B1: Write a program to demonstrate the working of the decision tree based ID3 algorithm. Use an appropriate data set for building the decision tree and apply this knowledge to classify a new sample.

In [ ]:
```
import pandas as pd
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier, export_text, plot_tree
```

In [14]:
```
import matplotlib.pyplot as plt
```

In [ ]:
```
# Load the Iris dataset
iris = load_iris()
X = iris.data # Independent variables
y = iris.target # dependent variable/class
```

In [16]:
```
# Create a DataFrame for better visualization
df = pd.DataFrame(data=X, columns=iris.feature_names)
df['target'] = y
print(df.head())
```

```
   sepal length (cm)  sepal width (cm)  petal length (cm)  petal width (cm)  \
0                5.1               3.5                1.4               0.2
1                4.9               3.0                1.4               0.2
2                4.7               3.2                1.3               0.2
3                4.6               3.1                1.5               0.2
4                5.0               3.6                1.4               0.2

   target
0       0
1       0
2       0
3       0
4       0
```

In [17]:
```
# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

In [27]:
```
X_train
```

Out[27]:
```
array([[4.6, 3.6, 1. , 0.2],
       [5.7, 4.4, 1.5, 0.4],
       [6.7, 3.1, 4.4, 1.4],
       [4.8, 3.4, 1.6, 0.2],
       [4.4, 3.2, 1.3, 0.2],
       [6.3, 2.5, 5. , 1.9],
       [6.4, 3.2, 4.5, 1.5],
       [5.2, 3.5, 1.5, 0.2],
       [5. , 3.6, 1.4, 0.2],
       [5.2, 4.1, 1.5, 0.1],
       [5.8, 2.7, 5.1, 1.9],
       [6. , 3.4, 4.5, 1.6],
       [6.7, 3.1, 4.7, 1.5],
       [5.4, 3.9, 1.3, 0.4],
       [5.4, 3.7, 1.5, 0.2],
       [5.5, 2.4, 3.7, 1. ],
       [6.3, 2.8, 5.1, 1.5],
       [6.4, 3.1, 5.5, 1.8],
       [6.6, 3. , 4.4, 1.4],
       [7.2, 3.6, 6.1, 2.5],
       [5.7, 2.9, 4.2, 1.3],
       [7.6, 3. , 6.6, 2.1],
       [5.6, 3. , 4.5, 1.5],
       [5.1, 3.5, 1.4, 0.2],
       [7.7, 2.8, 6.7, 2. ],
       [5.8, 2.7, 4.1, 1. ],
       [5.2, 3.4, 1.4, 0.2],
       [5. , 3.5, 1.3, 0.3],
       [5.1, 3.8, 1.9, 0.4],
       [5. , 2. , 3.5, 1. ],
       [6.3, 2.7, 4.9, 1.8],
       [4.8, 3.4, 1.9, 0.2],
       [5. , 3. , 1.6, 0.2],
       [5.1, 3.3, 1.7, 0.5],
       [5.6, 2.7, 4.2, 1.3],
       [5.1, 3.4, 1.5, 0.2],
       [5.7, 3. , 4.2, 1.2],
       [7.7, 3.8, 6.7, 2.2],
       [4.6, 3.2, 1.4, 0.2],
       [6.2, 2.9, 4.3, 1.3],
```

```
[5.7, 2.5, 5. , 2. ],
[5.5, 4.2, 1.4, 0.2],
[6. , 3. , 4.8, 1.8],
[5.8, 2.7, 5.1, 1.9],
[6. , 2.2, 4. , 1. ],
[5.4, 3. , 4.5, 1.5],
[6.2, 3.4, 5.4, 2.3],
[5.5, 2.3, 4. , 1.3],
[5.4, 3.9, 1.7, 0.4],
[5. , 2.3, 3.3, 1. ],
[6.4, 2.7, 5.3, 1.9],
[5. , 3.3, 1.4, 0.2],
[5. , 3.2, 1.2, 0.2],
[5.5, 2.4, 3.8, 1.1],
[6.7, 3. , 5. , 1.7],
[4.9, 3.1, 1.5, 0.2],
[5.8, 2.8, 5.1, 2.4],
[5. , 3.4, 1.5, 0.2],
[5. , 3.5, 1.6, 0.6],
[5.9, 3.2, 4.8, 1.8],
[5.1, 2.5, 3. , 1.1],
[6.9, 3.2, 5.7, 2.3],
[6. , 2.7, 5.1, 1.6],
[6.1, 2.6, 5.6, 1.4],
[7.7, 3. , 6.1, 2.3],
[5.5, 2.5, 4. , 1.3],
[4.4, 2.9, 1.4, 0.2],
[4.3, 3. , 1.1, 0.1],
[6. , 2.2, 5. , 1.5],
[7.2, 3.2, 6. , 1.8],
[4.6, 3.1, 1.5, 0.2],
[5.1, 3.5, 1.4, 0.3],
[4.4, 3. , 1.3, 0.2],
[6.3, 2.5, 4.9, 1.5],
[6.3, 3.4, 5.6, 2.4],
[4.6, 3.4, 1.4, 0.3],
[6.8, 3. , 5.5, 2.1],
[6.3, 3.3, 6. , 2.5],
[4.7, 3.2, 1.3, 0.2],
[6.1, 2.9, 4.7, 1.4],
[6.5, 2.8, 4.6, 1.5],
```

```
[6.2, 2.8, 4.8, 1.8],
[7. , 3.2, 4.7, 1.4],
[6.4, 3.2, 5.3, 2.3],
[5.1, 3.8, 1.6, 0.2],
[6.9, 3.1, 5.4, 2.1],
[5.9, 3. , 4.2, 1.5],
[6.5, 3. , 5.2, 2. ],
[5.7, 2.6, 3.5, 1. ],
[5.2, 2.7, 3.9, 1.4],
[6.1, 3. , 4.6, 1.4],
[4.5, 2.3, 1.3, 0.3],
[6.6, 2.9, 4.6, 1.3],
[5.5, 2.6, 4.4, 1.2],
[5.3, 3.7, 1.5, 0.2],
[5.6, 3. , 4.1, 1.3],
[7.3, 2.9, 6.3, 1.8],
[6.7, 3.3, 5.7, 2.1],
[5.1, 3.7, 1.5, 0.4],
[4.9, 2.4, 3.3, 1. ],
[6.7, 3.3, 5.7, 2.5],
[7.2, 3. , 5.8, 1.6],
[4.9, 3.6, 1.4, 0.1],
[6.7, 3.1, 5.6, 2.4],
[4.9, 3. , 1.4, 0.2],
[6.9, 3.1, 4.9, 1.5],
[7.4, 2.8, 6.1, 1.9],
[6.3, 2.9, 5.6, 1.8],
[5.7, 2.8, 4.1, 1.3],
[6.5, 3. , 5.5, 1.8],
[6.3, 2.3, 4.4, 1.3],
[6.4, 2.9, 4.3, 1.3],
[5.6, 2.8, 4.9, 2. ],
[5.9, 3. , 5.1, 1.8],
[5.4, 3.4, 1.7, 0.2],
[6.1, 2.8, 4. , 1.3],
[4.9, 2.5, 4.5, 1.7],
[5.8, 4. , 1.2, 0.2],
[5.8, 2.6, 4. , 1.2],
[7.1, 3. , 5.9, 2.1]])
```

In [ ]:
```
# Create a Decision Tree Classifier with the ID3 algorithm
# In sklearn, you can set the criterion to "entropy" to use the ID3 algorithm
clf = DecisionTreeClassifier(criterion='entropy', random_state=42) #
```

In [ ]:
```

```

In [19]:
```
# Train the model
clf.fit(X_train, y_train)
```

Out[19]:
```
▾ DecisionTreeClassifier i ?
(https://scikit-learn.org/1.7/modules/generated/

Parameters
```

In [ ]:
```

```

In [21]:
```
# Make predictions
y_pred = clf.predict(X_test)
```

In [ ]:
```
y_pred # model Predicted values
```

Out[ ]:
```
array([1, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 0, 0, 0, 1, 2, 1, 1, 2, 0, 2,
       0, 2, 2, 2, 2, 2, 0, 0])
```

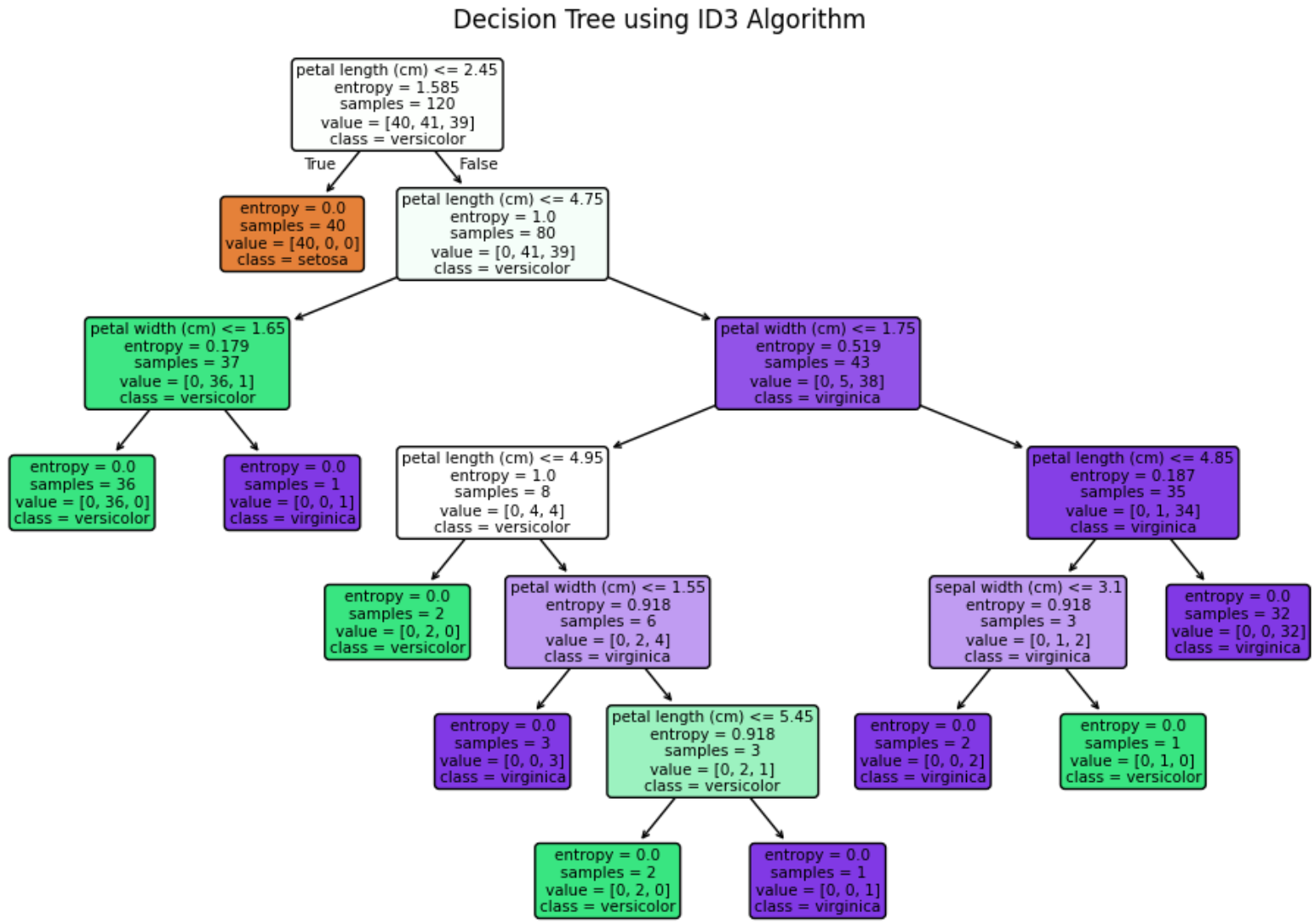
In [ ]:
```
y_test #Actual target values
```

Out[ ]:
```
array([1, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 0, 0, 0, 1, 2, 1, 1, 2, 0, 2,
       0, 2, 2, 2, 2, 2, 0, 0])
```

In [24]:
```
# Evaluate the model
accuracy = clf.score(X_test, y_test)
print(f"Accuracy: {accuracy:.2f}")
```

```
Accuracy: 1.00
```

In [25]:
```
# Display the decision tree
plt.figure(figsize=(12, 8))
plot_tree(clf, filled=True, feature_names=iris.feature_names, class_names=iris.target_names, rounded=True)
plt.title("Decision Tree using ID3 Algorithm")
plt.show()
```



In [26]:
```
# Print the tree structure
tree_structure = export_text(clf, feature_names=iris.feature_names)
print(tree_structure)
```

```
|--- petal length (cm) <= 2.45
|   |--- class: 0
|--- petal length (cm) >  2.45
|   |--- petal length (cm) <= 4.75
|   |   |--- petal width (cm) <= 1.65
|   |   |   |--- class: 1
|   |   |--- petal width (cm) >  1.65
|   |   |   |--- class: 2
|   |--- petal length (cm) >  4.75
|   |   |--- petal width (cm) <= 1.75
|   |   |   |--- petal length (cm) <= 4.95
|   |   |   |   |--- class: 1
|   |   |   |--- petal length (cm) >  4.95
|   |   |   |   |--- petal width (cm) <= 1.55
|   |   |   |   |   |--- class: 2
|   |   |   |   |--- petal width (cm) >  1.55
|   |   |   |   |   |--- petal length (cm) <= 5.45
|   |   |   |   |   |   |--- class: 1
|   |   |   |   |   |--- petal length (cm) >  5.45
|   |   |   |   |   |   |--- class: 2
|   |   |--- petal width (cm) >  1.75
|   |   |   |--- petal length (cm) <= 4.85
|   |   |   |   |--- sepal width (cm) <= 3.10
|   |   |   |   |   |--- class: 2
|   |   |   |   |--- sepal width (cm) >  3.10
|   |   |   |   |   |--- class: 1
|   |   |   |--- petal length (cm) >  4.85
|   |   |   |   |--- class: 2
```

In [ ]:
```
clf.predict([[4.5, 6.9, 6.7, 9.9]]) # to test the new sample
```

Out[ ]:
```
array([2])
```