

1. McCulloch-Pitts neuron XOR gate.

PROGRAM:

```
x1=[0,0,1,1]
x2=[0,1,0,1]
z=[0,1,1,0]
con=True
while con:
    w11=float(input("weight w11="))
    w12=float(input("weight w12="))
    w21=float(input("weight w21="))
    w22=float(input("weight w22="))
    v1=float(input("weight v1="))
    v2=float(input("weight v2="))
    theta=float(input("theta="))
    zin1=[x1[i]*w11+x2[i]*w21 for i in range(4)]
    zin2=[x1[i]*w21+x2[i]*w22 for i in range(4)]
    y1=[1 if zin1[i]>=theta else 0 for i in range(4)]
    y2=[1 if zin2[i]>=theta else 0 for i in range(4)]
    yin=[y1[i]*v1+y2[i]*v2 for i in range(4)]
    y=[1 if yin[i]>=theta else 0 for i in range(4)]
    print("output of net=")
```

```
print(y)
if y==z:
    con=False
else:
    print("net is not learning. enter another set of weights and threshold value")
print("mcculloch pitts neuron for xor function")
print("weights of neuron z1:")
print(w11)
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)
```

OUTPUT:

weight $w_{11}=1$

weight $w_{12}=-1$

weight $w_{21}=-1$

weight $w_{22}=1$

weight $v_1=1$

weight $v_2=1$

theta=1

output of net=

[0, 1, 1, 0]

mcculloch pitts neuron for xor function

weights of neuron z_1 :

1.0

-1.0

weights of neuron z_2 :

-1.0

1.0

weights of neuron y :

1.0

1.0

threshold value:

1.0

2. perceptron for an AND function with bipolar inputs and targets

PROGRAM:

```
def mcculloch_pitts_neuron():  
    y=[0,0,0,0]  
    x1=[1,1,0,0]  
    x2=[1,0,1,0]  
    z=[1,0,0,0]  
    con=True  
    print('enter the weights')  
    w1=float(input('weight w1='))  
    w2=float(input('weight w2='))  
    theta=float(input('enter the threshold value theta='))  
    while con:  
        zin=[x1[i]*w1+x2[i]*w2 for i in range(4)]  
        for i in range(4):  
            if zin[i]>=theta:  
                y[i]=1  
            else:  
                y[i]=0  
        print('output of net=')  
        print(y)
```

```

        if y==z:
            con=False
        else:
            print('net is not learning. enter another ser of weights and
threshold value')

            w1=float(input('weight w1='))
            w2=float(input('weight w2='))
            theta=float(input('enter threshold value theta'))

            print('mcculloch pitts neuron for and function')
            print('weights of neuron')
            print(w1)
            print(w2)
            print('threshold value=')
            print(theta)
mcculloch_pitts_neuron()

```

Output:

enter the weights

weight $w_1=1$

weight $w_2=1$

enter the threshold value $\theta=2$

output of net=

[1, 0, 0, 0]

output of net=

[1, 0, 0, 0]

output of net=

[1, 0, 0, 0]

mcculloch pitts neuron for and function

weights of neuron

1.0

1.0

threshold value=

2.0

3. XOR function with binary input and output

PROGRAM:

```
import numpy as np

from matplotlib import pyplot as plt

def sigmoid(z):

    return 1/(1+np.exp(-z))

def initializeParameters(inputFeatures,neuronsInHiddenLayers,outputFeatures):

    W1=np.random.randn(neuronsInHiddenLayers,inputFeatures)

    W2=np.random.randn(outputFeatures,neuronsInHiddenLayers)

    b1=np.zeros((neuronsInHiddenLayers,1))

    b2=np.zeros((outputFeatures,1))

    parameters={"W1":W1,"b1":b1,"W2":W2,"b2":b2}

    return parameters

def forwardPropagation(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)

cache=(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,cache,A2

def backwardPropagation(X,Y,cache):

    m=X.shape[1]

    (Z1,A1,W1,b1,Z2,A2,W2,b2)=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={"dz2":dz2,"dW2":dW2,"db2":db2,"dZ1":dZ1,"dW1":dW1,"db1":db1}

    return gradients

def updateParameters(parameters,gradients,learningRate):

    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=np.array([[0,0,1,1],[0,1,0,1]])
Y=np.array([[0,1,1,0]])
neuronsInHiddenLayers=2
inputFeatures=X.shape[0]
outputFeatures=Y.shape[0]

parameters=initializeParameters(inputFeatures,neuronsInHiddenLayers,outputFeatures)

epoch=100000
learningRate=0.01
losses=np.zeros((epoch,1))

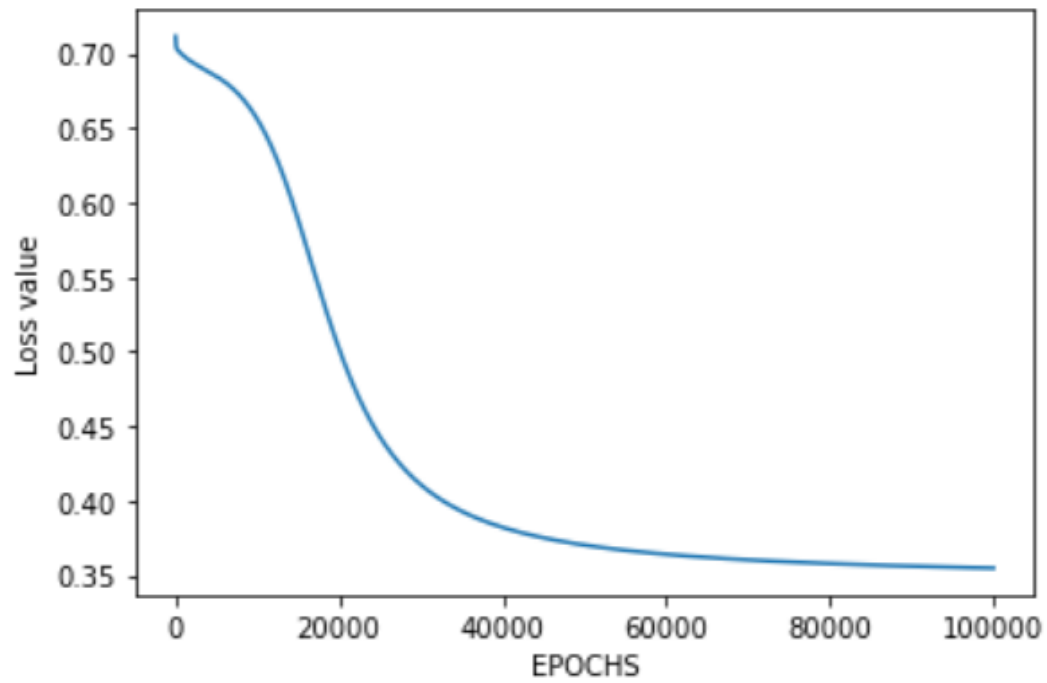
for i in range(epoch):
    losses[i,0],cache,A2=forwardPropagation(X,Y,parameters)
    gradients=backwardPropagation(X,Y,cache)
    parameters=updateParameters(parameters,gradients,learningRate)

plt.figure()
plt.plot(losses)
plt.xlabel("EPOCHS")
plt.ylabel("Loss value")
plt.show()

X=np.array([[1,1,0,0],[0,1,0,1]])
cost,_,A2=forwardPropagation(X,Y,parameters)
prediction=(A2>0.5)*1.0
print(prediction)

```

OUTPUT:



