10. Fuzzy Arithmetic operations

```
def fu_add(a,b):
  res={}
  for k_a,v_a in a.items():
     for k_b,v_b in b.items():
       if k_a == k_b:
         res[k_a]=max(v_a,v_b)
         break
  return res
def fu_sub(a,b):
  res={}
  for k_a,v_a in a.items():
     for k_b,v_b in b.items():
       if k_a == k_b:
         res[k_a]=max(v_a - v_b,0)
         break
  return res
def fu_mul(a,b):
  res={}
  for k_a,v_a in a.items():
     for k_b,v_b in b.items():
       if k_a == k_b:
         res[k_a]=v_a * v_b
         break
  return res
def fu_div(a,b):
  res={}
  for k_a,v_a in a.items():
     for k_b,v_b in b.items():
       if k_a == k_b and v_b!=0:
         res[k a]=v a/v b
```

```
break
```

return res

$$B {=} \{"a":0.9,"b":0.9,"c":0.4,"d":0.5\}$$

 $c=fu_add(A,B)$

print("Fuzzy Addition:",c)

d=fu_sub(A,B)

print("Fuzzy Subtraction:",d)

 $e=fu_mul(A,B)$

print("Fuzzy Multiplication:",e)

 $f=fu_div(A,B)$

```
9. Fuzzy Logical Operations:
A=dict()
B=dict()
Y=dict()
A={"a":0.2,"b":0.3,"c":0.6,"d":0.6}
B = \{"a":0.9,"b":0.9,"c":0.4,"d":0.5\}
print("The First Fuzzy Set is :",A)
print("The Second Fuzzy Set is :",B)
for k_a,k_b in zip(A,B):
  a_val=A[k_a]
  b_val=B[k_a]
  if a_val > b_val:
    Y[k_a]=a_val
  else:
    Y[k_b]=b_val
print("The Fuzzy set union is:",Y)
print("\nThe First Fuzzy Set is :",A)
print("The Second Fuzzy Set is :",B)
for k_a,k_b in zip(A,B):
  a_val=A[k_a]
  b_val=B[k_a]
  if a_val < b_val:
    Y[k_a]=a_val
  else:
    Y[k_b]=b_val
print("The Fuzzy set intersection is :",Y)
print("\nThe Fuzzy Set is :",A)
for k_a in A:
  Y[k_a]=1-A[k_a]
print("The Fuzzy set complement is :",Y)
```

6. De_Morgan's Law

```
if 'not (' in statement and ' and ' in statement:
    A, B = statement.split('not (')[1].split(' and ')
    return f'(not {A}) or (not {B})'

elif 'not (' in statement and ' or ' in statement:
    A, B = statement.split('not (')[1].split(' or ')
    return f'(not {A}) and (not {B})'

else:
    print("Invalid Statement")
    return statement

result = demorgan_law('not (A and B)')

print("De Morgan's Law to 'not (A and B)' is",result)

result = demorgan_law('not (A or B)')

print("De Morgan's Law to 'not (A or B)' is",result)
```

4.SOM

import math

```
class SOM:
  def winner(self,weight,sample):
     D0,D1=0,0
     for i in range(len(sample)):
       D0+=math.pow((sample[i]-weight[0][i]),2)
       D1+=math.pow((sample[i]-weight[1][i]),2)
    return 0 if D0>D1 else 0
  def update(self,weight,sample,J,alpha):
     for i in range(len(weight)):
       weight[J][i]+=alpha*(sample[i]-weight[J][i])
    return weight
def main():
  T = [[1,1,0,0],[0,0,0,1],[1,0,0,0],[0,0,1,1]]
  weight=[[0.2,0.6,0.8,0.9],[0.4,0.5,0.3,0.6]]
  som=SOM()
  alpha=0.5
  epoch=3000
  for i in range(epoch):
     for j in range(len(T)):
       sample=T[j]
       J=som.winner(weight,sample)
       weight=som.update(weight,sample,J,alpha)
  t_s=[0,0,0,1]
  J=som.winner(weight,t_s)
  print("The sample cluster is:",J)
  print("Train weight is:",weight)
if __name__=="__main__":
  main()
```

2.PERCEPTRON:

```
import numpy as np
def unitStep(v):
  return 1 if v \ge 0 else 0
def perceptrionModel(x,w,b):
  v=np.dot(w,x)+b
  y=unitStep(v)
  return y
def OR_logic(x):
  w=np.array([1,1])
  b=0.5
  return perceptrionModel(x,w,b)
tests=[
  (np.array([0,1]),1),
  (np.array([1,1]),1),
  (np.array([0,0]),0),
  (np.array([1,0]),1)
for test in tests:
  x,expected=test
  result=OR_logic(x)
  print(f"OR(\{x[0]\},\{x[1]\}) = \{result\} (expected \{expected\})")
```

1.ANN

```
import numpy as np
from matplotlib import pyplot as plt
def sigmoid(z):
  return 1/(1+np.exp(-z))
def inPa(inFe,neHid,ouFe):
  w1 = np.random.randn(neHid,inFe)
  w2 = np.random.randn(ouFe,neHid)
  b1 = np.zeros((neHid,1))
  b2 = np.zeros((ouFe,1))
  parameters = \{"w1":w1,"w2":w2,"b1":b1,"b2":b2\}
  return parameters
def foPa(X,Y,parameters):
  m=X.shape[1]
  w1=parameters["w1"]
  w2=parameters["w2"]
  b1=parameters["b1"]
  b2=parameters["b2"]
  z1=np.dot(w1,X)+b1
  a1=sigmoid(z1)
  z2=np.dot(w2,a1)+b2
  a2=sigmoid(z2)
  cahce=(z1,a1,w1,b1,z2,a2,w2,b2)
  logprobs=np.multiply(np.log(a2),Y)+np.multiply(np.log(1-a2),(1-Y))
  cost=-np.sum(logprobs)/m
  return cost,cahce,a2
def baPa(X,Y,cache):
  m=X.shape[1]
```

```
(z_1,a_1,w_1,b_1,z_2,a_2,w_2,b_2)=cache
  dz2=a2-Y
  dw2=np.dot(dz2,a1.T)/m
  db2=np.sum(dz2,axis=1,keepdims=True)
  da1=np.dot(w2.T,dz2)
  dz1=np.multiply(da1,a1*(1-a1))
  dw1=np.dot(dz1,X.T)/m
  db1=np.sum(dz1,axis=1,keepdims=True)/m
  gradients = {\text{"dz2":dz2,"dw2":dw2,"db2":db2,"dz1":dz1,"dw1":dw1,"db1":db1}}
  return gradients
def upPa(gradients,learningRate,parameters):
  parameters["w1"]=parameters["w1"]-learningRate*gradients["dw1"]
  parameters["w2"]=parameters["w2"]-learningRate*gradients["dw2"]
  parameters["b1"]=parameters["b1"]-learningRate*gradients["db1"]
  parameters["b2"]=parameters["b2"]-learningRate*gradients["db2"]
  return parameters
X = \text{np.array}([[0,0,1,1],[0,1,0,1]])
Y = np.array([[0,0,0,1]])
neHid=2
inFe=X.shape[0]
ouFe=Y.shape[0]
parameters = inPa(inFe,neHid,ouFe)
learningRate=0.01
epoch=100000
losses =np.zeros((epoch,1))
for i in range(epoch):
  losses[i,0],cache,a2=foPa(X,Y,parameters)
  gradients=baPa(X,Y,cache)
  parameters=upPa(gradients,learningRate,parameters)
```

```
plt.figure()
plt.plot(losses)
plt.xlabel("Epoch")
plt.ylabel("Loss Value")
plt.show()

X=np.array([[1,1,0,0,],[0,1,0,1]])
cost,__,a1=foPa(X,Y,parameters)
predict=(a2>0.5)*1.0

print(predict)
```