Soft Computing Techniques



JNAN VIKAS MANDAL'S

Mohanial Raichand Mehta College of Commerce
Diwali Maa College of Science
Amritial Raichand Mehta College of Arts
Dr. R.T. Doshi College of Computer Science
NAAC Re-Accredited Grade 'A+' (CGPA: 3.31) (3rd Cycle)

DEPARTMENT OF INFORMATION TECHNOLOGY

CERTIFICATE

This is to certify that <u>Janhavi Vijay Kharkar</u> bearing seat no. <u>1310657</u> has done the project work/journal work in the subject of Soft Computing of semester 1 Practical Examination during the academic year 2023-24 under the guidance of **Mrs. Sharayu Kadam** being the partial requirement for the fulfilment of the curriculum of Degree in Master of Science in Information Technology under University of Mumbai.

Place: Airoli	Date:
Sign of Subject in- Charge	Sign of External Examiner
Sign of Coordinator	

Table of Content

Practical No	Details	Date	Sign
1	Implement the following:		
A	Design a simple linear neural network model.		
В	Calculate the output of neural net using both		
	binary and bipolar sigmoidal function.		
2	Implement the following:		
A	Generate AND/NOT function		
	usingMcCulloch-Pitts neural net.		
В	Generate XOR function using McCulloch-Pitts		
	neural net.		
3	Implement the Following		
A	Write a program to implement Hebb's rule.		
В	Write a program to implement delta rule.		
4	Implement the Following		
A	Write a program for Back Propagation		
	Algorithm		
В	Write a program for error Backpropagation		
	algorithm.		
5.	Implement the Following		
\mathbf{A}	Write a program for Hopfield Network.		
В	Write a program for Radial Basis function		
6.	Implement the Following		
A	Kohonen Self organizing map		
В	Adaptive resonance theory		
7.	Implement the Following		
A	Write a program for Linear separation.		
В	Write a program for Hopfield network model		
	for associative memory		
8.	Implement the Following		
A	Membership and Identity Operators in, not in,		
В	Membership and Identity Operators is, is not		
9.	Implement the Following		
A	Find ratios using fuzzy logic		
В	Solve Tipping problem using fuzzy logic		
10.	Implement the Following		
A	Implementation of Simple genetic algorithm		
В	Create two classes: City and Fitness using Genetic algorithm		

Practical 1A

Aim: Design a simple linear neural network model.

Code:

```
w=float(input("Enter the weight value: "))
b=float(input("Enter the bias value: "))
x=float(input("Enter the input value: "))
yin=float(b+(w*x))
print("The net input yin is:",yin)
```

Output:

```
= RESTART: C:/Users/Janhavi/AppData/Local/Programs/Python/Python38/RunAllInOneGo.py
Enter the weight value: 1
Enter the bias value: 1
Enter the input value: 0.4
The net input yin is: 1.4
>>>
```

Practical 1B

Aim: Calculate the output of neural net using both binary and bipolar sigmoidal function.

```
import math bias = float(input("Enter the value of bias: ")) n = int(input("Enter the number of input neurons: ")) \\ w = [ ] \\ x = [  ] \\ for i in range(0,n): \\ a = float(input("Enter the input: ")) \\ x.append(a) \\ b = float(input("Enter the weight: "))
```

```
w.append(b)
print("The given weights are: ")
print(w)
print("The given input are: " )
print(x)
y = bias
for i in range(0, n):
   y = y + (w[i]*x[i])
print("The calculated net input y : ")
print(y)
# Applying Binary Sigmoidal function on the net input i.e y
binary = 1/(1+(math.exp(-y)))
print("The output after applying binary sigmoidal function activation ")
print (round(binary, 3))
# Applying Bipolar Sigmoidal function on the net input i.e y
bipolar = -1+(2/(1+(math.exp(-y))))
print("The output after applying bipolar sigmoidal function activation ")
print(round(bipolar, 3))
```

```
Enter the value of bias: 1
Enter the number of input neurons: 3
Enter the input: 0.4
Enter the weight: 1
Enter the input: 0.6
Enter the weight: 1
Enter the input: 0.2
Enter the weight: 1
The given weights are:
[1.0, 1.0, 1.0]
The given input are:
[0.4, 0.6, 0.2]
The calculated net input y:
The output after applying binary sigmoidal function activation
The output after applying bipolar sigmoidal function activation
0.8
>>>
```

Practical 2A

Aim- Generate AND/NOT function using McCulloch-Pitts neural net.

```
num_ip = int(input("Enter the number of inputs : "))
theta = 1
x1 = []
x^2 = []
for i in range(0,num_ip):
  a = int(input("Enter the input x1: "))
  x1.append(a)
  b = int(input("Enter the input x2:"))
  x2.append(b)
print("x1 = ", x1)
print("x2 = ", x2)
print("Value of theta is 1.")
print("Case 1: For calculating the net input the weight is considered as w1 =
w2 = 1")
w1 = w2 = 1
case_y1 =[]
case\_yin1 = []
print("x1 w1 x2 w2 case_y1 case_yin1")
for i in range(0,num_ip):
  case_y1.append(x1[i]*w1 + x2[i]*w2)
  if (case_y1[i] >= theta):
     case_yin1.append( 1 )
  else:
     case_yin1.append(0)
  print(x1[i]," ", w1," ",x2[i]," ",w2," ", case_y1[i],"
                                                                ",case_yin1[i])
print("From the calculated net inputs, it is not possible to fire the neuron form
input (1, 0) only."
"\n Hence, these weights are not suitable. ")
```

```
print("Case 2: For calculating the net input the weight is considered as w1 = 1,
w2 = -1"
w1 = 1
w2 = -1
case y2 = []
case\_yin2 = []
print("x1 w1 x2 w2 case y2 case yin2")
for i in range(0,num_ip):
  case_y2.append(x1[i]*w1 + x2[i]*w2)
  if (case_y2[i] >= theta):
    case_yin2.append(1)
  else:
    case yin2.append(0)
  print(x1[i]," ", w1," ",x2[i]," ",w2," ", case y2[i]," ",case yin2[i])
print("From the calculated net inputs, it is possible to fire the neuron form
input (1, 0) only."
"\n Hence, these weights are suitable. ")
```

```
Enter the number of inputs : 4
Enter the input x1: 1
Enter the input x2: 1
Enter the input x1: 1
Enter the input x2: 0
Enter the input x1: 0
Enter the input x2: 1
Enter the input x1: 0
Enter the input x2: 0
x1 = [1, 1, 0, 0]
x2 = [1, 0, 1, 0]
Value of theta is 1.
Case 1: For calculating the net input the weight is considered as w1 = w2 = 1
xl wl x2 w2 case yl case yinl
1 1 1 1
1 1 0 1
                  1
0 1 1 1
                             1
                 - 1
0 1 0 1
From the calculated net inputs, it is not possible to fire the neuron form input (1, 0) only.
Hence, these weights are not suitable.
Case 2: For calculating the net input the weight is considered as w1 = 1, w2 = -1
x1 w1 x2 w2 case_y2 case_yin2
1 1 1 -1 0
1 1 0 -1
                              1
  1 1 -1
From the calculated net inputs, it is possible to fire the neuron form input (1, 0) only.
Hence, these weights are suitable.
```

Practical 2B

Aim- Generate XOR function using McCulloch-Pitts neural net.

Go to command prompt:

pg. 8

```
cd C:\Users\Janhavi\AppData\Local\Programs\Python\Python38
python -m pip install numpy
Code:
import numpy as np
print('Enter weights')
w11=int(input('Weight w11='))
w12=int(input('weight w12='))
w21=int(input('Weight w21='))
w22=int(input('weight w22='))
v1=int(input('weight v1='))
v2=int(input('weight v2='))
print('Enter Threshold Value')
theta=int(input('theta='))
x1=np.array([1, 1, 0, 0])
x2=np.array([1, 0, 1, 0])
z=np.array([0, 1, 1, 0]) #This is target output
                    #covergence flag is set to 1
con=1
y1=np.zeros((4,))
                        #starts from 0
y2=np.zeros((4,))
y=np.zeros((4,))
while con==1:
  zin1=np.zeros((4,))
  zin2=np.zeros((4,))
  zin1=x1*w11+x2*w21
                             \#for pair(1,0)
  zin2=x1*w21+x2*w22
                             \#for pair(0,1)
  print("z1",zin1)
  print("z2",zin2)
  for i in range(0,4):
     if zin1[i]>=theta:
```

```
y1[i]=1
    else:
       y1[i]=0
    if zin2[i]>=theta:
       y2[i]=1
    else:
       y2[i]=0
  yin=np.array([])
  yin=y1*v1+y2*v2
  for i in range(0,4):
    if yin[i]>=theta:
       y[i]=1
    else:
       y[i]=0
  print("yin",yin)
  print('Output of Net')
  y=y.astype(int)
  print("y",y)
                    #calculated output
  print("z",z)
                   #target output
  if np.array_equal(y,z):
    con=0
  else:
    print("Net is not learning enter another set of weights and Threshold
value")
    w11=input("Weight w11=")
    w12=input("weight w12=")
    w21=input("Weight w21=")
    w22=input("weight w22=")
    v1=input("weight v1=")
    v2=input("weight v2=")
    theta=input("theta=")
print("McCulloch-Pitts Net for XOR function")
print("Weights of Neuron Z1")
print(w11)
pg. 9
```

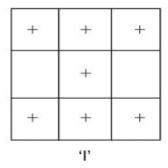
```
print(w21)
print("weights of Neuron Z2")
print(w12)
print(w22)
print("weights of Neuron Y")
print(v1)
print(v2)
print("Threshold value")
print(theta)
```

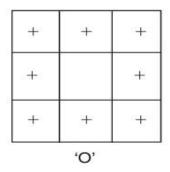
```
Enter weights
Weight wll=1
weight w12=-1
Weight w21=-1
weight w22=1
weight vl=1
weight v2=1
Enter Threshold Value
theta=1
z1 [ 0 1 -1 0]
z2 [ 0 -1 1 0]
yin [0. 1. 1. 0.]
Output of Net
y [0 1 1 0]
z [0 1 1 0]
McCulloch-Pitts Net for XOR function
Weights of Neuron Zl
weights of Neuron Z2
-1
weights of Neuron Y
Threshold value
>>>
```

Practical 3A

Aim- Write a program to implement Hebb's rule.

Using the Hebb rule, find the weights required to perform the following classifications of the given input patterns shown in Figure 16. The pattern is shown as 3 × 3 matrix form in the squares. The "+" symbols represent the value "1" and empty squares indicate "-1". Consider "I" belongs to the members of class (so has target value 1) and "O" does not belong to the members of class (so has target value -1).





```
import numpy as np
#first pattern
x1=np.array([1,1,1,-1,1,-1,1,1])
#second pattern
x2=np.array([1,1,1,1,-1,1,1,1,1])
#initialize bais value
b=0
#define target
y=np.array([1,-1])
wtold=np.zeros((9,))
wtnew=np.zeros((9,))
wtnew=wtnew.astype(int)
wtold=wtold.astype(int)
bais=0
print("First input with target =1")
for i in range(0,9):
pg. 11
```

Practical 3B

Aim- Write a program to implement of delta rule.

Code:

Bias value 0

```
import numpy as np
import time
np.set_printoptions(precision=2)
x=np.zeros((3,))
weights=np.zeros((3,))
desired=np.zeros((3,))
actual=np.zeros((3,))
for i in range(0,3):
    x[i]=float(input("Initial inputs:"))

for i in range(0,3):
    weights[i]=float(input("Initial weights:"))

for i in range(0,3):
    desired[i]=float(input("Desired output:"))
```

```
a=float(input("Enter learning rate:"))
actual=x*weights
print("actual",actual)
print("desired",desired)
while True:
  if np.array_equal(desired,actual):
     break #no change
  else:
    for i in range(0,3):
       weights[i]=weights[i]+a*(desired[i]-actual[i])
  actual=x*weights
  print("weights",weights)
  print("actual",actual)
  print("desired",desired)
print("*"*30)
print("Final output")
print("Corrected weights",weights)
print("actual",actual)
print("desired",desired)
```

```
Initial inputs:1
Initial inputs:1
Initial inputs:1
Initial weights:1
Initial weights:1
Initial weights:1
Desired output:2
Desired output:3
Desired output:4
Enter learning rate:1
actual [1. 1. 1.]
desired [2. 3. 4.]
weights [2. 3. 4.]
actual [2. 3. 4.]
desired [2. 3. 4.]
********
Final output
Corrected weights [2. 3. 4.]
actual [2. 3. 4.]
desired [2. 3. 4.]
```

Practical 4A

Aim- Write a program for Back Propagation Algorithm.

```
import numpy as np
import decimal
import math
np.set_printoptions(precision=2)
v1=np.array([0.6, 0.3])
v2=np.array([-0.1, 0.4])
w=np.array([-0.2,0.4,0.1])
b1 = 0.3
b2 = 0.5
x_1 = 0
x2=1
alpha=0.25
print("calculate net input to z1 layer")
zin1 = round(b1 + x1*v1[0] + x2*v2[0],4)
print("z1=",round(zin1,3))
print("calculate net input to z2 layer")
zin2 = round(b2 + x1*v1[1] + x2*v2[1],4)
print("z2=",round(zin2,4))
print("Apply activation function to calculate output")
z1=1/(1+math.exp(-zin1))
z1=round(z1,4)
z2=1/(1+math.exp(-zin2))
z2=round(z2,4)
print("z1=",z1)
print("z2=",z2)
print("calculate net input to output layer")
yin=w[0]+z1*w[1]+z2*w[2]
print("yin=",yin)
print("calculate net output")
y=1/(1+math.exp(-yin))
```

```
print("y=",y)
fyin=y*(1-y)
dk=(1-y)*fyin
print("dk=",dk)
dw1 = alpha * dk * z1
dw2 = alpha * dk * z2
dw0= alpha * dk
print("compute error portion in delta")
din1=dk*w[1]
din2=dk*w[2]
print("din1=",din1)
print("din2=",din2)
print("error in delta")
fzin1 = z1 * (1-z1)
print("fzin1=",fzin1)
d1=din1* fzin1
fzin2 = z2 * (1-z2)
print("fzin2=",fzin2)
d2=din2* fzin2
print("d1=",d1)
print("d2=",d2)
print("Changes in weights between input and hidden layer")
dv11=alpha * d1 * x1
print("dv11=",dv11)
dv21=alpha * d1 * x2
print("dv21=",dv21)
dv01=alpha * d1
print("dv01=",dv01)
dv12=alpha * d2 * x1
print("dv12=",dv12)
dv22=alpha * d2 * x2
print("dv22=",dv22)
dv02=alpha*d2
```

```
print("dv02=",dv02)

print("Final weights of network")

v1[0]=v1[0]+dv11

v1[1]=v1[1]+dv12

print("v1=",v1)

v2[0]=v2[0]+dv21

v2[1]=v2[1]+dv22

print("v2=",v2)

w[1]=w[1]+dw1

w[2]=w[2]+dw2

b1=b1+dv01

b2=b2+dv02

w[0]=w[0]+dw0

print("w=",w)

print("bias b1=",b1, " b2=",b2)
```

```
====== RESTART: C:/Users/Janhavi/AppData/Local/Programs/Pvthon/Pvthon38/RunAllInOneGo.pv =======
calculate net input to zl layer
z1 = 0.2
calculate net input to z2 layer
z2 = 0.9
Apply activation function to calculate output
z2= 0.7109
calculate net input to output layer
yin= 0.09101
calculate net output
y= 0.5227368084248941
dk= 0.11906907074145694
compute error portion in delta
din1= 0.04762762829658278
din2= 0.011906907074145694
error in delta
fzin1= 0.24751996
fzin2= 0.20552119000000002
d1= 0.011788788650865037
d2= 0.0024471217110978417
Changes in weights between input and hidden layer
dv11 = 0.0
dv21= 0.0029471971627162592
dv01= 0.0029471971627162592
dv12 = 0.0
dv22= 0.0006117804277744604
dv02= 0.0006117804277744604
Final weights of network
v1= [0.6 0.3]
v2= [-0.1 0.4]
w= [-0.17 0.42 0.12]
bias bl= 0.30294719716271623 b2= 0.5006117804277744
>>>
```

Practical 4B

Aim- Write a program for error Backpropagation algorithm.

Code:

```
import math
a0 = -1
t = -1
w10=float(input("Enter weight first network"))
b10=float(input("Enter base first network:"))
w20=float(input("Enter weight second network:"))
b20=float(input("Enter base second network:"))
c=float(input("Enter learning coefficient:"))
n1 = float(w10*c+b10)
a1=math.tanh(n1)
n2 = float(w20*a1+b20)
a2=math.tanh(float(n2))
e=t-a2
s2=-2*(1-a2*a2)*e
s1=(1-a1*a1)*w20*s2
w21=w20-(c*s2*a1)
w11=w10-(c*s1*a0)
b21=b20-(c*s2)
b11=b10-(c*s1)
print("The updated weight of first n/w w11=",w11)
print("The uploaded weight of second n/w w21= ",w21)
print("The updated base of first n/w b10=",b10)
print("The updated base of second n/w b20= ",b20)
```

Practical 5A

Aim-Write a program for Hopfield Network.

```
class Neuron:
  def __init__(self, j):
     self.activation = 0
     self.weightv = [0] * 4
     for i in range(4):
        self.weightv[i] = j[i]
  def act(self, m, x):
     a = 0
     for i in range(m):
       a += x[i] * self.weightv[i]
     return a
class Network:
  def __init__(self, a, b, c, d):
     self.nrn = [None] * 4
     self.output = [0] * 4
     self.nrn[0] = Neuron(a)
     self.nrn[1] = Neuron(b)
     self.nrn[2] = Neuron(c)
     self.nrn[3] = Neuron(d)
  def threshld(self, k):
     if k \ge 0:
       return 1
     else:
       return 0
  def activation(self, patrn):
     for i in range(4):
       for j in range(4):
          print("\n nrn[{}].weightv[{}] is {}".format(i, j,
self.nrn[i].weightv[j]))
        self.nrn[i].activation = self.nrn[i].act(4, patrn)
```

```
print("\nactivation is", self.nrn[i].activation)
       self.output[i] = self.threshld(self.nrn[i].activation)
       print("\noutput value is", self.output[i])
def main():
  patrn1 = [1, 0, 1, 0]
  wt1 = [0, -3, 3, -3]
  wt2 = [-3, 0, -3, 3]
  wt3 = [3, -3, 0, -3]
  wt4 = [-3, 3, -3, 0]
  print("\nTHIS PROGRAM IS FOR A HOPFIELD NETWORK WITH A
SINGLE LAYER OF")
  print("4 FULLY INTERCONNECTED NEURONS. THE NETWORK
SHOULD RECALL THE")
  print("PATTERNS 1010 AND 0101 CORRECTLY.\n")
  # create the network by calling its constructor.
  # the constructor calls neuron constructor as many times as the number of
  # neurons in the network.
  h1 = Network(wt1, wt2, wt3, wt4)
  # present a pattern to the network and get the activations of the neurons
  h1.activation(patrn1)
  # check if the pattern given is correctly recalled and give message
  for i in range(4):
    if h1.output[i] == patrn1[i]:
       print("\n pattern=", patrn1[i], " output =", h1.output[i], " component
matches")
    else:
       print("\n pattern=", patrn1[i], " output =", h1.output[i], " discrepancy
occurred")
  print("\n\n")
  patrn2 = [0, 1, 0, 1]
  h1.activation(patrn2)
  for i in range(4):
    if h1.output[i] == patrn2[i]:
       print("\n pattern=", patrn2[i], " output =", h1.output[i], " component
matches")
```

```
else:
    print("\n pattern=", patrn2[i], " output =", h1.output[i], " discrepancy
occurred")
if __name__ == "__main__":
    main()
```

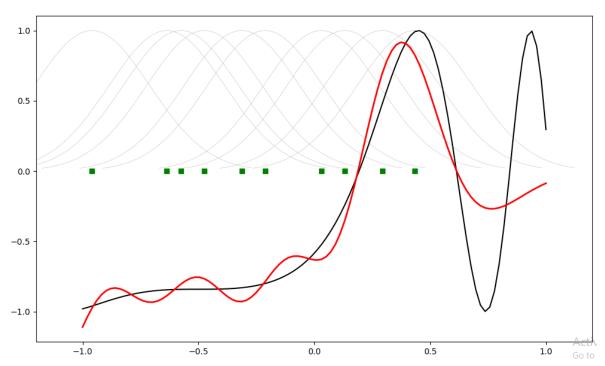
```
>>>
= RESTART: C:/Users/Janhavi/AppData/Local/Programs/Python/Python38/RunAllInOneGo.py
THIS PROGRAM IS FOR A HOPFIELD NETWORK WITH A SINGLE LAYER OF
4 FULLY INTERCONNECTED NEURONS, THE NETWORK SHOULD RECALL THE
PATTERNS 1010 AND 0101 CORRECTLY.
 nrn[0].weightv[0] is 0
 nrn[0].weightv[1] is -3
 nrn[0].weightv[2] is 3
nrn[0].weightv[3] is -3
activation is 3
 nrn[1].weightv[0] is -3
 nrn[1].weightv[1] is 0
 nrn[1].weightv[2] is -3
 nrn[1].weightv[3] is 3
activation is -6
output value is 0
nrn[2].weightv[0] is 3
 nrn[2].weightv[1] is -3
 nrn[2].weightv[2] is 0
nrn[2].weightv[3] is -3
 nrn[2].weightv[3] is -3
activation is 3
output value is 1
 nrn[3].weightv[0] is -3
 nrn[3].weightv[1] is 3
 nrn[3].weightv[2] is -3
 nrn[3].weightv[3] is 0
activation is -6
 pattern= 1 output = 1 component matches
 pattern= 0 output = 0 component matches
pattern= 1 output = 1 component matches
 pattern= 0 output = 0 component matches
 nrn[0].weightv[0] is 0
 nrn[0].weightv[1] is -3
 nrn[0].weightv[2] is 3
 nrn[0].weightv[3] is -3
activation is -6
output value is 0
nrn[1].weightv[2] is -3
nrn[1].weightv[3] is 3
activation is 3
nrn[2].weightv[0] is 3
nrn[2].weightv[1] is -3
nrn[2].weightv[2] is 0
nrn[2].weightv[3] is -3
output value is 0
nrn[3].weightv[0] is -3
nrn[3].weightv[1] is 3
nrn[3].weightv[3] is 0
activation is 3
output value is 1
pattern= 0 output = 0 component matches
pattern= 1 output = 1 component matches
pattern= 0 output = 0 component matches
pattern= 1 output = 1 component matches
```

Practical 5B

Aim-Write a program for Radial Basis function

```
from scipy import *
from scipy.linalg import norm, pinv
from matplotlib import pyplot as plt
from numpy import sin, exp, zeros, dot, array, random, arange
import numpy as np
class RBF:
  def __init__(self, indim, numCenters, outdim):
     self.indim = indim
     self.outdim = outdim
     self.numCenters = numCenters
     self.centers = [random.uniform(-1, 1, indim) for i in range(numCenters)]
     self.beta = 8
     self.W = random.random((self.numCenters, self.outdim))
  def _basisfunc(self, c, d):
     assert len(d) == self.indim
     return exp(-self.beta * norm(c-d)**2)
  def _calcAct(self, X):
     G = zeros((X.shape[0], self.numCenters), float)
     for ci, c in enumerate(self.centers):
       for xi, x in enumerate(X):
          G[xi, ci] = self.\_basisfunc(c, x)
     return G
  def train(self, X, Y):
     """ X: matrix of dimensions n x indim
        y: column vector of dimension n x 1 """
     # choose random center vectors from training set
     rnd_idx = random.permutation(X.shape[0])[:self.numCenters]
     self.centers = [X[i,:] for i in rnd_idx]
     print("center", self.centers)
     # calculate activations of RBFs
```

```
G = self.\_calcAct(X)
     print(G)
     # calculate output weights (pseudoinverse)
     self.W = dot(pinv(G), Y)
  def test(self, X):
     """ X: matrix of dimensions n x indim """
     G = self.\_calcAct(X)
     Y = dot(G, self.W)
    return Y
if __name__ == '__main__':
  # ---- 1D Example -----
  n = 100
  x = np.mgrid[-1:1:complex(0,n)].reshape(n, 1)
  # set y and add random noise
  y = \sin(3*(x+0.5)**3-1)
  # rbf regression
  rbf = RBF(1, 10, 1)
  rbf.train(x, y)
  z = rbf.test(x)
  # plot original data
  plt.figure(figsize=(12, 8))
  plt.plot(x, y, 'k-')
  # plot learned model
  plt.plot(x, z, 'r-', linewidth=2)
  # plot rbfs
  plt.plot(rbf.centers, zeros(rbf.numCenters), 'gs')
  for c in rbf.centers:
    # RF prediction lines
    cx = arange(c-0.7, c+0.7, 0.01)
    cy = [rbf._basisfunc(array([cx_]), array([c])) for cx_ in cx]
     plt.plot(cx, cy, '-', color='gray', linewidth=0.2)
  plt.xlim(-1.2, 1.2)
  plt.show()
```



Practical 6A

Aim- Kohonen Self organizing map (SOM)

Code:

from minisom import MiniSom

import matplotlib.pyplot as plt

data = [[0.80, 0.55, 0.22, 0.03],

[0.82, 0.50, 0.23, 0.03],

[0.80, 0.54, 0.22, 0.03],

[0.80, 0.53, 0.26, 0.03],

[0.79, 0.56, 0.22, 0.03],

[0.75, 0.60, 0.25, 0.03],

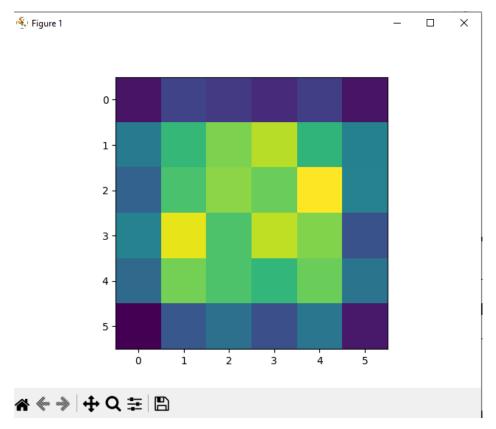
[0.77, 0.59, 0.22, 0.03]

som = MiniSom(6, 6, 4, sigma=0.3, learning_rate=0.5) # initialization of 6x6 SOM

som.train_random(data, 100) # trains the SOM with 100 iterations

plt.imshow(som.distance_map())

plt.show()

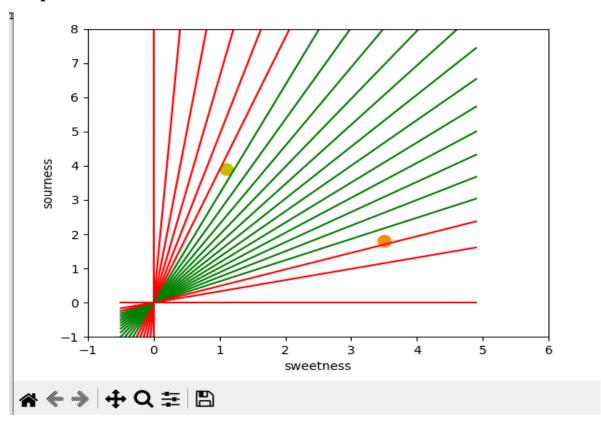


Practical 7A

Aim- Write a program for Linear separation.

```
import numpy as np
import matplotlib.pyplot as plt
def create_distance_function(a, b, c):
  """ 0 = ax + by + c """
  def distance(x, y):
     """ returns tuple (d, pos)
     d is the distance
     If pos == -1 point is below the line,
     0 on the line and +1 if above the line"""
     nom = a * x + b * y + c
     if nom == 0:
       pos = 0
     elif (nom<0 and b<0) or (nom>0 and b>0):
       pos = -1
     else:
       pos = 1
     return (np.absolute(nom) / np.sqrt( a ** 2 + b ** 2), pos)
  return distance
points = [(3.5, 1.8), (1.1, 3.9)]
fig, ax = plt.subplots()
ax.set_xlabel("sweetness")
ax.set_ylabel("sourness")
ax.set_xlim([-1, 6])
ax.set_ylim([-1, 8])
X = np.arange(-0.5, 5, 0.1)
colors = ["r", ""] # for the samples
size = 10
for (index, (x, y)) in enumerate(points):
  if index == 0:
```

```
ax.plot(x, y, "o", color="darkorange", markersize=size)
  else:
     ax.plot(x, y, "oy", markersize=size)
step = 0.05
for x in np.arange(0, 1+step, step):
  slope = np.tan(np.arccos(x))
  dist4line1 = create_distance_function(slope, -1, 0)
  Y = slope * X
  results = []
  for point in points:
     results.append(dist4line1(*point))
  if (results[0][1] != results[1][1]):
    ax.plot(X, Y, "g-")
  else:
     ax.plot(X, Y, "r-")
plt.show()
```

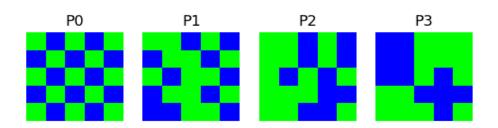


Practical 7B

Aim- Write a program for Hopfield network model for associative memory.

Code:

```
import numpy as np
# Patch np.int to use the built-in int
np.int = int
# Now import the rest of the modules
from neurodynex.hopfield_network import network, pattern_tools, plot_tools
pattern\_size = 5
hopfield_net = network.HopfieldNetwork(nr_neurons=pattern_size**2)
factory = pattern_tools.PatternFactory(pattern_size, pattern_size)
checkerboard = factory.create_checkerboard()
pattern_list = [checkerboard]
pattern_list.extend(factory.create_random_pattern_list(nr_patterns=3,
on_probability=0.5))
plot_tools.plot_pattern_list(pattern_list)
overlap matrix = pattern tools.compute overlap matrix(pattern list)
plot_tools.plot_overlap_matrix(overlap_matrix)
hopfield_net.store_patterns(pattern_list)
noisy init state = pattern tools.flip n(checkerboard, nr of flips=4)
hopfield_net.set_state_from_pattern(noisy_init_state)
states = hopfield_net.run_with_monitoring(nr_steps=4)
states_as_patterns = factory.reshape_patterns(states)
plot_tools.plot_state_sequence_and_overlap(states_as_patterns, pattern_list,
reference_idx=0, suptitle="Network dynamics")
```



Practical 8A

Aim- Membership and Identity Operators | in, not in.

```
# Python program to illustrate
# Finding common member in list
# without using 'in' operator
# Define a function() that takes two lists
def overlapping(list1, list2):
  c = 0
  d = 0
  for i in list1:
     c += 1
  for i in list2:
     d += 1
  for i in range(0, c):
     for j in range(0, d):
        if(list1[i] == list2[j]):
          return 1
  return 0
list1 = [1, 2, 3, 4, 5]
list2 = [6,7,8,9]
if(overlapping(list1, list2)):
  print("overlapping")
else:
  print("not overlapping")
```

Practical 8B

Aim- Membership and Identity Operators is, is not

Code:

```
# Python program to illustrate the use
# of 'is' identity operator
x = 5
if (type(x) is int):
    print("true")
else:
    print("false")
# Python program to illustrate the
# use of 'is not' identity operator
x = 5.2
if (type(x) is not int):
    print("true")
else:
    print("false")
```

```
>>> = RESTART: C:/Users/Janhavi/AppData/Local/Programs/Python/Python38/RunAllInOneGo
.py
not overlapping
>>> = RESTART: C:/Users/Janhavi/AppData/Local/Programs/Python/Python38/RunAllInOneGo
.py
true
true
>>> |
```

Practical 9A

Aim- Find ratios using fuzzy logic.

Code:

```
from fuzzywuzzy import fuzz

from fuzzywuzzy import process

s1 = "I love fuzzysforfuzzys"

s2 = "I am loving fuzzysforfuzzys"

print("FuzzyWuzzy Ratio:", fuzz.ratio(s1, s2))

print("FuzzyWuzzyPartialRatio: ", fuzz.partial_ratio(s1, s2))

print("FuzzyWuzzyTokenSortRatio: ", fuzz.token_sort_ratio(s1, s2))

print("FuzzyWuzzyTokenSetRatio: ", fuzz.token_set_ratio(s1, s2))

print("FuzzyWuzzyWRatio: ", fuzz.WRatio(s1, s2), '\n\n')

# for process library,

query = 'fuzzys for fuzzys'

choices = ['fuzzy for fuzzy', 'fuzzy fuzzy', 'g. for fuzzys']

print("List of ratios: ")

print(process.extract(query, choices), '\n')

print("Best among the above list: ", process.extractOne(query, choices))
```

Output:

```
Type "help", "copyright", "credits" or "license()" for more information.

>>>

RESTART: C:/Users/Janhavi/AppData/Local/Programs/Python/Python38/RunAllInOneGo.py
FuzzyWuzzy Ratio: 86
FuzzyWuzzyPartialRatio: 86
FuzzyWuzzyTokenSortRatio: 86
FuzzyWuzzyTokenSetRatio: 87
FuzzyWuzzyWatio: 86

List of ratios:
[('g. for fuzzys', 95), ('fuzzy for fuzzy', 94), ('fuzzy fuzzy', 86)]

Best among the above list: ('g. for fuzzys', 95)

>>> |
```

Practical 9B

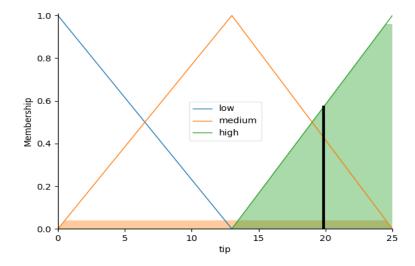
Aim- Solve Tipping problem using fuzzy logic.

install - pip install scikit-fuzzy

```
import numpy as np
import skfuzzy as fuzz
from skfuzzy import control as ctrl
```

```
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')
quality.automf(3)
service.automf(3)
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])
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'])
tipping_ctrl = ctrl.ControlSystem([rule1, rule2, rule3])
tipping = ctrl.ControlSystemSimulation(tipping_ctrl)
tipping.input['quality'] = 6.5
tipping.input['service'] = 9.8
tipping.compute()
print(tipping.output['tip'])
tip.view(sim=tipping)
```

```
>>>
= RESTART: C:/Users/Janhavi/AppData/Local/Programs/Python/Python38/RunAllInOneGo
.py
19.847607361963192
>>> |
```



Practical 10A

Aim- Implementation of Simple genetic algorithm.

```
import random
# Number of individuals in each generation
POPULATION SIZE = 100
# Valid genes
GENES =
"abcdefghijklmnopqrstuvwxyzABCDEFGHIJKLMNOPQRSTUVWXYZ
1234567890, .-;:_!"#%&/()=?@${[]}"
# Target string to be generated
TARGET = "Janhavi kharkar"
class Individual(object):
  "Class representing individual in population "
  def __init__(self, chromosome):
    self.chromosome = chromosome
    self.fitness = self.cal_fitness()
  @classmethod
  def mutated_genes(cls):
    "create random genes for mutation "
    global GENES
    gene = random.choice(GENES)
    return gene
  @classmethod
  def create_gnome(cls):
    "create chromosome or string of gene "
    global TARGET
    gnome\_len = len(TARGET)
    return [cls.mutated_genes() for _ in range(gnome_len)]
  def mate(self, par2):
    "" Perform mating and produce new offspring "
    # chromosome for offspring
    child_chromosome = []
    for gp1, gp2 in zip(self.chromosome, par2.chromosome):
```

```
# random probability
       prob = random.random()
       # if prob is less than 0.45, insert gene
       # from parent 1
       if prob < 0.45:
         child_chromosome.append(gp1)
       # if prob is between 0.45 and 0.90, insert
       # gene from parent 2
       elif prob < 0.90:
         child_chromosome.append(gp2)
       # otherwise insert random gene(mutate),
       # for maintaining diversity
       else:
         child_chromosome.append(self.mutated_genes())
    # create new Individual(offspring) using
    # generated chromosome for offspring
    return Individual(child_chromosome)
  def cal_fitness(self):
     "Calculate fittness score, it is the number of characters in string which
differ from target string. "
    global TARGET
    fitness = 0
    for gs, gt in zip(self.chromosome, TARGET):
       if gs != gt:
         fitness += 1
    return fitness
# Driver code
def main():
  global POPULATION_SIZE
  # current generation
  generation = 1
  found = False
  population = []
  # create initial population
pg. 33
```

```
for _ in range(POPULATION_SIZE):
    gnome = Individual.create_gnome()
    population.append(Individual(gnome))
  while not found:
    # sort the population in increasing order of fitness score
    population = sorted(population, key=lambda x: x.fitness)
    # if the individual having lowest fitness score ie.
    # 0 then we know that we have reached to the target
    # and break the loop
    if population[0].fitness \leq 0:
       found = True
       break
    # Otherwise generate new offsprings for new generation
    new_generation = []
    # Perform Elitism, that mean 10% of fittest population
    # goes to the next generation
    s = int((10 * POPULATION_SIZE) / 100)
    new_generation.extend(population[:s])
    # From 50% of fittest population, Individuals
    # will mate to produce offspring
    s = int((90 * POPULATION_SIZE) / 100)
    for _ in range(s):
       parent1 = random.choice(population[:50])
       parent2 = random.choice(population[:50])
       child = parent1.mate(parent2)
       new_generation.append(child)
    population = new_generation
    print("Generation: {}\tString: {}\tFitness: {}".format(generation,
"".join(population[0].chromosome), population[0].fitness))
    generation += 1
  print("Generation: {}\tString: {}\tFitness: {}".format(generation,
"".join(population[0].chromosome), population[0].fitness))
if name == ' main ':
  main()
```

```
D64)1 on win32
Type "help", "copyright", "credits" or "license()" for more information.
= RESTART: C:/Users/Janhavi/AppData/Local/Programs/Pvthon/Pvthon38/RunAllInOneGo
.py
Generation: 1
               String: XalhdKo4CAlM, $H Fitness: 13
Generation: 2
               String: XalhdKo4CAlM,$H Fitness: 13
Generation: 3
               String: DmnhTHq{OJaCh;% Fitness: 12
Generation: 4
                String: DmnhTHq{OJaCh;% Fitness:
Generation: 5
               String: DmnhTHq{OJaCh;% Fitness: 12
                String: F}khXyi Jhan%P& Fitness:
Generation:
               String: 5}nhWHi JhaC n% Fitness:
Generation: 7
Generation: 8
                String: Danhh@iQ&haK82r Fitness:
Generation: 9
                String: ]aihhHi &haKk;r Fitness:
Generation: 10
               String: laihhHi &haKk;r Fitness:
Generation: 11
              String: JanhfPi &haM8gr Fitness:
Generation: 12
               String: JanhfPi &haM8gr Fitness:
Generation: 13
              String: JanhqPi 3haM8ar Fitness:
Generation: 14
               String: JanhqPi 3haM8ar Fitness:
Generation: 15
               String: JanhqPi 3haM8ar Fitness:
               String: JanhqPi 3haM8ar Fitness:
Generation: 16
Generation: 17
               String: Janhhviy3haMkar Fitness:
Generation: 18
               String: Janhhviv3haMkar Fitness:
Generation: 19
               String: JanhMvi }ha#kar Fitness:
               String: JanhMvi }ha#kar Fitness:
Generation: 20
               String: JanhMvi }ha#kar Fitness:
Generation: 21
Generation: 22
               String: JanhMvi }ha#kar Fitness:
               String: Janhavi lha$kar Fitness:
Generation: 23
               String: Janhavi lha$kar Fitness:
Generation: 24
Generation: 25
               String: Janhavi lha$kar Fitness:
Generation: 26
               String: Janhavi lha$kar Fitness:
Generation: 27
               String: Janhavi lha$kar Fitness:
Generation: 28
               String: Janhavi lha$kar Fitness:
Generation: 29
               String: Janhavi lha$kar Fitness: 2
Generation: 30
               String: Janhavi lha$kar Fitness: 2
Generation: 31
               String: Janhavi lha$kar Fitness: 2
Generation: 32
               String: Janhavi lha$kar Fitness:
Generation: 33
               String: Janhavi lha$kar Fitness: 2
```

```
Generation: 26
               String: Janhavi lha$kar Fitness: 2
Generation: 27
                String: Janhavi lha$kar Fitness:
               String: Janhavi lha$kar Fitness:
Generation: 28
Generation: 29
                String: Janhavi lha$kar Fitness:
Generation: 30
               String: Janhavi lha$kar Fitness:
Generation: 31
               String: Janhavi lha$kar Fitness:
Generation: 32
               String: Janhavi lha$kar Fitness:
Generation: 33
               String: Janhavi lha$kar Fitness:
Generation: 34
               String: Janhavi lha$kar Fitness:
               String: Janhavi lha$kar Fitness:
Generation: 35
               String: Janhavi lha$kar Fitness:
Generation: 36
Generation: 37
               String: Janhavi lha$kar Fitness:
               String: Janhavi lha$kar Fitness:
Generation: 38
Generation: 39
               String: Janhavi lhaSkar Fitness:
               String: Janhavi lha$kar Fitness:
Generation: 40
               String: Janhavi Rharkar Fitness:
Generation: 41
Generation: 42
               String: Janhavi Rharkar Fitness:
Generation: 43
               String: Janhavi Rharkar Fitness:
Generation: 44
               String: Janhavi Rharkar Fitness:
Generation: 45
               String: Janhavi Rharkar Fitness:
Generation: 46
               String: Janhavi Rharkar Fitness:
Generation: 47
               String: Janhavi Rharkar Fitness:
Generation: 48
               String: Janhavi Rharkar Fitness:
Generation: 49
               String: Janhavi Rharkar Fitness:
Generation: 50
               String: Janhavi Rharkar Fitness:
Generation: 51
               String: Janhavi Rharkar Fitness:
Generation: 52
                String: Janhavi Rharkar Fitness:
Generation: 53
              String: Janhavi Rharkar Fitness:
Generation: 54
               String: Janhavi Rharkar Fitness:
Generation: 55
               String: Janhavi Rharkar Fitness:
               String: Janhavi Rharkar Fitness:
Generation: 56
Generation: 57
               String: Janhavi Rharkar Fitness:
Generation: 58
               String: Janhavi Rharkar Fitness:
Generation: 59
               String: Janhavi Rharkar Fitness:
Generation: 60
               String: Janhavi Rharkar Fitness:
Generation: 61
               String: Janhavi Rharkar Fitness:
Generation: 62
               String: Janhavi Rharkar Fitness:
Generation: 63
               String: Janhavi Rharkar Fitness: 1
Generation: 64 String: Janhavi kharkar Fitness: 0
```

Practical 10B

Aim- Create two classes: City and Fitness using Genetic algorithm.

```
import numpy as np
import random
import operator
import pandas as pd
import matplotlib.pyplot as plt
from tkinter import Tk, Canvas, Frame, BOTH, Text
import math
class City:
  def __init__(self, x, y):
     self.x = x
     self.y = y
  def distance(self, city):
     xDis = abs(self.x - city.x)
     yDis = abs(self.y - city.y)
     distance = np.sqrt((xDis ** 2) + (yDis ** 2))
     return distance
  def __repr__(self):
     return "(" + str(self.x) + "," + str(self.y) + ")"
class Fitness:
  def __init__(self, route):
     self.route = route
     self.distance = 0
     self.fitness = 0.0
  def routeDistance(self):
     if self.distance == 0:
       pathDistance = 0
       for i in range(0, len(self.route)):
          fromCity = self.route[i]
          toCity = None
```

```
if i + 1 < len(self.route):
            toCity = self.route[i + 1]
           else:
            toCity = self.route[0]
          pathDistance += fromCity.distance(toCity)
       self.distance = pathDistance
     return self.distance
  def routeFitness(self):
     if self.fitness == 0:
       self.fitness = 1 / float(self.routeDistance())
     return self.fitness
def createRoute(cityList):
  route = random.sample(cityList, len(cityList))
  return route
def initialPopulation(popSize, cityList):
  population = []
  for i in range(0, popSize):
     population.append(createRoute(cityList))
  return population
def rankRoutes(population):
  fitnessResults = {}
  for i in range(0, len(population)):
     fitnessResults[i] = Fitness(population[i]).routeFitness()
  return sorted(fitnessResults.items(), key=operator.itemgetter(1),
reverse=True)
def selection(popRanked, eliteSize):
  selectionResults = []
  df = pd.DataFrame(np.array(popRanked), columns=["Index", "Fitness"])
  df['cum_sum'] = df.Fitness.cumsum()
  df['cum_perc'] = 100 * df.cum_sum / df.Fitness.sum()
  for i in range(0, eliteSize):
     selectionResults.append(popRanked[i][0])
  for i in range(0, len(popRanked) - eliteSize):
     pick = 100 * random.random()
pg. 37
```

```
for i in range(0, len(popRanked)):
       if pick \leq df.iat[i, 3]:
          selectionResults.append(popRanked[i][0])
          break
  return selectionResults
def matingPool(population, selectionResults):
  matingpool = []
  for i in range(0, len(selectionResults)):
     index = selectionResults[i]
     matingpool.append(population[index])
  return matingpool
def breed(parent1, parent2):
  child = []
  childP1 = []
  childP2 = []
  geneA = int(random.random() * len(parent1))
  geneB = int(random.random() * len(parent1))
  startGene = min(geneA, geneB)
  endGene = max(geneA, geneB)
  for i in range(startGene, endGene):
     childP1.append(parent1[i])
  childP2 = [item for item in parent2 if item not in childP1]
  child = childP1 + childP2
  return child
def breedPopulation(matingpool, eliteSize):
  children = []
  length = len(matingpool) - eliteSize
  pool = random.sample(matingpool, len(matingpool))
  for i in range(0, eliteSize):
     children.append(matingpool[i])
  for i in range(0, length):
     child = breed(pool[i], pool[len(matingpool) - i - 1])
     children.append(child)
```

```
return children
def mutate(individual, mutationRate):
  for swapped in range(len(individual)):
    if(random.random() < mutationRate):</pre>
       swapWith = int(random.random() * len(individual))
       city1 = individual[swapped]
       city2 = individual[swapWith]
       individual[swapped] = city2
       individual[swapWith] = city1
  return individual
def mutatePopulation(population, mutationRate):
  mutatedPop = []
  for ind in range(0, len(population)):
    mutatedInd = mutate(population[ind], mutationRate)
    mutatedPop.append(mutatedInd)
  return mutatedPop
def nextGeneration(currentGen, eliteSize, mutationRate):
  popRanked = rankRoutes(currentGen)
  selectionResults = selection(popRanked, eliteSize)
  matingpool = matingPool(currentGen, selectionResults)
  children = breedPopulation(matingpool, eliteSize)
  nextGeneration = mutatePopulation(children, mutationRate)
  return nextGeneration
def geneticAlgorithm(population, popSize, eliteSize, mutationRate,
generations):
  pop = initialPopulation(popSize, population)
  print("Initial distance: " + str(1 / rankRoutes(pop)[0][1]))
  for i in range(0, generations):
    pop = nextGeneration(pop, eliteSize, mutationRate)
  print("Final distance: " + str(1 / rankRoutes(pop)[0][1]))
  bestRouteIndex = rankRoutes(pop)[0][0]
  bestRoute = pop[bestRouteIndex]
  return bestRoute
```

```
def geneticAlgorithmPlot(population, popSize, eliteSize, mutationRate,
generations):
  pop = initialPopulation(popSize, population)
  progress = []
  progress.append(1 / rankRoutes(pop)[0][1])
  for i in range(0, generations):
    pop = nextGeneration(pop, eliteSize, mutationRate)
    progress.append(1 / rankRoutes(pop)[0][1])
  plt.plot(progress)
  plt.ylabel('Distance')
  plt.xlabel('Generation')
  plt.show()
def main():
  cityList = []
  for i in range(0, 25):
    cityList.append(City(x=int(random.random() * 200),
y=int(random.random() * 200)))
  geneticAlgorithmPlot(population=cityList, popSize=100, eliteSize=20,
mutationRate=0.01, generations=500)
if __name__ == '__main__':
main()
```

