4.6 3.1 1.5 Iris-setosa 1.4 4 5 5.0 3.6 0.2 Iris-setosa **145** 146 6.7 3.0 5.2 2.3 Iris-virginica **146** 147 6.3 2.5 5.0 1.9 Iris-virginica 5.2 **147** 148 6.5 3.0 2.0 Iris-virginica **148** 149 6.2 3.4 5.4 2.3 Iris-virginica **149** 150 3.0 5.1 5.9 1.8 Iris-virginica 150 rows × 6 columns In [3]: df.describe() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[3]: count 150.000000 150.000000 150.000000 150.000000 150.000000 mean 75.500000 5.843333 3.054000 3.758667 1.198667 43.445368 0.828066 0.433594 1.764420 0.763161 std 1.000000 4.300000 2.000000 1.000000 0.100000 min 25% 38.250000 5.100000 2.800000 1.600000 0.300000 75.500000 5.800000 3.000000 4.350000 1.300000 1.800000 **75**% 112.750000 6.400000 3.300000 5.100000 max 150.000000 7.900000 4.400000 6.900000 2.500000 In [4]: **#Checking Null Values** df.isnull() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species Out[4]: 0 False False False False False False False 1 False 2 False False False False 3 False False False 4 False False False False **145** False **146** False **147** False False False False False False **148** False False False False False False **149** False False False False False False 150 rows × 6 columns In [5]: df.head() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[5]: **Species** 5.1 3.5 1.4 0.2 Iris-setosa **1** 2 4.9 0.2 Iris-setosa 3.0 1.4 4.7 3.2 1.3 0.2 Iris-setosa 1.5 0.2 Iris-setosa 4.6 3.1 5.0 3.6 1.4 0.2 Iris-setosa In [7]: #Shape of Dataset rows and columns Out[7]: (150, 6) In [9]: **#Data Visualization** sns.pairplot(df, hue='Species', palette='coolwarm') Out[9]: <seaborn.axisgrid.PairGrid at 0x1b965d4f9a0> 150 125 100 р 75 50 25 SepalLengthCm 4.5 SepalWidthCm Species 3.0 Iris-versicolor Iris-virginica 2.5 2.0 PetalLengthCm 2.5 2.0 PetalWidthCm 10 0.5 100 ld SepalWidthCm SepalLengthCm PetalLengthCm PetalWidthCm In [10]: X=df.iloc[:,[0,1,2,3]]y=df.Species df.shape Out[10]: (150, 6) In [11]: X.shape Out[11]: (150, 4) y.shape Out[12]: (150,) In [13]: **#Spliting Data** from sklearn.model\_selection import train\_test\_split X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y) In [14]: X\_train.shape Out[14]: (112, 4) X\_test.shape Out[15]: (38, 4) y\_train.shape Out[16]: (112,) In [17]: y\_test.shape

In [1]:

In [2]:

Out[2]:

#importing libraries import numpy as np import pandas as pd import seaborn as sns

import warnings

**#Loading dataset** 

df

0

1

2

1

3

import matplotlib.pyplot as plt

from sklearn.tree import plot\_tree

warnings.filterwarnings("ignore")

5.1

4.9

4.7

df = pd.read\_csv('iris.csv')

from sklearn.metrics import accuracy\_score

Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm

3.5

3.0

3.2

1.4

1.4

1.3

**Species** 

Iris-setosa

Iris-setosa

Iris-setosa

0.2

0.2

f\_n=["Sepal length", "Sepal width", "Petal length", "Petal width"] c\_n=["Setosa", "Versicolor", "Virginica"] plot\_tree(clf, feature\_names=f\_n, class\_names=c\_n, filled=True) Out[30]: [Text(2678.399999999996, 2265.0, 'Sepal length <= 102.5\ngini = 0.663\nsamples = 112\nvalue = [34, 43, 35]\nclass = Versicolor'), Text(1785.6, 1359.0, 'Sepal length <= 50.0 in = 0.493 in samples = 77 invalue = [34, 43, 0] inclass = Versicolor'),Text(892.8, 453.0, 'gini = 0.0\nsamples = 34\nvalue = [34, 0, 0]\nclass = Setosa'),  $Text(2678.399999999996, 453.0, 'gini = 0.0 \nsamples = 43 \nvalue = [0, 43, 0] \nclass = Versicolor'),$  $Text(3571.2, 1359.0, 'gini = 0.0 \nsamples = 35 \nvalue = [0, 0, 35] \nclass = Virginica')]$ Sepal length <= 102.5 gini = 0.663samples = 112value = [34, 43, 35]class = Versicolor Sepal length <= 50.0 gini = 0.0gini = 0.493samples = 35samples = 77value = [0, 0, 35]value = [34, 43, 0]class = Virginica class = Versicolor gini = 0.0gini = 0.0samples = 34samples = 43value = [0, 43, 0]value = [34, 0, 0]

class = Versicolor

Out[17]: (38,)

#Building DecisionTree Classifier Model

clf = DecisionTreeClassifier() clf.fit(X\_train, y\_train)

y\_pred\_train = clf.predict(X\_train) y\_pred\_test = clf.predict(X\_test)

pd.DataFrame(df.Species).value\_counts()

50

50

accuracy\_score(y\_train, y\_pred\_train)

accuracy\_score(y\_test, y\_pred\_test)

# Testing the accuracy of model prediction

class = Setosa

Out[20]: DecisionTreeClassifier()

Iris-setosa Iris-versicolor

Iris-virginica

dtype: int64

Out[24]: 0.9473684210526315

from sklearn.tree import DecisionTreeClassifier

In [20]:

In [21]:

In [22]:

In [23]:

Out[23]: 1.0

In [24]:

In [30]:

In [ ]:

Out[22]: Species