2_Regresion Lineal Multiple

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Creado por:

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1 Regresión Lineal Múltiple

```
[1]: | # pip install scikit-learn
[2]: import numpy as np
     from sklearn import datasets, linear_model
     import matplotlib.pyplot as plt
     import pandas as pd
     from sklearn.metrics import mean_squared_error, r2_score
[3]: dataset = datasets.load diabetes()
     print(dataset.DESCR)
     # Crear un DataFrame con los datos
     data = pd.DataFrame(dataset.data, columns=dataset.feature_names)
     data['level'] = dataset.target
    .. _diabetes_dataset:
    Diabetes dataset
    Ten baseline variables, age, sex, body mass index, average blood
    pressure, and six blood serum measurements were obtained for each of n = 1
    442 diabetes patients, as well as the response of interest, a
    quantitative measure of disease progression one year after baseline.
    **Data Set Characteristics:**
    :Number of Instances: 442
    :Number of Attributes: First 10 columns are numeric predictive values
    :Target: Column 11 is a quantitative measure of disease progression one year
    after baseline
```

:Attribute Information:

- age in years - age - sex - bmi body mass index average blood pressure - bp - s1 tc, total serum cholesterol - s2 ldl, low-density lipoproteins - s3 hdl, high-density lipoproteins - s4 tch, total cholesterol / HDL ltg, possibly log of serum triglycerides level – s5 glu, blood sugar level - s6
- Note: Each of these 10 feature variables have been mean centered and scaled by the standard deviation times the square root of `n_samples` (i.e. the sum of squares of each column totals 1).

Source URL:

https://www4.stat.ncsu.edu/~boos/var.select/diabetes.html

For more information see:

Bradley Efron, Trevor Hastie, Iain Johnstone and Robert Tibshirani (2004) "Least Angle Regression," Annals of Statistics (with discussion), 407-499. (https://web.stanford.edu/~hastie/Papers/LARS/LeastAngle_2002.pdf)

[4]: data

```
[4]:
                                   bmi
                                               bp
                                                         s1
                                                                   s2
                                                                              s3
                         sex
          0.038076 0.050680 0.061696 0.021872 -0.044223 -0.034821 -0.043401
     0
         -0.001882 -0.044642 -0.051474 -0.026328 -0.008449 -0.019163 0.074412
     1
          0.085299 \quad 0.050680 \quad 0.044451 \quad -0.005670 \quad -0.045599 \quad -0.034194 \quad -0.032356
         -0.089063 -0.044642 -0.011595 -0.036656 0.012191 0.024991 -0.036038
     4
          0.005383 - 0.044642 - 0.036385 \ 0.021872 \ 0.003935 \ 0.015596 \ 0.008142
     437 0.041708 0.050680 0.019662 0.059744 -0.005697 -0.002566 -0.028674
     438 -0.005515 0.050680 -0.015906 -0.067642 0.049341 0.079165 -0.028674
     439 0.041708 0.050680 -0.015906 0.017293 -0.037344 -0.013840 -0.024993
     440 -0.045472 -0.044642 0.039062 0.001215 0.016318 0.015283 -0.028674
     441 -0.045472 -0.044642 -0.073030 -0.081413 0.083740 0.027809 0.173816
                                    s6
                s4
                          s5
                                        level
         -0.002592 0.019907 -0.017646
                                        151.0
     0
         -0.039493 -0.068332 -0.092204
                                         75.0
     1
     2
         -0.002592 0.002861 -0.025930 141.0
     3
         0.034309 0.022688 -0.009362
                                        206.0
         -0.002592 -0.031988 -0.046641 135.0
```

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438 0.034309 -0.018114 0.044485
                                        104.0
     439 -0.011080 -0.046883 0.015491
                                        132.0
     440 0.026560 0.044529 -0.025930
                                        220.0
     441 -0.039493 -0.004222 0.003064
                                        57.0
     [442 rows x 11 columns]
[5]: # Load the diabetes dataset
     diabetes X, diabetes y = datasets.load_diabetes(return_X_y=True)
[6]: from sklearn.model_selection import train_test_split
     # Separo los datos de "train" entrenamiento y "test" prueba para probar los_
      →algoritmos
     X_train, X_test, y_train, y_test = train_test_split(diabetes_X, diabetes_y,_
      →test_size=0.2)
[7]: | lr_multiple = linear_model.LinearRegression()
[8]: lr_multiple.fit(X_train, y_train)
[8]: LinearRegression()
[9]: y_pred = lr_multiple.predict(X_test)
     y_pred
[9]: array([198.92942131, 114.84534815, 241.55686724, 199.64125541,
            112.61590082, 107.1934254, 107.96941096, 137.59993048,
            69.27761389, 222.33109311, 150.86244978, 145.40001303,
            255.34710296, 163.35076062, 142.81825513, 126.70611647,
            238.31372899, 186.66772647, 153.05948515, 244.01602178,
            65.51277224, 209.49948887, 96.87859477, 159.76543679,
            112.6826685 , 209.19870258, 105.15518047, 117.03331812,
            97.82531017, 78.06915422, 185.62907937, 218.51218553,
            157.96980347, 179.05651855, 137.4350862, 117.71019316,
            255.82640598, 66.88811217, 195.9950507, 86.4545535,
           254.58156976, 166.56049251, 253.55564875, 83.4338512,
            163.32869193, 95.08668772, 228.58659933, 237.04679853,
            105.48363302, 60.39818583, 83.11228058, 169.84090478,
            253.0277734 , 126.99207723 , 132.12846921 , 176.93674279 ,
           259.91047517, 120.86320122, 138.05650514, 104.2942936 ,
           190.68577909, 197.46431645, 161.17951365, 76.62152224,
            148.65018355, 143.67764528, 151.82205271, 79.41488423,
            164.88841741, 168.62825415, 166.3294742 , 190.99375231,
            242.50128692, 91.6317138, 200.91842072, 157.80520745,
```

178.0

437 -0.002592 0.031193 0.007207

```
92.35210501, 233.11088782, 120.85966053, 104.91747567,
             176.31137634, 210.97548991, 164.59101327, 90.39905682,
             177.96704449, 144.60286349, 123.53437884, 99.76696947,
             221.45945529])
[10]: y_test
[10]: array([220., 88., 259., 265., 199., 118., 63., 40., 134., 259., 246.,
             97., 273., 58., 103., 53., 270., 164., 86., 252., 158., 310.,
             87., 113., 102., 297.,
                                     65., 53., 49., 59., 217., 173., 155.,
             143., 170., 214., 310., 52., 202., 60., 233., 141., 281., 181.,
             206., 69., 99., 272., 88., 57., 71., 258., 243., 42., 49.,
             91., 242., 78., 115., 142., 170., 48., 185., 75., 150., 185.,
             85., 42., 178., 242., 131., 232., 132., 88., 186., 276., 101.,
             236., 182., 135., 126., 121., 120., 64., 311., 202., 71., 118.,
             180.])
[11]: print('DATOS DEL MODELO REGRESIÓN LINEAL MULTIPLE')
      print()
      print('Valor de las pendientes o coeficientes "a":')
      print(lr multiple.coef )
      print('Valor de la intersección o coeficiente "b":')
      print(lr_multiple.intercept_)
     DATOS DEL MODELO REGRESIÓN LINEAL MULTIPLE
     Valor de las pendientes o coeficientes "a":
         6.89443248 -226.02422036
                                   529.1542349
                                                              -886.09501482
                                                 344.9279648
       530.84496929 209.30837657
                                   238.80440808
                                                 765.83681449
                                                                74.64827602]
     Valor de la intersección o coeficiente "b":
     152.84136793062558
[12]: print("Precisión del modelo:")
      print(lr_multiple.score(X_train, y_train))
     Precisión del modelo:
     0.5314676813743083
     Creado por:
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```