



RĪGAS TEHNISKĀ UNIVERSITĀTE

Riga Technical University

Telecommunications Software (RAE411).

Sixth Practical Exercise.

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I. Introduction:

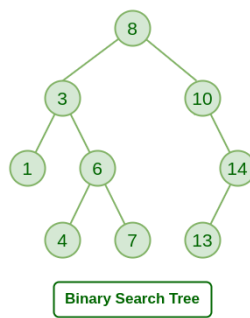
This report is the documentation for the sixth practical exercise. This exercise is divided into two sections: Binary search tree and SDN traffic classification with DT. Python is used to implement both sections using Anaconda.

II. Binary Search Tree:

Binary search tree is a node-based sorted data structure at which:

1. The left subtree of a node contains only nodes with lower keys' values than the parents' values.
2. The right subtree of a node contains only nodes with higher keys' values than the parents' values.
3. Each node has only a maximum value of two children.
4. There must be no duplicates for values in a tree.

Binary search trees are implemented to be used in the sorting algorithms. It is the basic data structure used in Microsoft Excel and Spreadsheets. Binary search trees have three main functions which are: searching, insertion, and deletion. There are also three ways to traverse through a binary tree which are: in-order, pre-order, and post-order traversing.



A. Node creation:

Nodes are the main building unit for a binary tree. A node has three properties: its value, its right child, and its left child. A class in python is done by the name of Node to create a node whenever it is needed in a tree. This class has a constructor which holds the data for the node, its left and right children.

```
class Node:
    def __init__(self, data):    #constructor to initialize the class whenever an object is created from it
        self.data=data         #self is the object created, data is the value for the node.
        self.lchild=None       #self is automatically passed to the function without writing it in the function init
        self.rchild=None
```

B. Tree creation:

Binary tree is mainly multiple nodes connected to each other. A class is created in python to create a tree from a given list. The constructor of this class mainly takes the first element in the tree as the root node and insert the other elements in the with the insertion function following the rules of the binary tree data structure.

```
class BinarySearchTree:
    def __init__(self, node_list):
        self.root=Node(node_list[0]) #starting with the first value in the list as the root
        for data in node_list[1:]:    #insertion of all of the other elements except the root which is already inserted
            self.insert(data)
```

C. Search function:

This function enables the user to insert a value for a node in the tree and return whether this value is found or not. It has 3 inputs: the tree root, the parent, and the value needed to be found. It is a recursive function that check whether the value is found in the tree or not.

```
def search(self, node, parent, data):
    if node is None:                #if node, not found
        return False, node, parent
    if node.data == data:           #if value of root is equal to data, the search value is the root
        return True, node, parent
    if node.data > data:            #if value of root is greater than data, child must be on left
        return self.search(node.lchild, node, data)
    else:                          #if value of root is smaller than data, child must be on right
        return self.search(node.rchild, node, data)
```

D. Insertion function:

This function starts by making sure that the value inserted is not already in the tree to avoid duplicates. It is a recursive function that operates on the idea of creating a node to the value inserted. Then, the value of this node is then compared to its root. If the value is bigger than the root's value, then the function recurs in the right direction. If it is smaller, then it recurs in the left direction.

```
def insert(self, data):
    flag, n, p = self.search(self.root, self.root, data) #making sure value is in the tree
    if not flag: #if node not found, create a new node with value given
        new_node = Node(data)
        if data > p.data: #if node value bigger than parent
            p.rchild = new_node #put it on right
        else: #if smaller, put child on left of parent
            p.lchild = new_node
```

E. Deletion function:

Deletion function has several scenarios and it depends whether the deleted node is a parent for one child or for two children. It makes sure at first that the needed node is present in the tree.

```
def delete(self, root, data):
    flag, n, p = self.search(root, root, data)
    if flag is False:
        print("No key value found")
    else:
        if n.lchild is None: #parent for one child only
            if n==p.lchild:
                p.lchild=n.rchild
            else:
                p.rchild=n.rchild
            del p
        elif n.rchild is None:
            if n==p.lchild:
                p.lchild=n.lchild
            else:
                p.rchild=n.lchild
            del p
        else: #parent of two children
            pre=n.rchild
            if pre.lchild is None:
                n.data=pre.data
                n.rchild=pre.rchild
                del pre
            else:
                next=pre.lchild
                while next.lchild is not None:
                    pre=next
                    next=next.lchild
                n.data=next.data
                pre.lchild=next.rchild
            del n
```

F. Traversing functions:

Traversing functions are used with the search trees to have an algorithm on how we extract the whole nodes in a tree. It is the process of visiting each node in the tree exactly once.

a. Preorder traversing:

This algorithm works by printing: the root, then printing the whole left subtree until reaching to the leaves, then traversing in the right subtree.

```
def preorder(self, node):  
    if node is not None:  
        print(node.data),  
        self.preorder(node.lchild)  
        self.preorder(node.rchild)
```

b. Inorder traversing:

This algorithm works by printing: the left node first, then the parent, then the right node.

```
def inorder(self, node):  
    if node is not None:  
        self.inorder(node.lchild)  
        print(node.data),  
        self.inorder(node.rchild)
```

c. Postorder traversing:

This algorithm works by printing: the left node, then the right node, then the parent.

```
def postorder(self, node):  
    if node is not None:  
        self.preorder(node.lchild)  
        self.preorder(node.rchild)  
        print(node.data),
```

G. Implementation on given lists:

In this report, implementation of all of the three traversal method, deletion of element and searching for them is applied for each list.

a. List a:

List a is given as: [49, 38, 65, 97, 60, 76, 13, 27, 5, 1]

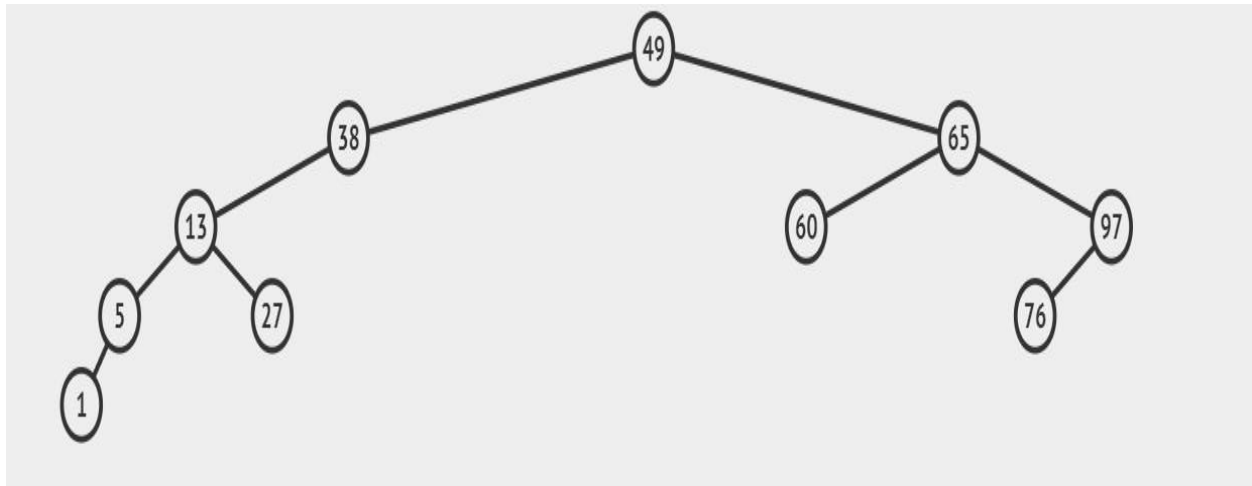


Figure 1 Original Tree for List a

```
a = [49, 38, 65, 97, 60, 100, 13, 27, 5, 1]
tree=BinarySearchTree(a)
print('preorder:')
tree.preorder(tree.root)
print()
print(tree.search(tree.root, tree.root, 13))
print('\ninorder:')
tree.inorder(tree.root)
print('\npostorder:')
tree.postorder(tree.root)
tree.delete(tree.root,65)
print('\npreorder after element deletion:')
tree.preorder(tree.root)
print()
print(tree.search(tree.root, tree.root, 65))
```

Figure 2 Code for operations on List a

```

preorder:
49
38
13
5
1
27
65
60
97
100

(True, <__main__.Node object at 0x000001E4D03FFFA0>, <__main__.Node object at 0x000001E4D03FF6A0>)

inorder:
1
5
13
27
38
49
60
65
97
100

postorder:
38
13
5
1
27
65
60
97
100
49

preorder after element deletion:
49
38
13
5
1
27
97
60
100

(False, None, <__main__.Node object at 0x000001E4D03FF1C0>)

```

Figure 3 Output for list a

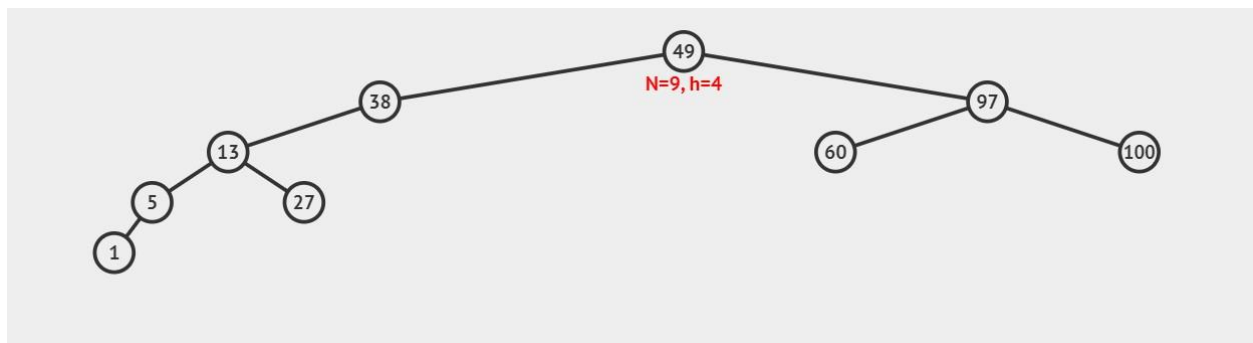


Figure 4 Tree for list a after element deletion

b. List b:

According to the given list: [149, 38, 65, 197, 60, 176, 13, 217, 5, 11].

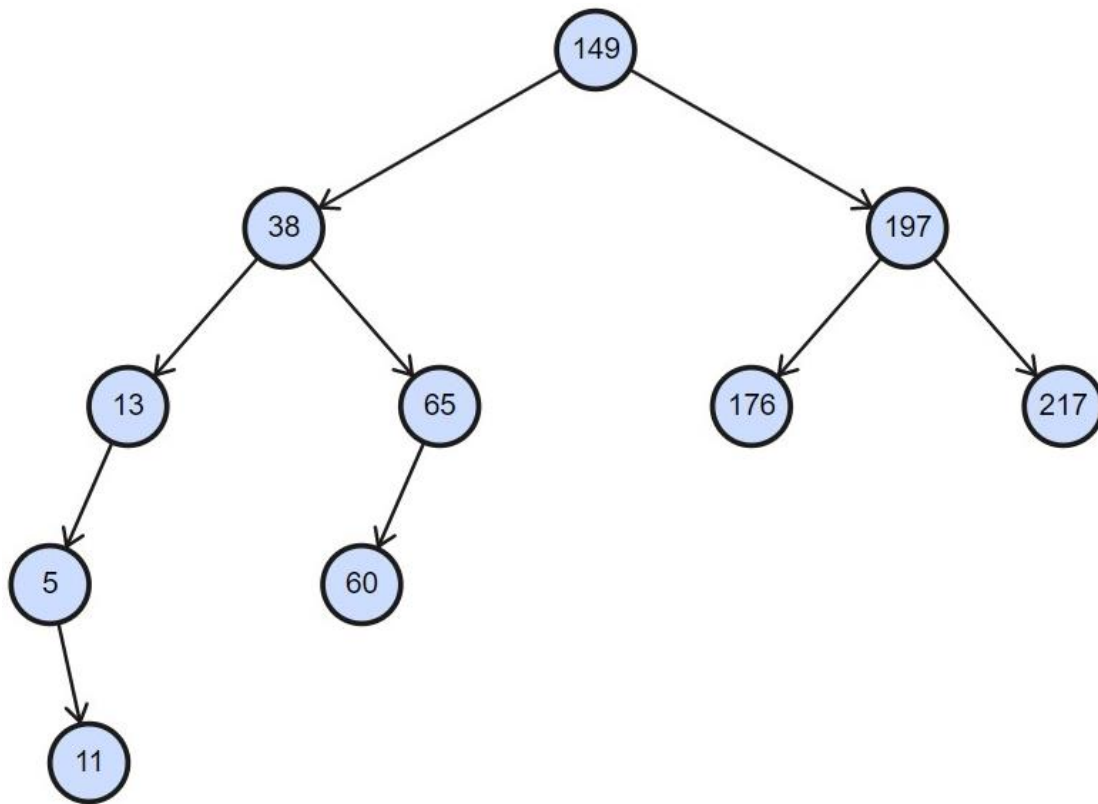


Figure 5 Original Tree for List b

```
b = [149, 38, 65, 197, 60, 176, 13, 217, 5, 11]
tree=BinarySearchTree(b)
print('preorder:')
tree.preorder(tree.root)
print()
print(tree.search(tree.root, tree.root, 217))
print('\ninorder:')
tree.inorder(tree.root)
print('\npostorder:')
tree.postorder(tree.root)
tree.delete(tree.root,197)
print('\npreorder after element deletion:')
tree.preorder(tree.root)
print()
print(tree.search(tree.root, tree.root, 197))
```

Figure 6 Code for operations on List b


```

preorder:
149
38
13
5
11
65
60
197
176
217

(True, <__main__.Node object at 0x000001E4D03F2220>, <__main__.Node object at 0x000001E4D03F21F0>)

inorder:
5
11
13
38
60
65
149
176
197
217

postorder:
38
13
5
11
65
60
197
176
217
149

preorder after element deletion:
149
38
13
5
11
65
60
217
176

(False, None, <__main__.Node object at 0x000001E4D03F2820>)

```

Figure 7 Output for list b

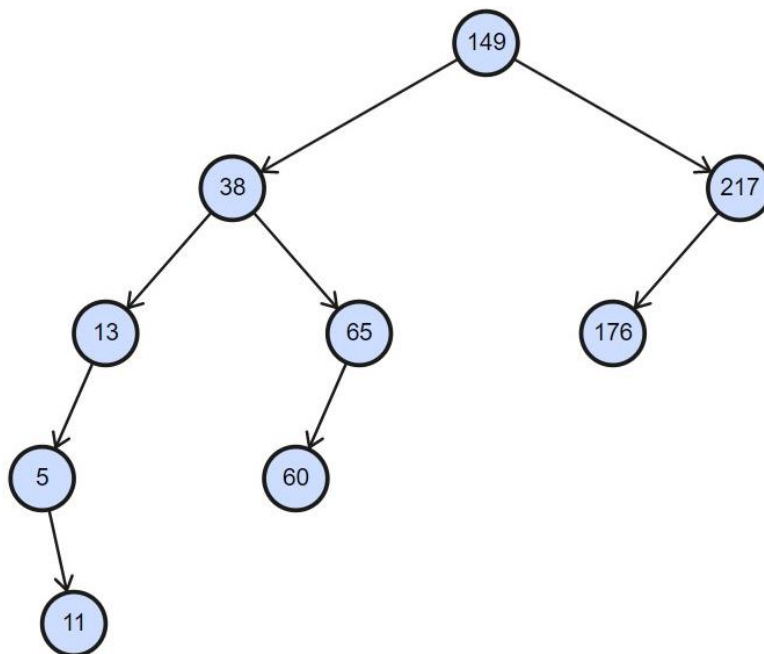


Figure 8 Tree for list b after element deletion

c. List c:

According to the given list: [49, 38, 65, 97, 64, 76, 13, 77, 5, 1, 55, 50].

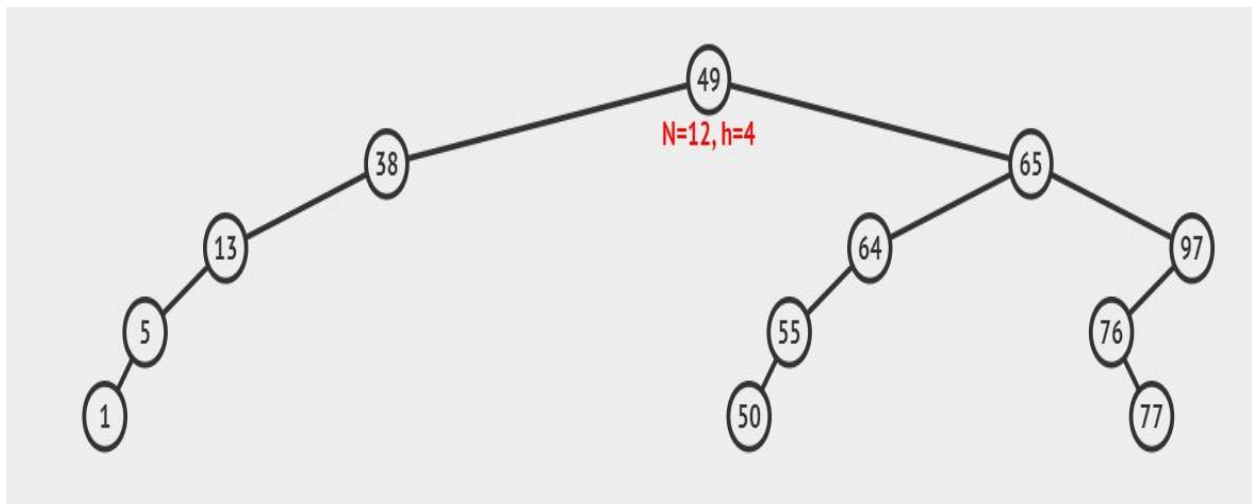


Figure 9 Original Tree for List c

```
c = [49, 38, 65, 97, 64, 76, 13, 77, 5, 1, 55, 50]
tree=BinarySearchTree(c)
print('preorder:')
tree.preorder(tree.root)
print()
print(tree.search(tree.root, tree.root, 77))
print('\ninorder:')
tree.inorder(tree.root)
print('\npostorder:')
tree.postorder(tree.root)
tree.delete(tree.root,38)
print('\npreorder after element deletion:')
tree.preorder(tree.root)
print()
print(tree.search(tree.root, tree.root, 38))
```

Figure 10 Code for operations on List c

```

preorder:
49
38
13
5
1
65
64
55
50
97
76
77

(True, <__main__.Node object at 0x000001E4D03FF970>, <__main__.Node object at 0x000001E4D03FF0D0>)

inorder:
1
5
13
38
49
50
55
64
65
76
77
97

postorder:
38
13
5
1
65
64
55
50
97
76
77
49

preorder after element deletion:
49
13
5
1
65
64
55
50
97
76
77

(False, None, <__main__.Node object at 0x000001E4D03FF730>)

```

Figure 11 Output for list c

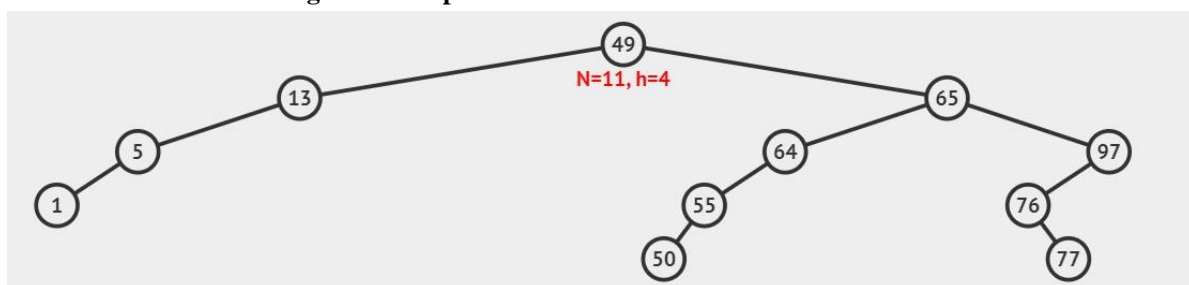


Figure 12 Tree for list c after element deletion

III. SDN classification with decision tree:

Decision tree builds classification or regression models in the form of a tree structure. It breaks down a dataset into smaller and smaller subsets while at the same time an associated decision tree is incrementally developed. The final result is a tree with decision nodes and leaf nodes.

A. Program Code:

This code is used to perform some data analysis and classification tasks using a decision tree classifier on a dataset stored in a CSV file.

```
import pandas as pd
import numpy as np
import scipy.stats as stats
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import confusion_matrix, recall_score, precision_score, f1_score, classification_report
from sklearn.model_selection import cross_val_score, KFold
import matplotlib.pyplot as plt
import seaborn as sn

# importing csv file from location
dataset = pd.read_csv("C:/Users/Zeyad Mohamed/Downloads/assignments-main (1)/assignments-main/SDN_traffic.csv")

print(dataset.head())
print(dataset.info())
print(dataset.describe())
print(dataset.duplicated())

#dataset needed for the analysis in csv file
x = dataset[['forward_bps_var',
            "tp.sre", "tp.dst", "nw_proto",
            "forward_pe", "forward_bc", "forward_pl",
            "forward_piat", "forward_pps", "forward_bps", "forward_pl_mean",
            "forward_piat_mean", "forward_pps_mean", "forward_bps_mean", "forward_pl_var", "forward_piat_var",
            "forward_pps_var", "forward_pl_q1", "forward_pl_q3",
            "forward_piat_q1", "forward_piat_q3", "forward_pl_max", "forward_pl_min",
            "forward_piat_max", "forward_piat_min", "forward_pps_max", "forward_pps_min",
            "forward_bps_max", "forward_bps_min", "forward_duration", "forward_size_packets",
            "forward_size_bytes", "reverse_pc", "reverse_bc", "reverse_pl", "reverse_piat", "reverse_pps",
            "reverse_bps", "reverse_pl_mean", "reverse_piat_mean", "reverse_pps_mean", "reverse_bps_mean", "reverse_pl_var",
            "reverse_piat_var", "reverse_pps_var", "reverse_bps_var",
            "reverse_piat_q1", "reverse_pl_q3", "reverse_piat_max", "reverse_piat_min", "reverse_pps_max", "reverse_pps_min",
            "reverse_piat_q3", "reverse_pl_max", "reverse_piat_max", "reverse_bps_min", "reverse_duration",
            "reverse_size_packets", "reverse_size_bytes"]]

x.loc[1877, 'forward_bps_var'] = float(11968865203349)
x.loc[9131, 'forward_bps_var'] = float(12880593884833)
x.loc[2381, 'forward_bps_var'] = float(39987497172945)
x.loc[2562, 'forward_bps_var'] = float(663388742992)
x.loc[1931, 'forward_bps_var'] = float(37770223877794)
x.loc[2078, 'forward_bps_var'] = float(9822747730895)
x.loc[2567, 'forward_bps_var'] = float(37778223877794)
x.loc[2586, 'forward_bps_var'] = float(97227875883751)
x.loc[2754, 'forward_bps_var'] = float(18789751483737)
x.loc[2765, 'forward_bps_var'] = float(33969277035759)
x.loc[2984, 'forward_bps_var'] = float(39284786962856)
x.loc[3844, 'forward_bps_var'] = float(9169996863653)
x.loc[3349, 'forward_bps_var'] = float(37123283690575)
x.loc[3507, 'forward_bps_var'] = float(61019864598464)
x.loc[3610, 'forward_bps_var'] = float(46849628984872)
x.loc[3717, 'forward_bps_var'] = float(97158873841506)
x.loc[3845, 'forward_bps_var'] = float(11968865203349)
x.loc[3868, 'forward_bps_var'] = float(85874278395372)

X = pd.DataFrame(x)
X["forward_bps_var"] = pd.to_numeric(X["forward_bps_var"])
print(X.info())

Y = dataset[["category"]]
Y = Y.to_numpy()
Y = Y.ravel()
Labels, uniques = pd.factorize(Y)
Y = Labels
Y = Y.ravel()

X = stats.zscore(X)
X = np.nan_to_num(X)

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, random_state=0, test_size=0.3)
```

```

clf = DecisionTreeClassifier(random_state=0, max_depth=2)
clf.fit(X_train, Y_train)

cv = KFold(n_splits=10, random_state=0, shuffle=True)
accuracy = clf.score(X_test, Y_test)
KFold10_accuracy = cross_val_score(clf, X_train, Y_train, scoring='accuracy', cv=cv, n_jobs=-1)
print(KFold10_accuracy.mean())

predict = clf.predict(X_test)
cm = confusion_matrix(Y_test, predict)
precision = precision_score(Y_test, predict, average='weighted', labels=np.unique(predict))
recall = recall_score(Y_test, predict, average='weighted', labels=np.unique(predict))
f1scoreMacro = f1_score(Y_test, predict, average='macro', labels=np.unique(predict))
print(classification_report(Y_test, predict, target_names=uniques))

importance = clf.feature_importances_
important_features_dict = {}
for idx, val in enumerate(importance):
    important_features_dict[idx] = val
important_features_list = sorted(important_features_dict,
                                key=important_features_dict.get,
                                reverse=True)
print(f'10 most important features: {important_features_list[:10]}')

fn = ['forward_bps_var',
      "tp.src", "tp.dst", "nw_proto",
      "forward_pe", "forward_bc", "forward_pl",
      "forward_piat", "forward_pps", "forward_bps", "forward_pl_mean",
      "forward_piat_mean", "forward_pps_mean", "forward_bps_mean", "forward_pl_var", "forward_piat_var",
      "forward_pps_var", "forward_pl_q1", "forward_pl_q3",
      "forward_piat_q1", "forward_piat_q3", "forward_pl_max", "forward_pl_min",
      "forward_plat_max", "forward_piat_win", "forward_pps_max", "forward_pps_min",
      "forward_bps_max", "forward_bps_min", "forward_duration", "forward_size_packets",
      "forward_size_bytes", "reverse_pc", "reverse_bc", "reverse_pl", "reverse_piat", "reverse_PRS",
      "reverse_bps", "reverse_pl_mean", "reverse_piat_mean", "reverse_pps_mean", "reverse_bps_mean", "reverse_pl_var",
      "reverse_plin", "reverse_pl", "reverse_piat", "reverse_piat_var", "reverse_pps_var", "reverse_bps_var",
      "reverse_piat_q1", "reverse_pl_q3", "reverse_piat_max", "reverse_piat_min", "reverse_pps_max", "reverse_pps_min",
      "reverse_piat_q3", "reverse_pl_max", "reverse_bps_max", "reverse_bps_min", "reverse_duration", "reverse_size_packets",
      "reverse_size_bytes"]

la = ['WWW', 'DNS', 'FTP', 'ICMP', 'P2P', 'VOIP']

plt.figure(1, dpi=300)
fig = tree.plot_tree(clf, filled=True, feature_names=fn, class_names=la)
plt.title("Decision tree trained on all the features")
plt.show()

import seaborn as sn
import matplotlib.pyplot as plt

labels = uniques
plt.figure(2, figsize=(5, 2))
plt.title("Confusion Matrix", fontsize=10)
cm_new = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
sn.heatmap(cm_new, annot=True, cmap="YlGnBu", fmt=".2f", xticklabels=labels, yticklabels=labels)
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.xticks(rotation=45)
plt.yticks(rotation=45)
plt.show()

```


B. Code Output:

```

      id_flow      nw_src  tp_src      nw_dst  \
0  b2bb77a570fcfa9325eb9e51b6116d2a  172.16.25.104  41402  34.107.221.82
1  f07977b0d1d6645c4fe1e9efea080ff3  172.16.25.104  41406  34.107.221.82
2  e4026ba9b6c1957516e92bdd0d04878f  172.16.25.104  38232  52.84.77.43
3  e2d747932e41500b1463fe8ae4299ecb  172.16.25.104  38234  52.84.77.43
4  56325703391225ad65e013e7a2b02fac  172.16.25.104  60166  52.32.34.32

      tp_dst  nw_proto  forward_pc  forward_bc  forward_pl  forward_piat  ...  \
0         80         6         5         300         60.00         6.0  ...
1         80         6         5         300         60.00         6.0  ...
2        443         6         3         198         66.00        10.0  ...
3        443         6         3         198         66.00        10.0  ...
4        443         6         4         265         66.25         7.5  ...

      reverse_piat_max  reverse_piat_min  reverse_pps_max  reverse_pps_min  \
0         10.333333         6.00         0.166667         0.096774
1         10.000000         6.20         0.161290         0.100000
2         10.333333         10.00         0.100000         0.096774
3         10.333333         10.00         0.100000         0.096774
4          7.750000          7.75         0.129032         0.129032

      reverse_bps_max  reverse_bps_min  reverse_duration  reverse_size_packets  \
0         15.133333         5.806452         121         15
1         15.133333         6.000000         121         15
2          6.000000         5.806452          91          9
3          6.000000         5.806452          91          9
4          8.548387         8.548387          31          4

      reverse_size_bytes  category
0          1114      WWW
1          1114      WWW
2           540      WWW
3           540      WWW
4           265      WWW

#   Column      Non-Null Count  Dtype
---  -
0   id_flow      4234 non-null    object
1   nw_src       4234 non-null    object
2   tp_src       4234 non-null    int64
3   nw_dst       4234 non-null    object
4   tp_dst       4234 non-null    int64
5   nw_proto     4234 non-null    int64
6   forward_pc   4234 non-null    int64
7   forward_bc   4234 non-null    int64
8   forward_pl   4234 non-null    float64
9   forward_piat 4234 non-null    float64
10  forward_pps  4234 non-null    float64
11  forward_bps  4234 non-null    float64
12  forward_pl_mean 4234 non-null    float64
13  forward_piat_mean 4234 non-null    float64
14  forward_pps_mean 4234 non-null    float64
15  forward_bps_mean 4234 non-null    float64
16  forward_pl_var 4234 non-null    float64
17  forward_piat_var 4234 non-null    float64
18  forward_pps_var 4234 non-null    float64
19  forward_bps_var 4234 non-null    object
20  forward_pl_q1  4234 non-null    float64
21  forward_pl_q3  4234 non-null    float64
22  forward_piat_q1 4234 non-null    float64
23  forward_piat_q3 4234 non-null    float64
24  forward_pl_max 4234 non-null    float64
25  forward_pl_min 4234 non-null    float64
26  forward_piat_max 4234 non-null    float64
27  forward_piat_min 4234 non-null    float64
28  forward_pps_max 4234 non-null    float64
29  forward_pps_min 4234 non-null    float64
30  forward_bps_max 4234 non-null    float64
31  forward_bps_min 4234 non-null    float64
32  forward_duration 4234 non-null    int64
33  forward_size_packets 4234 non-null    int64

```

34	forward_size_bytes	4234	non-null	int64
35	reverse_pc	4234	non-null	int64
36	reverse_bc	4234	non-null	float64
37	reverse_pl	4234	non-null	float64
38	reverse_piat	4234	non-null	float64
39	reverse_pps	4234	non-null	float64
40	reverse_bps	4234	non-null	float64
41	reverse_pl_mean	4234	non-null	float64
42	reverse_piat_mean	4234	non-null	float64
43	reverse_pps_mean	4234	non-null	float64
44	reverse_bps_mean	4234	non-null	float64
45	reverse_pl_var	4234	non-null	float64
46	reverse_piat_var	4234	non-null	float64
47	reverse_pps_var	4234	non-null	float64
48	reverse_bps_var	4234	non-null	float64
49	reverse_pl_q1	4234	non-null	float64
50	reverse_pl_q3	4234	non-null	float64
51	reverse_piat_q1	4234	non-null	float64
52	reverse_piat_q3	4234	non-null	float64
53	reverse_pl_max	4234	non-null	float64
54	reverse_pl_min	4234	non-null	float64
55	reverse_piat_max	4234	non-null	float64
56	reverse_piat_min	4234	non-null	float64
57	reverse_pps_max	4234	non-null	float64
58	reverse_pps_min	4234	non-null	float64
59	reverse_bps_max	4234	non-null	float64
60	reverse_bps_min	4234	non-null	float64
61	reverse_duration	4234	non-null	int64
62	reverse_size_packets	4234	non-null	int64
63	reverse_size_bytes	4234	non-null	int64
64	category	4234	non-null	object

dtypes: float64(48), int64(12), object(5)

memory usage: 2.1+ MB

None

	tp_src	tp_dst	nw_proto	forward_pc	forward_bc	\
count	4234.000000	4234.000000	4234.000000	4234.000000	4.234000e+03	
mean	39994.956542	8540.046528	6.660132	3835.848370	7.356521e+06	
std	17331.881734	17575.486397	3.815368	18375.794566	3.585172e+07	
min	0.000000	0.000000	1.000000	0.000000	0.000000e+00	
25%	35248.500000	80.000000	6.000000	2.000000	1.200000e+02	
50%	44009.000000	443.000000	6.000000	3.000000	1.980000e+02	
75%	52130.250000	443.000000	6.000000	6.000000	3.850000e+02	
max	65534.000000	60949.000000	17.000000	181104.000000	3.558093e+08	

	forward_pl	forward_piat	forward_pps	forward_bps	\
count	4234.000000	4234.000000	4.234000e+03	4.234000e+03	
mean	316.336560	15.261581	4.788105e+02	2.576202e+06	
std	3732.045349	182.065520	2.021312e+04	1.200390e+08	
min	0.000000	0.000000	0.000000e+00	0.000000e+00	
25%	60.000000	0.048051	6.451613e-02	4.000000e+00	
50%	66.000000	3.500000	1.666667e-01	1.260000e+01	
75%	79.811688	7.500000	5.161290e-01	4.600000e+01	
max	154375.000000	4125.000000	1.303625e+06	7.422774e+09	

	forward_pl_mean	...	reverse_pl_min	reverse_piat_max	\
count	4234.000000	...	4234.000000	4234.000000	
mean	1582.814224	...	54.418871	23.652912	
std	9644.341190	...	269.495303	229.416470	
min	0.000000	...	0.000000	0.000000	
25%	43.000000	...	0.000938	0.000433	
50%	61.250000	...	15.500000	7.500000	
75%	98.000000	...	60.000000	15.500000	
max	162975.000000	...	5573.208202	4125.000000	

	reverse_piat_min	reverse_pps_max	reverse_pps_min	reverse_bps_max	\
count	4.234000e+03	4.234000e+03	4.234000e+03	4.234000e+03	
mean	5.189081e+02	1.263424e+03	6.683260e+04	1.270755e+05	
std	2.792340e+04	4.689801e+04	2.674774e+06	4.139731e+06	
min	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	
25%	3.225807e-02	3.030303e-02	3.125000e-02	1.935484e+00	
50%	6.559140e-01	9.677419e-02	1.000000e-01	6.026316e+00	
75%	8.500000e+00	2.903226e-01	2.325000e+01	4.167742e+01	
max	1.816375e+06	2.316875e+06	1.556534e+08	1.707531e+08	

	reverse_bps_min	reverse_duration	reverse_size_packets	\
count	4.234000e+03	4234.000000	4.234000e+03	
mean	6.747949e+04	3224.000000	2.750047e+05	
std	3.034402e+06	20429.627234	1.519335e+06	
min	0.000000e+00	0.000000	0.000000e+00	
25%	2.242424e+00	5.000000	2.000000e+00	
50%	9.258333e+00	30.000000	1.800000e+01	
75%	4.900000e+01	60.000000	6.860000e+02	
max	1.488506e+08	232137.000000	1.717689e+07	

	reverse_size_bytes
count	4.234000e+03
mean	2.592156e+05
std	2.875554e+06
min	0.000000e+00
25%	0.000000e+00
50%	0.000000e+00
75%	2.460000e+02
max	1.214242e+08


```
[8 rows x 60 columns]
0      False
1      False
2      False
3      False
4      False
...
4229   False
4230   False
4231   False
4232   False
4233   False
Length: 4234, dtype: bool
```

IV. GitHub Link for the whole project:

For the link:

<https://github.com/ZeyadNashaat/Telecommunications-Software-RAE411-/tree/main/Sixth%20practical%20exercise>