha = 1)
         model_final.fit(X_train, y_train)
         y_final_test_prediction = model_final.predict(X_test)
         mean_squared_error(y_test, y_final_test_prediction)
         print(mean_squared_error(y_test, y_final_test_prediction))
        [ -1.59725436 \ -17.6553551 \ -25.94686869 \ -7.26031918 \ -13.86382377 ]
         -17.17616644 -9.27138728]
        fit_time
                                       0.000801
                                       0.000901
        score_time
        test_neg_mean_squared_error 3.323018
        test_neg_mean_absolute_error 1.308467
```

19.89997918056674 3.502569470564967

Tn []: