Roll No: 205229133

Lab8. Animal Classification using Decision Trees

Objectives

In this lab, you will build ID3 and CART Decision Tree to classify whether an animal is a Mammal or Reptile.

Learning Outcomes

After completing this lab, you will be able to

- Create and import training dataset and test dataset
- Create ID3 Decision Tree using Entropy metric
- Create CART Decision Tree using Gini metric
- · Visualize graph using graphviz

Import necessary Library

In [1]:

```
import csv
import pandas as pd
from sklearn import tree
from sklearn.tree import export_graphviz
from sklearn.preprocessing import LabelEncoder
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score,classification_report
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report,auc,roc_c
import warnings
warnings.filterwarnings('ignore')
```

Step1. [Create Dataset]

Create the following dataset using Excel and save it as CSV file.

```
In [2]:
```

```
import pandas as pd
animal_data = pd.read_csv('animal.csv')
```

Step2. [Model building using ID3]

- Import your data set
- · Create DT model using 'entropy' criterion
- · Perform training and testing
- Print accuracy and classification report.
- · Interpret your results

· Visualize your DT model using graphviz

In [3]:

animal_data

Out[3]:

	Toothed	Hair	Breathes	Legs	Species
0	True	True	True	True	Mammal
1	True	True	True	True	Mammal
2	True	False	True	False	Repite
3	False	True	True	True	Mammal
4	True	True	True	True	Mammal
5	True	True	True	True	Mammal
6	True	False	False	False	Repite
7	True	False	True	False	Repite
8	True	True	True	True	Mammal
9	False	False	True	True	Repite

In [4]:

```
from sklearn.preprocessing import LabelEncoder
label_encoder = LabelEncoder()
animal_data["Label"] = label_encoder.fit_transform(animal_data["Species"])
animal_data
```

Out[4]:

	Toothed	Hair	Breathes	Legs	Species	Label
0	True	True	True	True	Mammal	0
1	True	True	True	True	Mammal	0
2	True	False	True	False	Repite	1
3	False	True	True	True	Mammal	0
4	True	True	True	True	Mammal	0
5	True	True	True	True	Mammal	0
6	True	False	False	False	Repite	1
7	True	False	True	False	Repite	1
8	True	True	True	True	Mammal	0
9	False	False	True	True	Repite	1

In [5]:

```
categories = list(label_encoder.inverse_transform([0,1]))
categories
```

Out[5]:

```
['Mammal', 'Repite']
```

```
In [6]:
```

```
X = animal_data.drop(['Label','Species'],axis=1)
```

In [7]:

```
y = animal_data.Label
```

In [8]:

```
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score,classification_report

X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.33,random_state=0)
clf = DecisionTreeClassifier(criterion='entropy',max_depth=4, random_state=42)
clf.fit(X_train,y_train)
```

Out[8]:

DecisionTreeClassifier(criterion='entropy', max_depth=4, random_state=42)

In [9]:

```
y_pred = clf.predict(X_test)
y_pred
```

Out[9]:

array([1, 0, 0, 0])

In [10]:

```
print("Accuracy of train :",clf.score(X_train,y_train))
print("Accuracy of test :",clf.score(X_test,y_test))
print('\n')
print(classification_report(y_test,y_pred))
```

Accuracy of train : 1.0 Accuracy of test : 0.75

		precision	recall	f1-score	support
	0	0.67	1.00	0.80	2
	1	1.00	0.50	0.67	2
accurac	СУ			0.75	4
macro av	/g	0.83	0.75	0.73	4
weighted av	/g	0.83	0.75	0.73	4

In [11]:

```
from sklearn import tree
from sklearn.tree import export_graphviz

with open("tree1.dot", 'w') as f:
    f = tree.export_graphviz(clf,out_file=f,max_depth = 4,impurity = False,feature_names =
```

Now open tree1.txt file which will be created in your working directory then Copy and paste the code to http://webgraphviz.com/ (http://webgraphviz.com/)

Another Way to visualize

In [12]:

```
from sklearn import tree
%matplotlib inline
tree.plot_tree(clf)
```

Out[12]:

```
[Text(167.4, 163.07999999999999, 'X[3] <= 0.5\nentropy = 0.918\nsamples = 6
\nvalue = [4, 2]'),
  Text(83.7, 54.360000000000014, 'entropy = 0.0\nsamples = 2\nvalue = [0,
2]'),
  Text(251.10000000000002, 54.36000000000014, 'entropy = 0.0\nsamples = 4\nv
alue = [4, 0]')]</pre>
```

```
X[3] <= 0.5
entropy = 0.918

samples = 6

value = [4, 2]
entropy = 0.0
samples = 2

value = [0, 2]
entropy = 0.0
samples = 4

value = [4, 0]
```

Step3. [Create a Test Set]

```
In [13]:
```

In [14]:

```
test_set = pd.read_csv('testing.csv')
test_set
```

Out[14]:

	Name	Toothed	Hair	Breathes	Legs	Species
0	Turtile	False	False	True	False	Reptile
1	Blue Whales	False	True	True	True	Mammal
2	Crocodile	True	False	True	True	Reptile

In [15]:

```
test_set["Label"] = label_encoder.fit_transform(test_set["Species"])
test_set
```

Out[15]:

	Name	Toothed	Hair	Breathes	Legs	Species	Label
0	Turtile	False	False	True	False	Reptile	1
1	Blue Whales	False	True	True	True	Mammal	0
2	Crocodile	True	False	True	True	Reptile	1

Step4. [Perform prediction]

• Use your ID3 DT model that you created before and predict labels for this test set. Check your predictions. Correct?

```
In [16]:
stp4 = test_set.drop(['Name','Species','Label'],axis=1)
y_prd = clf.predict(stp4)
y_prd
Out[16]:
array([1, 0, 0])
In [17]:
accuracy_score(test_set.Label,y_prd)
Out[17]:
0.66666666666666
Step5. [Build CART Decision Tree Model]

    Now, you are going to build a new CART decision tree using criterion='gini'.

    Train you model with full training data (No, train test split, this time)

  · Predict samples for the test file

    Visualize your CART DT using graphviz

In [18]:
clf_1 = DecisionTreeClassifier(criterion='gini',max_depth=4, random_state=42)
clf_1.fit(X,y)
```

```
clf_1.fit(X,y)
Out[18]:
DecisionTreeClassifier(max_depth=4, random_state=42)

In [19]:
clf_1.predict(stp4)
Out[19]:
array([1, 0, 1])
In [20]:
```

Now open tree2.txt file which will be created in your working directory then Copy and paste the code to http://webgraphviz.com/ (http://webgraphviz.com/ (http://webgraphviz.com/ (http://webgraphviz.com/)

f = tree.export_graphviz(clf_1,out_file=f,max_depth = 4,impurity = False,feature_names

Another Way to visualize

with open("tree2.txt", 'w') as f:

In [21]:

```
from sklearn import tree
tree.plot_tree(clf_1)
```

Out[21]:

```
[Text(167.4, 163.0799999999999, 'X[1] <= 0.5\ngini = 0.48\nsamples = 10\nvalue = [6, 4]'),  
Text(83.7, 54.36000000000014, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),  
Text(251.10000000000002, 54.36000000000014, 'gini = 0.0\nsamples = 6\nvalue = [6, 0]')]
```

Step6. [Buid DT with Zoo dataset]

- Download full animal dataset at https://archive.ics.uci.edu/ml/datasets/Zoo
 (https://archive.ics.uci.edu/ml/datasets/Zoo)
- Import, build model using ID3 and CART, train and test accuracy. Print classification report. Visualize your trees.

In [22]:

```
animal_2 = pd.read_csv("zoo.csv")
```

```
In [23]:
```

```
animal_2.head()
```

Out[23]:

	name	hair	feathers	eggs	milk	airborne	aquatic	predator	toothed	backbone	breath
0	aardvark	1	0	0	1	0	0	1	1	1	
1	antelope	1	0	0	1	0	0	0	1	1	
2	bass	0	0	1	0	0	1	1	1	1	
3	bear	1	0	0	1	0	0	1	1	1	
4	boar	1	0	0	1	0	0	1	1	1	
4											•

In [24]:

```
animal_2.shape
```

Out[24]:

(101, 18)

In [25]:

```
animal_2.info()
```

2 feathers 101 non-null int64 3 101 non-null int64 eggs 4 101 non-null milk int64 5 airborne 101 non-null int64 6 aquatic 101 non-null int64 7 101 non-null predator int64 8 toothed 101 non-null int64 9 backbone 101 non-null int64 10 breathes 101 non-null int64 11 venomous 101 non-null int64 12 fins 101 non-null int64 13 legs 101 non-null int64 14 tail 101 non-null int64 15 domestic 101 non-null int64 16 catsize 101 non-null int64

101 non-null

int64

dtypes: int64(17), object(1)
memory usage: 14.3+ KB

In [26]:

17

type

```
X1 = animal_2.drop(['name','type'],axis=1)
y1 = animal_2.type
```

```
In [27]:
```

```
X__train,X__test,y__train,y__test = train_test_split(X1,y1,test_size=0.33,random_state=0)
```

In [28]:

```
clf_2 = DecisionTreeClassifier(criterion='entropy',max_depth=3, random_state=52)
clf_2.fit(X__train,y__train)
```

Out[28]:

DecisionTreeClassifier(criterion='entropy', max_depth=3, random_state=52)

In [29]:

```
clf_2.predict(X__test)
```

Out[29]:

```
array([4, 4, 4, 1, 1, 1, 2, 4, 1, 1, 7, 1, 2, 7, 4, 6, 1, 7, 2, 4, 2, 4, 1, 2, 1, 1, 1, 2, 4, 4, 4, 4, 4, 1], dtype=int64)
```

In [30]:

```
clf_3 = DecisionTreeClassifier(criterion='gini',max_depth=4, random_state=42)
clf_3.fit(X__train,y__train)
```

Out[30]:

DecisionTreeClassifier(max_depth=4, random_state=42)

In [31]:

```
y_pred=clf_3.predict(X__test)
y_pred
```

Out[31]:

```
array([7, 4, 4, 1, 1, 1, 2, 4, 1, 1, 7, 1, 2, 7, 4, 6, 1, 7, 2, 4, 2, 7, 1, 2, 1, 1, 1, 2, 4, 7, 4, 7, 7, 1], dtype=int64)
```

In [32]:

```
from sklearn.preprocessing import StandardScaler
scale = StandardScaler()
ss = scale.fit_transform(X__train)
ss1 = scale.transform(X__test)

print("model accuracy :",accuracy_score(y__test,y_pred))
print("Train accuracy :",clf_3.score(ss,y__train))
print("Test accuracy :",clf_3.score(ss1,y__test))
```

model accuracy : 0.8235294117647058 Train accuracy : 0.9253731343283582 Test accuracy : 0.8235294117647058

```
In [33]:
```

```
animal_2.type.value_counts(dropna=False)
Out[33]:
1
     41
     20
2
4
     13
7
     10
6
      8
3
      5
5
Name: type, dtype: int64
```

In [34]:

```
with open("tree3.txt", 'w') as f:
    f = tree.export_graphviz(clf_3,out_file=f,max_depth = 16,impurity = False,feature_names
```

In [35]:

```
print(classification_report(y__test,y_pred))
```

	precision	recall	f1-score	support
1	1.00	1.00	1.00	12
2	1.00	1.00	1.00	6
3	0.00	0.00	0.00	4
4	1.00	1.00	1.00	7
5	0.00	0.00	0.00	1
6	1.00	0.50	0.67	2
7	0.25	1.00	0.40	2
accuracy			0.82	34
macro avg	0.61	0.64	0.58	34
weighted avg	0.81	0.82	0.80	34

Now open tree3.txt file which will be created in your working directory then Copy and paste the code to http://webgraphviz.com/ (http://webgraphviz.com/)

In [36]:

```
from sklearn import tree
tree.plot_tree(clf_3)
```

Out[36]:

```
[Text(239.14285714285714, 195.696, 'X[3] <= 0.5\ngini = 0.736\nsamples = 67
\nvalue = [29, 14, 1, 6, 3, 6, 8]'),
  Text(191.31428571428572, 152.208, 'X[1] <= 0.5\ngini = 0.763\nsamples = 38
\nvalue = [0, 14, 1, 6, 3, 6, 8]'),
  Text(143.4857142857143, 108.72, 'X[11] \leftarrow 0.5 \neq 0.747 = 0.747 = 24 = 24 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.747 = 0.7
value = [0, 0, 1, 6, 3, 6, 8]'),
  Text(95.65714285714286, 65.232, 'X[4] \leftarrow 0.5 \cdot ngini = 0.66 \cdot nsamples = 18 \cdot nva
lue = [0, 0, 1, 0, 3, 6, 8]'),
  Text(47.82857142857143, 21.744, 'gini = 0.556\nsamples = 13\nvalue = [0, 0,
1, 0, 3, 1, 8]'),
  Text(143.4857142857143, 21.744, 'gini = 0.0 \nsamples = 5 \nvalue = [0, 0, 0, 0]
0, 0, 5, 0]'),
  Text(191.31428571428572, 65.232, 'gini = 0.0\nsamples = 6\nvalue = [0, 0,
0, 6, 0, 0, 0]'),
  Text(239.14285714285714, 108.72, 'gini = 0.0\nsamples = 14\nvalue = [0, 14,
0, 0, 0, 0, 0]'),
  Text(286.9714285714286, 152.208, 'gini = 0.0\nsamples = 29\nvalue = [29, 0,
0, 0, 0, 0, 0]')]
```

