UNIVERSITY OF MUMBAI

PRACTICAL REPORT M.Sc. I.T (PART 2) 2023-2024

PAPER 1: <u>APPLIED ARTIFICIAL INTELLIGENCE</u>

PAPER 2: MACHINE LEARNING

PAPER 3: ROBOTIC PROCESS AUTOMATION

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Padmashri Annasaheb Jadhav Bhartiya Samaj Unnati Mandal's B.N.N COLLEGE

(Arts, Science, Commerce & Self-Funded Course)
(Affiliated to University of Mumbai)
DEPARTMENT OF INFORMATION TECHNOLOGY
BHIWANDI, MAHARASHTRA-421302



Padmashri Annasaheb Jadhav Bhartiya Samaj Unnati Mandal's B.N.N College of Arts, Science and Commerce, Bhiwandi. (Self-Funded Course) (Department of Information Technology)

CERTIFICATE

This is to certify that Mr./Miss. Snehal Nagesh Bachewar .			
PRN No. <u>2016016402206486</u> Class: <u>MSc-IT</u>	Exam Seat No. 2294989 .		
has satisfactorily completed practical in Applied Artificial Intelligence .			
As laid down in the regulation of University of M	umbai for the purpose of		
Semester- III Practical Examination 2023-2024.			
Date:			
Place: Bhiwandi			
In-Charge Professor	Signature of External Examiner		
In-Charge Froiessor	Signature of External Examiner		
	_		
Signature of HOD			



Padmashri Annasaheb Jadhav Bhartiya Samaj Unnati Mandal's B.N.N College of Arts, Science and Commerce, Bhiwandi. (Self-Funded Course) (Department of Information Technology)

CERTIFICATE

This is to certify that Mr./Miss. Snehal Nagesh Bachewar.

PRN No. 2016016402206486 Class: MSc-IT Exam Seat No. 2294989 ... has satisfactorily completed practical in Machine Learning ... As laid down in the regulation of University of Mumbai for the purpose of Semester- III Practical Examination 2023-2024.

Date: Place: Bhiwandi ... Signature of External Examiner ... Signature of HOD



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PRN No. 2016016402206486 Class: MSc-IT Exam Seat No. 2294989

has satisfactorily completed practical in Robotic Process Automation

As laid down in the regulation of University of Mumbai for the purpose of Semester- III Practical Examination 2023-2024.

Date:
Place: Bhiwandi

In-Charge Professor Signature of External Examiner

MSC-IT PART-2	Applied Artificial Intelligence	SEAT NUMBER:- 2294989
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PRATICAL 1) Design a bot using AIML

```
CODE: We need to create following three files.
```

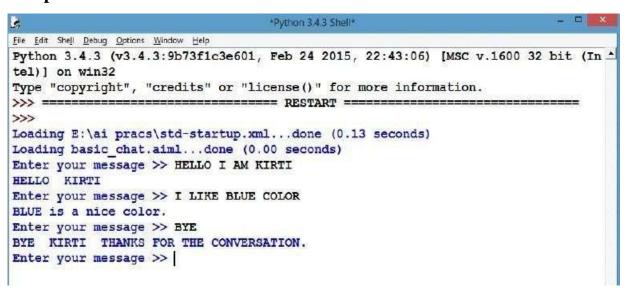
```
1) std-startup.xml
```

2) basic_chat.aiml

```
<aiml version="1.0.1" encoding="UTF-8">
<!-- basic_chat.aiml -->
<category>
       <pattern>HELLO I AM *</pattern>
       <template> HELLO <set name="username"> <star/> </set> </template>
</category>
<category>
       <pattern>I LIKE * COLOR</pattern>
       <template><star index="1"/> is a nice color.</template>
</category>
<category>
       <pattern>BYE</pattern>
       <template> BYE <get name="username"/> THANKS FOR THECONVERSATION.
</template>
</category>
</aiml>
```

3) testbot.py

```
import aiml
kernel = aiml.Kernel()
kernel.learn("F:\\aaipracts\\std-startup.xml")
kernel.learn("F:\\aaipracts\\basic_chat.aiml")
kernel.respond("LOAD")
# Press CTRL-C to break this loopwhile
True:
    print( kernel.respond(input("Enter your message >> "))))
```



PRATICAL 2) Implement Bayes Theorem using Python

```
# calculate the probability of cancer patient and diagnostic test
# calculate P(A|B) given P(A), P(B|A), P(B|not A)
 def bayes_theorem(p_a, p_b_given_a, p_b_given_not_a):
# calculate P(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_agiven_b = (p_bgiven_a * p_a) / pbreturn
        p_a_given_b
# P(A)
p_a = 0.0002
\# P(B|A)
p_b_given_a = 0.85
\# P(B|not A)
p_b_given_not_a = 0.05 #
calculate P(A|B)
result = bayes_theorem(p_a, p_b_given_a, p_b_given_not_a)
# summarize
print(P(A|B) = \%.3f\%\%'\% (result * 100))
```

Running this program calculates the probability that a patient has cancer.

```
File Edit Shell Debug Options Window Help

Python 3.7.0 (v3.7.0:1bf9cc5093, Jun 27 2018, 04:59:51) [MSC v.1914 64 bit (AMD64 Type "copyright", "credits" or "license()" for more information.

>>>

=== RESTART: C:/Users/Anam/AppData/Local/Programs/Python/Python37/bayes.py ===
P(A|B) = 0.339%

>>>
```

PRATICAL 3) Implement Conditional Probability and joint probability using Python.

A) Code for conditional probability:

```
pofB=float(input("Enter number of C programmers in percentage "))
pofAandB=float(input("Enter number of C and Java programmers in percentage "))
pofB=pofB/100
pofAandB=pofAandB/100
print("Event A that student is Java Progammer=?")
print("Event B that student is C Progammer=",pofB)
print("Event A and B that is student knowing both C and Java is =",pofAandB)
print("Lets Calculate P(A|B) = P(A \text{ and } B) / P(B)")
pAgivenB=pofAandB / pofB
print("P(A|B)=",pAgivenB)
print("There are ",pAgivenB *100," % chances that the student that knows C also knowsJava")
```

```
Python 3.43 Shell
File Edit Shell Debug Qotions Window Help
Python 3.4.3 (v3.4.3:9b73f1c3e601, Feb 24 2015, 22:43:06) [MSC v.1600 32 bit (Intel)] c
Type "copyright", "credits" or "license()" for more information.
>>>
Enter number of C programmers in percentage 80
Enter number of C and Java programmers in percentage 40
Event A that student is Java Progammer=?
Event B that student is C Progammer= 0.8
Event A and B that is student knowing both C and Java is = 0.4
Lets Calculate P(A|B) = P(A and B) / P(B)
P(A|B) = 0.5
There are 50.0 % chances that the student that knows C also knows Java
>>>
```

```
B)Code for Joint Probability:

cardnumber=input("Enter number of Card")

pofA=4/52

pofB=26/52

print("p(A)=>Probability of drawing card

with number

",cardnumber,"=",round(pofA,2))

print("p(B)=>Probability of drawing card with color ",cardcolor," =",round(pofB,2))

print("Joint Probablity of A and B=P(A) * P(B)")

pAandB=round(pofA pofB,2)

print("P(A and B)=",pAandB)

print("There are ",pAandB *100," % chances that of getting ",cardcolor, " card with number", cardnumber)
```

PRATICAL 4) Design a Fuzzy based application using Python / \boldsymbol{R} .

```
Code:
elt=['w','x','y','z']
A=[0.5,0.4,0.3,0.2]
B=[0.2,0.1,0.2,1]
U=[]
print("elements=",elt)
print("set A=",A)
 print("set B=",B)
for i in range(0,4):
   if A[i]>B[i]:
       U.append(A[i])
    else:
      U.append(B[i])
print("Union")
for i in range(0,3):
   print(U[i],"/",elt[i],end='+')
   for i in range(3,4):
   print(U[i],"/",elt[i],end='')
print()
I=[]
 for i in range(0,4):if
    A[i] < B[i]:
       I.append(A[i])
    else:
      I.append(B[i])
print()
print("Intersection")
      for i in range(0,3):
   print(I[i],"/",elt[i],end='+')
   for i in range(3,4):
    print(I[i],"/",elt[i],end=' ')
print()
J=[]
K=[]
C=[1,1,1,1]
print()
print("Complement of A")
for i in range(0,4):
J.append(C[i]-A[i])
output=round(J[i],2)
for i in range(0,3):
print(J[i] ,"/",elt[i],end=' + ')
for i in range(3,4):
print(J[i],"/",elt[i],end=' ')
print()
print()
print("Complement of B")
for i in range(0,4):
K.append(C[i]-B[i])
```

```
for i in range(0,3):
   print(K[i],"/",elt[i],end='+')
   for i in range(3,4):
   print(K[i],"/",elt[i],end=' ')
   L=[]
M=[]
print()
 for i in range(0,4):if
    A[i] < K[i]:
L.append(A[i])
       else:
      L.append(K[i])
print()
print("Difference of A/B")
for i in range(0,3):
   print(L[i],"/",elt[i],end=' + ')
   for i in range(3,4):
   print(L[i],"/",elt[i],end=' ')
   for i in range(0,4):
    if B[i] < J[i]:
       M.append(A[i])
    else:
      M.append(J[i])
print()
print("Difference of B/A")
for i in range(0,3):
   print(M[i],"/",elt[i],end=' + ')
   for i in range(3,4):
   print(M[i],"/",elt[i],end=' ')
   print()
Sum=[]
Sum1=[]
print()
print("Sum of A and B")
for i in range(0,4):
    Sum.append(A[i]+B[i])
    output=round(Sum[i],2)
    Sum1.append(output)
for i in range(0,3):
   print(Sum1[i],"/",elt[i],end=' + ')
   for i in range(3,4):
   print(Sum1[i] ,"/",elt[i],end=' ')
   print()
Prod=[]
Prod1=[]
print()
print("Product of A and B")
for i in range(0,4):
    Prod.append(A[i]*B[i])
```

```
output=round(Prod[i],2)
  Prod1.append(output)
for i in range(0,3):
  print(Prod1[i] ,"/",elt[i],end=' + ')
  for i in range(3,4):
    print(Prod1[i] ,"/",elt[i],end=' ')
```

```
Output:
Python 3.7.0 Shell
<u>File Edit Shell Debug Options Window Help</u>
Python 3.7.0 (v3.7.0:1bf9cc5093, Jun 27 2018, 04:59:51) [MSC v.
Type "copyright", "credits" or "license()" for more information
>>>
RESTART: C:\Users\Anam\AppData\Local\Programs\Python\Python37\
elements= ['w', 'x', 'y', 'z']
set A= [0.5, 0.4, 0.3, 0.2]
set B=[0.2, 0.1, 0.2, 1]
Union
0.5 / w + 0.4 / x + 0.3 / y + 1 / z
Intersection
0.2 / w + 0.1 / x + 0.2 / y + 0.2 / z
Complement of A
0.5 / w + 0.6 / x + 0.7 / y + 0.8 / z
Complement of B
0.8 / w + 0.9 / x + 0.8 / y + 0 / z
Difference of A/B
0.5 / w + 0.4 / x + 0.3 / y + 0 / z
Difference of B/A
0.5 / w + 0.4 / x + 0.3 / y + 0.8 / z
Sum of A and B
0.7 / w + 0.5 / x + 0.5 / y + 1.2 / z
Product of A and B
0.1 / w + 0.04 / x + 0.06 / y + 0.2 / z
>>>
```

PRATICAL 5) Write an application to implement clustering algorithm. K-Means Clustering:

Code:

```
newiris <- iris
newiris$Species <- NULL
(kc <- kmeans(newiris,3))
print(kc)
# Compare the Species label with the clustering result.table
(iris$Species,kc$cluster)
plot(newiris[c("Sepal.Length","Sepal.Width")],col=kc$cluster)
points(kc$centers[,c("Sepal.Length","Sepal.Width")],col=1:3,pch=8,cex=2)
```

```
R Console
                                                          > # Apply E mean to iris and store result
> newiris <- iris
> newiris$Species <- NULL
> (kc <- kmeans(newiris,3))
K-means clustering with 3 clusters of sizes 62, 38, 50
Cluster means:
 Sepal.Length Sepal.Width Petal.Length Petal.Width
   5.901613 2.748387 4.393548 1.433871
   6.850000 3.073684 5.742105 2.071053
   5.006000 3.428000 1.462000 0.246000
Clustering vector:
 [139] 1 2 2 2 1 2 2 2 1 2 2 1
Within cluster sum of squares by cluster:
[1] 39.82097 23.87947 15.15100
 (between SS / total SS = 88.4 %)
Available components:
[1] "cluster"
            "centers"
                     "totss"
                               "withinss"
                                         "tot.withinss" "betweenss"
[7] "size"
            "iter"
                     "ifault"
```

```
R Console

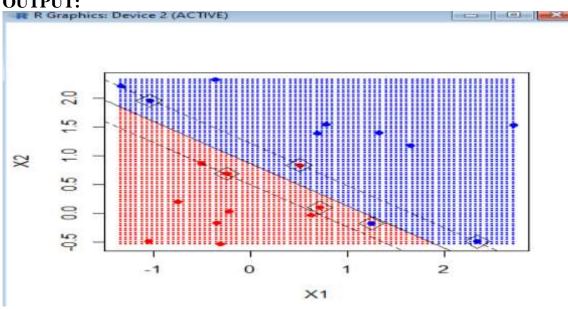
> #Compare the Species label with the clustering result
> table (iris$Species, kc$cluster)

1 2 3
setosa 0 0 50
versicolor 48 2 0
virginica 14 36 0
> |
```

PRATICAL 6) Write an application to implement clustering algorithm. Code:

```
set.seed(10111)
x=matrix(rnorm(40),20,2)
y=rep(c(-1,1),c(10,10))
x[y==1,]=x[y==1,]+1
plot(x,col=y+3,pch=19)
# Now you load the package e1071 which contains the sym function. > library(e1071)
dat = data.frame(x, y = as.factor(y))
svmfit = svm(y \sim ., data = dat, kernel = "linear", cost = 10, scale = FALSE)
print(svmfit)
plot(symfit, dat)
make.grid = function(x, n = 75)
\{grange = apply(x, 2, range)\}
x1 = seq(from = grange[1,1], to = grange[2,1], length = n) x2 = seq(from = grange[1,2], to = grange[2,2],
length = n)expand.grid(X1 = x1, X2 = x2)}
xgrid = make.grid(x)
xgrid[1:10,]
vgrid = predict(symfit, xgrid)
plot(xgrid, col = c("red","blue")[as.numeric(ygrid)], pch = 20, cex = .2)
points(x, col = y + 3, pch = 19)
points(x[svmfit$index,], pch = 5, cex = 2)
beta = drop(t(symfit$coefs)%*%x[symfit$index,])
beta0 = symfit$rho
plot(xgrid, col = c("red", "blue")[as.numeric(ygrid)], pch = 20, cex = .2)
points(x, col = y + 3, pch = 19)
points(x[svmfit$index,], pch = 5, cex = 2)
abline(beta0 / beta[2], -beta[1] / beta[2])
abline((beta 0 - 1) / beta[2], -beta[1] / beta[2], lty = 2)
abline((beta0 + 1) / beta[2], -beta[1] / beta[2], lty = 2)
```

OUTPUT:



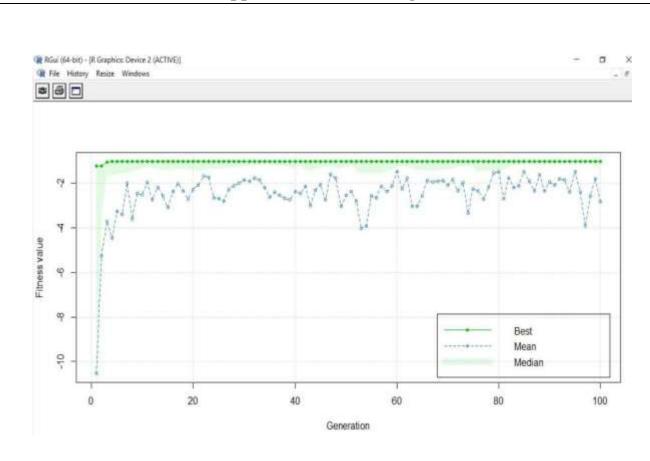
PRATICAL 7) Simulate genetic algorithm with suitable example using Python / R or another platform.

Code:

Packages: GA, Rccp, RcppArmadillo, Cli, Crayon, assertthat

```
f <- \ function(x) \ abs(x) + cos(x) curve(f, -20, 20) fitness <- \ function(x) - f(x) GA <- \ ga(type = "real-valued", fitness = fitness, lower = -20, upper = 20) summary(GA) plot(GA)
```

```
RGui (64-bit) - [R Console]
R File Edit View Misc Packages Windows Help
GA | iter = 93 | Mean = -1.845089 | Best = -1.000036
GA | iter = 94 | Mean = -2.382493 | Best = -1.000036
GA | iter = 95 | Mean = -1.476980 | Best = -1.000036
GA | iter = 96 | Mean = -2.386826 | Best = -1.000036
GA | iter = 97 | Mean = -3.910274 | Best = -1.000036
GA | iter = 98 | Mean = -2.573000 | Best = -1.000036
GA | iter = 99 | Mean = -1.793797 | Best = -1.000036
GA | iter = 100 | Mean = -2.833497 | Best = -1.000036
> summary (GA)
-- Genetic Algorithm -----
GA settings:
Type = real-valued Population size = 50
Number of generations = 100
Elitism = 2
Crossover probability = 0.8
Mutation probability = 0.1
Search domain =
       x I
lower -20
upper 20
GA results:
Iterations
                       = 100
Fitness function value = -1.000036
Solution =
[1,] -3.639934e-05
> plot (GA)
```



MSC-IT PART-2	Machine Learning	SEAT NUMBER:- 2294989
	MACHINE LEA	RNING
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	Matrix.		

PRATICAL 1A) Design a simple machine learning model to train the training instances and test the same.

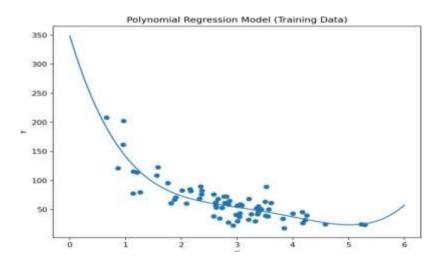
```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.metrics import r2_score
# Set a random seed for reproducibility
np.random.seed(2)
# Generate random data
x = np.random.normal(3, 1, 100)
y = np.random.normal(150, 40, 100) / x
# Visualize the data
plt.figure(figsize=(8, 6))
plt.scatter(x, y)
plt.title("Scatter Plot of Data")
plt.xlabel("X")
plt.ylabel("Y")
plt.show()
# Split the data into training and testing sets
train_x, test_x, train_y, test_y = train_test_split(x, y, test_size=0.3)
plt.scatter(train_x,train_y)
plt.show()
plt.scatter(test_x,test_y)
plt.show()
# Create and visualize a polynomial regression model for training data
degree = 4 # Adjust the polynomial degree as needed
train_model = np.poly1d(np.polyfit(train_x, train_y, degree))
myline = np.linspace(0, 6, 200)
plt.figure(figsize=(8, 6))
plt.scatter(train_x, train_y)
plt.plot(myline, train_model(myline))
plt.title("Polynomial Regression Model (Training Data)")
plt.xlabel("X")
plt.ylabel("Y")
plt.show()
# Calculate and print the R-squared score for the training data
r2_train = r2_score(train_y, train_model(train_x))
print("R-squared score for training data:", r2_train)
```

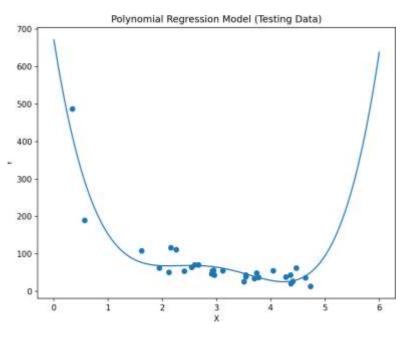
```
# Create and visualize a polynomial regression model for testing data test_model = np.poly1d(np.polyfit(test_x, test_y, degree))

plt.figure(figsize=(8, 6))
plt.scatter(test_x, test_y)
plt.plot(myline, test_model(myline))
plt.title("Polynomial Regression Model (Testing Data)")
plt.xlabel("X")
plt.ylabel("Y")
plt.ylabel("Y")
plt.show()

# Calculate and print the R-squared score for the testing data r2_test = r2_score(test_y, test_model(test_x))
print("R-squared score for testing data:", r2_test)
```

Make predictions using the model prediction = test_model(5) print("Prediction for x = 5:", prediction)





PRATICAL 1B) Implement and demonstrate the FIND-S algorithm for finding the most specific hypothesis based on a given set of training data samples. Read the training data from a .CSV file.

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.metrics import r2_score
# Set a random seed for reproducibility
np.random.seed(2)
# Generate random data
x = np.random.normal(3, 1, 100)
y = np.random.normal(150, 40, 100) / x
# Visualize the data
plt.figure(figsize=(8, 6))
plt.scatter(x, y)
plt.title("Scatter Plot of Data")
plt.xlabel("X")
plt.ylabel("Y")
plt.show()
# Split the data into training and testing sets
train_x, test_x, train_y, test_y = train_test_split(x, y, test_size=0.3)
plt.scatter(train_x,train_y)
plt.show()
plt.scatter(test_x,test_y)
plt.show()
# Create and visualize a polynomial regression model for training data
degree = 4 # Adjust the polynomial degree as needed
train_model = np.poly1d(np.polyfit(train_x, train_y, degree))
myline = np.linspace(0, 6, 200)
plt.figure(figsize=(8, 6))
plt.scatter(train_x, train_y)
plt.plot(myline, train_model(myline))
plt.title("Polynomial Regression Model (Training Data)")
plt.xlabel("X")
plt.ylabel("Y")
plt.show()
# Calculate and print the R-squared score for the training data
r2_train = r2_score(train_y, train_model(train_x))
print("R-squared score for training data:", r2_train)
```

```
# Create and visualize a polynomial regression model for testing data
test model = np.poly1d(np.polyfit(test x, test y, degree))
plt.figure(figsize=(8, 6))
plt.scatter(test_x, test_y)
plt.plot(myline, test_model(myline))
plt.title("Polynomial Regression Model (Testing Data)")
plt.xlabel("X")
plt.ylabel("Y")
plt.show()
# Calculate and print the R-squared score for the testing data
r2_test = r2_score(test_y, test_model(test_x))
print("R-squared score for testing data:", r2_test)
# Make predictions using the model
prediction = test_model(5)
print("Prediction for x = 5:", prediction)
```

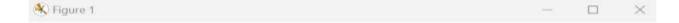
```
= RESTART: C:\Users\kamle\OneDrive\Desktop\ml\mlplb.py
< csv.reader object at 0x00000243F94B36A0>
The given training examples are:
['sky', 'airtemp', 'humidity', 'wind ', 'water', 'forecast', 'enjoysport']
['Sunny', 'Warm', 'Normal', 'Strong', 'Warm', 'Same', 'Yes']
['Sunny', 'Warm', 'High', 'Strong', 'Warm', 'Same', 'Yes']
['Rainy', 'Cold', 'High', 'Strong', 'Warm', 'Change', 'No']
['Sunny', 'Warm', 'High', 'Strong', 'Cool', 'Change', 'Yes']
The positive examples are:
['Sunny', 'Warm', 'Normal', 'Strong', 'Warm', 'Same', 'Yes']
['Sunny', 'Warm', 'High', 'Strong', 'Warm', 'Same', 'Yes']
['Sunny', 'Warm', 'High', 'Strong', 'Cool', 'Change', 'Yes']
The steps of the Find-s algorithm are :
 [181, 181, 181, 181, 181, 181]
['Sunny', 'Warm', 'Normal', 'Strong', 'Warm', 'Same']
['Sunny', 'Warm', '?', 'Strong', 'Warm', 'Same']
['Sunny', 'Warm', '?', 'Strong', '?', '?']
The maximally specific Find-s hypothesis for the given training examples is :
['Sunny', 'Warm', '?', 'Strong', '?', '?']
```

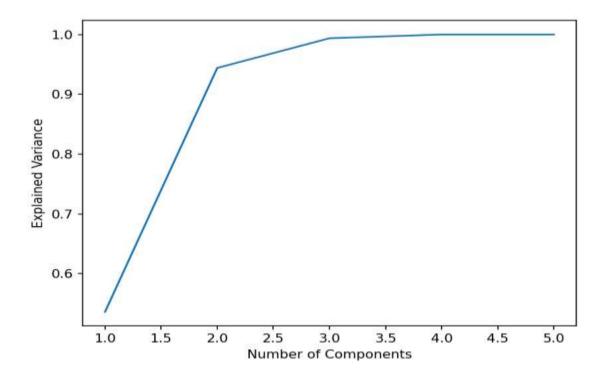
MSC-IT PART-2

PRATICAL 2A) Perform Data Loading, Feature selection (Principal Component analysis) and Feature Scoring and Ranking.

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.decomposition import PCA
# Data Loading
data = pd.read_csv("C:/Users/kamle/OneDrive/Desktop/ml/p2/data.csv") # Replace "data.csv" with your
dataset file
X = data.drop('target_column', axis=1) # Features
y = data['target_column'] # Target variable (if applicable)
# Feature Selection using PCA
pca = PCA()
pca.fit(X)
explained_var_ratio = pca.explained_variance_ratio_
# Plot the explained variance to decide on the number of components to keep
plt.plot(range(1, len(explained_var_ratio) + 1), explained_var_ratio.cumsum())
plt.xlabel('Number of Components')
plt.ylabel('Explained Variance')
plt.show()
# Choose the number of components that explain most of the variance
num_components = 5 # Adjust as needed
# Transform the data with the selected number of components
pca = PCA(n components=num components)
X_pca = pca.fit_transform(X)
# Feature Scoring and Ranking
loadings = pca.components_
feature_scores = abs(loadings).mean(axis=0) # Use mean absolute loading values
# Create a DataFrame to display feature scores
feature_scores_df = pd.DataFrame({ 'Feature': X.columns, 'Score': feature_scores})
# Sort features by score in descending order
feature scores df = feature scores df.sort values(by='Score', ascending=False)
# Display the ranked features
print(feature_scores_df)
```

Output:





```
= RESTART: C:\Users\kamle\OneDrive\Desktop\ml\p2\mlp2a.py
```

Feature Score
0 feature1 0.431647
4 feature5 0.384937
1 feature2 0.374804
3 feature4 0.328244
2 feature3 0.317932

·>>

PRATICAL 2B) For a given set of training data examples stored in a .CSV file, implement and demonstrate the CandidateElimination algorithm to output a description of the set of all hypotheses consistent with the training examples.

```
import numpy as np
import pandas as pd
# Loading Data from a CSV File
data = pd.DataFrame(data=pd.read_csv('C:/Users/kamle/OneDrive/Desktop/ml/p2/trainingData.csv'))
print(data)
# Separating concept features from Target
concepts = np.array(data.iloc[:,0:-1])
print(concepts)
# Isolating target into a separate DataFrame
# copying last column to target array
target = np.array(data.iloc[:,-1])
print(target)
def learn(concepts, target):
  learn() function implements the learning method of the Candidate elimination algorithm.
  Arguments:
     concepts - a data frame with all the features
     target - a data frame with corresponding output values
  # Initialise S0 with the first instance from concepts
  #.copy() makes sure a new list is created instead of just pointing to the same memory location
  specific_h = concepts[0].copy()
  print("\nInitialization of specific_h and general_h")
  print(specific_h)
  \#h = ["\#" \text{ for i in range}(0,5)]
  #print(h)
  general_h = [["?" for i in range(len(specific_h))] for i in range(len(specific_h))]
  print(general_h)
  # The learning iterations
  for i, h in enumerate(concepts):
     # Checking if the hypothesis has a positive target
     if target[i] == "Yes":
       for x in range(len(specific_h)):
          # Change values in S & G only if values change
          if h[x] != specific_h[x]:
            specific_h[x] = '?'
            general_h[x][x] = '?'
     # Checking if the hypothesis has a positive target
     if target[i] == "No":
       for x in range(len(specific_h)):
```

```
# For negative hyposthesis change values only in G
          if h[x] != specific h[x]:
            general_h[x][x] = specific_h[x]
          else:
             general h[x][x] = '?'
     print("\nSteps of Candidate Elimination Algorithm",i+1)
     print(specific_h)
     print(general_h)
  # find indices where we have empty rows, meaning those that are unchanged
  indices = [i for i, val in enumerate(general_h) if val == ['?', '?', '?', '?', '?', '?']]
  for i in indices:
     # remove those rows from general_h
     general_h.remove(['?', '?', '?', '?', '?', '?'])
  # Return final values
  return specific_h, general_h
s_final, g_final = learn(concepts, target)
print("\nFinal Specific_h:", s_final, sep="\n")
print("\nFinal General_h:", g_final, sep="\n")
```

```
= RESTART: C:\Users\kamle\OneDrive\Desktop\m1\p2\m1p2b.py
  sky airtemp humidity
              wind water forecast enjoysport
0 Sunny
                       Same
      Warm Normal Strong Warm
1 Sunny
      Warm
          High Strong Warm
                       Same
2 Rainy Cold
          High Strong Warm Change
         High Strong Cool Change
3 Sunny
     Warm
[['Sunny' 'Warm' 'Hormal' 'Strong' 'Warm' 'Same']
['Rainy' 'Cold' 'High' 'Strong' 'Warm' 'Change']
['Sunny' 'Warn' 'High' 'Strong' 'Cool' 'Change']]
['Yes' 'Yes' 'No' 'Yes']
Initialization of specific h and general h
Steps of Candidate Elimination Algorithm 1
Steps of Candidate Elimination Algorithm 2
['Sunny' 'Warm' '?' 'Strong' 'Warm' (['?', '?', '?', '?', '?', '?']) [''' '?', '?', '?', '?']]
              Steps of Candidate Elimination Algorithm 3
Steps of Candidate Elimination Algorithm 4
Final Specific h:
['Sunny' 'Warm' '?' 'Strong' '?' '?']
[['Sunny', '7', '7', '7', '7', '7'], ['7', 'Warn', '7', '7', '7', '7']]
```

PRATICAL 3A) Write a program to implement the naïve Bayesian classifier for a sample training data set stored as a .CSV file. Compute the accuracy of the classifier, considering few test data sets.

```
# import necessary libarities
import pandas as pd
from sklearn import tree
from sklearn.preprocessing import LabelEncoder
from sklearn.naive bayes import GaussianNB
# load data from CSV
data = pd.read_csv("C:/Users/kamle/OneDrive/Desktop/ml/p3/trainingData.csv")
print("THe first 5 values of data is :\n",data.head())
# obtain Train data and Train output
X = data.iloc[:,:-1]
print("\nThe First 5 values of train data is\n",X.head())
y = data.iloc[:,-1]
print("\nThe first 5 values of Train output is\n",y.head())
# Convert then in numbers
le outlook = LabelEncoder()
X.Outlook = le\_outlook.fit\_transform(X.Outlook)
le_Temperature = LabelEncoder()
X.Temperature = le_Temperature.fit_transform(X.Temperature)
le_Humidity = LabelEncoder()
X.Humidity = le_Humidity.fit_transform(X.Humidity)
le_Windy = LabelEncoder()
X.Windy = le_Windy.fit_transform(X.Windy)
print("\nNow the Train data is :\n",X.head())
le_PlayTennis = LabelEncoder()
y = le_PlayTennis.fit_transform(y)
print("\nNow the Train output is\n",y)
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.20)
classifier = GaussianNB()
classifier.fit(X_train,y_train)
from sklearn.metrics import accuracy_score
print("Accuracy is:",accuracy_score(classifier.predict(X_test),y_test))
```

```
THe first 5 values of data is :
    sky airtemp humidity wind water forecast enjoysport
         Warm Normal Strong Warm Same
0 Sunny
1 Sunny Warm High Strong Warm Same
2 Rainy Cold High Strong Warm Change
3 Sunny Warm High Strong Cool Change
                                                 Yes
                                                  No
                                                 Yes
The First 5 values of train data is
   sky airtemp humidity wind water forecast
0 Sunny Warm Normal Strong Warm Same
         Warm High Strong Warm
1 Sunny
                                      Same
2 Rainy Cold
                 High Strong Warm Change
3 Sunny Warm High Strong Cool Change
The first 5 values of Train output is
0
    Yes
1
    Yes
2
    No
3
    Yes
Name: enjoysport, dtype: object
```

PRATICAL 3B) Write a program to implement Decision Tree and Random forest with Prediction, Test Score and Confusion Matrix.

```
# Import pandas library
import pandas as pd
# Loading dataset
df = pd.read_csv("C:/Users/kamle/OneDrive/Desktop/ml/p3/diabetes_dataset.csv")
df.head()
# Feature variables
x = df.drop(['Outcome'], axis=1)
# Target variable
y = df.Outcome
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split
# Split the dataset
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42)
# Create Decision Tree classifier object
model = DecisionTreeClassifier()
# Train Decision Tree Classifier
model.fit(x_train, y_train)
# Predict the response for the test dataset
y_pred = model.predict(x_test)
# Evaluation using Accuracy score
from sklearn import metrics
print("Accuracy:", metrics.accuracy_score(y_test, y_pred) * 100)
# Evaluation using Confusion matrix
from sklearn.metrics import confusion_matrix
cm = confusion matrix(y test, y pred)
print("Confusion Matrix:")
print(cm)
# Accuracy calculation from Confusion matrix
accuracy = (cm[0, 0] + cm[1, 1]) / sum(sum(cm))
print("Accuracy from Confusion Matrix:", accuracy * 100)
# Evaluation using Classification report
from sklearn.metrics import classification_report
print("Classification Report:")
print(classification_report(y_test, y_pred))
```

```
# Checking prediction value
prediction = model.predict([[6, 148, 72, 35, 0, 33.6, 0.627, 50]])
print("Prediction for input:", prediction)
```

Import modules for Visualizing Decision trees from sklearn.tree import export_graphviz from sklearn.tree import plot_tree import matplotlib.pyplot as plt

Visualizing Decision Tree plt.figure(figsize=(20, 10)) plot_tree(model, feature_names=x.columns, class_names=['0', '1'], filled=True, rounded=True) plt.show()

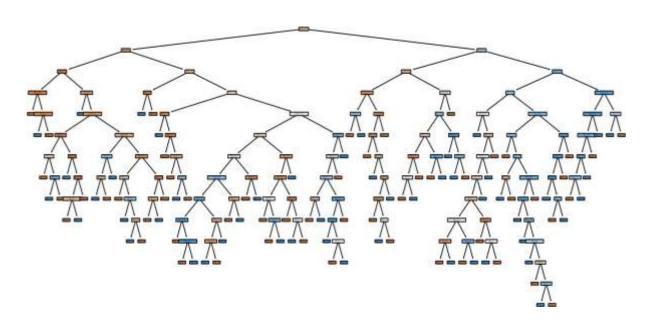
Create Decision Tree classifier object with entropy and max_depth model = DecisionTreeClassifier(criterion="entropy", max_depth=3)

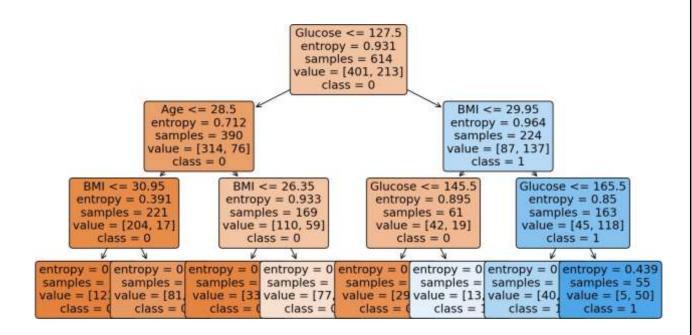
Train Decision Tree Classifier
model.fit(x_train, y_train)

Predict the response for the test dataset
y_pred = model.predict(x_test)

Model Accuracy
print("Accuracy:", metrics.accuracy_score(y_test, y_pred) * 100)

Better Decision Tree Visualization plt.figure(figsize=(20, 10)) plot_tree(model, feature_names=x.columns, class_names=['0', '1'], filled=True, rounded=True) plt.show()





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Accuracy: 75.32467532467533

Confusion Matrix:

[[76 23] [15 40]]

Accuracy from Confusion Matrix: 75.32467532467533

Classification Report:

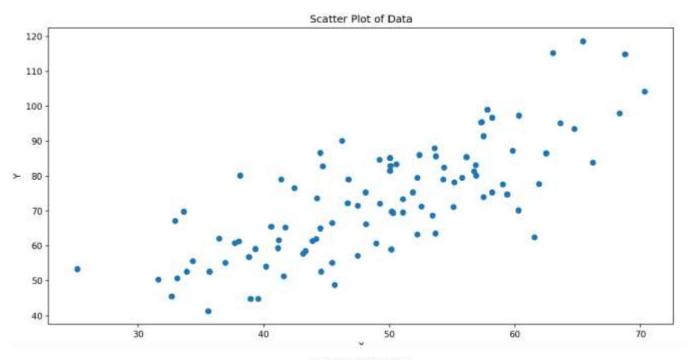
	precision	recall	f1-score	support
0	0.84	0.77	0.80	99
1	0.63	0.73	0.68	55
accuracy			0.75	154
macro avg	0.74	0.75	0.74	154
weighted avg	0.76	0.75	0.76	154

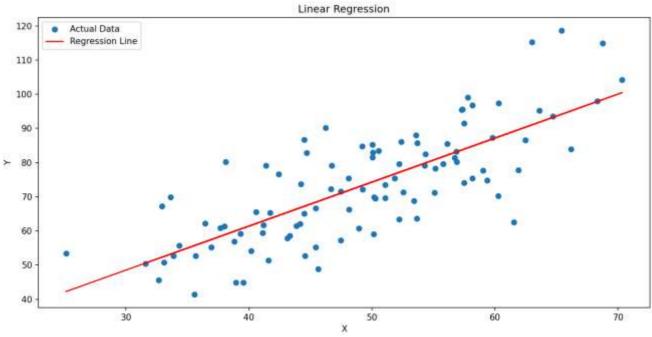
Prediction for input: [1]
Accuracy: 76.62337662337663

PRATICAL 4A) For a given set of training data examples stored in a .CSV file implement Least Square Regression algorithm.

```
# Importing Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
# Set the plot size
plt.rcParams['figure.figsize'] = (12.0, 9.0)
# Preprocessing Input Data
data = pd.read_csv('C:/Users/kamle/Downloads/ML/prac4/data.csv')
# Extracting the features X and Y
X = data.iloc[:, 0].values
Y = data.iloc[:, 1].values
# Plotting the data points
plt.scatter(X, Y)
plt.xlabel('X')
plt.ylabel('Y')
plt.title('Scatter Plot of Data')
plt.show()
# Building the model
X_{mean} = np.mean(X)
Y_mean = np.mean(Y)
num = 0
den = 0
for i in range(len(X)):
  num += (X[i] - X_mean) * (Y[i] - Y_mean)
  den += (X[i] - X_mean) ** 2
m = num / den
c = Y_mean - m * X_mean
print("Slope (m):", m)
print("Intercept (c):", c)
# Making Predictions
Y_pred = m * X + c
# Plotting the regression line
plt.scatter(X, Y, label='Actual Data')
plt.plot(X, Y_pred, color='red', label='Regression Line')
plt.xlabel('X')
```

plt.ylabel('Y')
plt.title('Linear Regression')
plt.legend()
plt.show()





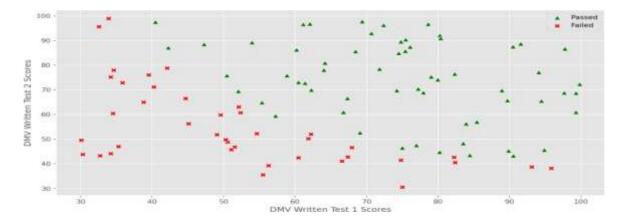
= RESTART: C:\Users\kamle\Downloads\ML\prac4\a. For a given set of training dat examples stored in a .CSV file implement Least Square Regression algorithm.py Slope (m): 1.2873573699494276
Intercept (c): 9.90860619348318

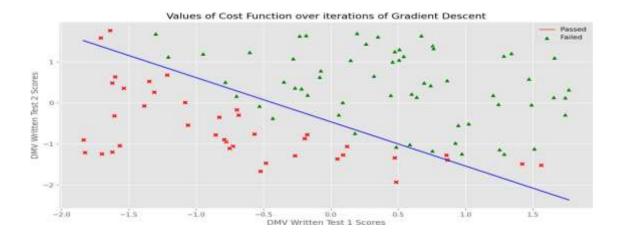
PRATICAL 4B) For a given set of training data examples stored in a .CSV file implement Logistic Regression algorithm.

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
plt.style.use("ggplot")
#matplotlib inline
from pylab import rcParams
rcParams['figure.figsize'] = 12, 8
data = pd.read csv("C:/Users/kamle/Downloads/ML/prac4/DMV Written Tests.csv")
data.head()
scores = data[['DMV_Test_1', 'DMV_Test_2']].values
results = data['Results'].values
passed = (results == 1).reshape(100, 1)
failed = (results == 0).reshape(100, 1)
ax = sns.scatterplot(x=scores[passed[:, 0], 0],
             y=scores[passed[:, 0], 1],
             marker="^",
             color="green",
             s=60)
sns.scatterplot(x=scores[failed[:, 0], 0],
         y=scores[failed[:, 0], 1],
         marker="X",
         color="red",
         s=60)
ax.set(xlabel="DMV Written Test 1 Scores", ylabel="DMV Written Test 2 Scores")
ax.legend(["Passed", "Failed"])
plt.show();
def logistic_function(x):
  return 1/(1 + np.exp(-x))
logistic_function(0)
def compute_cost(theta, x, y):
  m = len(y)
  y_pred = logistic_function(np.dot(x, theta))
  error = (y * np.log(y_pred)) + ((1 - y) * np.log(1 - y_pred))
  cost = -1 / m * sum(error)
  gradient = 1 / m * np.dot(x.transpose(), (y_pred - y))
```

```
return cost[0], gradient
mean_scores = np.mean(scores, axis=0)
std_scores = np.std(scores, axis=0)
scores = (scores - mean scores) / std scores # standardization
rows = scores.shape[0]
cols = scores.shape[1]
X = np.append(np.ones((rows, 1)), scores, axis=1) # include intercept
y = results.reshape(rows, 1)
theta_init = np.zeros((cols + 1, 1))
cost, gradient = compute_cost(theta_init, X, y)
print("Cost at initialization", cost)
print("Gradient at initialization:", gradient)
def gradient_descent(x, y, theta, alpha, iterations):
  costs = []
  for i in range(iterations):
     cost, gradient = compute_cost(theta, x, y)
     theta -= (alpha * gradient)
  return theta, costs
theta, costs = gradient_descent(X, y, theta_init, 1, 200)
print("Theta after running gradient descent:", theta)
plt.plot(costs)
plt.xlabel("Iterations")
plt.ylabel("$J(\\Theta)$")
plt.title("Values of Cost Function over iterations of Gradient Descent");
sns.scatterplot(x=X[passed[:, 0], 1],
          y=X[passed[:, 0], 2],
          marker="^",
          color="green",
          s=60)
ax = sns.scatterplot(x=X[failed[:, 0], 1],
            y=X[failed[:, 0], 2],
            marker="X",
            color="red",
            s=60)
ax.legend(["Passed", "Failed"])
ax.set(xlabel="DMV Written Test 1 Scores", ylabel="DMV Written Test 2 Scores")
x_boundary = np.array([np.min(X[:, 1]), np.max(X[:, 1])])
y_boundary = -(theta[0] + theta[1] * x_boundary) / theta[2]
sns.lineplot(x=x_boundary, y=y_boundary, color="blue")
plt.show();
```

Output:





```
Cost at initialization 0.693147180559946

Gradient at initialization: [[-0.1 ]
    [-0.28122914]
    [-0.25098615]]

Theta after running gradient descent: [[1.50850586]
    [3.5468762 ]
    [3.29383709]]

Training Accuracy: 89 %

A person who scores 50 and 79 on their DMV written tests have a 0.71 probability of passing.
```

PRATICAL 5A) 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.

```
import numpy as np
import math
import csv
def read_data(filename):
  with open(filename, 'r') as csvfile:
     datareader = csv.reader(csvfile, delimiter=',')
     headers = next(datareader)
     metadata = []
     traindata = []
     for name in headers:
       metadata.append(name)
     for row in datareader:
       traindata.append(row)
  return (metadata, traindata)
class Node:
  def init (self, attribute):
     self.attribute = attribute
     self.children = []
     self.answer = ""
  def__str_(self):
     return self.attribute
def subtables(data, col, delete):
  dict = \{ \}
  items, counts = np.unique(data[:, col], return_counts=True)
  for x in range(items.shape[0]):
     dict[items[x]] = data[data[:, col] == items[x]]
     if delete:
       dict[items[x]] = np.delete(dict[items[x]], col, 1)
  return items, dict
def entropy(S):
  items, counts = np.unique(S, return_counts=True)
  total size = S.size
  entropies = np.zeros(counts.shape)
  for x in range(items.shape[0]):
     ratio = counts[x] / total_size
```

```
entropies[x] = ratio * math.log(ratio, 2)
  return -np.sum(entropies)
def gain_ratio(data, col):
  items, dict = subtables(data, col, delete=False)
  total\_size = data.shape[0]
  entropies = np.zeros((items.shape[0], 1))
  intrinsic = np.zeros((items.shape[0], 1))
  for x in range(items.shape[0]):
     ratio = dict[items[x]].shape[0] / total_size
     entropies[x] = ratio * entropy(dict[items[x]][:, -1])
     intrinsic[x] = ratio * math.log(ratio, 2)
  total_entropy = entropy(data[:, -1])
  iv = -np.sum(intrinsic)
  for x in range(entropies.shape[0]):
     total_entropy -= entropies[x]
  return total_entropy / iv
def create_node(data, metadata):
  if data.size == 0:
     node = Node("")
     node.answer = "No data"
     return node
  if (np.unique(data[:, -1])).shape[0] == 1:
     node = Node("")
     node.answer = np.unique(data[:, -1])[0]
     return node
  gains = np.zeros((data.shape[1] - 1, 1))
  for col in range(data.shape[1] - 1):
     gains[col] = gain_ratio(data, col)
  split = np.argmax(gains)
  node = Node(metadata[split])
  metadata = np.delete(metadata, split, 0)
  items, dict = subtables(data, split, delete=True)
  for x in range(items.shape[0]):
     child = create_node(dict[items[x]], metadata)
     node.children.append((items[x], child))
  return node
```

```
def empty(size):
  s = ""
  for x in range(size):
    s += " "
  return s
def print_tree(node, level):
  if node.answer != "":
    print(empty(level), "Answer:", node.answer)
    return
  print(empty(level), "Attribute:", node.attribute)
  for value, n in node.children:
    print(empty(level + 1), f"Value: {value}")
    print_tree(n, level + 2)
metadata, traindata = read_data("C:/Users/kamle/Downloads/ML/prac5/tennisdata.csv")
data = np.array(traindata)
node = create_node(data, metadata)
print_tree(node, 0)
Output:
ppropriate data see for sarraing
 Attribute: Outlook
     Value: Overcast
        Answer: Yes
     Value: Rainy
        Attribute: Windy
            Value: False
                Answer: Yes
            Value: True
                Answer: No
     Value: Sunny
        Attribute: Humidity
            Value: High
                Answer: No
            Value: Normal
                Answer: Yes
```

PRATICAL 5B) Write a program to implement k-Nearest Neighbour algorithm to classify the iris data set.

```
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsClassifier
from sklearn import datasets
iris=datasets.load iris()
print("Iris Data set loaded...")
x_train, x_test, y_train, y_test = train_test_split(iris.data,iris.target,test_size=0.1)
#random_state=0
for i in range(len(iris.target_names)):
  print("Label", i , "-",str(iris.target_names[i]))
classifier = KNeighborsClassifier(n_neighbors=2)
classifier.fit(x train, y train)
y_pred=classifier.predict(x_test)
print("Results of Classification using K-nn with K=1")
for r in range(0, len(x_test)):
  print(" Sample:", str(x_test[r]), " Actual-label:", str(y_test[r])," Predicted-label:", str(y_pred[r]))
  print("Classification Accuracy:", classifier.score(x_test,y_test));
```

Output:

```
tranize trenizeaar tim (prace tri nrree a pregran e
Iris Data set loaded...
Label 0 - setosa
Label 1 - versicolor
Label 2 - virginica
Results of Classification using K-nn with K=1
Sample: [4.4 3.2 1.3 0.2] Actual-label: 0 Predicted-label: 0
Classification Accuracy: 1.0
Sample: [4.6 3.4 1.4 0.3]
                           Actual-label: 0 Predicted-label: 0
Classification Accuracy: 1.0
                           Actual-label: 1 Predicted-label: 1
Sample: [6.1 2.8 4.7 1.2]
Classification Accuracy: 1.0
                           Actual-label: 0 Predicted-label: 0
Sample: [5.4 3.7 1.5 0.2]
Classification Accuracy: 1.0
Sample: [6.3 3.3 4.7 1.6] Actual-label: 1 Predicted-label: 1
Classification Accuracy: 1.0
Sample: [5.7 4.4 1.5 0.4] Actual-label: 0 Predicted-label: 0
Classification Accuracy: 1.0
Sample: [6.4 2.8 5.6 2.1] Actual-label: 2 Predicted-label: 2
Classification Accuracy: 1.0
                          Actual-label: 1 Predicted-label: 1
Sample: [5.7 2.9 4.2 1.3]
Classification Accuracy: 1.0
Sample: [5. 3.5 1.6 0.6]
                           Actual-label: 0 Predicted-label: 0
Classification Accuracy: 1.0
 Sample: [5.2 3.4 1.4 0.2]
                           Actual-label: 0 Predicted-label: 0
Classification Accuracy: 1.0
Sample: [5.4 3.9 1.7 0.4]
                           Actual-label: 0 Predicted-label: 0
Classification Accuracy: 1.0
Sample: [6.3 2.7 4.9 1.8]
                          Actual-label: 2 Predicted-label: 2
Classification Accuracy: 1.0
Sample: [6.7 3. 5.2 2.3]
                           Actual-label: 2 Predicted-label: 2
Classification Accuracy: 1.0
Sample: [6.5 3. 5.5 1.8] Actual-label: 2 Predicted-label: 2
Classification Accuracy: 1.0
Sample: [5.7 3.8 1.7 0.3] Actual-label: 0 Predicted-label: 0
Classification Accuracy: 1.0
```

PRATICAL 6A) Implement the different Distance methods (Euclidean) with Prediction, Test Score and Confusion Matrix.

```
# Importing the libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
# Importing the dataset
dataset = pd.read_csv('C:/Users/kamle/Downloads/ML/prac6/Social_Network_Ads.csv')
X = dataset.iloc[:, [2, 3]].values
y = dataset.iloc[:, -1].values
# Splitting the dataset into the Training set and Test set
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20, random_state=0)
# Feature Scaling
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X train = sc.fit transform(X train)
X_{test} = sc.transform(X_{test})
# Training the K-NN model on the Training set
from sklearn.neighbors import KNeighborsClassifier
classifier = KNeighborsClassifier(n_neighbors=5, metric='minkowski', p=2)
classifier.fit(X train, y train)
# Predicting the Test set results
y_pred = classifier.predict(X_test)
# Making the Confusion Matrix
from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
ac = accuracy_score(y_test, y_pred)
# Print the results
print("Confusion Matrix:")
print(cm)
print("\nAccuracy Score:", ac)
Output:
 Confusion Matrix:
  [[55 3]
   [ 1 21]]
 Accuracy Score: 0.95
```

PRATICAL 6B) Implement the classification model using clustering for the following techniques with K means clustering with Prediction, Test Score and Confusion Matrix

```
# read in the iris data
from sklearn.datasets import load_iris
iris = load_iris()
# create X (features) and y (response)
X = iris.data
y = iris.target
# Import the class
from sklearn.linear_model import LogisticRegression
# instantiate the model (using the default parameters)
logreg = LogisticRegression()
# fit the model with data
logreg.fit(X, y)
# predict the response values for the observations in X
logreg.predict(X)
# Store the predicted response values
y_pred = logreg.predict(X)
# Check how many Predictions were generated
len(y_pred)
# Computer classification accuracy for the logistic regression model
from sklearn import metrics
print(metrics.accuracy_score(y, y_pred))
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(X, y)
y_pred = knn.predict(X)
print(metrics.accuracy_score(y, y_pred))
knn = KNeighborsClassifier(n_neighbors=1)
knn.fit(X, y)
y_pred = knn.predict(X)
print(metrics.accuracy_score(y, y_pred))
# print the shapes of X and y
# X is our features matrix with 150 x 4 dimensions
print(X.shape)
# y is our response vector with 150 x 1 dimension
print(y.shape)
```

```
# STEP 1: split X and y into training and testing sets
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=4)
# print the shapes of the new X objects
print(X_train.shape)
print(X_test.shape)
# print the shapes of the new y objects
print(y_train.shape)
print(y_test.shape)
# STEP 2: train the model on the training set
logreg = LogisticRegression()
logreg.fit(X_train, y_train)
# STEP 3: make predictions on the testing set
y_pred = logreg.predict(X_test)
# Computer actual response values (y_test) with predicted response values (y_pred)
print(metrics.accuracy_score(y_test, y_pred))
knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(X_train, y_train)
y_pred = knn.predict(X_test)
print(metrics.accuracy_score(y_test, y_pred))
# try k-1 through k-25 and record testing accuracy
k_range = range(1, 26)
# we can create python dictionary using [] or dict()
scores = []
# we use a loop through the range 1 to 26
# we append the scores in the dictionary
for k in k_range:
  knn = KNeighborsClassifier(n_neighbors=k)
  knn.fit(X_train, y_train)
  y_pred = knn.predict(X_test)
  scores.append(metrics.accuracy_score(y_test, y_pred))
print(scores)
# import Matplotlib (scientific plotting library)
import matplotlib.pyplot as plt
# plot the relationship between k and testing accuracy
plt.plot(k_range, scores)
plt.xlabel('Value of k for KNN')
plt.ylabel('Testing Accuracy')
```

```
# instantiate the model with the best-known parameters
knn = KNeighborsClassifier(n_neighbors=11)

# train the model with X and y (not X_train and y_train)
knn.fit(X, y)

# make a prediction for an out-of-sample observation
knn.predict([[3, 5, 4, 3]])
```

Output:

```
0.9666666666667
1.0
(150, 4)
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```

PRATICAL 7A) Implement the classification model using clustering for the following techniques with hierarchical clustering with Prediction, Test Score and Confusion Matrix.

```
import matplotlib.pyplot as plt
import pandas as pd
import scipy.cluster.hierarchy as sch
from sklearn.cluster import AgglomerativeClustering
dataset = pd.read_csv('C:/Users/kamle/Downloads/ML/prac7/abalone_csv.csv')
X = dataset.iloc[:, [3, 4]].values
# Dendrogram
dendrogram = sch.dendrogram(sch.linkage(X, method="ward"))
plt.title('Dendrogram')
plt.xlabel('Sample Index')
plt.ylabel('Euclidean Distances')
plt.show()
# Hierarchical Clustering
hc = AgglomerativeClustering(n clusters=5, affinity='euclidean', linkage='ward')
y hc = hc.fit predict(X)
print("Prediction Values: ", y_hc)
# Scatter Plot
plt.scatter(X[y_hc == 0, 0], X[y_hc == 0, 1], s=100, c='red', label='Cluster 1')
plt.scatter(X[y_hc == 1, 0], X[y_hc == 1, 1], s=100, c='blue', label='Cluster 2')
plt.scatter(X[y_hc == 2, 0], X[y_hc == 2, 1], s=100, c='green', label='Cluster 3')
plt.scatter(X[y_hc == 3, 0], X[y_hc == 3, 1], s=100, c='cyan', label='Cluster 4')
plt.scatter(X[y_hc == 4, 0], X[y_hc == 4, 1], s=100, c='magenta', label='Cluster 5')
# Plot Settings
plt.title('Clusters of Abalones (Hierarchical Clustering Model)')
plt.xlabel('Annual income (k$)') # Update with appropriate label
plt.ylabel('Spending score(1 to 100)') # Update with appropriate label
plt.legend()
plt.show()
Output:
```

PRATICAL 7B) Perform Text pre-processing, Text clustering, classification with Prediction, Test Score and Confusion Matrix.

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
# csv_file = 'Restaurant_Reviews.tsv'
csv_file = 'C:/Users/kamle/Downloads/ML/prac7/Restaurant_Reviews.tsv'
dataset = pd.read_csv(csv_file, delimiter='\t', quoting=3)
# Text preprocessing using Natural Language Toolkit (nltk)
import re
import nltk
nltk.download('stopwords')
from nltk.corpus import stopwords
from nltk.stem.porter import PorterStemmer
corpus = []
for i in range(0, 1000):
  review = re.sub('[^a-zA-Z]', ", dataset['Review'][i]) # Remove non-alphabetic characters
  review = review.lower() # Convert to lowercase
  review = review.split() # Tokenize the words
  ps = PorterStemmer() # Stemming and removing stopwords
  review = [ps.stem(word) for word in review if not word in set(stopwords.words('english'))]
  # Join the words to form the processed review
  review = ' '.join(review)
  corpus.append(review)
# Creating the bag of words model
from sklearn.feature extraction.text import CountVectorizer
cv = CountVectorizer(max_features=1500)
X = \text{cv.fit\_transform(corpus).toarray()}
Y = dataset.iloc[:, 1].values
# Splitting the dataset into the training set and test set
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.25, random_state=100)
# Fitting Naive Bayes to the training set
from sklearn.naive_bayes import GaussianNB
classifier = GaussianNB()
classifier.fit(X_train, Y_train)
# Predicting the test set results
```

```
Machine Learning
MSC-IT PART-2
Y_pred = classifier.predict(X_test)
# Model Accuracy
from sklearn import metrics
from sklearn.metrics import confusion_matrix
print("Accuracy:", metrics.accuracy_score(Y_test, Y_pred))
# Making the confusion matrix
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(Y_test, Y_pred)
print(cm)
Output:
  Accuracy: 0.54
  [[ 1 115]
   [ 0 134]]
PRATICAL 8A) Write a program to construct a Bayesian network considering
medical data. Use this model to demonstrate the diagnosis of heart patients using
standard Heart Disease Data Set.
import numpy as np
import pandas as pd
from pgmpy.models import BayesianModel
from pgmpy.estimators import MaximumLikelihoodEstimator
from pgmpy.inference import VariableElimination
# Read Cleveland Heart Disease data
heartDisease = pd.read_csv('C:/Users/kamle/Downloads/ML/prac8/heart.csv')
heartDisease = heartDisease.replace('?', np.nan)
# Display the data
print('Few examples from the dataset are given below')
print(heartDisease.head())
```

Model Bayesian Network

Inferencing with Bayesian Network

print(q['heartdisease'])

print('\nInferencing with Bayesian Network:') HeartDisease_infer = VariableElimination(model)

model = BayesianModel([('age', 'trestbps'), ('age', 'fbs'),

Learning CPDs using Maximum Likelihood Estimators print('\nLearning CPD using Maximum likelihood estimators') model.fit(heartDisease, estimator=MaximumLikelihoodEstimator)

Computing the Probability of HeartDisease given Age print('\n1. Probability of HeartDisease given Age=28')

Computing the Probability of HeartDisease given cholesterol

('sex', 'trestbps'), ('exang', 'trestbps'), ('trestbps', 'heartdisease'),

('fbs', 'heartdisease'), ('heartdisease', 'restecg'), ('heartdisease', 'thalach'), ('heartdisease', 'chol')])

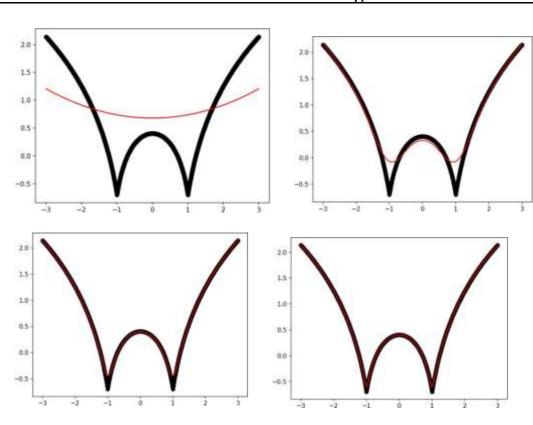
q = HeartDisease_infer.query(variables=['heartdisease'], evidence={'age': 28})

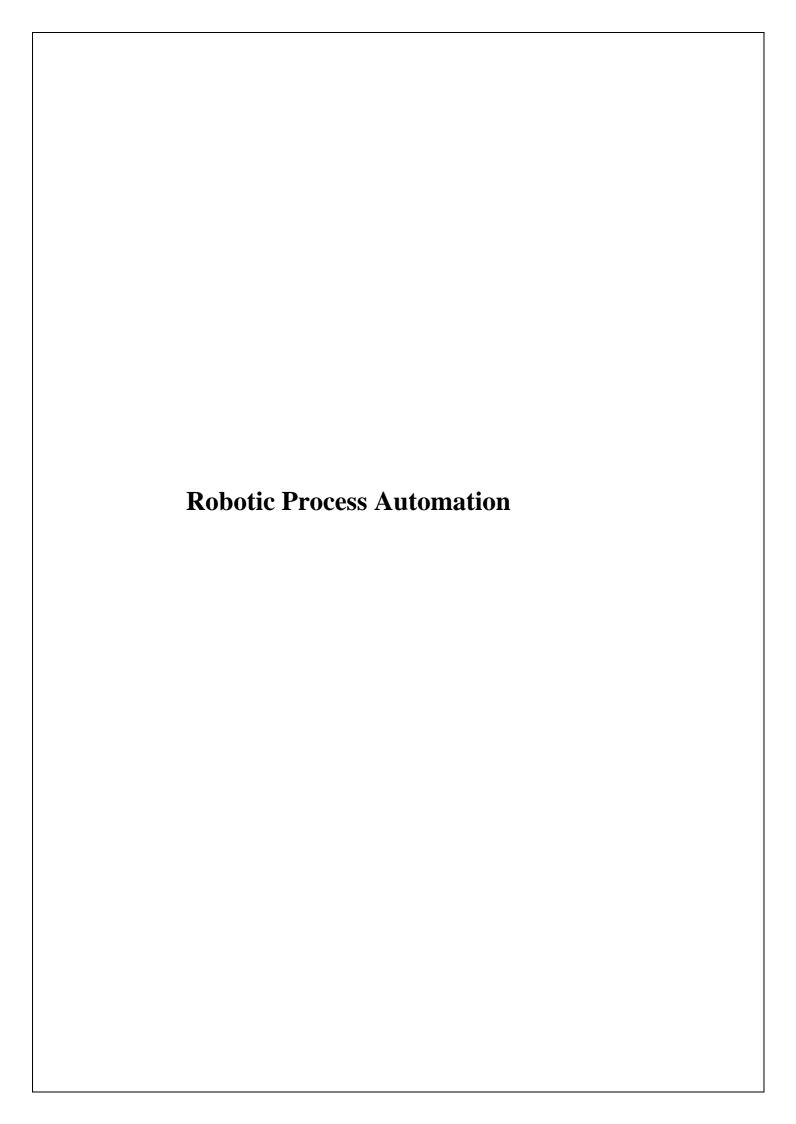
```
\label{eq:print} \begin{split} & print('\n2.\ Probability\ of\ HeartDisease\ given\ cholesterol = 100')\\ & q = HeartDisease\_infer.query(variables = ['heartdisease'],\ evidence = \{'chol':\ 100\})\\ & print(q['heartdisease'])\\ & \textbf{Output:} \end{split}
```

PRATICAL 8B) Implement the non-parametric Locally Weighted Regression algorithm in order to fit data points. Select appropriate data set for your experiment and draw graphs.

```
import numpy as np
import matplotlib.pyplot as plt
def local_regression(x0, X, Y, tau):
  x0 = [1, x0]
  X = [[1, i] \text{ for } i \text{ in } X]
  X = np.asarray(X)
  xw = (X.T) * np.exp(np.sum((X - x0) ** 2, axis=1) / (-2 * tau))
  beta = np.linalg.pinv(xw @ X) @ xw @ Y @ x0
  return beta
def draw(tau):
  prediction = [local\_regression(x0, X, Y, tau) for x0 in domain]
  plt.plot(X, Y, 'o', color='black')
  plt.plot(domain, prediction, color='red')
  plt.show()
X = np.linspace(-3, 3, num=1000)
domain = X
Y = np.log(np.abs(X ** 2 - 1) + .5)
draw(10)
draw(0.1)
draw(0.01)
draw(0.001)
```

Output:





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Practical No	Practical Aim	Page No			
	Practical No. 1				
A	Use two input dialogs for First Name and Last Name store in a variable and show in Message Box.				
В	Use two input dialogs for First Name and Last Name store in a variable and show in Message Box.				
C	Use Web Recorder to empty trash in Gmail.				
	Use two Input dialogs for Numbers and do following calculations and show in Message Box.				
	Practical No. 2				
	Use two Input dialogs for Numbers and do following calculations and show in Message Box.				
	Create Different type of variable (number, datetime, Boolean, generic, array, data table) and provide default value and show in Message Box.				
	Practical No. 3				
	Create an automation UiPath Project using decision statements.				
	Create an automation UiPath Project using looping statements (Dummy list of fruits).				
	Practical No. 4				
A	Automate any process using basic recording (Existing Notepad).				
В	Automate any process using desktop recording (Double Ui).				
C	Automate any process using web recording.				
	Practical No. 5				
A	Consider an array of names. We have to find out how many of these start with theletter "a" Create an automation where the number of names starting with "a" is counted and the result is displayed.				
	Practical No. 6				
A	Create an application automating the read, write and append operation on excel file.				
В	Create automate the process to extract data from an excel file into a data table and vice versa.				
	Practical No. 7				
A	Install and automate any process using UiPath with the following plug-ins: Any two				
В	Automate the following screen scraping methods using UiPath i. Full Test (Invoice PDF) 3 PDS (invoice) ii. Extract and put value in Excel				
С	Automate the process of send mail event (use gmail setting). SMTP server address: smtp.gmail.com. Gmail SMTP port (TLS): 587. Add 3 different attachments as input.				

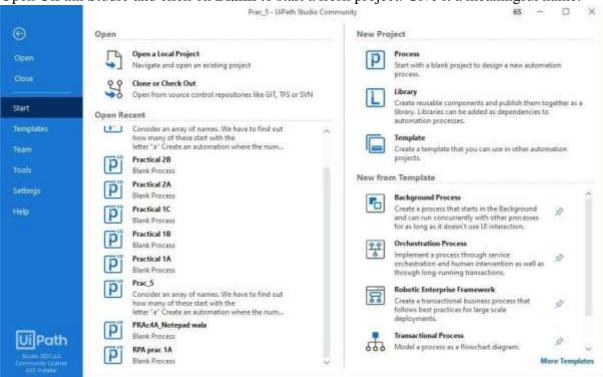
Practical no 1

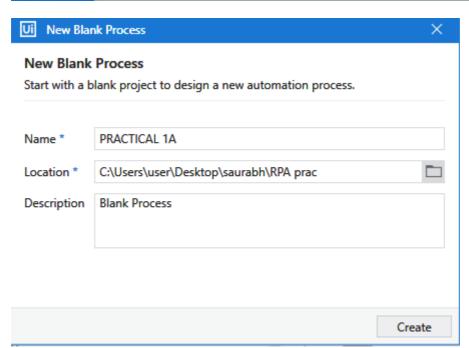
1 A) Create a simple sequence-based project.

Aim: Use two input dialogs for First Name and Last Name store in a variable and show in Message Box.

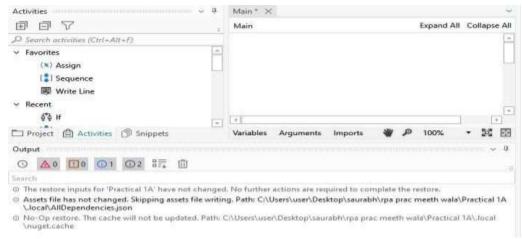
Steps:

1. Open UiPath Studio and click on **Blank** to start a fresh project. Give it a meaningful name.





2. On the Designer panel, drag and drop a **Sequence** activity from the **Activities** panel.



3. Search for **Input dialog** in the Search panel of the **Activities** panel. Drag and drop the **Input dialog** activity inside the **Sequence**.

Write the appropriate message on the **Label** of this **Input dialog** to ask for the user's name. In our case, we have put in "Enter First Name"

4. Drag another **Input dialog** activity into the **Sequence**.



Write the Label as "Enter Last Name"

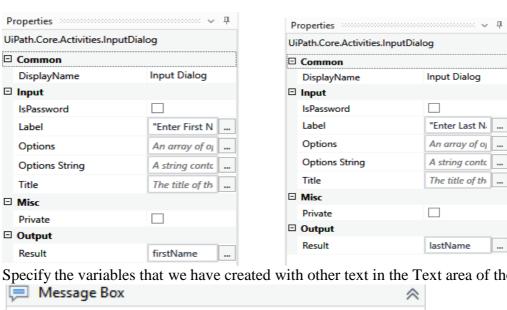
5. Drag and drop a **Message box** activity into the **Sequence**.



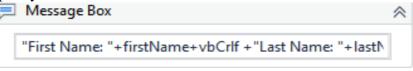
6. Next, create two **variables** and give them the desired names. These variables will receive the text that the user has entered in the **Input dialog** boxes.



7. We now have to specify the **Result** property (in the **Properties** panel) of the **Input dialog** box. On specifying the variable name there, it will receive the text that the user entered.



8. Specify the variables that we have created with other text in the Text area of the **Message box**.



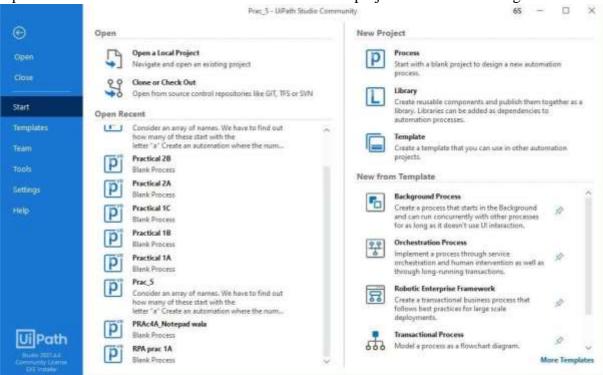
9. Hit the **Debug** button and see the result. × **Enter First Name** Saurabh Ok × Enter Last Name Yadav Ok Message Box X First Name: Saurabh

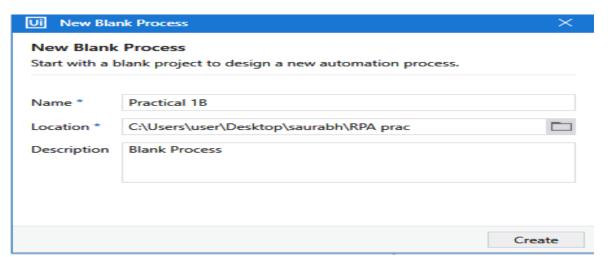


1 B) Create a simple flowchart-based project

Aim: Use two input dialogs for First Name and Last Name store in a variable and show in Message Box. **Steps:**

1. Open UiPath Studio and click on **Blank** to start a fresh project. Give it a meaningful name.

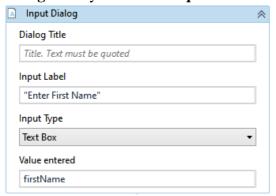




2. On the Designer panel, drag and drop a **Flowchart** activity from the **Activities** panel. Double click into Flowchart and drag **Sequence** activity.



3. Search for **Input dialog** in the Search panel of the **Activities** panel. Drag and drop the **Input dialog** activity inside the **Sequence**.



Write the appropriate message on the **Label** of this **Input dialog** to ask for the user's name. In our case, we have put in "Enter First Name"

4. Drag another **Input dialog** activity into the **Sequence**.



Write the Label as "Enter Last Name"

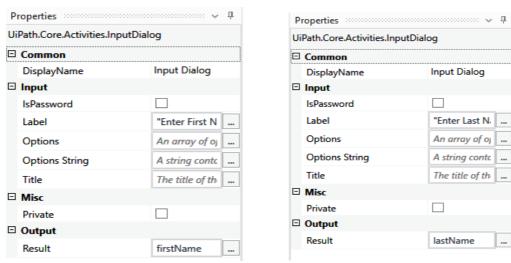
5. Drag and drop a **Message box** activity into the **Sequence**.



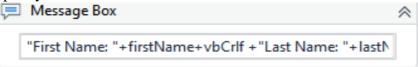
6. Next, create two **variables** and give them the desired names. These variables will receive the text that the user has entered in the **Input dialog** boxes.



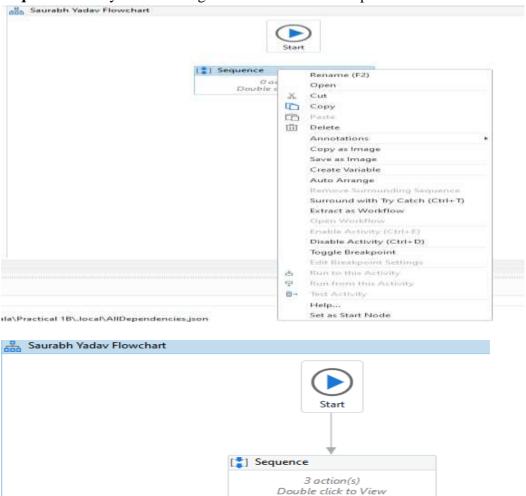
7. We now have to specify the **Result** property (in the **Properties** panel) of the **Input dialog** box. On specifying the variable name there, it will receive the text that the user entered.



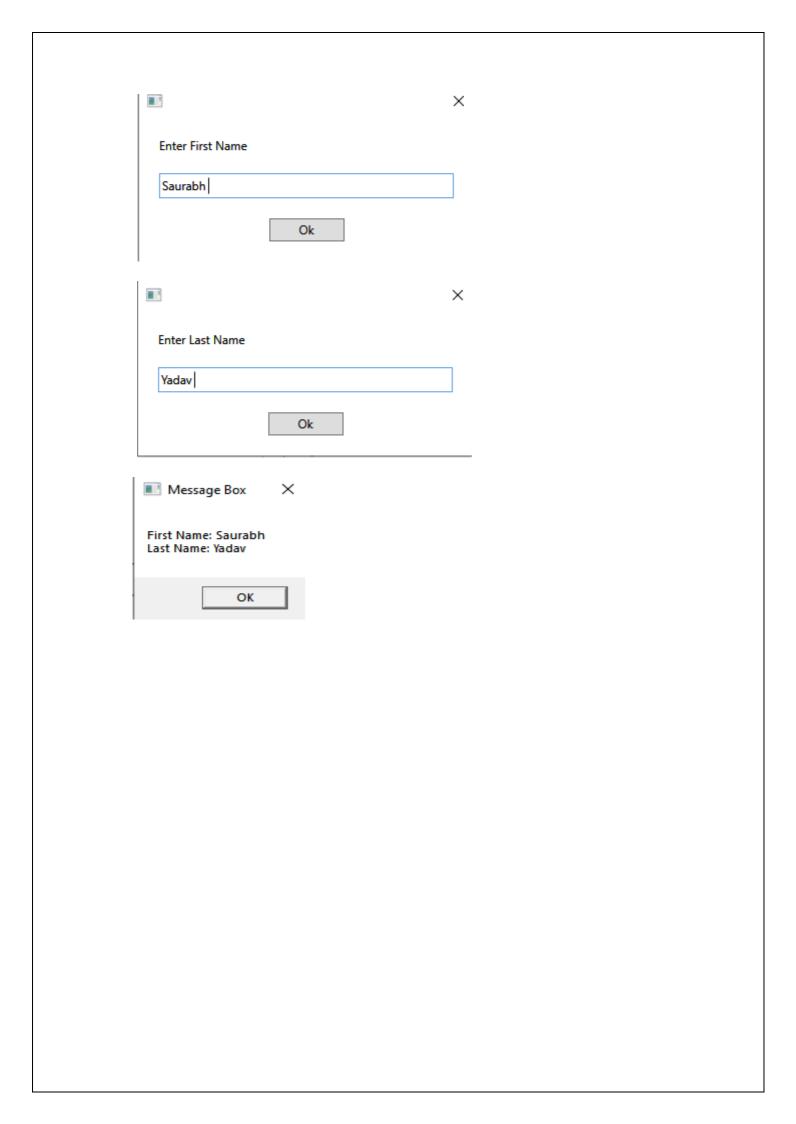
8. Specify the variables that we have created with other text in the Text area of the **Message box**.



9. We need to connect the **Sequence** to the **Start** icon. This can be done by right-clicking on the **Sequence** activity and choosing the **Set as Start node** option.



10. Hit the **Debug** button and see the result.

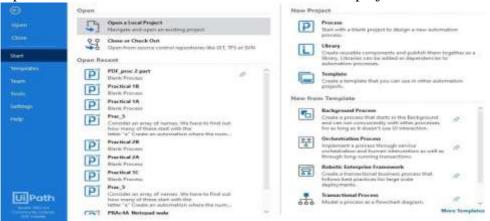


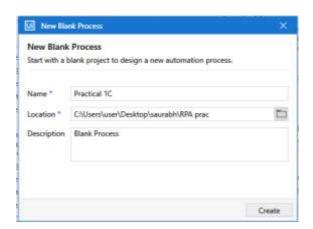
1 C) Create an UiPath Robot which can empty recycle bin in Gmail solely on basis of recording.

Aim: Use Web Recorder to empty trash in gmail.

Steps:

1. Open UiPath Studio and click on **Blank** to start a fresh project. Give it a meaningful name.

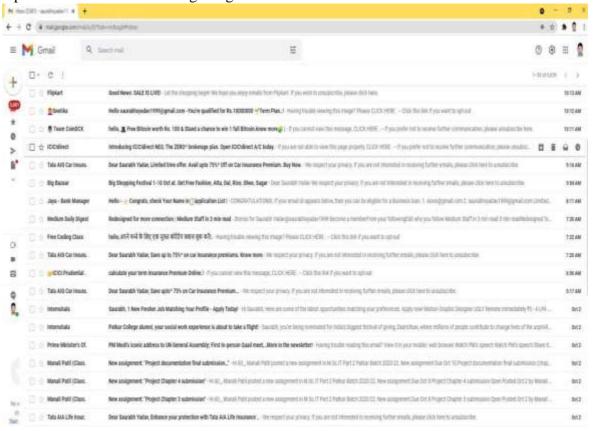




2. Choose Web recorder from the Recording drop-down list:



3. Open a Chrome browser and go to gmail.com.



4. On the web recording panel click on "Open Browser" option and select the chrome window. In the url prompt type gmail.com.

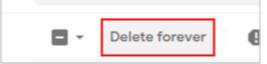




5. From the drop down click on the in:trash option.



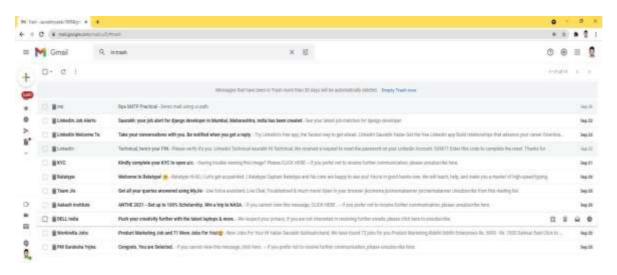
6. Once in the trash section of Gmail. Click on the checkbox button to select all the mail of the current page. Then click on Delete Forever button that appears.



7. Press Esc key and click on Save and Exit option in the web recording panel.



8. Now Run the program by clicking on Debug and see the output.



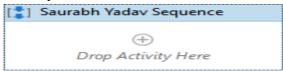
Practical no 2

2 A) Automate UiPath Number Calculation (Addition, Subtraction, Multiplication, Division of numbers).

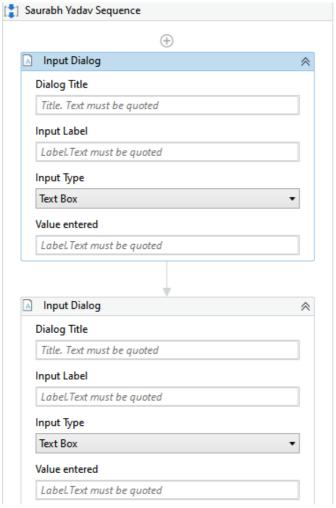
Aim: Use two Input dialogs for Numbers and do following calculations and show in Message Box.

Steps:

1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.



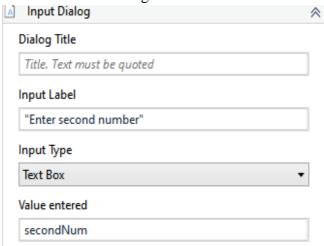
2. Search Input Dialog activity and drag and drop the activity twice.



Give the first input dialog label as "Enter first number".

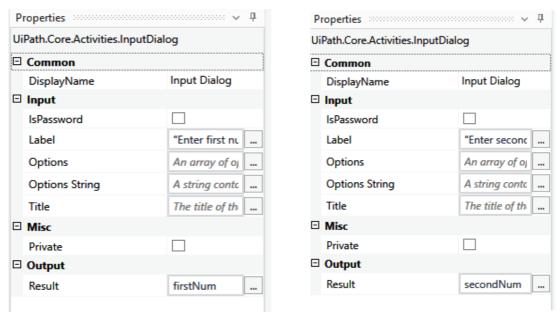


Give the second dialog label as "Enter second number".

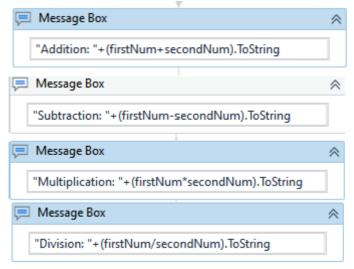


3. Create the variables to store both inputs, then enter the variables in the input dialog Result attributes.

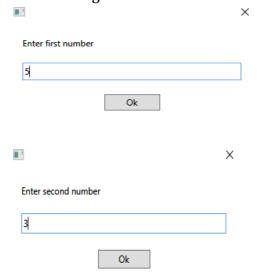


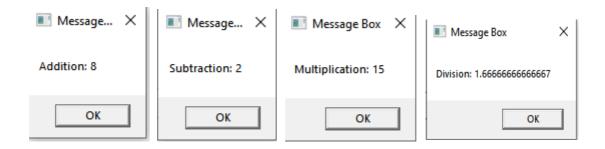


4. Drag and drop message box four times and perform the addition, subtraction, multiplication and division operation in each of the message boxes.



5. Hit the **Debug** button and see the result





2 B) Create an automation UiPath project using different types of variables (number, datetime, Boolean, generic, array, data table)

Aim: Create Different type of variable and provide default value and show in Message Box.

Steps:

1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.



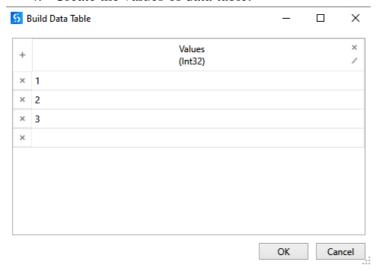
2. Create number(Int32), datetime, Boolean, string, generic, array variables and assign them default value.



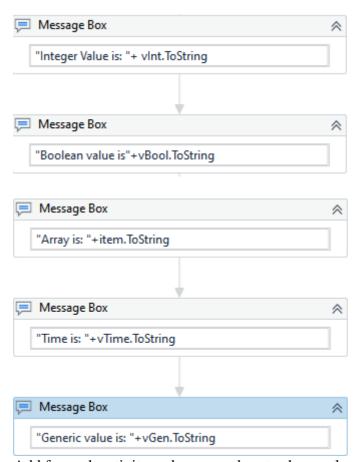
3. Create a datable variable and add a build dataTable activity.



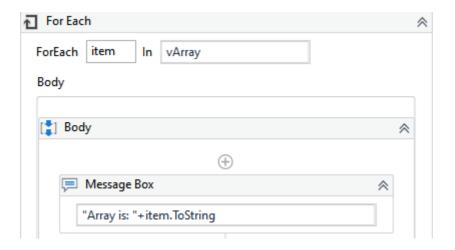
4. Create the values of data table.



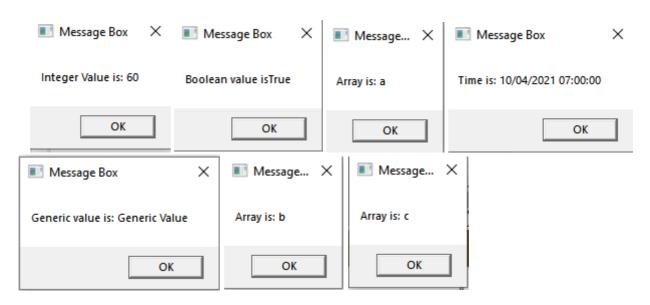
5. Add message box to show the number, datetime, Boolean, string, and generic variable.



6. Add for each activity and message box to show values of array variable.



Output:



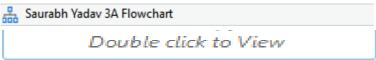
Practical no 3

3 A) Create an automation UiPath Project using decision statements.

Aim: Use one input dialog for Number, create a process to find provided Number is odd or even using condition (Flow Chart).

Steps:

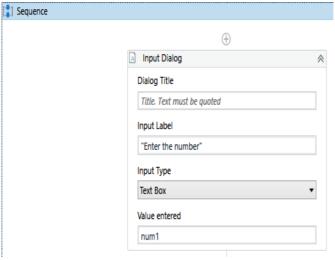
6. Create a new Blank Project and give it an appropriate name. Drag a Flowchart activity from Activity tab.



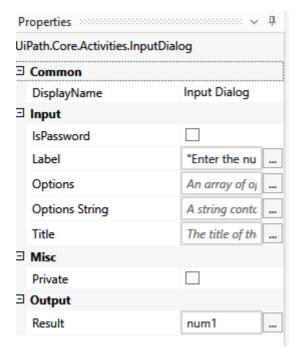
7. Double click on flowchart and drag a sequence activity.



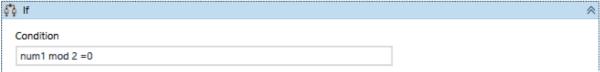
8. Drag an Input Dialog activity and give label "Enter the number".



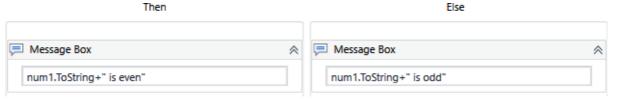
9. Create a variable of Int32 type and enter the variable in result attribute of Input Dialog.



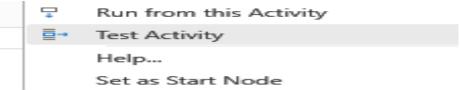
10. Write the following expression in Condition: num mod 2 = 0.

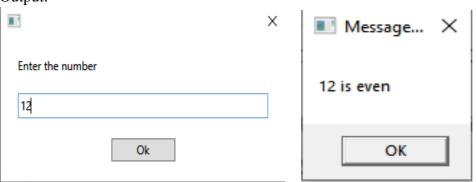


11. In the Message Box in then part write "Number is even" and in the Message Box in else part write "Number is odd".



12. Connect the sequence to the start node by right clicking on sequence and selecting the "Set as start node" option.





3 B) Create an automation UiPath Project using looping statements.

Aim: Create a dummy List of fruits. Loop each and print in Message Box.

Steps:

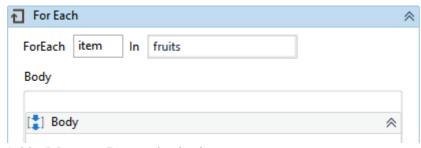
1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.



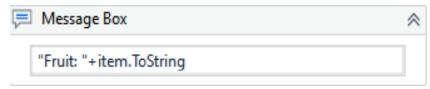
2. Create an Array of String Type named fruits. Give default value as {"Apple", "Mango", "Banana", "Orange", "Watermelon"}.

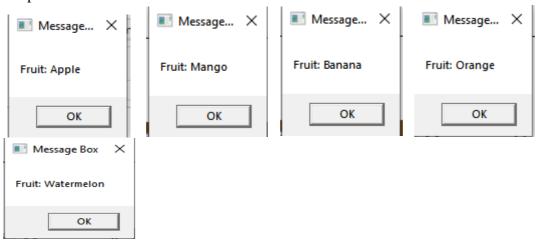


3. Add a for each activity and enter the fruits array as the iterating variable.



4. Add a Message Box and print item.





Practical: 4

4 A) Automate any process using basic recording.

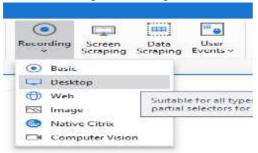
Aim: Automate Existing Notepad using basic Recorder

Steps:

1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.



2. Select Desktop Recording under Recording.



3. Open a new notepad, then on desktop recording click on open application and select notepad the click ok on the prompt that appears.



4. Start recording by clicking on record button in recording panel, then click in the text area of notepad.



5. In the type into prompt enter the text you want to be typed.

Output:



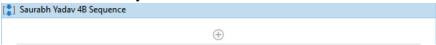
4.B) Automate any process using desktop recording.

Aim: Use "double Ui" and automate stuff (Scrape Text, Input Text, Click Button) Insert the data from the following excel file.

Α	В	С	D	E	F	G
Sr no.	Cash In	Check_1	Check_2	Total	Transactio	Status
1	100	200	400			
2	200	300	600			
3	300	400	800			
4	400	500	1000			
5	500	600	1200			
6	600	700	1400			
7	700	800	1600			
8	800	900	1800			

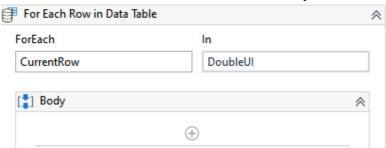
Steps:

1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.

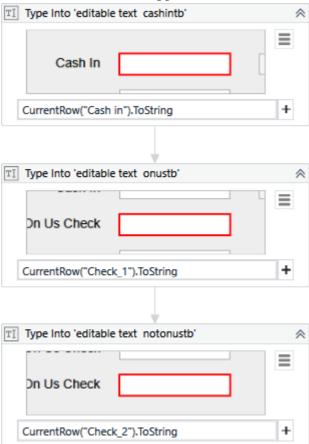


Add Excel Application Scope Activity and enter the path of the excel file. Excel Application Scope \wedge "C:\Users\user\Desktop\saurabh\RPA prac my own\Double UI.xIsx" 3. Add Read Range and create a DataTable to store the data that will be read by read range. Read Range \wedge "Sheet1" Name Variable type Default Scope DoubleUl DataTable Saurabh Yadav 4B Sequence Enter a VB expression transactionid Saurabh Yadav 4B Sequence Enter a VB expression Saurabh Yadav 4B Sequence Enter a VB expression 4. Add Open Application activity, start Double UI then click on "Indicate window on screen" option and select Double UI. Open Application 'doubleui.exe DoubleUI' DoubleUI 10-07-2021 **UiPath** 07:27:51 F [**†**] Do (+)

5. Add For each row in DataTable activity and iterate over DoubleUI datatable.



6. Add three type into activity in the for each activity. Indicate to the three textboxes in the Double UI app. Enter the DataTable element you wan to enter in the Text attribute.



7. Add two get text activities and create two variables to store the transaction is andtotal. Indicate these elements to the Get text activities.



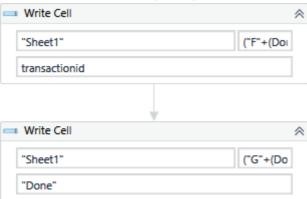


In case of Total remove the name attribute from the Edit Selector.

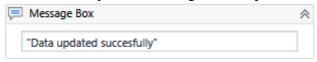
8. Add Click activity and indicate to Accept Button in Double UI.

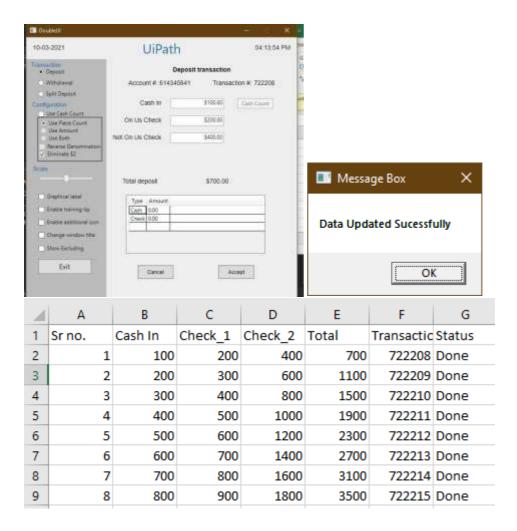


9. Add three Write Cell Activities to enter the transaction id and total and to Update the status column. ("E"+(DoubleUI.Rows.IndexOf(CurrentRow)+2).ToString)



10. Finally add a Message Box to print the success of the updation of Excel file.





4 C) Automate any process using web recording.

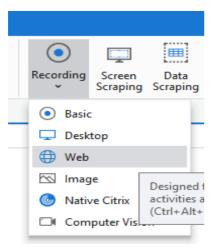
Aim: Use "google.com" and automate stuff (Scrape Text, Input Text, Click Button)

Steps:

1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.



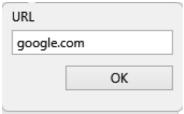
2. Select web recording under the recording option.



3. Open a Chrome window and from the web recording dialog select open browser and click on the Chrome window.



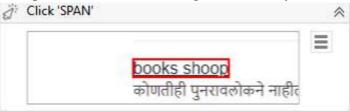
4. In the url dialog enter google.com and press enter.



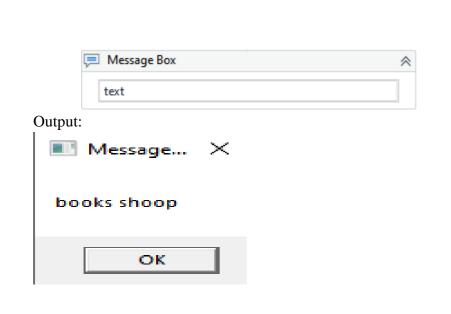
5. Now start recording by clicking on Record and click on the search button. Type the desired search and click on the Google search button. The press Esc.



6. Now press save and exit. Add get text activity and select the text you want to scrape.



7. Add message box to print scrapped text.

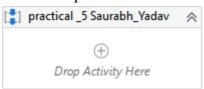


Practical no 5

Aim: Consider an array of names. We have to find out how many of these start with the letter "a" Create an automation where the number of names starting with "a" is counted and the result is displayed.

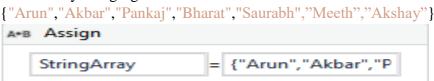
Steps:

1. Create a sequence.



2. Use 1 Assign Activity

Create array string Eg:



3. In Variable section take datatype as Array of[T] -> String



4. Take For each loop where each item in stringArray will occur repeatedly.



5. Take another Assign activity and create new variable as "First_letter_name" with datatype as String.



item.ToString.SubString(0,1)

Where 0 is the starting index value.

1 is the length of the name.

6. Now Take If condition -First_letter_name.ToLower.Equals("a") this condition states that take the letter from the name list of array which starts with alphabet "A" or "a".



7. Then: take assign activity create Counter variable to count the number of names in the array starting with "A", give default value of Counter variable as 0.

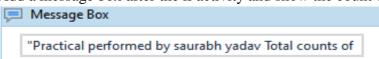


Then

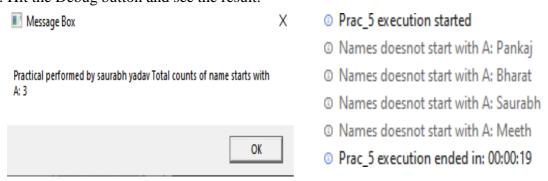
8. Else: take write line activity and put as "Names doesnot start with A:"+item.ToString



9. Add a message box after the if activity and show the count of Names starting with "A".



10. Hit the Debug button and see the result.



Practical no 6

6A. Create an application automating the read, write and append operation on excel file.

Aim: Using Excel sheet data perform some calculation on the data writing it into the cells

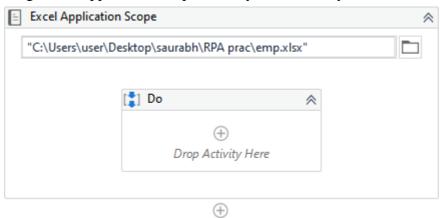
Steps: Create a sample excel as shown below

1	Emp_Id	F_name	L_name	Salary	Expenses	Saving
2	1001	Saurabh	Yadav	100000	10000	
3	1002	Pankaj	Gavali	200000	20000	
4	1003	Bharat	Bhagat	300000	30000	
5	1004	Virat	Kohli	400000	40000	
6	1005	Mahendra	Dhoni	500000	50000	
7						

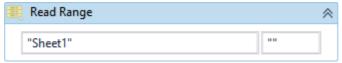
1. Start project as sequence



2. Drag Excel Application Scope Activity from Activity Panel and enter the excel file path.

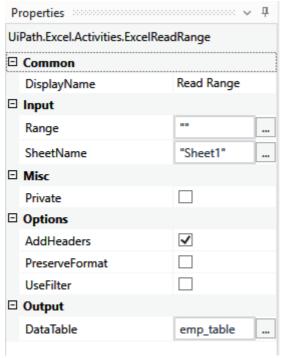


3. Add "Read Range" from activities and add the range of the cells if required.

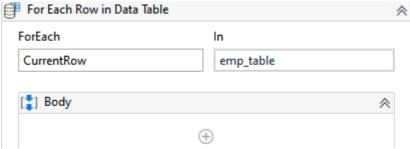


4. Create a variable of type DataTable (here variable name: emp_table variable type: DataTable). Enter the variable in the Read Range Output>DatableTable attribute.





5. Select "For Each Row" and enter Input DataTable (here emp_table).



- 6. Inside Body insert "Assign" variable (Saving) with int32 as variable type.
- 7. Enter the following in value part of assign CInt(CurrentRow("Salary"))-CInt(CurrentRow("Expenses")).



8. A "Message Box" to print the result converting it to string.

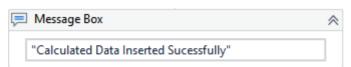


9. To insert the Saving value in the excel use Write Cell Activity. Give the Savings Variable in Value. In Range type:

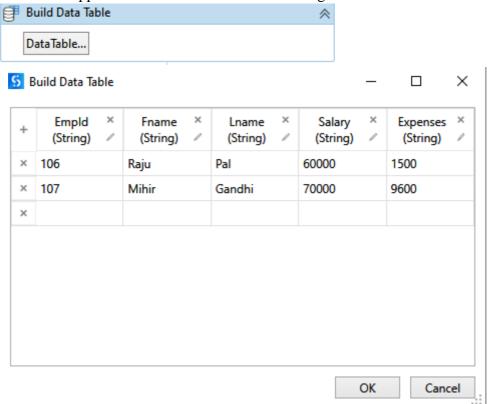
"F"+(emp_table.Rows.IndexOf(CurrentRow)+2).ToString.



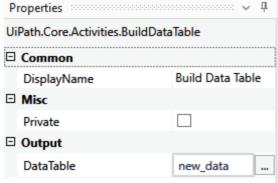
10. Add a Message Box to indicate that the calculated values have been successfully inserted.



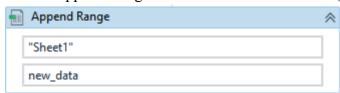
11. In order to append new data create data table using "Build Data Table" and insert the values.



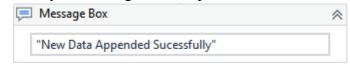
12. Create a new variable with name: new_data and variable type: DataTable and pass it to the DataTable field in Output attribute of Build datatable activity.

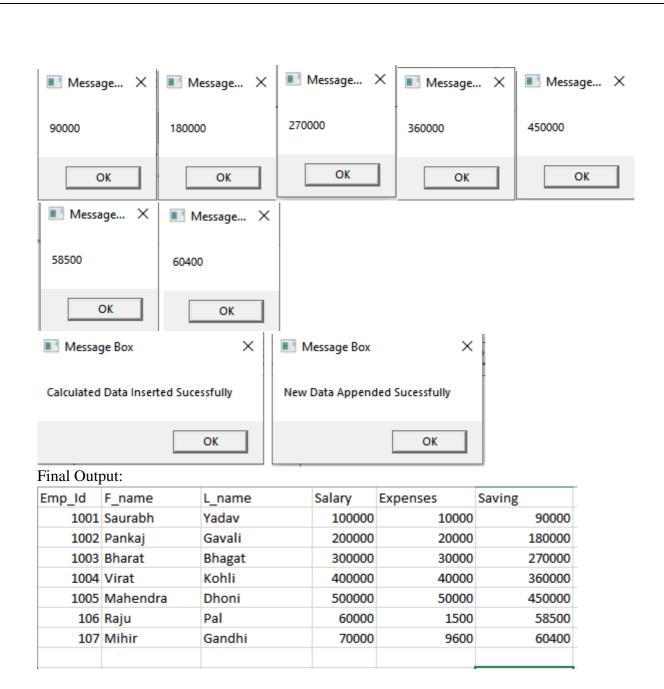


13. Select "Append Range" and insert DataTable name ("new data").



14. Finally, a "Message Box" to print a confirmation message.





6B. Create automate the process to extract data from an excel file into a data table and vice versa.

Aim: Using one excel sheet data to copy and writing it into other excel sheet. (Using excel data given below)

	(L	sing	exe	cel	data	given	be	low)	ļ
ı	_			_			٦.		

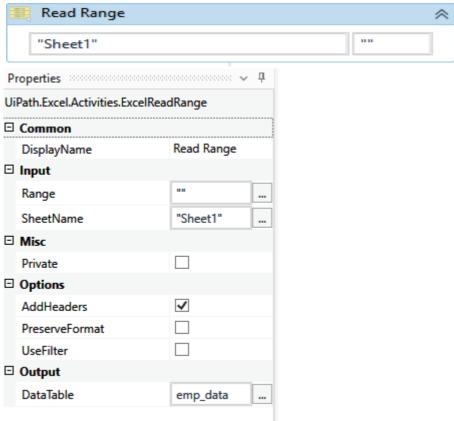
Emp_Id	F_name	L_name	Salary	Expenses	Saving
1001	Saurabh	Yadav	100000	10000	90000
1002	Pankaj	Gavali	200000	20000	180000
1003	Bharat	Bhagat	300000	30000	270000
1004	Virat	Kohli	400000	40000	360000
1005	Mahendra	Dhoni	500000	50000	450000
106	Raju	Pal	60000	1500	58500
107	Mihir	Gandhi	70000	9600	60400

Step:

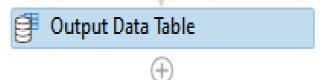
1. Start project with sequence.

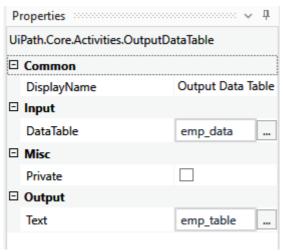


3. Inside body add "Read Range" variable name and datatype as datatable(emp_data) and range to extract data from excel (blank means entire sheet).



4. Add "Output Data Table" input as "emp data" output as "emp table".





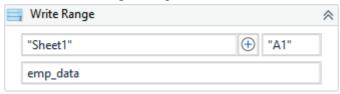
5. "Message Box" to print the result adding variable "emp_table" (here extracted data will be printed).



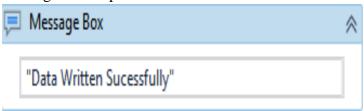
6. Insert "excel application scope" and new excel sheet path.



7. Insert "Write Range" range from where the data should be pasted and variable "emp_data".



8. Message Box to print confirmation







Emp_Id,F_name,L_name,Salary,Expenses,Saving 1001,Saurabh,Yadav,100000,10000,90000 1002,Pankaj,Gavali,200000,20000,180000 1003,Bharat,Bhagat,300000,30000,270000 1004,Virat,Kohli,40000,40000,360000 1005,Mahendra,Dhoni,500000,50000,450000 106,Raju,Pal,60000,1500,58500 107,Mihir,Gandhi,70000,9600,60400



A	Α	В	С	D	E	F	G
1	1001	Saurabh	Yadav	100000	10000	90000	
2	1002	Pankaj	Gavali	200000	20000	180000	
3	1003	Bharat	Bhagat	300000	30000	270000	
4	1004	Virat	Kohli	400000	40000	360000	
5	1005	Mahendra	Dhoni	500000	50000	450000	
6	106	Raju	Pal	60000	1500	58500	
7	107	Mihir	Gandhi	70000	9600	60400	
8	106	Raju	Pal	60000	1500		
9	107	Mihir	Gandhi	70000	9600		
10							

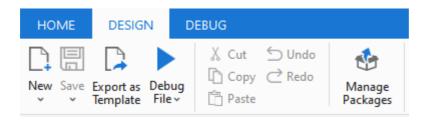
Practical no 7

7A: Install and automate any process using UiPath.

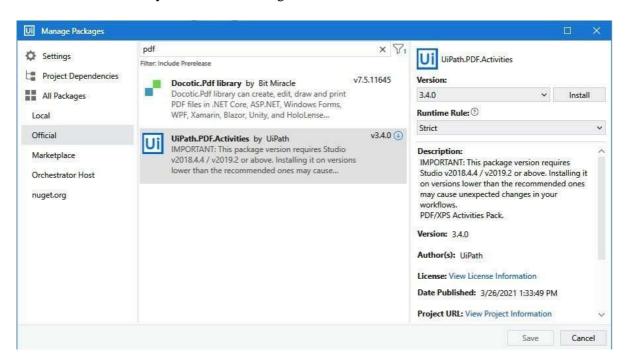
Aim: Install and automate any process using UiPath with the following plug-ins: Any two.

Steps:

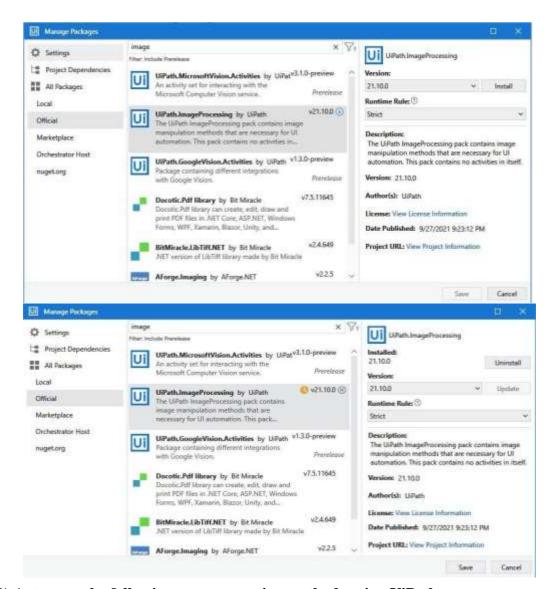
1. Create a Process→Open Ui path→Click on manage package.



2. Click on All Packages on Manage Package Pop-up widow and Search the Required Activity or Package you want to install→Select that particular activity or package and click on install→ Click on it will Automatically will save the changes.



3. Click on All Packages on Manage Package Pop-up widow and Search the Required Activity or Package you want to install→Select that particular activity or package and click on install→ Click on it will Automatically will save the changes.



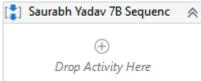
7 B) Automate the following screen scraping methods using UiPath

i. Full Test (Invoice PDF) 3 PDS (invoice)

Extract and put value in Excel

Steps:

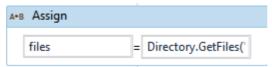
1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.



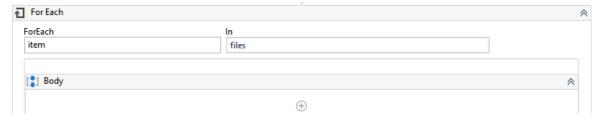
2. Add Excel Application Scope Activity give path and in an assign activity start a counter from 2.



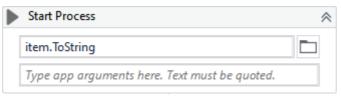
3. Add a assign activity to get the files in the given folder path.



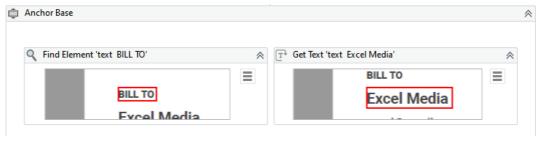
4. Add a for each activity to iterate over the files.



5. Add an Start process activity and pass the iterator variable.



6. Open one pdf and using anchor base, find element and get text extract the elements you want to get.



7 C) Automate the process of send mail event (use gmail setting).

SMTP server address: smtp.gmail.com.

Gmail SMTP port (TLS): 587.

Add 3 different attachments as input.

Step:

1. Create a new Blank Project and give it an appropriate name. Drag a Sequence activity from Activity tab.



2. Add Send SMTP Mail Message Activity enter the recipient email, subject and body of the email to be sent.



3. Enter the files variable in AttachmentsCollection Attribute. Enter smtp port number in port attribute of Host and the hostname in server field. Enter the email and password of the sender in the Logon email and password.

