Exercise 2.2

1.1 Problem Statement:

To Implement ID3 (information gain) algorithm for decision tree learning for transforming continuous variables into discrete variables.

1.2 Description of Machine learning Algorithm:

ID3 (Iterative Dichotomiser 3) is named such because the algorithm iteratively (repeatedly) dichotomizes(divides) features into two or more groups at each step.

Invented by Ross Quinlan ID3 algorithm builds decision trees using a top-down greedy search approach through the space of possible branches with no backtracking. Classification algorithm that follows a greedy approach by selecting a best attribute that yields maximum Information Gain(IG) or minimum Entropy(H). Typically used in Machine Learning and Natural Language Processing domains.

Some major benefits of ID3 are: **Understandable prediction rules are created from the training data**. Builds a short tree in relatively small time. It only needs to test enough attributes until all data is classified.

1.3 Description of the Dataset:

Title of the dataset: Iris Plants Database

The Iris Dataset contains information of three species of Iris flowers (Iris setosa, Iris virginica and Iris versicolor). The data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.

Data Set Characteristics: Multivariate

Area: Life Sciences

Number of samples (or instances) in the dataset: 150

Number of attributes (or features): 05 Attribute

Information:

- 1. sepal length in cm
- 2. sepal width in cm
- 3. petal length in cm
- 4. petal width in cm
- 5. class:
 - -- Iris Setosa
 - -- Iris Versicolour
 - -- Iris Virginica
- 6. Number of samples of each species of iris flowers:

Class Distribution: 33.3% for each of 3 classes.

50 (Setosa), 50 (Versicolor), 50 (Virginica)

- 7. Predicted attribute: class of iris plant.
- 8. Missing Attribute Values: None

Feature Name	Units of measurement	Datatype	Description	
sepal length	Centimeters	Real (Numerical)	Length of Iris flower's sepal	
sepal width length	Centimeters	Real (Numerical)	Width of Iris flower's sepal	
petal length	Centimeters	Real (Numerical)	Length of Iris flower's petal	
petal width length	Centimeters	Real (Numerical)	Width of Iris flower's petal	
variety	Variety of species [Setosa, Virginica, Versicolor]	Object (Categorical)	Variety of the species of Iris flower	

1.4 Data Preprocessing and Exploratory Data Analysis (EDA):

Data preprocessing is the process of transforming raw data into an understandable format. It is also an important step in data mining as we cannot work with raw data. The quality of the data should be checked before applying machine learning or data mining algorithms.

Major Tasks in Data Preprocessing:

- 1. Data cleaning
- 2. Data integration
- 3. Data reduction
- 4. Data transformation

Exploratory data analysis (EDA) is a technique that data professionals can use to understand a dataset before they start to model it. Some people refer to EDA as data exploration. The goal of conducting EDA is to determine the characteristics of the dataset. Conducting EDA can help data analysts make predictions and assumptions about data. Often, EDA involves data visualization, including creating graphs like histograms, scatter plots and box plots.

Major Tasks in EDA:

- 1. Observe your dataset
- 2. Find any missing values
- 3. Categorize your values
- 4. Find the shape of your dataset
- 5. Identify relationships in your dataset
- 6. Locate any outliers in your dataset

1.5 Machine Learning Package Used for Model building:

The scikit-learn (formerly scikits.learn and also known as sklearn) is a free software machine learning library for the Python programming language. It provides simple and efficient tools for predictive data analysis. It features various classification, regression and clustering algorithms including support-vector machines, random forests, gradient boosting, k-means and DBSCAN, and is designed to interoperate with the Python numerical and scientific libraries NumPy and SciPy.

The sklearn.model_selection.train_test_split() method splits arrays or matrices into random train and test subsets. The parameters for the method are as follows:

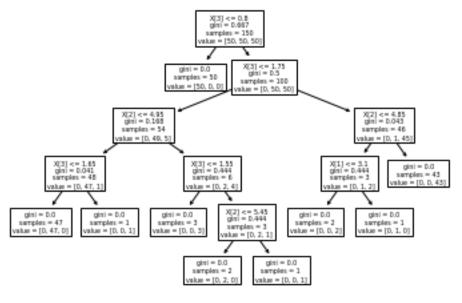
sklearn.model_selection.train_test_split(*arrays, test_size=None, train_size=None, random_state=Non e, shuffle=True, stratify=None)

It returns lists containing train-test split of inputs.

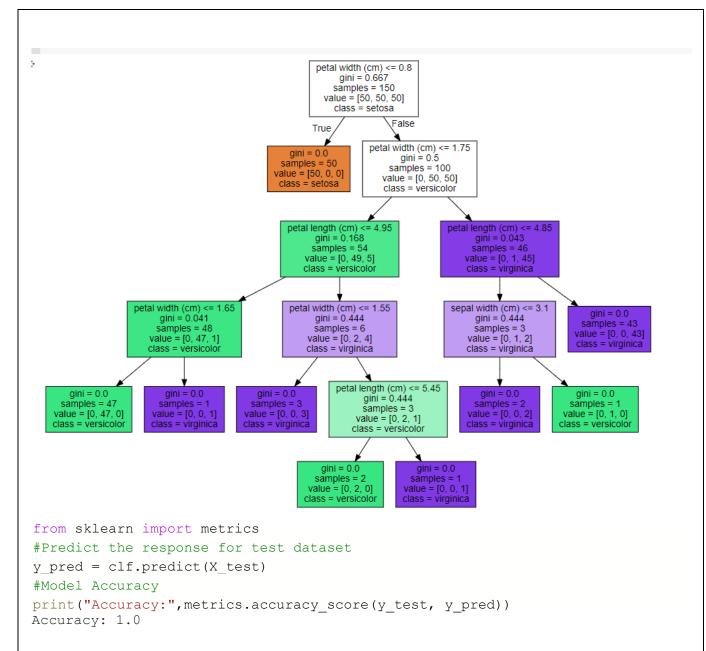
1.6 Implementation:

```
from google.colab import files
uploaded=files.upload()
from sklearn.datasets import load iris
iris = load iris()
X = iris.data
y = iris.target
feature names = iris.feature names
target names = iris.target names
print("Feature names:", feature names)
print("Target names:", target names)
print("\nFirst 10 rows of X:\n", X[:10])
Feature names: ['sepal length (cm)', 'sepal width (cm)', 'petal length (cm)',
'petal width (cm)']
Target names: ['setosa' 'versicolor' 'virginica']
First 10 rows of X:
 [[5.1 3.5 1.4 0.2]
 [4.9 3. 1.4 0.2]
 [4.7 3.2 1.3 0.2]
 [4.6 3.1 1.5 0.2]
 [5. 3.6 1.4 0.2]
 [5.4 3.9 1.7 0.4]
 [4.6 3.4 1.4 0.3]
 [5. 3.4 1.5 0.2]
 [4.4 2.9 1.4 0.2]
 [4.9 3.1 1.5 0.1]]
from sklearn.model selection import train test split
X train, X test, y train, y test = train test split(
   X, y, test size = 0.3, random state = 1
print(X train.shape)
print(X test.shape)
print(y train.shape)
print(y test.shape)
(105, 4)
(45, 4)
(105,)
(45,)
from sklearn import tree
clf = tree.DecisionTreeClassifier()
clf = clf.fit(X, y)
tree.plot tree(clf)
[\text{Text}(0.5, 0.9166666666666666, 'X[3] <= 0.8 \text{ ngini} = 0.667 \text{ nsamples} = 150 \text{ nvalue} =
[50, 50, 50]'),
 Text(0.4230769230769231, 0.75, 'gini = 0.0 \nsamples = 50 \nvalue = [50, 0, 0]'),
```

```
Text(0.5769230769230769, 0.75, 'X[3] <= 1.75\ngini = 0.5\nsamples = 100\nvalue =
 [0, 50, 50]'),
   Text(0.3076923076923077, 0.583333333333333334, 'X[2] \le 4.95 \ngini = 0.168 \nsamples
= 54 \nvalue = [0, 49, 5]'),
  Text(0.15384615384615385, 0.4166666666666667, 'X[3] <= 1.65 \nqini =
0.041 \times = 48 \times = [0, 47, 1]'),
   Text(0.07692307692307693, 0.25, 'gini = 0.0 \nsamples = 47 \nvalue = [0, 47, 0]'),
    Text(0.23076923076923078, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 1]'),
    Text(0.46153846153846156, 0.416666666666667, 'X[3] <= 1.55 
0.444 \times = 6 \times = [0, 2, 4]'),
   Text(0.38461538461538464, 0.25, 'gini = 0.0 \nsamples = 3 \nvalue = [0, 0, 3]'),
    [0, 2, 1]'),
    Text(0.46153846153846156, 0.0833333333333333, 'qini = 0.0 \nsamples = 2 \nvalue = 0.0 \nsamples = 0.0 \nsamp
 [0, 2, 0]'),
   Text(0.6153846153846154, 0.08333333333333333, 'gini = 0.0 \nsamples = 1 \nvalue = 1 \nva
[0, 0, 1]'),
  = 46 \ln e = [0, 1, 45]'),
  = 3 \text{ nvalue} = [0, 1, 2]'),
    Text(0.6923076923076923, 0.25, 'gini = 0.0 \nsamples = 2 \nvalue = [0, 0, 2]'),
   Text(0.8461538461538461, 0.25, 'gini = 0.0 \nsamples = 1 \nvalue = [0, 1, 0]'),
   [0, 0, 43]')]
```



Graphical Representation:



Implement ID3 (information gain) algorithm for decision tree learning for transforming continuous variables into discrete variables.

```
import itertools
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from pandas import DataFrame, Series
from IPython.display import Image
try:
    from StringIO import StringIO
except ImportError:
    from io import StringIO
import pydotplus
from sklearn import preprocessing
from sklearn import tree
%matplotlib inline
```

```
plotting decision tree using method
def plot decision tree(clf, feature name, target name):
    dot data = StringIO()
    tree.export graphviz(clf, out file=dot data,
                            feature names=feature name,
                            class names=target name,
                            filled=True, rounded=True,
                            special characters=True)
    graph = pydotplus.graph from dot data(dot data.getvalue())
    return Image(graph.create png())
df=pd.read csv("Iris.csv")
df
  Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species
  0
                  1
                                5.1
                                               3.5
                                                             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
  3
                  4
                                4.6
                                               3.1
                                                             1.5
                                                                      0.2
                                                                            Iris-setosa
  4
                  5
                                5.0
                                               3.6
                                                             1.4
                                                                      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
 147
                148
                                6.5
                                               3.0
                                                             5.2
                                                                      2.0 Iris-virginica
 148
                                6.2
                                               3.4
                                                             5.4
                                                                      2.3 Iris-virginica
                149
 149
                150
                                5.9
                                               3.0
                                                             5.1
                                                                      1.8 Iris-virginica
150 rows × 6 columns
df['Species'].unique()
df['Species']=df['Species'].replace(['Iris-setosa','Iris-versicolor','Iris-
virginica'], [0,1,2])
df['Species']
        0
0
1
        0
2
        0
3
        0
4
        0
145
       2
146
       2
       2
147
148
       2
149
Name: Species, Length: 150, dtype: int64
```

```
X train = df[['SepalLengthCm']]
Y train = df.Species
Method to Calculate Entropy
def entropy(target col):
    elements, counts = np.unique(target col, return counts=True)
    for (e, c) in zip(elements, counts):
        print("Node has {} elements of Class {}".format(c, e))
    entropy = np.sum(
        Γ
             ( –
counts[i] / np.sum(counts)) * np.log2(counts[i] / np.sum(counts))
            for i in range(len(elements))
    )
    return entropy
entropy before split = entropy(df.Species)
print ("Initial Entropy = {}".format(entropy before split))
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Initial Entropy = 1.584962500721156
a=df['SepalLengthCm'].to numpy()
a=np.unique(a)
a=np.sort(a)
array([4.3, 4.4, 4.5, 4.6, 4.7, 4.8, 4.9, 5., 5.1, 5.2, 5.3, 5.4, 5.5,
       5.6, 5.7, 5.8, 5.9, 6. , 6.1, 6.2, 6.3, 6.4, 6.5, 6.6, 6.7, 6.8,
       6.9, 7., 7.1, 7.2, 7.3, 7.4, 7.6, 7.7, 7.9])
1=[]
for i in range (len(a)-1):
  k = (a[i] + a[i+1])/2
  l.append(k)
split values=np.array(1)
left node data = df.loc[df.SepalLengthCm < split values[0]]</pre>
right node data = df.loc[df.SepalLengthCm > split values[0]]
left node data
Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species
1314 4.3
                 3.0
                              1.1
                                          0.1
right node data
```

Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species	
0	1	5.1	3.5	1.4	0.2	0
1	2	4.9	3.0	1.4	0.2	0
2	3	4.7	3.2	1.3	0.2	0
3	4	4.6	3.1	1.5	0.2	0
4	5	5.0	3.6	1.4	0.2	0
•••						
145	146	6.7	3.0	5.2	2.3	2
146	147	6.3	2.5	5.0	1.9	2
147	148	6.5	3.0	5.2	2.0	2
148	149	6.2	3.4	5.4	2.3	2
149	150	5.9	3.0	5.1	1.8	2

