



## A PRACTICAL REPORT

"Applied Artificial intelligence Practical"  
FOR  
UNIVERSITY OF MUMBAI

SUBMITTED BY

**NAME: Rupesh Patil**

**Roll NO: 17**

M.Sc. (INFORMATION TECHNOLOGY)

PART-2 SEM-III

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**CONDUCTED AT**

VALIA C.L. COLLEGE OF COMMERCE

&

VALIA L.C. COLLEGE OF ARTS

ANDHERI (W), MUMBAI - 400053

*Cosmopolitans*

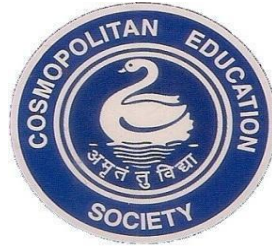
**VALIA C. L. COLLEGE OF COMMERCE**

**&**

**VALIA L.C. COLLEGE OF ARTS**

**(Affiliated to University of Mumbai)**

**D.N. Nagar, Andheri (W), Mumbai-400053.**



**CERTIFICATE**

This is to certify that **Mr. rupesh patil**, Roll No.: **17**, studying in Master of Science (Information Technology) SEM: III has satisfactorily completed the practicals in the **Course: Applied Artificial Intelligence**", as prescribed by the University of Mumbai, during the academic year 2020-2021.

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**Course In charge**

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**Head, Dept. Of Information  
Technology**

**DATE & COLLEGE SEAL**

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**Examiner**

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## **Practical 1**

**Aim:** Design an Expert system using AIML.

**Description:** An expert system is a computer program that uses artificial intelligence (AI) technologies to simulate the judgment and behavior of a human or an organization that has expert knowledge and experience in a particular field.

Typically, an expert system incorporates a knowledge base containing accumulated experience and an inference or rules engine -- a set of rules for applying the knowledge base to each particular situation that is described to the program. The system's capabilities can be enhanced with additions to the knowledge base or to the set of rules. Current systems may include machine learning capabilities that allow them to improve their performance based on experience, just as humans do.

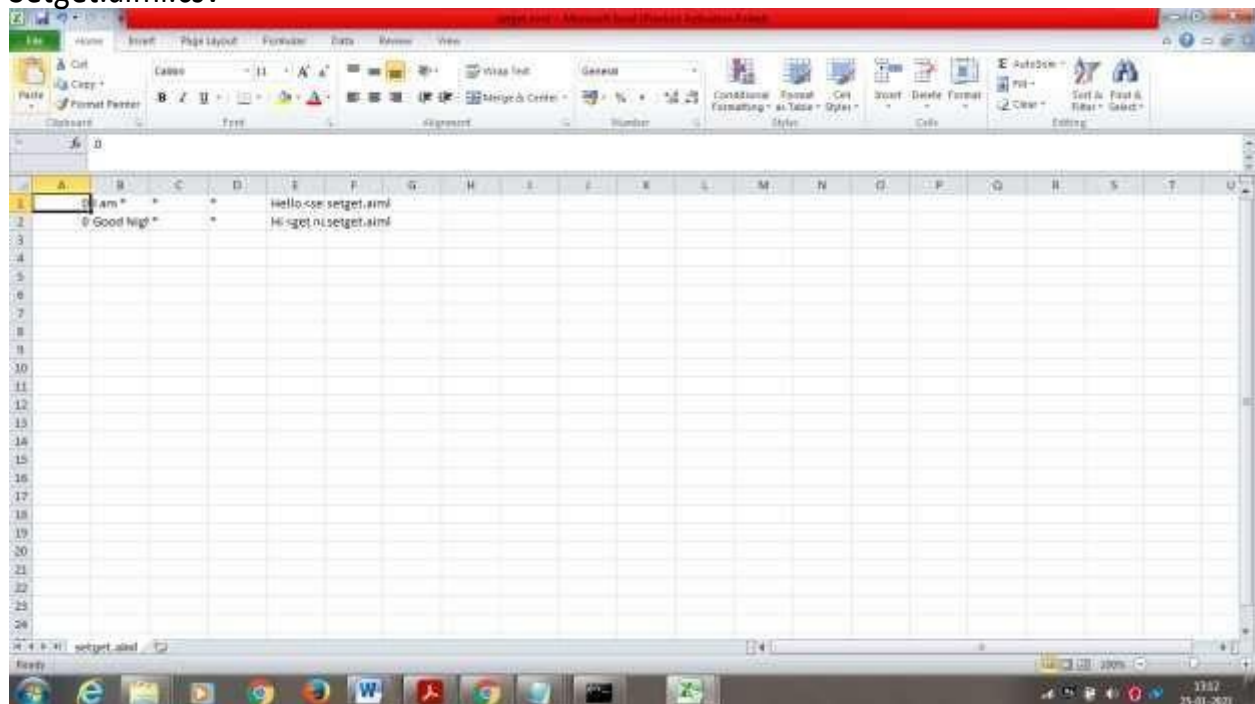
### **Code:**

```
setget.aiml
<?xml version = "1.0" encoding = "UTF-8"?>
<aiml version = "1.0.1" encoding = "UTF-8"?>
  <category>
    <pattern>I am *</pattern>
    <template>
      Hello <set name = "username"> Welcome to our clinic <star/>! </set>
    </template>
  </category>
  <category>
    <pattern>Good Night</pattern>
    <template>
      Hi <get name = "username"/> Thanks for the conversation! take the medicine
properly
    </template>
  </category>
</aiml>
```

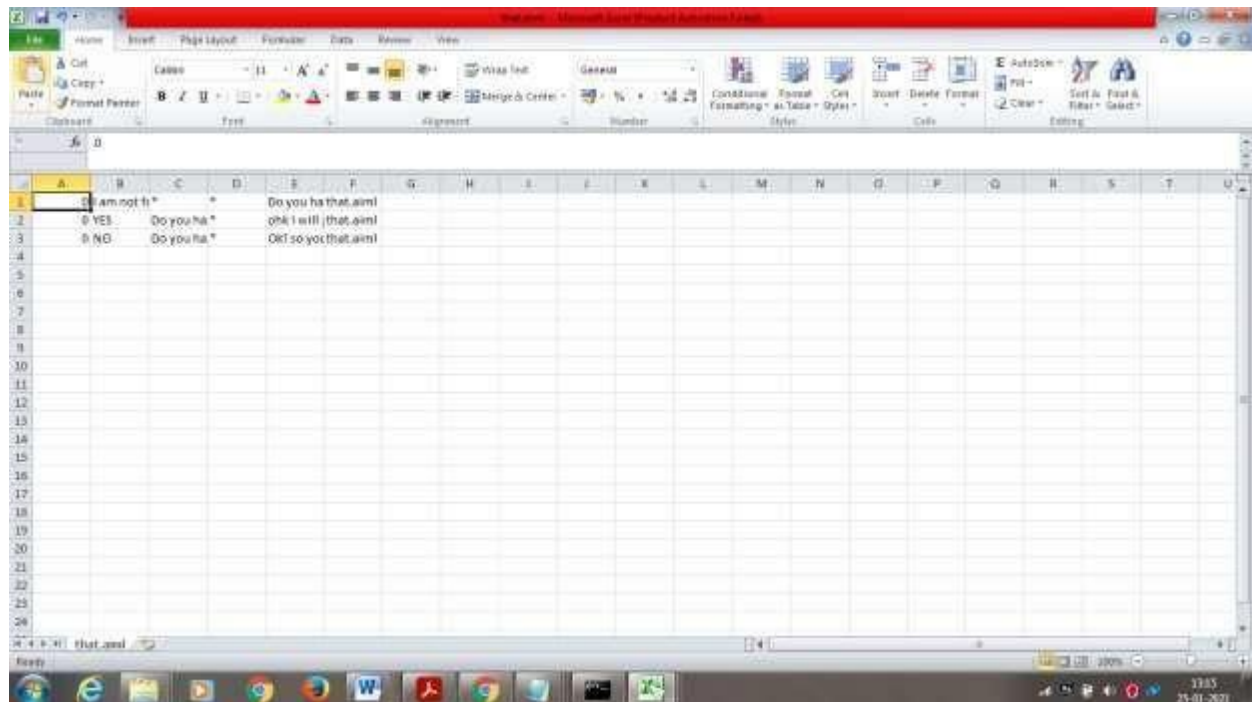
that.aiml

```
<?xml version = "1.0" encoding = "UTF-8"?>
<aiml version = "1.0.1" encoding = "UTF-8"?>
  <category>
    <pattern>I am not feeling well</pattern>
    <template>Do you have cough or cold</template>
  </category>
  <category>
    <pattern>YES</pattern>
    <that>Do you have cough or cold</that>
    <template>ohk I will prescribe you some medicene and syrup too.</template>
  </category>
  <category>
    <pattern>NO</pattern>
    <that>Do you have cough or cold</that>
    <template>Ok! so you have viral infection dont worry you will be
fine.</template>
  </category>
</aiml>
```

Setget.aiml.csv



That.aiml.csv



Execution command:

java -cp lib/Ab.jar Main bot = test action = chat trace = false

## Output:-

```
Command Prompt - java -cp lib\Abjar Main Bot - test action - chat trace - false
28228 nodes 22843 singletons 4881 leaves 8 shortcuts 1374 n-ary 28219 branches 8
.99996454 average branching
Human: I am Aamina
STATE:I am Aamina:THAT=unknown:TOPIC=unknown
normalized - I am Aamina
Matched: I AM * <THAT> * <TOPIC> * setget.aiml
Setting predicate username to Welcome to our clinic Aamina!
writeCertainIFCategories learnf.aiml size= 8
Robot: Hello Welcome to our clinic Aamina!
Human: I am not feeling well
STATE:I am not feeling well:THAT=Hello Welcome to our clinic Aamina:TOPIC=unknown
normalized - I am not feeling well
Matched: I AM NOT FEELING WELL <THAT> * <TOPIC> * that.aiml
writeCertainIFCategories learnf.aiml size= 8
Robot: Do you have cough or cold
Human: YES
STATE=YES:THAT=Do you have cough or cold:TOPIC=unknown
normalized - YES
Matched: YES <THAT> DO YOU HAVE COUGH OR COLD <TOPIC> * that.aiml
writeCertainIFCategories learnf.aiml size= 8
Robot: ohk I will prescribe you some medicine and syrup too.
Human: Ohk
STATE=Ohk:THAT=ohk I will prescribe you some medicine and syrup too:TOPIC=unknown
normalized - Ohk
Matched: OHK <THAT> * <TOPIC> * reductions1.aiml
0. <srail>OK</srail> from OHK <THAT> * <TOPIC> * topic=unknown)
Matched: OK <THAT> * <TOPIC> * reductions1.aiml
1. <srail>YES</srail> from OK <THAT> * <TOPIC> * topic=unknown)
Matched: YES <THAT> * <TOPIC> * reductions1.aiml
2. <srail>INTERJECTION</srail> from YES <THAT> * <TOPIC> * topic=unknown)
Matched: INTERJECTION <THAT> * <TOPIC> * personality.aiml
writeCertainIFCategories learnf.aiml size= 8
Robot: So.
Human: Thank You
STATE=Thank You:THAT=So:TOPIC=unknown
normalized - Thank You
Matched: THANK YOU <THAT> * <TOPIC> * reductions1.aiml
0. <srail>THANKS</srail> from THANK YOU <THAT> * <TOPIC> * topic=unknown)
Matched: THANKS <THAT> * <TOPIC> * personality.aiml
writeCertainIFCategories learnf.aiml size= 8
Robot: Any time.
Human: Good Night
STATE=Good Night:THAT=Any time:TOPIC=unknown
normalized - Good Night
Matched: GOOD NIGHT <THAT> * <TOPIC> * setget.aiml
writeCertainIFCategories learnf.aiml size= 8
Robot: Hi Welcome to our clinic Aamina! Thanks for the conversation! take the medicine properly
Human:
```

## **Practical 2: Bot using AIML**

**Aim:** Design a bot using AIML

**Description:** AIML stands for Artificial Intelligence Modelling Language. AIML is an XML based markup language meant to create artificial intelligent applications. AIML makes it possible to create human interfaces while keeping the implementation simple to program, easy to understand and highly maintainable. This tutorial will teach you the basics of AIML. All the basic components of AIML with suitable examples have been discussed in this tutorial.

### AIML Tags/Description

S.No.	AIML Tag / Description
1	<aiml> Defines the beginning and end of a AIML document.
2	<category> Defines the unit of knowledge in Alicebot's knowledge base.
3	<pattern> Defines the pattern to match what a user may input to an Alicebot.
4	<template> Defines the response of an Alicebot to user's input.

S.No.	AIML Tag / Description
1	<star> Used to match wild card * character(s) in the <pattern> Tag.
2	<srai> Multipurpose tag, used to call/match the other categories.



3	<code>&lt;random&gt;</code> Used <code>&lt;random&gt;</code> to get random responses.
4	<code>&lt;li&gt;</code> Used to represent multiple responses.
5	<code>&lt;set&gt;</code> Used to set value in an AIML variable.
6	<code>&lt;get&gt;</code> Used to get value stored in an AIML variable.
7	<code>&lt;that&gt;</code> Used in AIML to respond based on the context.
8	<code>&lt;topic&gt;</code> Used in AIML to store a context so that later conversation can be done based on that context.
9	<code>&lt;think&gt;</code> Used in AIML to store a variable without notifying the user.
10	<code>&lt;condition&gt;</code> Similar to switch statements in programming language. It helps ALICE to respond to matching input.

### **Code:**

#### **Inquiry.aiml**

```
<?xml version="1.0" encoding="UTF-8"?>
<aiml>
<category><pattern>*</pattern><that>HOW OLD ARE YOU</that>
```

```

<template><think><set    name="age"><star/></set></think>
<srai>MY AGE IS <star/></srai></template>
</category>
<category><pattern>INQUIRY  LOCATION</pattern>
<template><srai>INQUIRYLOCATION<get name="location"/></srai></template>
</category>
<category><pattern>INQUIRY LOCATION WHERE</pattern>
<template>Where are you?</template>
</category>
<category><pattern>INQUIRY LOCATION *</pattern>
<template><srai>RANDOM PICKUP LINE</srai></template>
</category>
<category><pattern>INQUIRY  NAME</pattern>
<template><srai>INQUIRY NAME <get name="username"/></srai></template>
</category>
<category><pattern>INQUIRY NAME *</pattern>
<template><srai>RANDOM PICKUP LINE</srai></template>
</category>
<category><pattern>INQUIRY NAME FRIEND</pattern>
<template>What is your name?</template>
</category>
<category><pattern>INQUIRY  AGE</pattern>
<template><srai>INQUIRY AGE <get name="age"/></srai></template>
</category>
<category><pattern>INQUIRY AGE HOW MANY</pattern>
<template>How old are you?</template>
</category>
<category><pattern>INQUIRY AGE *</pattern>
<template><srai>RANDOM PICKUP LINE</srai></template>
</category>
<category><pattern>INQUIRY  GENDER</pattern>
<template><srai>INQUIRY GENDER <get name="gender"/></srai></template>
</category>
<category><pattern>INQUIRY GENDER UNKNOWN</pattern>
<template>Are you a man or a woman?</template>
</category>
<category><pattern>INQUIRY GENDER *</pattern>

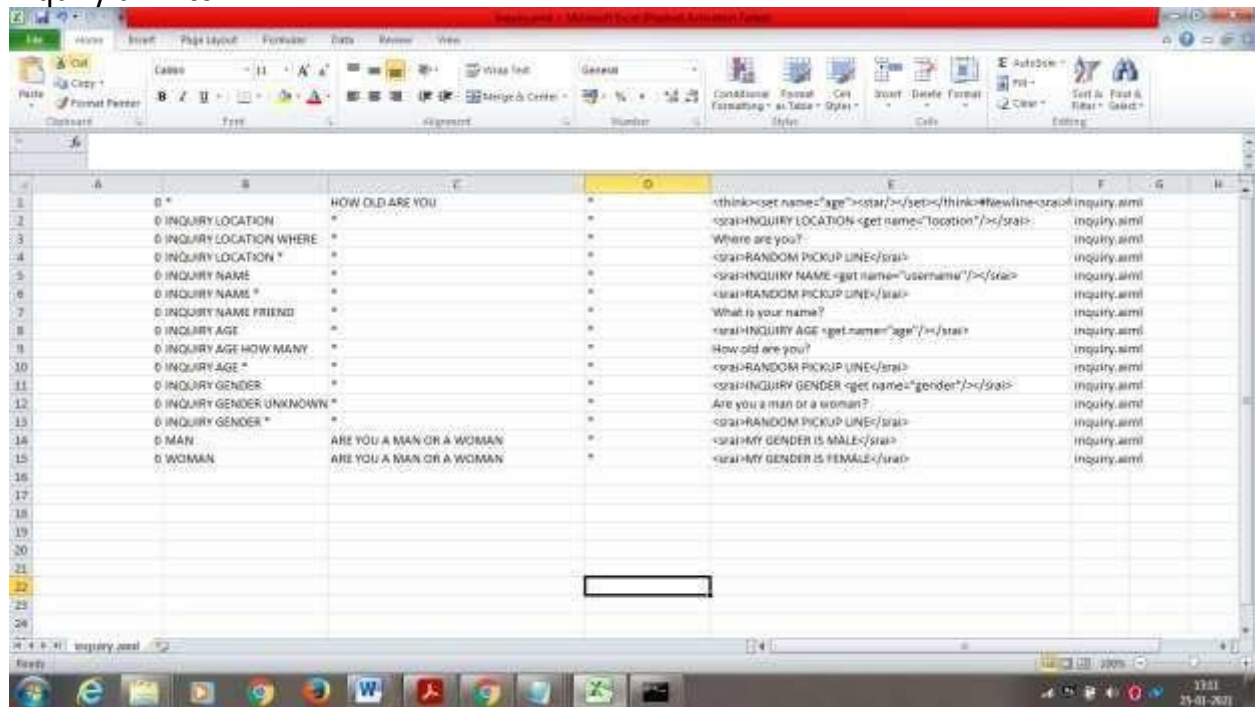
```

```

<template><srai>RANDOM PICKUP LINE</srai></template>
</category>
<category><pattern>MAN</pattern><that>ARE YOU A MAN OR A WOMAN</that>
<template><srai>MY GENDER IS MALE</srai></template>
</category>
<category><pattern>WOMAN</pattern><that>ARE YOU A MAN OR A WOMAN</that>
<template><srai>MY GENDER IS FEMALE</srai></template>
</category>
</aiml>

```

Inquiry.aiml.csv



## **Execution command:**

java -cp lib/Ab.jar Main bot = test action = chat trace = false

**Output:**

```

# Corvax Python Compiler: A Python to Machine Language Compiler - Part 2: Code Generation
# James Jacob (jacob@corvax.com) - jacob@corvax.com
# Where possible options include:
# (filename)
# Options to pass to annotation processors
# --add-module <module>[:<module>:]
# Heat modules to resolve in addition to the initial module, or all modules
#
# Use the module path if <module> is BIL-PODFILE-FILE.
# --out-<long-path> <path> --outlongpath <path>
# Override location of bootstrap class files.
# --class-path <path> --classpath <path> --cp <path>
# Specify where to find user class files and annotation processors.
# -d <dirpath> Specify where to place generated class files.
# --deprecation
# Output source locations where deprecated APIs are used.
# --enable-preview
# Enable preview language features. To be used in conjunction with either
# source or --release.
# --encoding <encoding> Specify character encoding used by source files.
# --endoredir <dir> Override location of endored standard path.
# --outdir <dir> Override location of installed extensions.
# --generate-all-debugging-info
# Generate all debugging info.
# --generate-only-class-debugging-info
# Generate only class debugging info.
# --no-debug
# Generate no debugging info.
# -h <dirpath> Specify where to place generated native module files.
# --help, --help, -? Print this help message.
# --help-extra, -x Print help on extra options.
# --implicit-imports <class>
# Specify whether or not to generate class files for implicitly referenced files.
# --iflag <flag> Pass <flag> directly to the runtime system.
# --limit-modules <module>[:<module>:]
# Limit the number of modules/modules modules.
# --module <module>[:<module>:] --m <module>[:<module>:]
# Compile only the specified module(s), check timestamps.
# --module-path <path> --mp <path>
# Specify where to find application modules.
# --module-source-path <module>[:<source>:]
# Specify where to find input source files for multiple modules.
# --module-version <version>
# Specify version of modules that are being compiled.
# --no-assert
# Generate no assertions.
# --parameters
# Generate metadata for reflection on method parameters.
# --proc <name>[:<name>:]
# Control whether annotation processing and/or compilation is done.
# --processor <class>[:<class>[:<class>[:<class>]]...
# Names of the annotation processors to run; bypasses default discovery process.
#
# --processor-module-path <path>
# Specify the module path where to find annotation processors.
# --processor-path <path> --processorspath <path>
# Specify where to find annotation processors.
# --profile <profile>
# Check that API used is available in the specified profile.
# --release <release>

```

### **Practical 3: Bayes theorem**

**Aim:** Implement Bayes Theorem using Python

**Description:** Bayes Theorem is a method of calculating conditional probability. The traditional method of calculating conditional probability (the probability that one event occurs given the occurrence of a different event) is to use the conditional probability formula, calculating the joint probability of event one and event two occurring at the same time, and then dividing it by the probability of event two occurring. However, conditional probability can also be calculated in a slightly different fashion by using Bayes Theorem.

When calculating conditional probability with Bayes theorem, you use the following steps:

- Determine the probability of condition B being true, assuming that condition A is true.
- Determine the probability of event A being true.
- Multiply the two probabilities together.
- Divide by the probability of event B occurring.

This means that the formula for Bayes Theorem could be expressed like this:

$$P(A|B) = P(B|A) * P(A) / P(B)$$

**Code:**

```
print("Aamina53004190028")

def bayes_theorem(p_a, p_b_given_a, p_b_given_not_a):
    not_a = 1 - p_a
    # calculate P(B)
    p_b = p_b_given_a * p_a + p_b_given_not_a * not_a
```

```

        # calculate P(A|B)
        p_a_given_b = (p_b_given_a * p_a) / p_b
        return p_a_given_b

Prob_a = (35 + 5) / (35 + 5 + 277 + 78)
print('prob_a', Prob_a)

prob_b = (78 + 5) / (35 + 5 + 277 + 78)
print('prob_b', prob_b)

Prob_of_a_intersec_b = 5 / (35 + 5 + 277 + 78)
print('Prob_of_a_intersec_b', Prob_of_a_intersec_b)

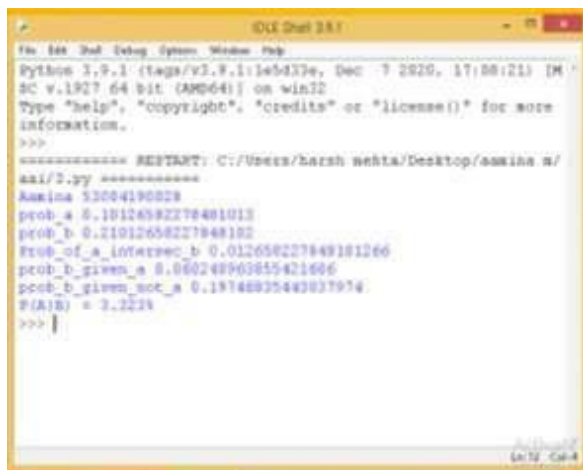
prob_b_given_a = Prob_of_a_intersec_b / prob_b
print("prob_b_given_a", prob_b_given_a)

prob_b_given_not_a = prob_b - Prob_of_a_intersec_b
print("prob_b_given_not_a", prob_b_given_not_a)

result = bayes_theorem(Prob_a, prob_b_given_a, prob_b_given_not_a)
print('P(A|B) = %.3f%%' % (result * 100))

```

### **Output:**



```

Python 3.9.1 (tags/v3.9.1:1le5d33e, Dec 7 2020, 17:08:21) [M
32 v.1927 64 bit (AMD64)] on win32
Type "help()", "copyright()", "credits()" or "license()" for more
information.
>>>
===== RESTART: C:/Users/harsh wehta/Desktop/assina m/
aai/2.py =====
Assina 53004190028
prob_a 0.10126582278481013
prob_b 0.21012658227848102
Prob_of_a_intersec_b 0.012658227848101266
prob_b_given_a 0.060248963855421666
prob_b_given_not_a 0.197468325443027974
P(A|B) = 3.22%
>>>

```

## **PRACTICAL 4**

**AIM:** To implement conditional probability and joint probability using python

### **DESCRIPTION:**

Conditional probability is the [probability](#) of one event occurring with some relationship to one or more other events. For example: Event A is that it is raining outside, and it has a 0.3 (30%) chance of raining today. Event B is that you will need to go outside, and that has a probability of 0.5 (50%). A conditional probability would look at these two events in relationship with one another, such as the probability that it is both raining and you will need to go outside.

Joint probability is a statistical measure that calculates the likelihood of two events occurring together and at the same point in time. Joint probability is the probability of event Y occurring at the same time that event X occurs.

### **CODE:**

```
import pandas as pd
import numpy as np
df = pd.read_csv('student-mat.csv')
df.head(3)
print(len(df))
df['grade_A'] = np.where(df['G3']*5 >= 80, 1, 0)
#Make another boolean column called high_absences with a value of 1 if a
student missed 10 or more classes.
df['high_absences'] = np.where(df['absences'] >= 10, 1, 0)
#Add one more column to make building a pivot table easier.
df['count'] = 1
#And drop all columns we don't care about.
df = df[['grade_A', 'high_absences', 'count']]
print(df.head())
#Now we'll create a pivot table from this.
pt = pd.pivot_table(
    df,
    values='count',
    index=['grade_A'],
```

```

columns=['high_absenses'],
aggfunc=np.size,
fill_value=0
)
print(pt)
#P(A) = (35 + 5) / (35 + 5 + 277 + 78) = 0.10126582278481013
#P(B) = (78 + 5) / (35 + 5 + 277 + 78) = 0.21012658227848102
#P(A ∩ B) = 5 / (35 + 5 + 277 + 78) = 0.012658227848101266
#And per the formula, P(A|B) = P(A ∩ B) / P(B), put it together.
#P(A|B) = 0.012658227848101266/ 0.21012658227848102= 0.06
p_A = (pt[0][1]+pt[1][1])/(pt[0][0]+pt[0][1]+pt[1][0]+pt[1][1])
print(p_A)
p_B = (pt[1][0]+pt[1][1])/(pt[0][0]+pt[0][1]+pt[1][0]+pt[1][1])
print(p_B)
p_A_intersection_B = (pt[1][1])/(pt[0][0]+pt[0][1]+pt[1][0]+pt[1][1])
print(p_A_intersection_B)
p_A_given_B = p_A_intersection_B/p_B
print('Conditional probability = ',p_A_given_B)
print('Joint probability = ',p_A_intersection_B)

```

## OUTPUT:

```

IDLE Shell 3.9.1
File Edit Shell Debug Options Window Help
Python 3.9.1 (tags/v3.9.1:1.2345678, Dec 1 2020, 11:00:21) [AMD64] on win32
Type "help", "copyright", "credits" or "license()" for more information.
>>>
===== RESTART: C:/Users/harsh mehta/Desktop/aamina m/4.py =====
395
  grade_A  high_absenses  count
0         0             0       1
1         0             0       1
2         0             1       1
3         0             0       1
4         0             0       1
high_absenses    0    1
grade_A
0          277   78
1           35    5
0.10126582278481013
0.21012658227848102
0.012658227848101266
Conditional probability =  0.060240963855421686
Joint probability =  0.012658227848101266
>>> |

```

Ln: 21 Col: 4



## **Practical 6: Fuzzy based application**

**Aim:** Design a Fuzzy based application using Python/R.

**Discription:** Let's create a fuzzy control system which models how you might choose to tip at a restaurant. When tipping, you consider the service and food quality, rated between 0 and 10. You use this to leave a tip of between 0 and 25%. We would formulate this problem as:

- **Antecedents (Inputs)**
  - **service**
    - Universe (i.e., crisp value range): How good was the service of the wait staff, on a scale of 0 to 10?
    - Fuzzy set (i.e., fuzzy value range): poor, acceptable, amazing
  - **food quality**
    - Universe: How tasty was the food, on a scale of 0 to 10?
    - Fuzzy set: bad, decent, great
- **Consequents (Outputs)**
  - **tip**
    - Universe: How much should we tip, on a scale of 0% to 25%
    - Fuzzy set: low, medium, high
- **Rules**
  - IF the *service* was good *or* the *food quality* was good, THEN the tip will be high.
  - IF the *service* was average, THEN the tip will be medium.
  - IF the *service* was poor *and* the *food quality* was poor THEN the tip will be low.
- **Usage**
  - **If I tell this controller that I rated:**
    - the service as 9.8, and
    - the quality as 6.5,
  - **it would recommend I leave:**
    - a 20.2% tip.

**Code:**

```
# -*- coding: utf-8 -*-  
"""
```

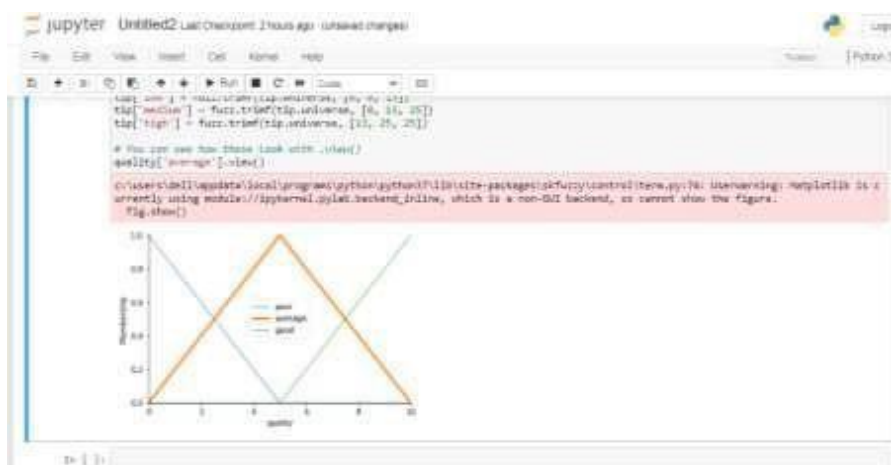
Created on Thu Jan 21 15:02:50 2021

@author: aamina

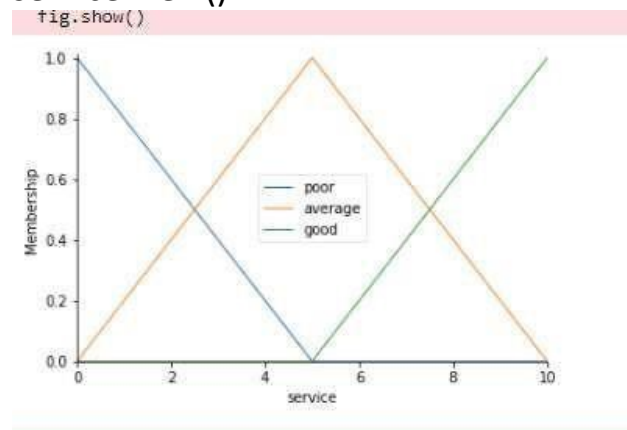
"""

```
import numpy as np
import skfuzzy as fuzz
from skfuzzy import control as ctrl
# New Antecedent/Consequent objects hold universe variables and membership
# functions
quality = ctrl.Antecedent(np.arange(0, 11, 1), 'quality')
service = ctrl.Antecedent(np.arange(0, 11, 1), 'service')
tip = ctrl.Consequent(np.arange(0, 26, 1), 'tip')
# Auto-membership function population is possible with .automf(3, 5, or 7)
quality.automf(3)
service.automf(3)
# Custom membership functions can be built interactively with a familiar,
# Pythonic API
tip['low'] = fuzz.trimf(tip.universe, [0, 0, 13])
tip['medium'] = fuzz.trimf(tip.universe, [0, 13, 25])
tip['high'] = fuzz.trimf(tip.universe, [13, 25, 25])
# You can see how these look with .view()
quality['average'].view()
```

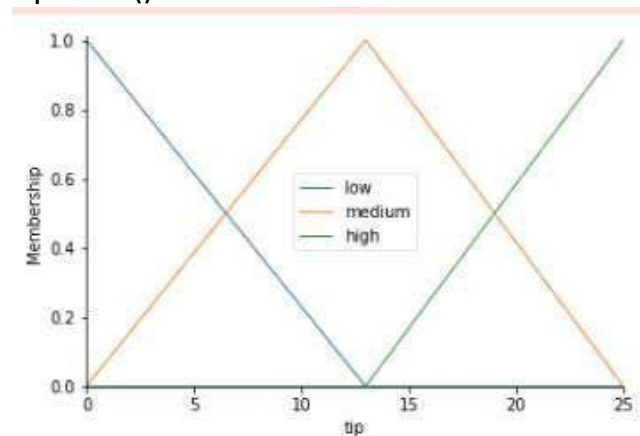
## **Output:**



service.view()



tip.view()



Now, to make these triangles useful, we define the *fuzzy relationship* between input and output variables. For the purposes of our example, consider three simple rules:

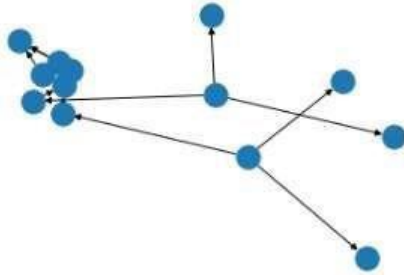
1. If the food is poor OR the service is poor, then the tip will be low
2. If the service is average, then the tip will be medium
3. If the food is good OR the service is good, then the tip will be high.

Most people would agree on these rules, but the rules are fuzzy. Mapping the imprecise rules into a defined, actionable tip is a challenge. This is the kind of task at which fuzzy logic excels.

```
rule1 = ctrl.Rule(quality['poor'] | service['poor'], tip['low'])
rule2 = ctrl.Rule(service['average'], tip['medium'])
rule3 = ctrl.Rule(service['good'] | quality['good'], tip['high'])
```

```
rule1.view()
```

```
Out[12]: (<Figure size 432x288 with 1 Axes>, <AxesSubplot:>)
```



```
tipping_ctrl=ctrl.ControlSystem([rule1,rule2,rule3])
```

```
tipping=ctrl.ControlSystemSimulation(tipping_ctrl)
```

**Suppose we rated the quality 6.5 out of 10 and the service 9.8 of 10.**

```
# Pass inputs to the ControlSystem using Antecedent labels with Pythonic API
```

```
# Note: if you like passing many inputs all at once, use .inputs(dict_of_data)
```

```
tipping.input['quality'] = 6.5
```

```
tipping.input['service'] = 9.8
```

```
# Crunch the numbers
```

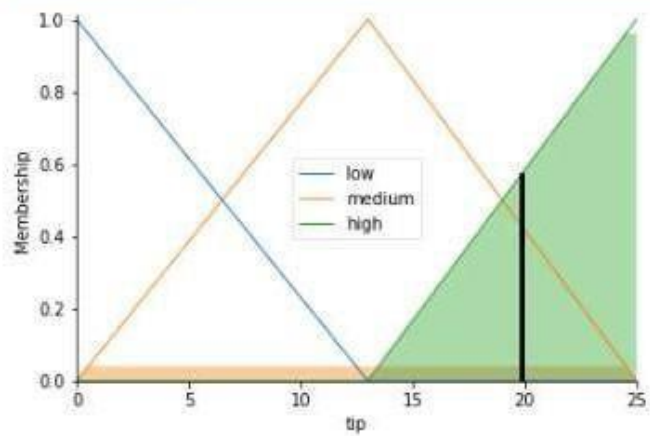
```
tipping.compute()
```

```
print (tipping.output['tip'])
```

```
tip.view(sim=tipping)
```

```
19.847607361963192
```

```
c:\users\dell\appdata\local\programs\python\python3;  
otlib is currently using module://ipykernel.pylab.ba  
fig.show()
```



The resulting suggested tip is **19.84%**

## Practical 7

### **Supervised learning model**

**Aim:** Write an application to simulate supervised and un-supervised learning model.

#### **Description:**

Supervised learning model

DecisionTreeClassifier:-

A decision tree is a flowchart-like tree structure where an internal node represents feature(or attribute), the branch represents a decision rule, and each leaf node represents the outcome. The topmost node in a decision tree is known as the root node. It learns to partition on the basis of the attribute value. It partitions the tree in recursively manner call recursive partitioning. This flowchart-like structure helps you in decision making. It's visualization like a flowchart diagram which easily mimics the human level thinking. That is why decision trees are easy to understand and interpret.

Import the packages. Let's first load the dataset using pandas' read CSV function. You can download the data [here](#). Here, you need to divide given columns into two types of variables dependent(or target variable) and independent variable(or feature variables).

To understand model performance, dividing the dataset into a training set and a test set is a good strategy.

Let's split the dataset by using function `train_test_split()`. You need to pass 3 parameters features, target, and test\_set size.

Let's estimate, how accurately the classifier or model can predict the type of cultivars. Accuracy can be computed by comparing actual test set values and predicted values.

#### **Code:**

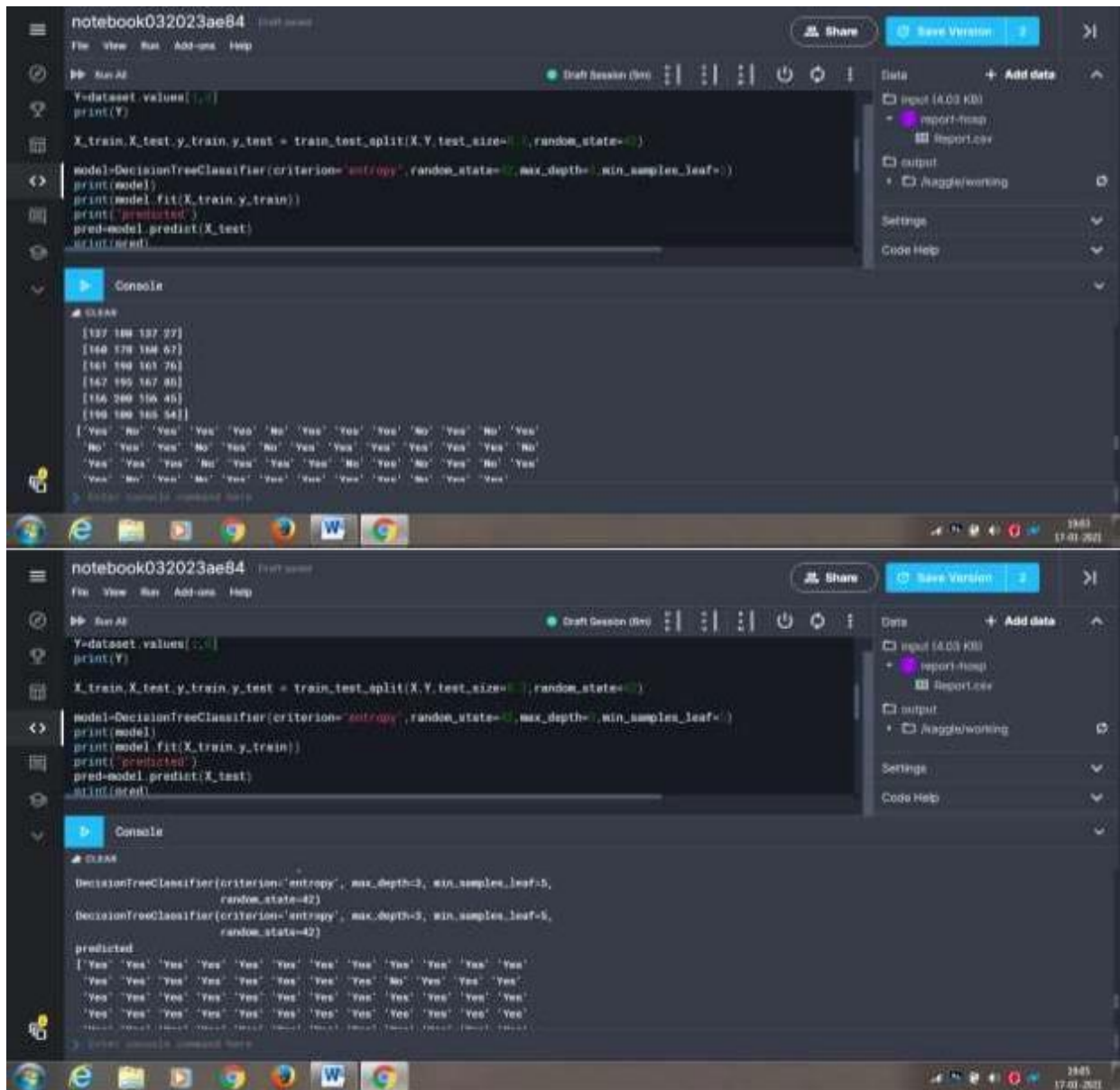
```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
from sklearn import metrics
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn import tree
from matplotlib import pyplot as plt
```

```

dataset=pd.read_csv("../input/report-hosp/Report.csv")
print(dataset)
size=len(dataset)
print(size)
print(dataset.head())
X=dataset.values[:,1:5]
print(X)
Y=dataset.values[:,0]
print(Y)
X_train,X_test,y_train,y_test
train_test_split(X,Y,test_size=0.3,random_state=42)
model=DecisionTreeClassifier(criterion="entropy",random_state=42,max_depth=
3,min_samples_leaf=5)
print(model)
print(model.fit(X_train,y_train))
print("predicted")
pred=model.predict(X_test)
print(pred)
accuracy=metrics.accuracy_score(y_test,pred)
print("accuracy")
print(accuracy)
print("confusion matrix")
print(metrics.confusion_matrix(y_test,pred))
print(metrics.classification_report(y_test,pred))
text_format=tree.export_text(model)
print(text_format)
fig=plt.figure(figsize=(25,20))
tree.plot_tree(model,feature_names=X,class_names=Y,filled=True)
=

```

### **Output:**





notebook032023ae84

```
Y=dataset.values[:,1:]
print(Y)

X_train,X_test,y_train,y_test = train_test_split(X,Y,test_size=0.3,random_state=42)

model=DecisionTreeClassifier(criterion='entropy',random_state=42,max_depth=3,min_samples_leaf=1)
print(model)
print(model.fit(X_train,y_train))
print('predicted')
pred=model.predict(X_test)
print(pred)
```

Console

```
CLEAR
'Yes' 'Yes' 'Yes' 'No' 'Yes' 'Yes'
accuracy
0.6666666666666667
confusion matrix
[[ 1 28]
 [ 1 44]]
precision    recall  f1-score   support

No           0.00      0.00      0.00         21
Yes          0.00      0.00      0.00         45
```

notebook032023ae84

```
Y=dataset.values[:,1:]
print(Y)

X_train,X_test,y_train,y_test = train_test_split(X,Y,test_size=0.3,random_state=42)

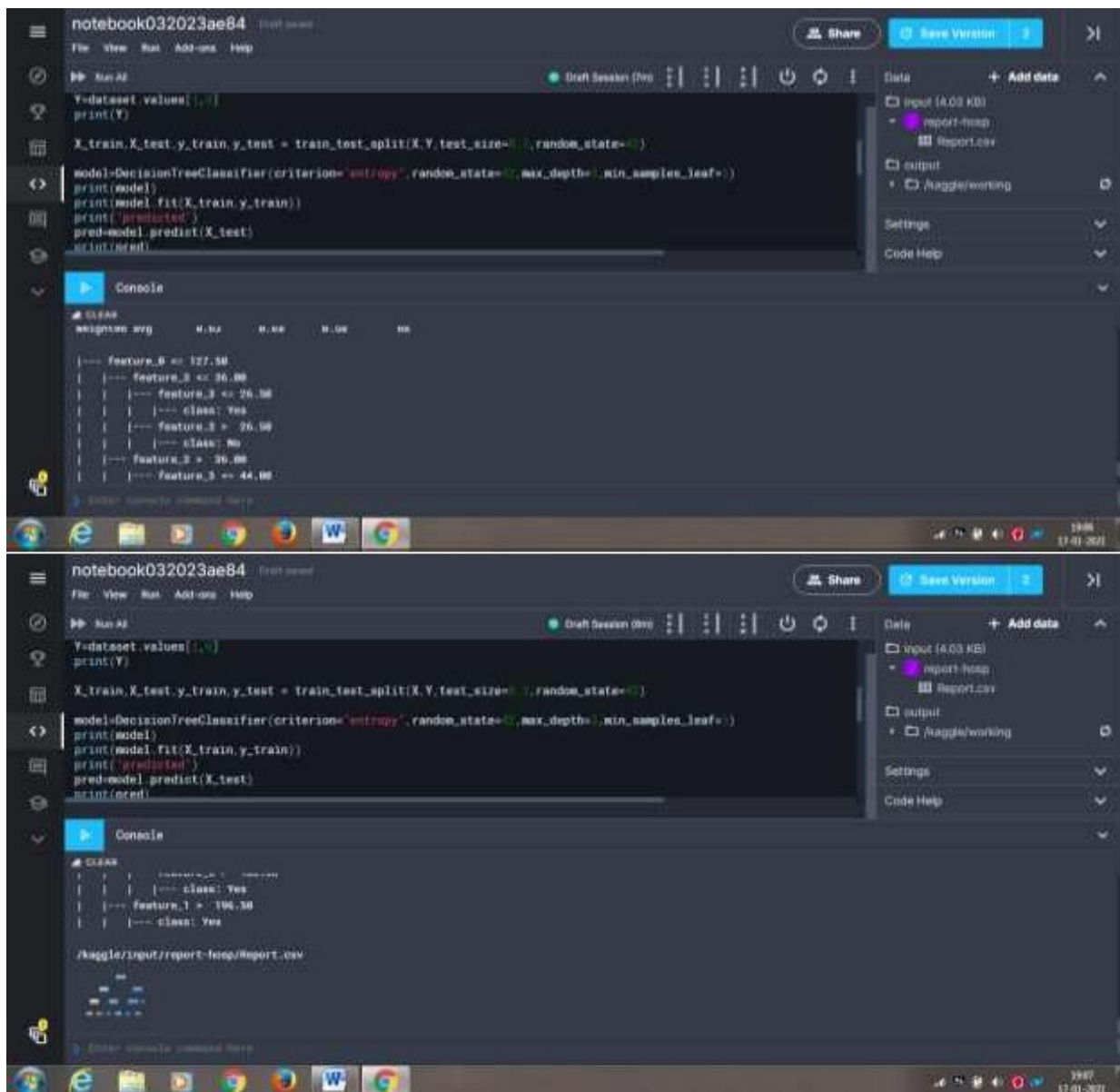
model=DecisionTreeClassifier(criterion='entropy',random_state=42,max_depth=3,min_samples_leaf=1)
print(model)
print(model.fit(X_train,y_train))
print('predicted')
pred=model.predict(X_test)
print(pred)
```

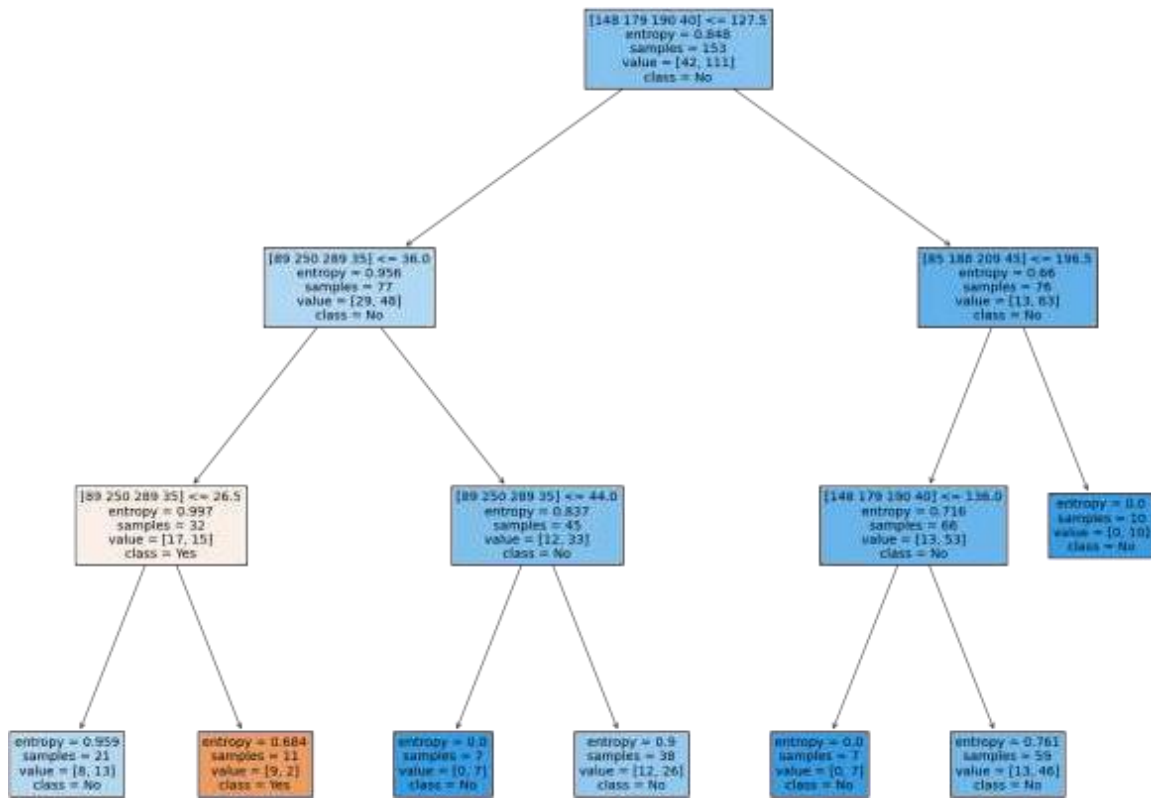
Console

```
CLEAR
[[ 1 28]
 [ 1 44]]
precision    recall  f1-score   support

No           0.00      0.00      0.00         21
Yes          0.00      0.00      0.00         45

accuracy
macro avg    0.00      0.00      0.00         66
weighted avg 0.00      0.00      0.00         66
```





## **PRACTICAL 8**

**AIM:** To implement clustering Algorithm (K-means Algorithm)

**DESCRIPTION:** Kmeans algorithm is an iterative algorithm that tries to partition the dataset into K pre-defined distinct non-overlapping subgroups (clusters) where each data point belongs to **only one group**. It tries to make the intra-cluster data points as similar as possible while also keeping the clusters as different (far) as possible. It assigns data points to a cluster such that the sum of the squared distance between the data points and the cluster's centroid (arithmetic mean of all the data points that belong to that cluster) is at the minimum. The less variation we have within clusters, the more homogeneous (similar) the data points are within the same cluster.

### **CODE:**

```
from copy import deepcopy
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
from matplotlib import pyplot as plt

# Set three centers, the model should predict similar results
center_1 = np.array([1,1])
center_2 = np.array([5,5])
center_3 = np.array([8,1])

# Generate random data and center it to the three centers
data_1 = np.random.randn(200,2) + center_1
data_2 = np.random.randn(200,2) + center_2
data_3 = np.random.randn(200,2) + center_3
data = np.concatenate((data_1, data_2, data_3), axis = 0)

plt.scatter(data[:,0], data[:,1], s=7)
plt.show()

#create K-mean algorithm
# Number of clusters
k = 3
```

```

# Number of training data
n = data.shape[0]
# Number of features in the data
c = data.shape[1]

mean = np.mean(data, axis = 0)
std = np.std(data, axis = 0)
centers = np.random.randn(k,c)*std + mean

# Plot the data and the centers generated as random
plt.scatter(data[:,0], data[:,1], s=7)
plt.scatter(centers[:,0], centers[:,1], marker='*', c='g', s=150)
plt.show()
centers_old = np.zeros(centers.shape) #to store old centers
centers_new = deepcopy(centers) #Store new centers

data.shape
clusters = np.zeros(n)
distances = np.zeros((n,k))

error = np.linalg.norm(centers_new - centers_old)

# When, after an update, the estimate of that center stays the same, exit loop
while error != 0:
    # Measure the distance to every center
    for i in range(k):
        distances[:,i] = np.linalg.norm(data - centers[i], axis=1)
    # Assign all training data to closest center
    clusters = np.argmin(distances, axis = 1)
    centers_old = deepcopy(centers_new)
    for i in range(k):
        centers_new[i] = np.mean(data[clusters == i], axis=0)
    error = np.linalg.norm(centers_new - centers_old)
centers_new

# Plot the data and the centers generated as random
plt.scatter(data[:,0], data[:,1], s=7)

```

```
plt.scatter(centers_new[:,0],centers_new[:,1],marker='*',c='g',s=150)
plt.show()
```

## OUTPUT:

