Task - 3

Problem Statement: Decision tree classifier for prediction

Dataset Used: <a href="https://www.kaggle.com/datasets/uciml/iris">https://www.kaggle.com/datasets/uciml/iris</a> (https://www.kaggle.com/datasets/uciml/iris)

About DataSet: The Iris dataset contains measurements of iris flowers belonging to three different species: Setosa, Versicolor, and Virginica. The dataset is well-structured and contains a total of 150 samples, with 50 samples for each of the three iris species.

```
In [38]: #import necessary libraries
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier,plot_tree
from sklearn.metrics import accuracy_score,classification_report,confusion_mat
from sklearn import tree
import matplotlib.pyplot as plt
```

```
In [39]: #Loading dataset
df=pd.read_csv("iris.csv")
df
```

## Out[39]:

	ld	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	149	6.2	3.4	5.4	2.3	Iris-virginica
149	150	5.9	3.0	5.1	1.8	Iris-virginica

150 rows × 6 columns

```
In [58]: #preparing data
    x=df.drop('Species',axis=1)
    y=df['Species']
    x
```

## Out[58]:

	I	d	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
	0	1	5.1	3.5	1.4	0.2
	1	2	4.9	3.0	1.4	0.2
	2	3	4.7	3.2	1.3	0.2
	3	4	4.6	3.1	1.5	0.2
	4	5	5.0	3.6	1.4	0.2
,						
14	<b>5</b> 14	6	6.7	3.0	5.2	2.3
14	<b>6</b> 14	7	6.3	2.5	5.0	1.9
14	7 14	8	6.5	3.0	5.2	2.0
14	<b>8</b> 14	9	6.2	3.4	5.4	2.3
14	<b>9</b> 15	0	5.9	3.0	5.1	1.8

150 rows × 5 columns

```
In [41]:
Out[41]: 0
                   Iris-setosa
                   Iris-setosa
         1
         2
                   Iris-setosa
         3
                   Iris-setosa
         4
                   Iris-setosa
         145
                Iris-virginica
         146
                Iris-virginica
         147
                Iris-virginica
                Iris-virginica
         148
         149
                Iris-virginica
         Name: Species, Length: 150, dtype: object
In [70]: #categorical data preprocessing
         x_en=pd.get_dummies(x,drop_first=True)
In [60]: #training set and testing set
         x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=
In [61]: #initializing decision tree classifire
         clf=DecisionTreeClassifier(random_state=42)
```

```
#train the classifier
In [62]:
         clf.fit(x_train,y_train)
Out[62]:
                   DecisionTreeClassifier
          DecisionTreeClassifier(random_state=42)
In [63]:
         #prediction
         y_pred=clf.predict(x_test)
         print('Number of features in x_train:',len(x_train.columns))
In [64]:
         print('Number of features in x_test:',len(x_test.columns))
         Number of features in x_train: 5
         Number of features in x_test: 5
In [65]: | print('Feature name:',x_train.columns)
         Feature name: Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm',
         'PetalWidthCm'], dtype='object')
In [72]:
         #evaluation of model
         accuracy=accuracy_score(y_test,y_pred)
         conf_matrix=confusion_matrix(y_test,y_pred)
         classification_rep=classification_report(y_test,y_pred)
         print(f'Accuracy:{accuracy:.2f}')
```

Accuracy: 1.00

Interpretation: An accuracy score of 1.00, or 100%, indicates that the model correctly predicted all instances in the dataset, meaning there were no misclassifications. In other words, every data point in the dataset was classified correctly according to the model's predictions.

```
print('Confusion Matrix')
In [73]:
         print(conf_matrix)
         print(classification_rep)
         Confusion Matrix
         [[10 0 0]
          [0 9 0]
          [0 0 11]]
                           precision
                                        recall f1-score
                                                            support
              Iris-setosa
                                1.00
                                          1.00
                                                     1.00
                                                                 10
                                                                  9
         Iris-versicolor
                                1.00
                                          1.00
                                                     1.00
          Iris-virginica
                                1.00
                                                     1.00
                                          1.00
                                                                 11
                                                     1.00
                                                                 30
                 accuracy
               macro avg
                                1.00
                                          1.00
                                                     1.00
                                                                 30
            weighted avg
                                1.00
                                          1.00
                                                     1.00
                                                                 30
```

Interpretation: All the instances are correctly classified according to the diagonal element.

```
In [69]: #visualizationof decision tree
    plt.figure(figsize=(15,10))
    plot_tree(clf,filled=True, feature_names=list(x.columns),rounded=True,fontsize
    plt.show()
```