149 rows × 6 columns

Method to Calculate Information Gain

```
def calc info gain numeric(df, left, right, entropy before split, split values
):
   print("Split Value = {} \n".format(split values))
   total elements = df.shape[0]
   print("Left Node")
   ent left = entropy(left)
   print("Entropy of Left Node {} \n".format(ent left))
   print("Right Node")
   ent right = entropy(right)
   print("Entropy of Right Node {} \n".format(ent_right))
   weighted entropy = ((left.shape[0] / total elements) * ent left) + (
        (right.shape[0] / total elements) * ent right
    info_gain = entropy_before_split - weighted_entropy
   print("Info gain at split {} is {} \n".format(split_values, info_gain))
   return info gain
calc info gain numeric(
   df, left node data. Species, right node data. Species, entropy before split,
split values[0]
)
```

```
Split Value = 4.35
Left Node
Node has 1 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 49 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5848973705351974
Info gain at split 4.35 is 0.01063111265619332
0.01063111265619332
info_gains = []
for sp value in split values:
    left data = df.loc[df.SepalLengthCm < sp value]</pre>
    right data = df.loc[df.SepalLengthCm > sp value]
    info gains.append(
         calc_info_gain_numeric(
             df, left data. Species, right data. Species, entropy before split, s
p value
    )
Split Value = 4.35
Left Node
Node has 1 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 49 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5848973705351974
Info gain at split 4.35 is 0.01063111265619332
Split Value = 4.45
Left Node
Node has 4 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 46 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5838692920574484
```

```
Info gain at split 4.45 is 0.04332972311857297
Split Value = 4.55
Left Node
Node has 5 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 45 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5832262740679837
Info gain at split 4.55 is 0.054510435788771794
Split Value = 4.65
Left Node
Node has 9 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 41 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5789467341944396
Info gain at split 4.65 is 0.10075257057838294
Split Value = 4.75
Left Node
Node has 11 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 39 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5756582992479666
Info gain at split 4.75 is 0.12485247675137368
Split Value = 4.85
Left Node
Node has 16 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 34 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5634074323981904
Info gain at split 4.85 is 0.18831852777877267
```

```
Split Value = 4.95
Left Node
Node has 20 elements of Class 0
Node has 1 elements of Class 1
Node has 1 elements of Class 2
Entropy of Left Node 0.5304060778306042
Right Node
Node has 30 elements of Class 0
Node has 49 elements of Class 1
Node has 49 elements of Class 2
Entropy of Right Node 1.5511852922535474
Info gain at split 4.95 is 0.18349149324964031
Split Value = 5.05
Left Node
Node has 28 elements of Class 0
Node has 3 elements of Class 1
Node has 1 elements of Class 2
Entropy of Left Node 0.6449742087569881
Right Node
Node has 22 elements of Class 0
Node has 47 elements of Class 1
Node has 49 elements of Class 2
Entropy of Right Node 1.5072705578320558
Info gain at split 5.05 is 0.2616484973584483
Split Value = 5.15
Left Node
Node has 36 elements of Class 0
Node has 4 elements of Class 1
Node has 1 elements of Class 2
Entropy of Left Node 0.6229837106687364
Right Node
Node has 14 elements of Class 0
Node has 46 elements of Class 1
Node has 49 elements of Class 2
Entropy of Right Node 1.424077891299181
Info gain at split 5.15 is 0.37985035212762996
Split Value = 5.25
Left Node
Node has 39 elements of Class 0
Node has 5 elements of Class 1
Node has 1 elements of Class 2
Entropy of Left Node 0.6531791627726858
Right Node
Node has 11 elements of Class 0
```

Node has 45 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 1.3779796176519241 Info gain at split 5.25 is 0.42442301953300343 Split Value = 5.35Left Node Node has 40 elements of Class 0 Node has 5 elements of Class 1 Node has 1 elements of Class 2 Entropy of Left Node 0.6434149067106801 Right Node Node has 10 elements of Class 0 Node has 45 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 1.359348827320379 Info gain at split 5.35 is 0.4451667423877512 Split Value = 5.45Left Node Node has 45 elements of Class 0 Node has 6 elements of Class 1 Node has 1 elements of Class 2 Entropy of Left Node 0.6496096346956632 Right Node Node has 5 elements of Class 0 Node has 44 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 1.2377158231343603 Info gain at split 5.45 is 0.5511234895788775 Split Value = 5.55Left Node Node has 47 elements of Class 0 Node has 11 elements of Class 1 Node has 1 elements of Class 2 Entropy of Left Node 0.8128223064150747 Right Node Node has 3 elements of Class 0 Node has 39 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 1.167065448996099 Info gain at split 5.55 is 0.5572326878069267 Split Value = 5.65Left Node Node has 47 elements of Class 0

Node has 16 elements of Class 1 Node has 2 elements of Class 2 Entropy of Left Node 0.9905881818153937 Right Node Node has 3 elements of Class 0 Node has 34 elements of Class 1 Node has 48 elements of Class 2 Entropy of Right Node 1.1646047697931075 Info gain at split 5.65 is 0.4957649190517246 Split Value = 5.75Left Node Node has 49 elements of Class 0 Node has 21 elements of Class 1 Node has 3 elements of Class 2 Entropy of Left Node 1.0923679048777892 Right Node Node has 1 elements of Class 0 Node has 29 elements of Class 1 Node has 47 elements of Class 2 Entropy of Right Node 1.0466940096877775 Info gain at split 5.75 is 0.5160405287075729 Split Value = 5.85Left Node Node has 50 elements of Class 0 Node has 24 elements of Class 1 Node has 6 elements of Class 2 Entropy of Left Node 1.2251570385077257 Right Node Node has 26 elements of Class 1 Node has 44 elements of Class 2 Entropy of Right Node 0.9517626756348311 Info gain at split 5.85 is 0.48738949822078115 Split Value = 5.95Left Node Node has 50 elements of Class 0 Node has 26 elements of Class 1 Node has 7 elements of Class 2 Entropy of Left Node 1.2659342914094807 Right Node Node has 24 elements of Class 1 Node has 43 elements of Class 2 Entropy of Right Node 0.9411864371816836 Info gain at split 5.95 is 0.4640822508667579

```
Split Value = 6.05
Left Node
Node has 50 elements of Class 0
Node has 30 elements of Class 1
Node has 9 elements of Class 2
Entropy of Left Node 1.3304654268094023
Right Node
Node has 20 elements of Class 1
Node has 41 elements of Class 2
Entropy of Right Node 0.9127341558073343
Info gain at split 6.05 is 0.4243744574525947
Split Value = 6.15
Left Node
Node has 50 elements of Class 0
Node has 34 elements of Class 1
Node has 11 elements of Class 2
Entropy of Left Node 1.378063041001916
Right Node
Node has 16 elements of Class 1
Node has 39 elements of Class 2
Entropy of Right Node 0.8698926856041563
Info gain at split 6.15 is 0.39322859003175203
Split Value = 6.25
Left Node
Node has 50 elements of Class 0
Node has 36 elements of Class 1
Node has 13 elements of Class 2
Entropy of Left Node 1.4130351465796736
Right Node
Node has 14 elements of Class 1
Node has 37 elements of Class 2
Entropy of Right Node 0.8478617451660526
Info gain at split 6.25 is 0.3640863106221135
Split Value = 6.35
Left Node
Node has 50 elements of Class 0
Node has 39 elements of Class 1
Node has 19 elements of Class 2
Entropy of Left Node 1.486053069017246
Right Node
Node has 11 elements of Class 1
Node has 31 elements of Class 2
Entropy of Right Node 0.8296071030882032
```

```
Info gain at split 6.35 is 0.2827143021640419
Split Value = 6.45
Left Node
Node has 50 elements of Class 0
Node has 41 elements of Class 1
Node has 24 elements of Class 2
Entropy of Left Node 1.5246940362018935
Right Node
Node has 9 elements of Class 1
Node has 26 elements of Class 2
Entropy of Right Node 0.8224042259549891
Info gain at split 6.45 is 0.2241360869102067
Split Value = 6.55
Left Node
Node has 50 elements of Class 0
Node has 42 elements of Class 1
Node has 28 elements of Class 2
Entropy of Left Node 1.5462566034163765
Right Node
Node has 8 elements of Class 1
Node has 22 elements of Class 2
Entropy of Right Node 0.8366407419411673
Info gain at split 6.55 is 0.1806290695998214
Split Value = 6.65
Left Node
Node has 50 elements of Class 0
Node has 44 elements of Class 1
Node has 28 elements of Class 2
Entropy of Left Node 1.5453788250514908
Right Node
Node has 6 elements of Class 1
Node has 22 elements of Class 2
Entropy of Right Node 0.74959525725948
Info gain at split 6.65 is 0.1881299416575073
Split Value = 6.75
Left Node
Node has 50 elements of Class 0
Node has 47 elements of Class 1
Node has 33 elements of Class 2
Entropy of Left Node 1.562956340286807
Right Node
Node has 3 elements of Class 1
Node has 17 elements of Class 2
```

```
Entropy of Right Node 0.6098403047164004
Info gain at split 6.75 is 0.14908829851040317
Split Value = 6.85
Left Node
Node has 50 elements of Class 0
Node has 48 elements of Class 1
Node has 35 elements of Class 2
Entropy of Left Node 1.5680951037987416
Right Node
Node has 2 elements of Class 1
Node has 15 elements of Class 2
Entropy of Right Node 0.5225593745369408
Info gain at split 6.85 is 0.1353614462387518
Split Value = 6.95
Left Node
Node has 50 elements of Class 0
Node has 49 elements of Class 1
Node has 38 elements of Class 2
Entropy of Left Node 1.5744201314186457
Right Node
Node has 1 elements of Class 1
Node has 12 elements of Class 2
Entropy of Right Node 0.39124356362925566
Info gain at split 6.95 is 0.11308433851092436
Split Value = 7.05
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 38 elements of Class 2
Entropy of Left Node 1.573692105413468
Right Node
Node has 12 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 7.05 is 0.1371657637407655
Split Value = 7.15
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 39 elements of Class 2
Entropy of Left Node 1.5756582992479669
Right Node
Node has 11 elements of Class 2
```

```
Entropy of Right Node 0.0
Info gain at split 7.15 is 0.12485247675137345
Split Value = 7.25
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 42 elements of Class 2
Entropy of Left Node 1.5802897199682002
Right Node
Node has 8 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 7.25 is 0.08895489915125987
Split Value = 7.35
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 43 elements of Class 2
Entropy of Left Node 1.5814447393681608
Right Node
Node has 7 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 7.35 is 0.0773185158568428
Split Value = 7.5
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 44 elements of Class 2
Entropy of Left Node 1.5824207639595793
Right Node
Node has 6 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 7.5 is 0.06583856731995996
Split Value = 7.65
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 45 elements of Class 2
Entropy of Left Node 1.5832262740679837
Right Node
Node has 5 elements of Class 2
Entropy of Right Node 0.0
```

```
Info gain at split 7.65 is 0.054510435788771794
Split Value = 7.80000000000001
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 49 elements of Class 2
Entropy of Left Node 1.5848973705351974
Right Node
Node has 1 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 7.8000000000001 is 0.01063111265619332
info gains.index(max(info gains))
split values[info gains.index(max(info gains))]
clf = tree.DecisionTreeClassifier(criterion='entropy')
f = clf.fit(X train, Y train)
plot decision tree(clf, X train.columns,["Iris-setosa", "Iris-
versicolor", "Iris-virginica"])
```

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```
X train = df[['SepalWidthCm']]
Y train = df.Species
entropy before split = entropy(df.Species)
print ("Initial Entropy = {}".format(entropy before split))
a=df['SepalWidthCm'].to numpy()
a=np.unique(a)
a=np.sort(a)
1=[]
for i in range(len(a)-1):
  k = (a[i] + a[i+1])/2
  1.append(k)
split values=np.array(1)
info gains = []
for sp value in split values:
    left data = df.loc[df.SepalWidthCm < sp value]</pre>
    right data = df.loc[df.SepalWidthCm > sp_value]
    info gains.append(
        calc info gain numeric(
            df, left data. Species, right data. Species, entropy before split, s
p_value
    )
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Initial Entropy = 1.584962500721156
Split Value = 2.1
Left Node
Node has 1 elements of Class 1
Entropy of Left Node 0.0
Right Node
Node has 50 elements of Class 0
Node has 49 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5848973705351974
Info gain at split 2.1 is 0.01063111265619332
Split Value = 2.25
Left Node
Node has 3 elements of Class 1
Node has 1 elements of Class 2
Entropy of Left Node 0.8112781244591328
Right Node
Node has 50 elements of Class 0
Node has 47 elements of Class 1
Node has 49 elements of Class 2
Entropy of Right Node 1.5844871076524119
```

```
Info gain at split 2.25 is 0.02109429928723161
Split Value = 2.3499999999999999
Left Node
Node has 1 elements of Class 0
Node has 6 elements of Class 1
Node has 1 elements of Class 2
Entropy of Left Node 1.061278124459133
Right Node
Node has 49 elements of Class 0
Node has 44 elements of Class 1
Node has 49 elements of Class 2
Entropy of Right Node 1.583151655115024
Info gain at split 2.34999999999996 is 0.02964410057444633
Split Value = 2.45
Left Node
Node has 1 elements of Class 0
Node has 9 elements of Class 1
Node has 1 elements of Class 2
Entropy of Left Node 0.8658566174572235
Right Node
Node has 49 elements of Class 0
Node has 41 elements of Class 1
Node has 49 elements of Class 2
Entropy of Right Node 1.5800835409779546
Info gain at split 2.45 is 0.05725560080138847
Split Value = 2.55
Left Node
Node has 1 elements of Class 0
Node has 13 elements of Class 1
Node has 5 elements of Class 2
Entropy of Left Node 1.1050139971135322
Right Node
Node has 49 elements of Class 0
Node has 37 elements of Class 1
Node has 45 elements of Class 2
Entropy of Right Node 1.5753776105088113
Info gain at split 2.55 is 0.06916428124241358
Split Value = 2.65000000000000004
Left Node
Node has 1 elements of Class 0
Node has 16 elements of Class 1
Node has 7 elements of Class 2
Entropy of Left Node 1.0994839817876882
```

```
Right Node
Node has 49 elements of Class 0
Node has 34 elements of Class 1
Node has 43 elements of Class 2
Entropy of Right Node 1.5691537059098613
Info gain at split 2.6500000000000004 is 0.09095595067084261
Split Value = 2.75
Left Node
Node has 1 elements of Class 0
Node has 21 elements of Class 1
Node has 11 elements of Class 2
Entropy of Left Node 1.0961391592565068
Right Node
Node has 49 elements of Class 0
Node has 29 elements of Class 1
Node has 39 elements of Class 2
Entropy of Right Node 1.5529893541477424
Info gain at split 2.75 is 0.13248018944948536
Split Value = 2.8499999999999999
Left Node
Node has 1 elements of Class 0
Node has 27 elements of Class 1
Node has 19 elements of Class 2
Entropy of Left Node 1.1058104407448335
Right Node
Node has 49 elements of Class 0
Node has 23 elements of Class 1
Node has 31 elements of Class 2
Entropy of Right Node 1.514240401232251
Info gain at split 2.84999999999996 is 0.19869682044162928
Split Value = 2.95
Left Node
Node has 2 elements of Class 0
Node has 34 elements of Class 1
Node has 21 elements of Class 2
Entropy of Left Node 1.1449531285355223
Right Node
Node has 48 elements of Class 0
Node has 16 elements of Class 1
Node has 29 elements of Class 2
Entropy of Right Node 1.453571188599267
Info gain at split 2.95 is 0.2486661749461121
Split Value = 3.05
```

```
Left Node
Node has 8 elements of Class 0
Node has 42 elements of Class 1
Node has 33 elements of Class 2
Entropy of Left Node 1.3516377723645525
Right Node
Node has 42 elements of Class 0
Node has 8 elements of Class 1
Node has 17 elements of Class 2
Entropy of Right Node 1.2905041149657768
Info gain at split 3.05 is 0.2606310953280566
Split Value = 3.1500000000000004
Left Node
Node has 13 elements of Class 0
Node has 45 elements of Class 1
Node has 37 elements of Class 2
Entropy of Left Node 1.4331305536052563
Right Node
Node has 37 elements of Class 0
Node has 5 elements of Class 1
Node has 13 elements of Class 2
Entropy of Right Node 1.1910845982784386
Info gain at split 3.1500000000000004 is 0.24058213073573298
Split Value = 3.25
Left Node
Node has 18 elements of Class 0
Node has 48 elements of Class 1
Node has 42 elements of Class 2
Entropy of Left Node 1.4806821149663847
Right Node
Node has 32 elements of Class 0
Node has 2 elements of Class 1
Node has 8 elements of Class 2
Entropy of Right Node 0.9637459942073318
Info gain at split 3.25 is 0.24902249956730627
Split Value = 3.3499999999999999
Left Node
Node has 20 elements of Class 0
Node has 49 elements of Class 1
Node has 45 elements of Class 2
Entropy of Left Node 1.4934800703557258
Right Node
Node has 30 elements of Class 0
Node has 1 elements of Class 1
```

```
Node has 5 elements of Class 2
Entropy of Right Node 0.7583594919230799
Info gain at split 3.34999999999996 is 0.26791136918926517
Split Value = 3.45
Left Node
Node has 29 elements of Class 0
Node has 50 elements of Class 1
Node has 47 elements of Class 2
Entropy of Left Node 1.5475979046405137
Right Node
Node has 21 elements of Class 0
Node has 3 elements of Class 2
Entropy of Right Node 0.5435644431995964
Info gain at split 3.45 is 0.19800994991118914
Split Value = 3.55
Left Node
Node has 35 elements of Class 0
Node has 50 elements of Class 1
Node has 47 elements of Class 2
Entropy of Left Node 1.5687623685201277
Right Node
Node has 15 elements of Class 0
Node has 3 elements of Class 2
Entropy of Right Node 0.6500224216483541
Info gain at split 3.55 is 0.12644892582564116
Split Value = 3.65000000000000004
Left Node
Node has 37 elements of Class 0
Node has 50 elements of Class 1
Node has 48 elements of Class 2
Entropy of Left Node 1.5729579374091802
Right Node
Node has 13 elements of Class 0
Node has 2 elements of Class 2
Entropy of Right Node 0.5665095065529053
Info gain at split 3.6500000000000004 is 0.11264940639760335
Split Value = 3.75
Left Node
Node has 40 elements of Class 0
Node has 50 elements of Class 1
Node has 48 elements of Class 2
Entropy of Left Node 1.5784655196850803
```

```
Right Node
Node has 10 elements of Class 0
Node has 2 elements of Class 2
Entropy of Right Node 0.6500224216483541
Info gain at split 3.75 is 0.0807724288790137
Split Value = 3.8499999999999999
Left Node
Node has 44 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Left Node 1.5824207639595793
Right Node
Node has 6 elements of Class 0
Entropy of Right Node 0.0
Info gain at split 3.84999999999996 is 0.06583856731995996
Split Value = 3.95
Left Node
Node has 46 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Left Node 1.5838692920574484
Right Node
Node has 4 elements of Class 0
Entropy of Right Node 0.0
Info gain at split 3.95 is 0.04332972311857297
Split Value = 4.05
Left Node
Node has 47 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Left Node 1.5843574128205722
Right Node
Node has 3 elements of Class 0
Entropy of Right Node 0.0
Info gain at split 4.05 is 0.03229223615699528
Split Value = 4.15
Left Node
Node has 48 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Left Node 1.5846978317634905
Right Node
```

```
Node has 2 elements of Class 0
Entropy of Right Node 0.0
Info gain at split 4.15 is 0.021393973381178766
Split Value = 4.30000000000001
Left Node
Node has 49 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Left Node 1.5848973705351974
Right Node
Node has 1 elements of Class 0
Entropy of Right Node 0.0
Info gain at split 4.3000000000001 is 0.01063111265619332
print(info gains.index(max(info gains)))
split values[info gains.index(max(info gains))]
clf = tree.DecisionTreeClassifier(criterion='entropy')
clf = clf.fit(X train, Y train)
plot decision tree(clf, X train.columns,["Iris-setosa", "Iris-versicol
or","Iris-virginica"])
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X train = df[['PetalLengthCm']]
Y train = df.Species
entropy before split = entropy(df.Species)
print ("Initial Entropy = {}".format(entropy before split))
a=df['PetalLengthCm'].to numpy()
```

```
a=np.unique(a)
a=np.sort(a)
1=[]
for i in range (len(a)-1):
  k = (a[i] + a[i+1])/2
  l.append(k)
split values=np.array(1)
info gains = []
for sp value in split values:
    left data = df.loc[df.PetalLengthCm < sp value]</pre>
    right data = df.loc[df.PetalLengthCm > sp value]
    info gains.append(
        calc info gain numeric(
            df, left data. Species, right data. Species, entropy before split, s
p value
    )
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Initial Entropy = 1.584962500721156
Split Value = 1.05
Left Node
Node has 1 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 49 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5848973705351974
Info gain at split 1.05 is 0.01063111265619332
Split Value = 1.15
Left Node
Node has 2 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 48 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5846978317634905
Info gain at split 1.15 is 0.021393973381178766
Split Value = 1.25
Left Node
Node has 4 elements of Class 0
```

Entropy of Left Node 0.0 Right Node Node has 46 elements of Class 0 Node has 50 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 1.5838692920574484 Info gain at split 1.25 is 0.04332972311857297 Split Value = 1.35Left Node Node has 11 elements of Class 0 Entropy of Left Node 0.0 Right Node Node has 39 elements of Class 0 Node has 50 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 1.5756582992479666 Info gain at split 1.35 is 0.12485247675137368 Split Value = 1.45Left Node Node has 23 elements of Class 0Entropy of Left Node 0.0 Right Node Node has 27 elements of Class 0 Node has 50 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 1.5338218400329051 Info gain at split 1.45 is 0.286326676159963 Split Value = 1.55Left Node Node has 37 elements of Class 0 Entropy of Left Node 0.0 Right Node Node has 13 elements of Class 0 Node has 50 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 1.3999016587576067 Info gain at split 1.55 is 0.5303699177904257 Split Value = 1.65Left Node Node has 44 elements of Class 0 Entropy of Left Node 0.0

```
Right Node
Node has 6 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.2572091905839593
Info gain at split 1.65 is 0.6965346727084915
Split Value = 1.7999999999999999
Left Node
Node has 48 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 2 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.119625155917844
Info gain at split 1.7999999999999 is 0.8236173946970221
Split Value = 2.45
Left Node
Node has 50 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.0
Info gain at split 2.45 is 0.9182958340544894
Split Value = 3.15
Left Node
Node has 50 elements of Class 0
Node has 1 elements of Class 1
Entropy of Left Node 0.13923299905509887
Right Node
Node has 49 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 0.999926399368686
Info gain at split 3.15 is 0.8776718574590896
Split Value = 3.4
Left Node
Node has 50 elements of Class 0
Node has 3 elements of Class 1
Entropy of Left Node 0.3138129641688651
Right Node
Node has 47 elements of Class 1
```

Node has 50 elements of Class 2 Entropy of Right Node 0.999309898706868 Info gain at split 3.4 is 0.8278615188843824 Split Value = 3.55Left Node Node has 50 elements of Class 0 Node has 5 elements of Class 1 Entropy of Left Node 0.4394969869215134 Right Node Node has 45 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9980008838722996 Info gain at split 3.55 is 0.7917463790641448 Split Value = 3.65000000000000004Left Node Node has 50 elements of Class 0 Node has 6 elements of Class 1 Entropy of Left Node 0.49123734182433315 Right Node Node has 44 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9970590569034106 Info gain at split 3.6500000000000004 is 0.7767435507806011 Split Value = 3.75Left Node Node has 50 elements of Class 0 Node has 7 elements of Class 1 Entropy of Left Node 0.5373760853377336 Right Node Node has 43 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9959094138937685 Info gain at split 3.75 is 0.7632957516786808 Split Value = 3.8499999999999999Left Node Node has 50 elements of Class 0 Node has 8 elements of Class 1 Entropy of Left Node 0.5787946246321198 Right Node Node has 42 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.994538681650011

Info gain at split 3.84999999999996 is 0.7511781877847297 Split Value = 3.95Left Node Node has 50 elements of Class 0 Node has 11 elements of Class 1 Entropy of Left Node 0.6807937753703206 Right Node Node has 39 elements of Class 1Node has 50 elements of Class 2 Entropy of Right Node 0.9889525767600615 Info gain at split 3.95 is 0.7213278365262558 Split Value = 4.05Left Node Node has 50 elements of Class 0 Node has 16 elements of Class 1 Entropy of Left Node 0.7990485210442682 Right Node Node has 34 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9736680645496202 Info gain at split 4.05 is 0.6881270353138906 Split Value = 4.15Left Node Node has 50 elements of Class 0 Node has 19 elements of Class 1 Entropy of Left Node 0.8490544242540479 Right Node Node has 31 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9599377175669781 Info gain at split 4.15 is 0.6760310980781259 Split Value = 4.25Left Node Node has 50 elements of Class 0 Node has 23 elements of Class 1 Entropy of Left Node 0.8989377852081333 Right Node Node has 27 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9346466439786691

Info gain at split 4.25 is 0.6676941680108144

Split Value = 4.35Left Node Node has 50 elements of Class 0 Node has 25 elements of Class 1 Entropy of Left Node 0.9182958340544896 Right Node Node has 25 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9182958340544896 Info gain at split 4.35 is 0.6666666666666665 Split Value = 4.45Left Node Node has 50 elements of Class 0 Node has 29 elements of Class 1 Entropy of Left Node 0.9484103893488013 Right Node Node has 21 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.8760643678555242 Info gain at split 4.45 is 0.6707958948791727 Split Value = 4.55Left Node Node has 50 elements of Class 0 Node has 36 elements of Class 1 Node has 1 elements of Class 2 Entropy of Left Node 1.0600686735422975 Right Node Node has 14 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 0.7642045065086203 Info gain at split 4.55 is 0.6491567773330031 Split Value = 4.65Left Node Node has 50 elements of Class 0 Node has 39 elements of Class 1 Node has 1 elements of Class 2 Entropy of Left Node 1.0660364738367423 Right Node Node has 11 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 0.6873150928309273 Info gain at split 4.65 is 0.6704145792867398

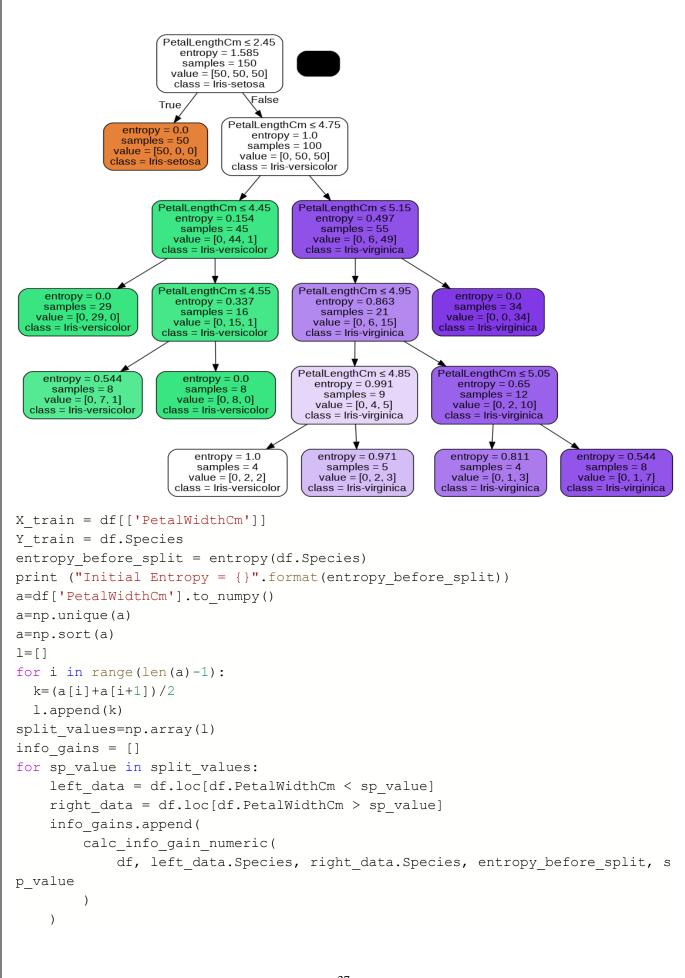
Split Value = 4.75Left Node Node has 50 elements of Class 0 Node has 44 elements of Class 1 Node has 1 elements of Class 2 Entropy of Left Node 1.0708261271859234 Right Node Node has 6 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 0.4971677614160753 Info gain at split 4.75 is 0.7244777743175104 Split Value = 4.85Left Node Node has 50 elements of Class 0 Node has 46 elements of Class 1 Node has 3 elements of Class 2 Entropy of Left Node 1.1643914991753441 Right Node Node has 4 elements of Class 1 Node has 47 elements of Class 2 Entropy of Right Node 0.39662777277837885 Info gain at split 4.85 is 0.6816106685207801 Split Value = 4.95Left Node Node has 50 elements of Class 0 Node has 48 elements of Class 1 Node has 6 elements of Class 2 Entropy of Left Node 1.2602398669134125 Right Node Node has 2 elements of Class 1 Node has 44 elements of Class 2 Entropy of Right Node 0.2580186686648155 Info gain at split 4.95 is 0.6320704679373134 Split Value = 5.05Left Node Node has 50 elements of Class 0 Node has 49 elements of Class 1 Node has 9 elements of Class 2 Entropy of Left Node 1.330416049724929 Right Node Node has 1 elements of Class 1 Node has 41 elements of Class 2 Entropy of Right Node 0.1623261801753929

Info gain at split 5.05 is 0.5816116144700971 Split Value = 5.15Left Node Node has 50 elements of Class 0 Node has 50 elements of Class 1 Node has 16 elements of Class 2 Entropy of Left Node 1.4408635901493612 Right Node Node has 34 elements of Class 2 Entropy of Right Node 0.0 Info gain at split 5.15 is 0.4706946576723168 Split Value = 5.25Left Node Node has 50 elements of Class 0 Node has 50 elements of Class 1 Node has 18 elements of Class 2 Entropy of Left Node 1.4636238205191796 Right Node Node has 32 elements of Class 2 Entropy of Right Node 0.0 Info gain at split 5.25 is 0.43357842857940154 Split Value = 5.35Left Node Node has 50 elements of Class 0 Node has 50 elements of Class 1 Node has 20 elements of Class 2 Entropy of Left Node 1.4833557549816876 Right Node Node has 30 elements of Class 2 Entropy of Right Node 0.0 Info gain at split 5.35 is 0.39827789673580605 Split Value = 5.45Left Node Node has 50 elements of Class 0 Node has 50 elements of Class 1 Node has 22 elements of Class 2 Entropy of Left Node 1.5004659065178614 Right Node Node has 28 elements of Class 2 Entropy of Right Node 0.0 Info gain at split 5.45 is 0.3645835634199621

```
Split Value = 5.55
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 25 elements of Class 2
Entropy of Left Node 1.5219280948873621
Right Node
Node has 25 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 5.55 is 0.31668908831502085
Split Value = 5.65
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 31 elements of Class 2
Entropy of Left Node 1.5527687679537436
Right Node
Node has 19 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 5.65 is 0.22887777670821996
Split Value = 5.75
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 34 elements of Class 2
Entropy of Left Node 1.5634074323981904
Right Node
Node has 16 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 5.75 is 0.18831852777877267
Split Value = 5.85
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 37 elements of Class 2
Entropy of Left Node 1.5714963639405002
Right Node
Node has 13 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 5.85 is 0.149662488322166
Split Value = 5.95
```

```
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 39 elements of Class 2
Entropy of Left Node 1.5756582992479669
Right Node
Node has 11 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 5.95 is 0.12485247675137345
Split Value = 6.05
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 41 elements of Class 2
Entropy of Left Node 1.5789467341944396
Right Node
Node has 9 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 6.05 is 0.10075257057838294
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 44 elements of Class 2
Entropy of Left Node 1.5824207639595793
Right Node
Node has 6 elements of Class 2
Entropy of Right Node 0.0
Split Value = 6.35
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 45 elements of Class 2
Entropy of Left Node 1.5832262740679837
Right Node
Node has 5 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 6.35 is 0.054510435788771794
Split Value = 6.5
Left Node
```

```
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 46 elements of Class 2
Entropy of Left Node 1.5838692920574484
Right Node
Node has 4 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 6.5 is 0.04332972311857297
Split Value = 6.65
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 47 elements of Class 2
Entropy of Left Node 1.5843574128205722
Right Node
Node has 3 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 6.65 is 0.03229223615699528
Split Value = 6.80000000000001
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 49 elements of Class 2
Entropy of Left Node 1.5848973705351974
Right Node
Node has 1 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 6.8000000000001 is 0.01063111265619332
info gains.index(max(info gains))
split values[info gains.index(max(info gains))]
clf = tree.DecisionTreeClassifier(criterion='entropy')
clf = clf.fit(X train, Y train)
plot decision tree(clf, X train.columns,["Iris-setosa", "Iris-
versicolor","Iris-virginica"])
```



```
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Initial Entropy = 1.584962500721156
Split Value = 0.15000000000000002
Left Node
Node has 6 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 44 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.5824207639595793
Info gain at split 0.1500000000000000 is 0.06583856731995996
Split Value = 0.25
Left Node
Node has 34 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 16 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.4408635901493612
Info gain at split 0.25 is 0.4706946576723168
Split Value = 0.35
Left Node
Node has 41 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 9 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.3285976826626762
Info gain at split 0.35 is 0.6195148513196114
Split Value = 0.45
Left Node
Node has 48 elements of Class 0
Entropy of Left Node 0.0
Right Node
Node has 2 elements of Class 0
Node has 50 elements of Class 1
Node has 50 elements of Class 2
Entropy of Right Node 1.119625155917844
```

Info gain at split 0.45 is 0.8236173946970221 Split Value = 0.55Left Node Node has 49 elements of Class 0 Entropy of Left Node 0.0 Right Node Node has 1 elements of Class 0 Node has 50 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 1.0702350572322654 Info gain at split 0.55 is 0.864337562184764 Split Value = 0.8Left Node Node has 50 elements of Class 0 Entropy of Left Node 0.0 Right Node Node has 50 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 1.0 Info gain at split 0.8 is 0.9182958340544894 Split Value = 1.05Left Node Node has 50 elements of Class 0 Node has 7 elements of Class 1Entropy of Left Node 0.5373760853377336 Right Node Node has 43 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9959094138937685 Info gain at split 1.05 is 0.7632957516786808 Split Value = 1.15 Left Node Node has 50 elements of Class 0 Node has 10 elements of Class 1 Entropy of Left Node 0.6500224216483541 Right Node Node has 40 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9910760598382222 Info gain at split 1.15 is 0.7303078961588811 Split Value = 1.25

Left Node Node has 50 elements of Class 0 Node has 15 elements of Class 1 Entropy of Left Node 0.7793498372920852 Right Node Node has 35 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.9774178175281716 Info gain at split 1.25 is 0.6933741412952886 Split Value = 1.35Left Node Node has 50 elements of Class 0 Node has 28 elements of Class 1 Entropy of Left Node 0.9418285354475155 Right Node Node has 22 elements of Class 1 Node has 50 elements of Class 2 Entropy of Right Node 0.8879763195151349 Info gain at split 1.35 is 0.6689830289211832 Split Value = 1.45Left Node Node has 50 elements of Class 0 Node has 35 elements of Class 1 Node has 1 elements of Class 2 Entropy of Left Node 1.0574541142159344 Right Node Node has 15 elements of Class 1 Node has 49 elements of Class 2 Entropy of Right Node 0.7855602922535472 Info gain at split 1.45 is 0.6435164172091734 Split Value = 1.55Left Node Node has 50 elements of Class 0 Node has 45 elements of Class 1 Node has 3 elements of Class 2 Entropy of Left Node 1.1649028407913813 Right Node Node has 5 elements of Class 1 Node has 47 elements of Class 2 Entropy of Right Node 0.4566836315394428 Info gain at split 1.55 is 0.6655756524704468 Split Value = 1.65

```
Left Node
Node has 50 elements of Class 0
Node has 48 elements of Class 1
Node has 4 elements of Class 2
Entropy of Left Node 1.1991801505660866
Right Node
Node has 2 elements of Class 1
Node has 46 elements of Class 2
Entropy of Right Node 0.24988229283318544
Info gain at split 1.65 is 0.6895576646295978
Split Value = 1.75
Left Node
Node has 50 elements of Class 0
Node has 49 elements of Class 1
Node has 5 elements of Class 2
Entropy of Left Node 1.2300240227101475
Right Node
Node has 1 elements of Class 1
Node has 45 elements of Class 2
Entropy of Right Node 0.15109697051711368
Info gain at split 1.75 is 0.6858094406835389
Split Value = 1.85
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 16 elements of Class 2
Entropy of Left Node 1.4408635901493612
Right Node
Node has 34 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 1.85 is 0.4706946576723168
Split Value = 1.95
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 21 elements of Class 2
Entropy of Left Node 1.492216254995037
Right Node
Node has 29 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 1.95 is 0.38124138835849286
Split Value = 2.05
```

```
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 27 elements of Class 2
Entropy of Left Node 1.5338218400329051
Right Node
Node has 23 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 2.05 is 0.286326676159963
Split Value = 2.15000000000000004
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 33 elements of Class 2
Entropy of Left Node 1.5601649549274947
Right Node
Node has 17 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 2.150000000000000 is 0.20161624068544404
Split Value = 2.25
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 36 elements of Class 2
Entropy of Left Node 1.5690590248577239
Right Node
Node has 14 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 2.25 is 0.16234898485015314
Left Node
Node has 50 elements of Class 0
Node has 50 elements of Class 1
Node has 44 elements of Class 2
Entropy of Left Node 1.5824207639595793
Right Node
Node has 6 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 2.34999999999996 is 0.06583856731995996
Split Value = 2.45
Left Node
Node has 50 elements of Class 0
```

```
Node has 50 elements of Class 1
Node has 47 elements of Class 2
Entropy of Left Node 1.5843574128205722
Right Node
Node has 3 elements of Class 2
Entropy of Right Node 0.0
Info gain at split 2.45 is 0.03229223615699528
print(info gains.index(max(info gains)))
split values[info gains.index(max(info gains))]
clf = tree.DecisionTreeClassifier(criterion='entropy')
clf = clf.fit(X train, Y train)
plot_decision_tree(clf, X train.columns,["Iris-setosa", "Iris-
versicolor", "Iris-virginica"])
5
                                                               PetalWidthCm ≤ 0.8
entropy = 1.585
samples = 150
                                                                value = [50, 50, 50]
class = Iris-setosa
                                                                                 PetalWidthCm \leq 1.75
                                                                                     entropy = 1.0
samples = 100
                                              samplés = 50
value = [50, 0, 0]
class = Iris-setosa
                                                                                 value = [0, 50, 50]
class = Iris-versicolor
                                                               retalWidthCm ≤ 1.
                                                                                                     PetalWidthCm ≤ 1.85
entropy = 0.151
samples = 46
value = [0, 1, 45]
class = Iris-virginica
                                                             entropy = 0.445
samples = 54
value = [0, 49, 5]
class = Iris-versicolor
                                                               PetalWidthCm ≤ 1.65
entropy = 0.706
samples = 26
value = [0, 21, 5]
class = Iris-versicolor
                                                                                                     entropy = 0.414
samples = 12
value = [0, 1, 11]
class = Iris-virginica
                      entropy = 0.0
samples = 28
value = [0, 28, 0]
class = Iris-versicolo
                                                                                                                                           entropy = 0.0
samples = 34
value = [0, 0, 34]
class = Iris-virginica
                                            PetalWidthCm ≤ 1.55
entropy = 0.65
                                                                                       entropy = 1.0
                                               entropy =
samples
                                                                                 samples = 2
value = [0, 1, 1]
class = Iris-versicolor
                                          value = [0, 20, 4]
class = Iris-versicolo
                      PetalWidthCm ≤ 1.45
entropy = 0.61
samples = 20
value = [0, 17, 3]
class = Iris-versicolor
                                                             entropy = 0.811
samples = 4
value = [0, 3, 1]
class = Iris-versicolor
                                          entropy = 0.65
samples = 12
value = [0, 10, 2]
class = Iris-versicolor
  entropy = 0.544
samples = 8
value = [0, 7, 1]
class = Iris-versicolor
```

1.7 Results and Discussion

To Implement ID3 (information gain) algorithm for decision tree learning for transforming continuous variables into discrete variables done successfully by calculating Entropy and Information Gain.

