Tanvi Bhosle Data Science And Business Analytics Intern @ TSF GRIP JULY2021 Task 6: Pridiction using Decision Tree Algorithm Dataset: https://bit.ly/3kXTdox In []: Import Libraries¶ In [103... import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns %matplotlib inline from sklearn.model_selection import train_test_split from sklearn.tree import DecisionTreeClassifier from sklearn import metrics import graphviz from sklearn import tree Import Dataset In [8]: data=pd.read_csv("C:\\Users\\Admin\\Desktop\\Iris.csv", encoding=("ISO-8859-1"), low_memory=False) data.head(5) Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[8]: **Species** 0 1 5.1 1.4 0.2 Iris-setosa **1** 2 4.9 3.0 1.4 0.2 Iris-setosa **2** 3 4.7 3.2 1.3 0.2 Iris-setosa 1.5 0.2 Iris-setosa 4.6 3.1 4 5 5.0 3.6 1.4 0.2 Iris-setosa In [65]: data.shape <class 'pandas.core.frame.DataFrame'> RangeIndex: 150 entries, 0 to 149 Data columns (total 6 columns): Column Non-Null Count Dtype -----0 Ιd 150 non-null int64 SepalLengthCm 150 non-null float64 SepalWidthCm 150 non-null float64 PetalLengthCm 150 non-null float64 PetalWidthCm 150 non-null float64 5 Species 150 non-null object dtypes: float64(4), int64(1), object(1) memory usage: 7.2+ KB In [11]: data["Species"].value_counts() Iris-setosa Iris-versicolor 50 Iris-virginica 50 Name: Species, dtype: int64 data.isnull().sum() SepalLengthCm 0 SepalWidthCm 0 PetalLengthCm PetalWidthCm Species 0 dtype: int64 **Sepal Dimensions** In [15]: plt.figure(figsize=(13,7)) sns.scatterplot(x=data["SepalLengthCm"], y=data["SepalWidthCm"], hue=data["Species"], s=100) Out[15]: <AxesSubplot:xlabel='SepalLengthCm', ylabel='SepalWidthCm'> Species Iris-setosa Iris-versicolor Iris-virginica 4.0 SepalWidthCm 0.0 2.5 2.0 4.5 5.0 5.5 6.5 7.0 7.5 8.0 6.0 SepalLengthCm In [28]: fig, axes=plt.subplots(figsize=(6,4)) plt.title("Distribution of Sepal Length") plt.hist(data["SepalLengthCm"]) fig, axes=plt.subplots(figsize=(6,4)) plt.title("Distribution of Sepal Width") plt.hist(data["SepalWidthCm"]) Out[28]: (array([4., 7., 22., 24., 38., 31., 9., 11., 2., 2.]), array([2. , 2.24, 2.48, 2.72, 2.96, 3.2 , 3.44, 3.68, 3.92, 4.16, 4.4]), <BarContainer object of 10 artists>) Distribution of Sepal Length 25 20 15 10 4.5 6.0 6.5 7.0 7.5 Distribution of Sepal Width 35 30 25 20 15 10 3.5 2.5 3.0 2.0 **Petal Dimensions** In [29]: plt.figure(figsize=(13,7)) sns.scatterplot(x=data["PetalLengthCm"], y=data["PetalWidthCm"], hue=data["Species"], s=100) Out[29]: <AxesSubplot:xlabel='PetalLengthCm', ylabel='PetalWidthCm'> Species Iris-setosa Iris-versicolor Iris-virginica 2.0 PetalWidthCm 1.0 0.5 0.0 PetalLengthCm In [30]: fig, axes=plt.subplots(figsize=(6,4)) plt.title("Distribution of Petal Length") plt.hist(data["PetalLengthCm"]) fig, axes=plt.subplots(figsize=(6,4)) plt.title("Distribution of Petal Width") plt.hist(data["PetalWidthCm"]) Out[30]: (array([41., 8., 1., 7., 8., 33., 6., 23., 9., 14.]), array([0.1 , 0.34, 0.58, 0.82, 1.06, 1.3 , 1.54, 1.78, 2.02, 2.26, 2.5]), <BarContainer object of 10 artists>) Distribution of Petal Length 35 30 25 20 15 10 Distribution of Petal Width 40 35 30 25 20 15 10 0.5 1.0 1.5 2.0 2.5 Correlation between Dependent and Independent variables In [70]: data1=data.drop("Id", axis=1) data1.head() data1["Species"].replace("Iris-setosa", "0", inplace=True) data1["Species"].replace("Iris-versicolor", "1", inplace=True) data1["Species"].replace("Iris-virginica", "2", inplace=True) data1["Species"]=data1.Species.astype(int) data1.head() Out[70]: SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species 3.5 0 5.1 0.2 0 1.4 4.9 3.0 1.4 0.2 2 4.7 3.2 0.2 1.3 0 4.6 3.1 1.5 0.2 0 5.0 3.6 1.4 0.2 0 In [73]: plt.figure(figsize=(8,4)) sns.heatmap(data1.corr(), annot=True) Out[73]: <AxesSubplot:> - 1.0 0.87 -0.11 0.82 0.78 SepalLengthCm · - 0.8 0.6 -0.42 -0.42 -0.11 1 -0.36 SepalWidthCm 0.4 -0.42 0.87 0.96 PetalLengthCm 1 0.95 0.2 -0.36 0.82 0.96 1 0.96 PetalWidthCm · 0.0 -0.20.78 -0.42 0.95 0.96 1 Species SepalWidthCm PetalLengthCm **Model Building** In [75]: train, test=train_test_split(data1, test_size=0.3) train.shape, test.shape Out[76]: ((105, 5), (45, 5)) In [79]: train_x=train[["SepalLengthCm", "SepalWidthCm", "PetalLengthCm", "PetalWidthCm"]] train_y=train.Species test_x=test[["SepalLengthCm", "SepalWidthCm", "PetalLengthCm", "PetalWidthCm"]] test_y=test.Species In [83]: dtree=DecisionTreeClassifier() dtree.fit(train_x, train_y) predictions=dtree.predict(test_x) accuracy=metrics.accuracy_score(predictions, test_y) accuracy Out[83]: 0.95555555555556 In [85]: x=data1[["SepalLengthCm", "SepalWidthCm", "PetalLengthCm", "PetalWidthCm"]] dtree1=DecisionTreeClassifier() dtree1.fit(x,y) Out[85]: DecisionTreeClassifier() Visualization In [109... fig = plt.figure(figsize=(25,20)) tree.plot_tree(dtree1, feature_names=["SepalLengthCm", "SepalWidthCm", "PetalLengthCm", "PetalWidthCm"], class_names=["0", "1", "2"], filled=True) Out[109... [Text(697.5, 996.6, 'PetalLengthCm <= 2.45\ngini = 0.667\nsamples = 150\nvalue = [50, 50, 50]\nclass = 0'), Text(590.1923076923077, 815.40000000000001, 'gini = 0.0\nsamples = 50\nvalue = [50, 0, 0]\nclass = 0'),
Text(804.8076923076923, 815.4000000000001, 'PetalWidthCm <= 1.75\ngini = 0.5\nsamples = 100\nvalue = [0, 50, 50]\nclass = 1'), Text(107.3076923076923, 271.7999999999995, 'gini = 0.0\nsamples = 47\nvalue = [0, 47, 0]\nclass = 1'), Text(321.9230769230769, 271.799999999995, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 1]\nclass = 2'), Text(751.1538461538462, 271.7999999999995, 'SepalLengthCm <= 6.95\ngini = 0.444\nsamples = 3\nvalue = [0, 2, 1]\nclass = 1'),
Text(643.8461538461538, 90.5999999999991, 'gini = 0.0\nsamples = 2\nvalue = [0, 2, 0]\nclass = 1'),
Text(858.4615384615385, 90.5999999999991, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 1]\nclass = 2'), Text(1180.3846153846155, 634.2, 'PetalLengthCm <= 4.85\ngini = 0.043\nsamples = 46\nvalue = [0, 1, 45]\nclass = 2'),
Text(1073.076923076923, 453.0, 'SepalWidthCm <= 3.1\ngini = 0.444\nsamples = 3\nvalue = [0, 1, 2]\nclass = 2'),
Text(965.7692307692307, 271.7999999999995, 'gini = 0.0\nsamples = 2\nvalue = [0, 0, 2]\nclass = 2'),
Text(1180.3846153846155, 271.7999999999995, 'gini = 0.0\nsamples = 1\nvalue = [0, 1, 0]\nclass = 1'), $Text(1287.6923076923076, 453.0, 'gini = 0.0 \nsamples = 43 \nvalue = [0, 0, 43] \nclass = 2')]$ PetalLengthCm <= 2.45 gini = 0.667samples = 150value = [50, 50, 50]class = 0PetalWidthCm \leq 1.75 gini = 0.0gini = 0.5samples = 50samples = 100value = [50, 0, 0]value = [0, 50, 50]class = 0class = 1PetalLengthCm <= 4.85 PetalLengthCm <= 4.95 gini = 0.168gini = 0.043samples = 54samples = 46value = [0, 49, 5]value = [0, 1, 45]class = 1class = 2PetalWidthCm <= 1.65 PetalWidthCm <= 1.55 SepalWidthCm <= 3.1 gini = 0.0gini = 0.444gini = 0.444gini = 0.041samples = 43samples = 6 samples = 48samples = 3value = [0, 0, 43]value = [0, 47, 1]value = [0, 2, 4]value = [0, 1, 2]class = 2class = 1class = 2class = 2SepalLengthCm <= 6.95 gini = 0.0gini = 0.0gini = 0.0gini = 0.0gini = 0.0gini = 0.444samples = 47samples = 1samples = 3samples = 2samples = 1samples = 3value = [0, 47, 0]value = [0, 0, 1]value = [0, 0, 3]value = [0, 0, 2]value = [0, 1, 0]value = [0, 2, 1]class = 1class = 2class = 2class = 2class = 1class = 1gini = 0.0gini = 0.0samples = 2samples = 1value = [0, 2, 0]value = [0, 0, 1]class = 1class = 2Trial In [116.. dtree.predict([[6.8,2,5,1.5]]) Out[116... array([2]) In [118... dtree.predict([[2,2,2,2]]) Out[118... array([0]) In [124.. dtree.predict([[0.1,2.5,8.9,2]]) Out[124... array([2]) In [125... dtree.predict([[5,3.4,1.5,0.2]]) Out[125... array([0]) In [126.. dtree.predict([[5.7,2.8,4.5,1.3]]) Out[126... array([1]) In []: