In [1]: **import** pandas **as** pd import numpy as np import seaborn as sns from matplotlib import pyplot as plt from sklearn.linear_model import LinearRegression import statsmodels.formula.api as smf from sklearn.metrics import accuracy_score from sklearn.model_selection import train_test_split In [2]: iris=pd.read_csv("C:\\Users\\Pranav\\Desktop\\DATA SCIENCE DATA\\CVC file\\Iris.csv") iris.head() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[2]: **Species** 0 1 5.1 3.5 1.4 0.2 Iris-setosa **1** 2 0.2 Iris-setosa 4.9 3.0 1.4 **2** 3 4.7 3.2 1.3 0.2 Iris-setosa **3** 4 4.6 3.1 1.5 0.2 Iris-setosa **4** 5 5.0 3.6 1.4 0.2 Iris-setosa In [3]: iris.tail() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** Out[3]: **145** 146 6.7 3.0 5.2 2.3 Iris-virginica **146** 147 6.3 2.5 5.0 1.9 Iris-virginica **147** 148 6.5 3.0 5.2 2.0 Iris-virginica **148** 149 6.2 3.4 5.4 2.3 Iris-virginica **149** 150 5.9 3.0 5.1 1.8 Iris-virginica In [4]: #Shape of dataset iris.shape (150, 6)Out[4]: #information about datast iris.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 150 entries, 0 to 149 Data columns (total 6 columns): Column Non-Null Count Dtype -----0 Id 150 non-null int64 SepalLengthCm 150 non-null float64 1 SepalWidthCm 150 non-null float64 2 PetalLengthCm 150 non-null float64 3 PetalWidthCm 150 non-null float64 4 150 non-null Species object dtypes: float64(4), int64(1), object(1) memory usage: 7.2+ KB In [6]: # mathamathical information about data iris.describe() Out[6]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **count** 150.000000 150.000000 150.000000 150.000000 150.000000 mean 75.500000 5.843333 3.054000 1.198667 3.758667 43.445368 0.828066 0.433594 1.764420 0.763161 1.000000 4.300000 2.000000 1.000000 0.100000 min 38.250000 5.100000 1.600000 0.300000 2.800000 5.800000 3.000000 4.350000 **75**% 112.750000 6.400000 3.300000 1.800000 5.100000 6.900000 2.500000 max 150.000000 7.900000 4.400000 In [7]: # Find out null value in dataset iris.isnull().sum() Id Out[7]: SepalLengthCm 0 SepalWidthCm 0 PetalLengthCm 0 PetalWidthCm 0 0 Species dtype: int64 In [8]: iris['Species'].value_counts() Iris-setosa 50 Out[8]: Iris-versicolor 50 50 Iris-virginica Name: Species, dtype: int64 In [9]: # find the value in iris dataset the sepalwidth is greater than 1 iris[iris['SepalWidthCm']>4] Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** Out[9]: **15** 16 5.7 4.4 1.5 0.4 Iris-setosa **32** 33 5.2 4.1 1.5 0.1 Iris-setosa **33** 34 5.5 4.2 1.4 0.2 Iris-setosa In [10]: # find the value in iris dataset the petalwidth is greater than 1 iris[iris['PetalWidthCm']>1] Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[10]: **Species** 50 51 7.0 3.2 4.7 1.4 Iris-versicolor 51 52 6.4 3.2 4.5 1.5 Iris-versicolor 52 53 6.9 4.9 1.5 Iris-versicolor 3.1 53 5.5 2.3 4.0 1.3 Iris-versicolor 55 6.5 4.6 54 2.8 1.5 Iris-versicolor **145** 146 6.7 3.0 5.2 2.3 Iris-virginica **146** 147 6.3 2.5 5.0 1.9 Iris-virginica 6.5 **147** 148 3.0 5.2 2.0 Iris-virginica **148** 149 6.2 3.4 5.4 Iris-virginica 5.9 **149** 150 3.0 5.1 1.8 Iris-virginica 93 rows × 6 columns In [11]: # find the value in iris dataset the petalwidth is greater than 2 iris[iris['PetalWidthCm']>2] Out[11]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species 100** 101 6.3 3.3 2.5 Iris-virginica **102** 103 7.1 3.0 5.9 2.1 Iris-virginica **104** 105 6.5 3.0 5.8 2.2 Iris-virginica **105** 106 7.6 3.0 6.6 2.1 Iris-virginica **109** 110 7.2 3.6 6.1 2.5 Iris-virginica **112** 113 6.8 3.0 5.5 2.1 Iris-virginica **114** 115 5.1 2.4 Iris-virginica 5.8 2.8 **115** 116 6.4 3.2 5.3 2.3 Iris-virginica **117** 118 7.7 3.8 6.7 2.2 Iris-virginica **118** 119 7.7 2.6 6.9 2.3 Iris-virginica **120** 121 6.9 3.2 5.7 2.3 Iris-virginica **124** 125 6.7 3.3 5.7 2.1 Iris-virginica **128** 129 6.4 2.8 5.6 2.1 Iris-virginica **132** 133 6.4 2.8 5.6 2.2 Iris-virginica **135** 136 7.7 3.0 6.1 2.3 Iris-virginica **136** 137 6.3 3.4 5.6 2.4 Iris-virginica **139** 140 6.9 3.1 5.4 2.1 Iris-virginica **140** 141 6.7 5.6 3.1 2.4 Iris-virginica **141** 142 6.9 3.1 5.1 2.3 Iris-virginica **143** 144 6.8 3.2 5.9 2.3 Iris-virginica **144** 145 6.7 3.3 5.7 2.5 Iris-virginica **145** 146 6.7 3.0 5.2 2.3 Iris-virginica **148** 149 6.2 3.4 5.4 2.3 Iris-virginica # Visualization sns.scatterplot(x='SepalLengthCm', y='PetalLengthCm', data=iris, hue='Species') plt.grid(True) plt.show() Species Iris-setosa lris-versicolor Iris-virginica 4.5 5.0 5.5 In [13]: X=iris[['SepalWidthCm']] y=iris[['SepalLengthCm']] In [14]: print(X) SepalWidthCm 3.5 1 3.0 2 3.2 3 3.1 3.6 145 3.0 146 2.5 147 3.0 148 3.4 149 3.0 [150 rows x 1 columns] In [15]: print(y) SepalLengthCm 5.1 1 4.9 4.7 3 4.6 5.0 . . . 6.7 145 146 6.3 147 6.5 148 6.2 149 5.9 [150 rows x 1 columns] In [16]: #training and test data X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.3, random_state=2) print("shape of X_train= ", X_train.shape) In [17]: print("shape of X_test= ", X_test.shape) print("shape of y_train= ",y_train.shape) print("shape of y_test= ",y_test.shape) shape of $X_{train} = (105, 1)$ shape of $X_{test} = (45, 1)$ shape of $y_{train} = (105, 1)$ shape of $y_{test} = (45, 1)$ In [18]: X_test.head() ${\bf SepalWidthCm}$ Out[18]: 3.4 3.1 113 2.5 12 3.0 24 3.4 In [19]: y_test.head() Out[19]: SepalLengthCm 4.6 4.6 113 5.7 4.8 24 4.8 In [20]: # creating the model lr=LinearRegression() In [21]: # fitbthe model lr.fit(X_train,y_train) Out[21]: ▼ LinearRegression LinearRegression() In [22]: y_pred=lr.predict(X_test) In [23]: y_test.head() Out[23]: SepalLengthCm 4.6 4.6 113 5.7 12 4.8 24 4.8 In [24]: y_pred[0:5] array([[5.79624971], Out[24]: [5.87619788], [6.03609422], [5.90284727], [5.79624971]]) In [25]: # Error in predction findout from sklearn.metrics import mean_squared_error In [26]: mean_squared_error(y_test,y_pred) 0.7284298865085145 Out[26]: In [27]: # Creatinf the model 2 y=iris[['SepalLengthCm']] In [28]: X=iris[['SepalWidthCm', 'PetalWidthCm', 'PetalLengthCm']] In [29]: #training and test data X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.3, random_state=2) print("shape of X_train= ",X_train.shape) In [30]: print("shape of X_test= ", X_test.shape) print("shape of y_train= ",y_train.shape) print("shape of y_test= ",y_test.shape) shape of $X_{train} = (105, 3)$ shape of $X_{test} = (45, 3)$ shape of $y_{train} = (105, 1)$ shape of $y_{test} = (45, 1)$ In [31]: lr2=LinearRegression() In [32]: lr2.fit(X_train,y_train) Out[32]: ▼ LinearRegression LinearRegression() In [35]: y_pred2=lr2.predict(X_test) mean_squared_error(y_test,y_pred2) 0.08942496370919266 Out[36]: