In [1]: import findspark findspark.init() In [2]: from pyspark.sql import SparkSession spark=SparkSession.builder.appName("DfApp").getOrCreate() In [3]: df=spark.read.option("header", "true").csv('D:\weight-height.csv', inferSchema=True) df.show() Weight| |Gender| Height| Male | 73.84701702 | 241.8935632 | Male|68.78190405|162.3104725 Male | 74.11010539 | 212.7408556 Male| 71.7309784|220.0424703| Male|69.88179586|206.3498006| Male | 67.25301569 | 152.2121558 Male| 65.865443|183.9278886 |Female|68.34851551|167.9711105 Male|67.01894966|175.9294404 Male| 56.234555|156.3996764 Male|71.19538228|186.6049256 Male|71.64080512|213.7411695 Male | 64.76632913 | 167.1274611 Male | 65.3456677 | 189.4461814 Female | 69.24373223 | 186.434168 Male| 67.6456197|172.1869301 Male|72.41831663|196.0285063 Male|63.97432572|172.8834702 |Female| 69.6400599|185.9839576| Male| 65.654333| 182.426648| +----+ only showing top 20 rows df.printSchema() root |-- Gender: string (nullable = true) |-- Height: double (nullable = true) |-- Weight: double (nullable = true) In [5]: df.count() Out[5]: In [6]: df.na.drop(how='any') DataFrame[Gender: string, Height: double, Weight: double] Out[6]: In [7]: df.count() 10000 Out[7]: In [8]: from pyspark.ml.feature import VectorAssembler In [9]: featureassembler=VectorAssembler(inputCols=['Height'], outputCol='HeightData') v=featureassembler.transform(df) v.show() Height| Weight| HeightData| Male|73.84701702|241.8935632|[73.84701702]| Male | 68.78190405 | 162.3104725 | [68.78190405] | Male | 74.11010539 | 212.7408556 | [74.11010539] | Male | 71.7309784 | 220.0424703 | [71.7309784] | Male | 69.88179586 | 206.3498006 | [69.88179586] | Male | 67.25301569 | 152.2121558 | [67.25301569] | Male| 65.865443|183.9278886| [65.865443]| |Female|68.34851551|167.9711105|[68.34851551]| Male|67.01894966|175.9294404|[67.01894966]| Male| 56.234555|156.3996764| [56.234555]| Male | 71.19538228 | 186.6049256 | [71.19538228] | Male|71.64080512|213.7411695|[71.64080512]| Male | 64.76632913 | 167.1274611 | [64.76632913] | Male | 65.3456677 | 189.4461814 | [65.3456677] | |Female|69.24373223| 186.434168|[69.24373223]| Male | 67.6456197 | 172.1869301 | [67.6456197] Male | 72.41831663 | 196.0285063 | [72.41831663] | Male | 63.97432572 | 172.8834702 | [63.97432572] | |Female| 69.6400599|185.9839576| [69.6400599]| Male| 65.654333| 182.426648| [65.654333]| +----+ only showing top 20 rows In [10]: v.select('HeightData','Weight').show() | HeightData| Weight| +----+ |[73.84701702]|241.8935632| |[68.78190405]|162.3104725| |[74.11010539]|212.7408556| | [71.7309784]|220.0424703| |[69.88179586]|206.3498006| |[67.25301569]|152.2121558| [65.865443]|183.9278886| |[68.34851551]|167.9711105| |[67.01894966]|175.9294404| [56.234555] | 156.3996764 | |[71.19538228]|186.6049256| |[71.64080512]|213.7411695| |[64.76632913]|167.1274611| [65.3456677] | 189.4461814 | |[69.24373223]| 186.434168| [67.6456197]|172.1869301| |[72.41831663]|196.0285063| |[63.97432572]|172.8834702| | [69.6400599]|185.9839576| | [65.654333]| 182.426648| +----+ only showing top 20 rows In [11]: v.printSchema() |-- Gender: string (nullable = true) |-- Height: double (nullable = true) |-- Weight: double (nullable = true) |-- HeightData: vector (nullable = true) In [12]: fd=v.select('HeightData','Weight') fd.show() | HeightData| |[68.78190405]|162.3104725| |[74.11010539]|212.7408556| | [71.7309784]|220.0424703| |[69.88179586]|206.3498006| |[67.25301569]|152.2121558| [65.865443]|183.9278886| |[68.34851551]|167.9711105| |[67.01894966]|175.9294404| [56.234555]|156.3996764| |[71.19538228]|186.6049256| |[71.64080512]|213.7411695| [64.76632913]|167.1274611| [65.3456677]|189.4461814| |[69.24373223]| 186.434168| | [67.6456197]|172.1869301| | [72.41831663] | 196.0285063 | |[63.97432572]|172.8834702| | [69.6400599]|185.9839576| | [65.654333]| 182.426648| +----+ only showing top 20 rows In [13]: # spliting of data into train and test data from pyspark.ml.regression import LinearRegression In [14]: train, test=fd.randomSplit([.70,.30]) lr = Linear Regression (features Col = 'Height Data', label Col = 'Weight')tm=lr.fit(train) In [15]: # coefficient and intercept tm.intercept -345.5104174679869 Out[15]: In [16]: tm.coefficients # correlation bet weight and height DenseVector([7.6367]) Out[16]: In [17]: # evaluate the train data res=tm.evaluate(test) res.predictions.show() C:\Users\user\anaconda3\lib\site-packages\pyspark\sql\context.py:125: FutureWarning: Deprecated in 3.0.0. Use SparkSession.builder.getOrCreate() instead. warnings.warn(HeightData| Weight| prediction| |[55.97919788]|85.41753362|81.98789608042966| |[56.06663635]|89.57120474|82.65564052622159| |[56.10536959]|87.29886913|82.95143602535967| |[56.63041198]|89.48048027|86.96104575123644| [56.765443]|201.7607769|87.99224190110743| [56.765544] | 166.5035316 | 87.99301321136738 | |[56.78543437]|83.99307747|88.14491070102207| |[56.78938641]|95.32808768|88.17509138418757| |[56.85608213]|97.36497833|88.68442893993495| |[56.94941514]|107.1718559| 89.3971884271096| |[56.97513323]|89.16984997|89.59359067149785| |[56.97527896]|90.34178426|89.59470357292543| |[57.02885744]|101.2025509|90.00386823962157| |[57.14819808]|91.64547294| 90.9152411115079| [57.2330564]|99.37128426|91.56328163485227| |[57.27014705]|94.49963415|91.84653310904099| |[57.31302352]| 93.8764374| 92.1739693587565| |[57.31390274]| 95.1390468| 92.1806837291204| |[57.35309276]|72.75014469|92.47996753618463| |[57.37575853]|114.1922086|92.65306002084378| +----+ only showing top 20 rows In [18]: print("R2", res.r2) print("Mean Absolute Error is", res.meanAbsoluteError) print("Root Mean Square Error(RMSE)", res. rootMeanSquaredError) R2 0.8490377509012588 Mean Absolute Error is 9.752730559572562 Root Mean Square Error(RMSE) 12.5096748871693 In []: