

21/01/2024

**Telecommunication Software**  
**Report 5**

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You will find the code from the Task 1 and 2 starting page 3. As always, all can also be found on the github repository : <https://github.com/Niennaaa/TelecommunicationSoftware>

### **Task 1 : Intrusion detection**

You can find the ML algorithm used the Annex and in the PDF "TASK1&2". The notebook containing the corresponding code is the file "TASK1&2.ipynb".

For this task I downloaded the files in the dataset shown in class and reproduced the Machine Learning code used to analyse it. I had to tweak the code a little as some of the code didn't work on my end.

### **Task 2 : Binary Search Tree**

You can find the code used to build the required class in the Annex and in the PDF "TASK1&2". The notebook containing the corresponding code is the file "TASK1&2.ipynb".

The code works, as shown by the tests used at the end, but I believe the code used to delete a node within a tree could be improved. I chose to replace the value to remove by the next greater numerical value within the tree as, this way, the property of the tree would be conserved. However, this led me to just erase the given tree and build a new one in the new order of value calculated by my function. So, I believe this function can definitely be improved.

### **Task 3 : SDN Traffic classification with DT**

Unfortunately, I didn't have the time to complete this task, but I still analysed and compared the ID3 and the CART algorithm. ID3 and CART are two different decision tree algorithms that can be used for classification. They work by learning simple decision rules inferred from the data features and splitting the data into smaller and smaller subsets based on the feature values.

The ID3 algorithm only works for binary classification and base its splitting of the data on the difference of value between a node and its parent. The CART algorithm however is a Classification And Regression Tree and is not limited to binary classification. Its splitting decision is based on the goal to minimize the Gini Impurity which measures how often an element would be incorrectly labelled if labelled randomly inside the set.

I found this published article : [\(PDF\) Performance Evaluation Among ID3, C4.5, and CART Decision Tree Algorithm \(researchgate.net\)](#) comparing the performances of these algorithms, and found that the CART algorithm seems to be better in every way to the ID3 algorithm : it is faster, more accurate and can handle situations such as missing values inside the set.

I plan on coding these algorithms and test their performances by myself, but I will not have the time to do it before the end of the semester. I might add my results to my github later this year, separately from this class.

### **Annex:**

```
In [1]: #TASK 1 : INTRUSION DETECTION
# /\ I used the code SEEN IN THE LECTURE for this task
# /\ This is NOT original content
# The code was tweaked a little as it wouldn't run right on my side but this is it.

#Necessary imports
!pip install --upgrade pandas --user

import pandas as pd
import numpy as np
import matplotlib
import matplotlib.pyplot as plt
from sklearn.manifold import TSNE
matplotlib.use('TkAgg')

#path = 'C:\\Users\\Jehanne\\OneDrive - De Vinci\\RTU\\TelecomSoftware\\labeled_flow
path = "C:\\Users\\Jehanne\\OneDrive - De Vinci\\RTU\\TelecomSoftware\\labeled_flow
df = pd.read_xml(path)

print(df.info)
```

Requirement already satisfied: pandas in c:\users\jehanne\appdata\roaming\python\python38\site-packages (2.0.3)

Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\jehanne\appdata\roaming\python\python38\site-packages (from pandas) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in c:\users\jehanne\anaconda3\lib\site-packages (from pandas) (2023.3.post1)

Requirement already satisfied: tzdata>=2022.1 in c:\users\jehanne\anaconda3\lib\site-packages (from pandas) (2022.1)

Requirement already satisfied: numpy>=1.20.3 in c:\users\jehanne\appdata\roaming\python\python38\site-packages (from pandas) (1.24.4)

Requirement already satisfied: six>=1.5 in c:\users\jehanne\anaconda3\lib\site-packages (from python-dateutil>=2.8.2->pandas) (1.16.0)

WARNING: Ignoring invalid distribution -umpy (c:\users\jehanne\anaconda3\lib\site-packages)

WARNING: Ignoring invalid distribution -umpy (c:\users\jehanne\anaconda3\lib\site-packages)

<bound method DataFrame.info of destinationBytes \		appName	totalSourceBytes	totalDe
0	HTTPWeb		38007	1273547
1	HTTPWeb		51524	1705876
2	DNS		2845	18948
3	HTTPWeb		4291	92920
4	HTTPWeb		4540	113303
...	...		...	...
142366	DNS		104	348
142367	DNS		93	463
142368	HTTPImageTransfer		251	66
142369	DNS		87	526
142370	HTTPImageTransfer		709	6264

	totalDestinationPackets	totalSourcePackets \
0	894	569
1	1192	771
2	52	35
3	71	47
4	83	61
...	...	...
142366	1	1
142367	1	1
142368	1	3
142369	1	1
142370	7	7

	sourcePayloadAsBase64 \
0	None
1	None
2	None
3	None
4	R0VUIC9kZXNpZ24wNS9pbWFnZXMvMjAwOS8xMjA3L2Rhdm...
...	...
142366	lM0BAAABAAAAAABDmQycmRmbml6ZW41YXBsCmNsb3VkZn...
142367	qx8BAAABAAAAAABBBWExNzg0AWwGYWthbWpA25ldAAAAQ...
142368	None
142369	sm0BAAABAAAAAABAmExBXR3aw1nA2NvbQAAAQABAApEA...
142370	None

	sourcePayloadAsUTF \
0	None
1	None
2	None
3	None
4	GET /design05/images/2009/1207/david20091207_2...
...	...
142366	.....d2rdfnizen5aplcloudfront.net..)...
142367	.....a1784.l.akamai.net..)...
142368	None
142369	.m....a1.twimg.com..)...
142370	None

	destinationPayloadAsBase64 \
0	None
1	None

2		None
3		None
4		None
...		...
142366	lM2BgAABAaAAgADDmQycmRmbml6ZW41YXBsCmNsb3VkZn...	
142367	qx+BgAABAAQACQAKBWExNzg0AWwGYWthbWFpA25ldAAAAQ...	
142368		None
142369	sm2BgAABAAYACQAKAmExBXR3aW1nA2NvbQAAAQABwAwABQ...	
142370		None

	destinationPayloadAsUTF	direction	\
0		None	L2R
1		None	L2R
2		None	L2L
3		None	L2R
4		None	L2R
...		...	...
142366	.....d2rdfnizen5aplcloudfront.net.....,..d...		L2R
142367	.....a1784.l.akamai.net.....H.....H..3...		L2R
142368		None	L2R
142369	.m.....a1.twimg.com.....a1.twimg.comedgesu...		L2R
142370		None	L2R

	sourceTCPFlagsDescription	destinationTCPFlagsDescription	\
0	R,P,A	P,A	
1	R,P,A	P,A	
2	NaN	NaN	
3	P,A	P,A	
4	F,P,A	F,P,A	
...	...	...	
142366	NaN	NaN	
142367	NaN	NaN	
142368	S,P,A	S,A	
142369	NaN	NaN	
142370	S,P,A	S,P,A	

	source	protocolName	sourcePort	destination	\
0	192.168.1.101	tcp_ip	4646	89.234.1.43	
1	192.168.1.101	tcp_ip	4722	89.234.1.43	
2	192.168.2.111	udp_ip	1654	192.168.5.122	
3	192.168.1.101	tcp_ip	1169	142.166.14.72	
4	192.168.2.111	tcp_ip	4345	142.166.14.85	
...	...	...	...	...	
142366	192.168.5.122	udp_ip	60428	198.164.30.2	
142367	192.168.5.122	udp_ip	59052	198.164.30.2	
142368	192.168.3.115	tcp_ip	2756	203.73.24.75	
142369	192.168.5.122	udp_ip	7320	198.164.30.2	
142370	192.168.1.101	tcp_ip	2080	72.246.31.72	

	destinationPort	startDateTime	stopDateTime	Tag
0	80	2010-06-16T10:36:37	2010-06-16T11:02:07	Normal
1	80	2010-06-16T10:38:51	2010-06-16T11:03:33	Normal
2	53	2010-06-16T10:48:35	2010-06-16T11:04:24	Normal
3	80	2010-06-16T10:51:32	2010-06-16T11:00:16	Normal
4	80	2010-06-16T10:52:32	2010-06-16T11:03:25	Normal
...	...	...	...	...

142366	53	2010-06-16T15:58:58	2010-06-16T15:58:58	Normal
142367	53	2010-06-16T15:58:59	2010-06-16T15:58:59	Normal
142368	80	2010-06-16T15:58:59	2010-06-16T15:58:59	Normal
142369	53	2010-06-16T15:58:59	2010-06-16T15:58:59	Normal
142370	80	2010-06-16T15:58:59	2010-06-16T15:58:59	Normal

[142371 rows x 20 columns]>

In [9]: *#Building and training*

```

AppCount = pd.value_counts(df['appName'])
AttackCount = pd.value_counts(df['Tag'])
AttackDataframe = pd.DataFrame(df.loc[df['Tag']=='Attack'])
AttackCount2 = pd.value_counts(AttackDataframe['appName'])
NormalDataframe = pd.DataFrame(df.loc[df['Tag']=="Normal"])
NormalDataframeY = NormalDataframe[["Tag"]]
AttackDataframeY = AttackDataframe[["Tag"]]

AttackDataframe = AttackDataframe[["totalSourceBytes", "totalDestinationBytes", \
    "totalDestinationPackets", "totalSourcePackets", "sourcePort", "destinationPort"]]

NormalDataframe = NormalDataframe[["totalSourceBytes", "totalDestinationBytes", \
    "totalDestinationPackets", "totalSourcePackets", "sourcePort", "destinationPort"]]

NormalDataframeY = NormalDataframeY.to_numpy()
NormalDataframeY = NormalDataframeY.ravel()
labels, uniques = pd.factorize(NormalDataframeY)
NormalDataframeY = labels
NormalDataframeY = NormalDataframeY.ravel()

AttackDataframeY = AttackDataframeY.to_numpy()
AttackDataframeY = AttackDataframeY.ravel()
labelsS, uniquesS = pd.factorize(AttackDataframeY)
AttackDataframeY = labelsS
AttackDataframeY = AttackDataframeY.ravel()

indices_zero = AttackDataframeY == 0
AttackDataframeY[indices_zero] = 1

from sklearn.model_selection import train_test_split
X_train_N, X_test_N, Y_train_N, Y_test_N = train_test_split(NormalDataframe, \
    NormalDataframeY, random_state = 0, test_size = 8000)
X_train_A, X_test_A, Y_train_A, Y_test_A = train_test_split(AttackDataframe, \
    AttackDataframeY, random_state = 0, test_size = 0.3)

X_train = pd.concat([X_train_N, X_train_A])
X_train = X_train.sample(frac=1, random_state = 42)

X_test = pd.concat([X_test_N, X_test_A])
X_test = X_train.sample(frac=1, random_state = 42)

Y_train_N = pd.DataFrame(Y_train_N)
Y_train_A = pd.DataFrame(Y_train_A)
Y_train = pd.concat([Y_train_N, Y_train_A])

```

```

Y_train = pd.DataFrame(Y_train)
Y_train = Y_train.sample(frac=1, random_state = 42)

Y_test_N = pd.DataFrame(Y_test_N)
Y_test_A = pd.DataFrame(Y_test_A)
Y_test = pd.concat([Y_test_N, Y_test_A])
Y_test = pd.DataFrame(Y_test)
Y_test = Y_train.sample(frac=1, random_state = 42)

from sklearn.model_selection import train_test_split
X_train_N, X_test_N, Y_train_N, Y_test_N = train_test_split(NormalDataframe, \
    NormalDataframeY, random_state = 0, test_size = 8000)
X_train_A, X_test_A, Y_train_A, Y_test_A = train_test_split(AttackDataframe, \
    AttackDataframeY, random_state = 0, test_size = 0.3)

X_train = pd.concat([X_train_N, X_train_A])
X_train = X_train.sample(frac=1, random_state = 42)

X_test = pd.concat([X_test_N, X_test_A])
X_test = X_train.sample(frac=1, random_state = 42)

Y_train_N = pd.Series(Y_train_N, name='Tag')
Y_train_A = pd.Series(Y_train_A, name='Tag')

Y_train = pd.concat([Y_train_N, Y_train_A])
Y_train = pd.DataFrame(Y_train)
Y_train = Y_train.sample(frac=1, random_state = 42)

Y_test_N = pd.Series(Y_test_N, name='Tag')
Y_test_A = pd.Series(Y_test_A, name='Tag')

Y_test = pd.concat([Y_test_N, Y_test_A])
Y_test = pd.DataFrame(Y_test)
Y_test = Y_train.sample(frac=1, random_state = 42)

```

In [10]: *#Graph*

```

transform = TSNE
X = X_test
trans = transform(n_components=2)
X_reduced = trans.fit_transform(X)
Y = pd.DataFrame(Y_test)
fig, ax = plt.subplots(figsize=(7,7))

ax.scatter(X_reduced[:,0],X_reduced[:,1], c = Y.iloc[:, 0].astype('category').cat.c
    cmap="jet", alpha=0.7)
#ax.scatter(X_reduced[:,0],X_reduced[:,1], c = Y[0].astype('category').cat.codes, \
#cmap="jet", alpha=0.7, aspect="equal",)

ax.set(xlabel="$X_1$", ylabel="$X_2$", title=f"{transform.__name__} visualization o

```

Out[10]: [Text(0.5, 0, '\$X\_1\$'),  
Text(0, 0.5, '\$X\_2\$'),  
Text(0.5, 1.0, 'TSNE visualization of IDS testing dataset')]

```
In [11]: 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
clf = DecisionTreeClassifier(random_state=0)
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)
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(
recall = recall_score(Y_test, predict, average = "weighted", labels=np.unique(predi
f1scoreMacro = f1_score(Y_test, predict, average = "macro", labels=np.unique(predic
print(classification_report(Y_test, predict, target_names=["Normal", "Attacks"]))
```

0.9999851157252362

	precision	recall	f1-score	support
Normal	1.00	1.00	1.00	134368
Attacks	1.00	1.00	1.00	2
accuracy			1.00	134370
macro avg	1.00	1.00	1.00	134370
weighted avg	1.00	1.00	1.00	134370

In [48]: #TASK 2 : BINARY SEARCH TREE

```
class Node: #Node class, with a value in root and another node in left and right
    def __init__(self, root=None, left=None, right=None):
        self.root = root
        self.left = left
        self.right = right

class BinarySearchTree: #BinarySearchTree class
    def __init__(self, numbers=None): #With a root and a size. A list of numbers
        #can be given to fill the initial tree
        self.root = None
        self.size = 0
        if numbers is not None:
            for number in numbers:
                self.insert(number)
                self.size = self.size + 1

    def search(self, searching, node = None): #Search of a value within the tree
        if node is None:
            return False, node
        if node.root == searching: #If root is value, we found it
            return True, node
        if node.root > searching : #Reccurence Logic
            return self.search(searching, node.left)
        else :
            return self.search(searching, node.right)
```



```

def insert(self, number, node = None): #Insert a new value
    if self.root == None: #If tree is empty we can just insert it
        self.root = Node(number)
        return
    if node==None :
        node = self.root
    alreadyExist, r= self.search(number, self.root)
    if not alreadyExist: #if the value already exist in the tree it's ignored.
        if number < node.root:
            if node.left == None: #if it fits it sits
                self.size = self.size+1
                node.left = Node(number)
                return
            else : #Reccurence logic
                self.insert(number, node.left)
        else :
            if node.right == None:
                self.size = self.size+1
                node.right = Node(number)
                return
            else :
                self.insert(number, node.right)

def reset(self, new_numbers=None):
    # Reset the tree with new numbers
    self.root = None
    self.size = 0
    if new_numbers!=None :
        for number in new_numbers:
            self.insert(number)
            self.size += 1

def delete(self, value, root=None): #probably not the most efficient method
    neworder = []
    if root==None:
        root = self.root.root
    exist, node = self.search(value, self.root)
    if not exist :
        print("Node doesnt exist")
        return
    #we're going to switch the node to delete with the
    #one that comes next numerically
    allValues = self.inorder(self.root)
    preorder = self.preorder(self.root)
    for i in range(len(allValues)):
        if allValues[i]==value:
            if i+1<len(allValues):
                new = allValues[i+1]
            else :
                new = "HighestUp"
            break
    neworder = []
    for i in preorder:#and rebuild the tree from scratch
        if i==value and new!="HighestUp":
            neworder.append(new)
        elif i==new:

```

```

        1
    else :
        neworder.append(i)
    self.reset(neworder)

def preorder(self, node=None, res = None): #Create a list of the
#value of the tree in preorder order
    if node == None :
        node = self.root
    if res == None :
        res = []
    res.append(node.root)
    if node.left is not None:
        res = self.preorder(node.left, res)
    if node.right is not None:
        res = self.preorder(node.right, res)
    return res

    #pb mon inorder ne prend que les trucs >root
def inorder(self, node=None, res = None): #Create a list of the
#value of the tree in inorder order
    if node == None :
        node = self.root
    if res == None :
        res = []
    if node.left is not None:
        res = self.inorder(node.left, res)
    res.append(node.root)
    if node.right is not None:
        res = self.preorder(node.right, res)
    return res

def postorder(self, node=None, res = None): #Create a list of the
#value of the tree in postorder order
    if node == None :
        node = self.root
    if res == None :
        res = []
    if node.left is not None:
        res = self.postorder(node.left, res)
    if node.right is not None:
        res = self.postorder(node.right, res)
    res.append(node.root)
    return res

```

```

In [50]: #Entry Lists
a = [49, 38, 65, 97, 60, 76, 13, 27, 5, 1]
b = [149, 38, 65, 197, 60, 176, 13, 217, 5, 11]
c = [49, 38, 65, 97, 64, 76, 13, 77, 5, 1, 55, 50, 24]

#Creation of the trees
treeA = BinarySearchTree(a)
treeB = BinarySearchTree(b)
treeC = BinarySearchTree(c)

#FUNCTION TEST

```

#### *#Orders*

```
print("TreeA, B, C ; preorder, inorder, postorder")
print(treeA.preorder())
print(treeA.inorder())
print(treeA.postorder())

print(treeB.preorder())
print(treeB.inorder())
print(treeB.postorder())

print(treeC.preorder())
print(treeC.inorder())
print(treeC.postorder())
```

#### *#Search*

```
print("\n TreeA, B, C; searching 50 then 60")
print(treeA.search(50, treeA.root))
print(treeA.search(60, treeA.root))
print(treeB.search(50, treeB.root))
print(treeB.search(60, treeB.root))
print(treeC.search(50, treeC.root))
print(treeC.search(60, treeC.root))
```

#### *#Insert*

```
print("\n Tree A, B, C; attempting to insert 25 twice")
treeA.insert(25)
print(treeA.preorder())
treeA.insert(25)
print(treeA.preorder())

treeB.insert(25)
print(treeB.preorder())
treeB.insert(25)
print(treeB.preorder())

treeC.insert(25)
print(treeC.preorder())
treeC.insert(25)
print(treeC.preorder())
```

#### *#Delete*

```
print("\n Tree A, B, C; attempting to delete 50 then 60")
treeA.delete(50)
treeA.delete(60)
print(treeA.preorder())

treeB.delete(50)
treeB.delete(60)
print(treeB.preorder())

treeC.delete(50)
treeC.delete(60)
print(treeC.preorder())
```

```
TreeA, B, C ; preorder, inorder, postorder
[49, 38, 13, 5, 1, 27, 65, 60, 97, 76]
[1, 5, 13, 27, 38, 49, 65, 60, 97, 76]
[1, 5, 27, 13, 38, 60, 76, 97, 65, 49]
[149, 38, 13, 5, 11, 65, 60, 197, 176, 217]
[5, 11, 13, 38, 65, 60, 149, 197, 176, 217]
[11, 5, 13, 60, 65, 38, 176, 217, 197, 149]
[49, 38, 13, 5, 1, 24, 65, 64, 55, 50, 97, 76, 77]
[1, 5, 13, 24, 38, 49, 65, 64, 55, 50, 97, 76, 77]
[1, 5, 24, 13, 38, 50, 55, 64, 77, 76, 97, 65, 49]
```

```
TreeA, B, C; searching 50 then 60
(False, None)
(True, <__main__.Node object at 0x000001DB31C5F9A0>)
(False, None)
(True, <__main__.Node object at 0x000001DB31C61A60>)
(True, <__main__.Node object at 0x000001DB31C5F670>)
(False, None)
```

```
Tree A, B, C; attempting to insert 25 twice
[49, 38, 13, 5, 1, 27, 25, 65, 60, 97, 76]
[49, 38, 13, 5, 1, 27, 25, 65, 60, 97, 76]
[149, 38, 13, 5, 11, 25, 65, 60, 197, 176, 217]
[149, 38, 13, 5, 11, 25, 65, 60, 197, 176, 217]
[49, 38, 13, 5, 1, 24, 25, 65, 64, 55, 50, 97, 76, 77]
[49, 38, 13, 5, 1, 24, 25, 65, 64, 55, 50, 97, 76, 77]
```

```
Tree A, B, C; attempting to delete 50 then 60
Node doesnt exist
[49, 38, 13, 5, 1, 27, 25, 65, 97, 76]
Node doesnt exist
[38, 13, 5, 11, 25, 65, 149, 197, 176, 217]
Node doesnt exist
[49, 38, 13, 5, 1, 24, 25, 65, 64, 55, 97, 76, 77]
```