`[[0. 1. 0. 1.]]`

4) Perceptron learning law

PROGRAM:

```
import numpy as np

import matplotlib.pyplot as plt

np.random.seed(0)

class_0=np.random.randn(50,2)+np.array([2,2])

class_1=np.random.randn(50,2)+np.array([-2,-2])

X=np.vstack((class_0,class_1))

y=np.hstack((np.zeros(50),np.ones(50)))

X_augmented=np.c_[X,np.ones(100)]

np.random.seed(1)

w = np.random.randn(3)

def perceptron_learning(X,y,w,epochs=50,learning_rate=0.1):

    errors=[]

    for _ in range(epochs):

        error_count=0

        for xi,target in zip(X,y):

            output=np.dot(xi,w)>=0

            error=target - output

            if error !=0:

                w+=learning_rate*error*xi
```

```
        error_count+=1

    errors.append(error_count)

    return w,errors

trained_weights,errors= perceptron_learning(X_augmented,y,w)

plt.scatter(class_0[:,0],class_0[:,1],marker='p',label='class 0')

plt.scatter(class_1[:,0],class_1[:,1],marker='*',label='class 1')

x_boundary=np.linspace(-5,5,100)

y_boundary=(-trained_weights[0]*
x_boundary-trained_weights[2])/trained_weights[1]

plt.plot(x_boundary,y_boundary,'p-',label='Decision Boundary')

plt.xlabel('Feature 1')

plt.ylabel('Feature 2')

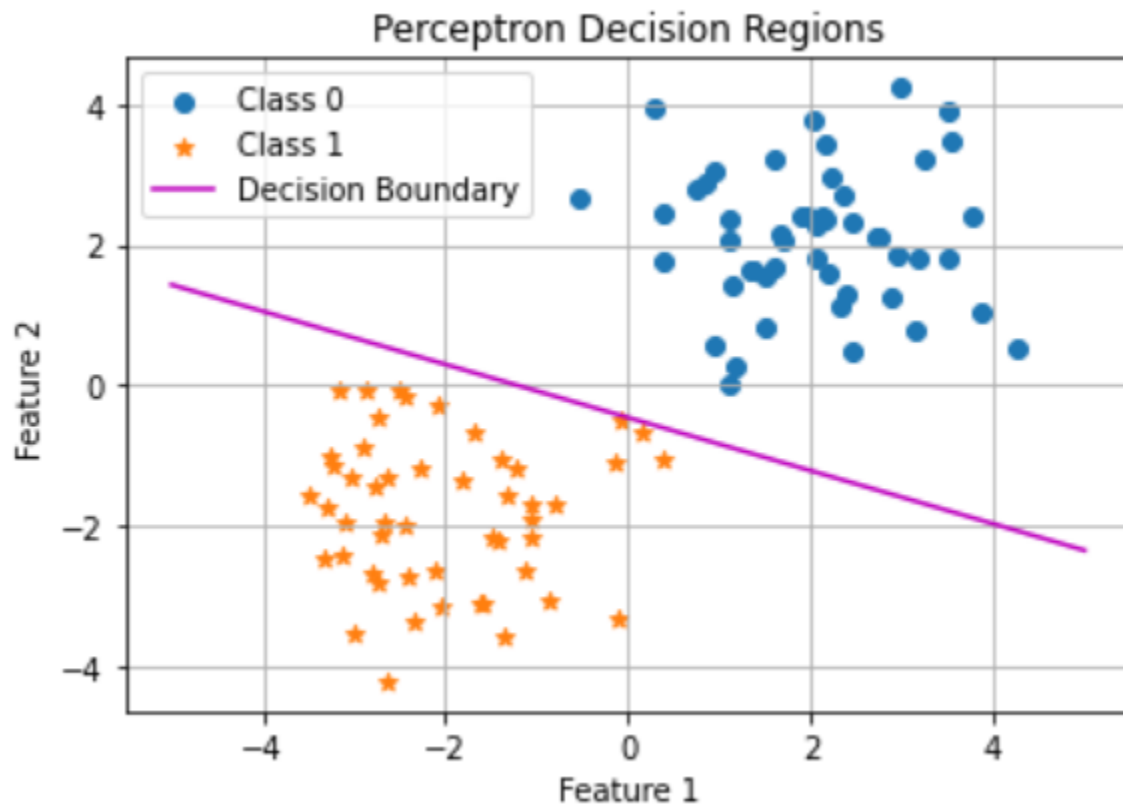
plt.legend()

plt.grid()

plt.title('Perceptron Decision Regions')

plt.show()
```

OUTPUT:



5) XOR function with Bipolar input and output

PROGRAM:

```
import numpy as np

def sigmoid(x):
    return 2/(1+np.exp(-x))-1

def sigmoid_derivative(x):
    return 0.5*(1+x)*(1-x)

x = np.array([[ -1,-1],[ -1,1],[ 1,-1],[ 1,1]])
t = np.array([[ -1],[ 1],[ 1],[ -1]])

input_size = 2
hidden_size = 2
output_size = 1

np.random.seed(0)

w_input_hidden = np.random.rand(input_size,hidden_size)
b_hidden = np.random.rand(1,hidden_size)
w_hidden_output = np.random.rand(hidden_size,output_size)
b_output = np.random.rand(1,output_size)

learning_rate = 0.1

epochs = 10000

for epoch in range(epochs):
    hidden_input = np.dot(x,w_input_hidden)+b_hidden
    hidden_output = sigmoid(hidden_input)
```

```

output = np.dot(hidden_output,w_hidden_output)+b_output

output = sigmoid(output)

output_error = t-output

hidden_error =
output_error.dot(w_hidden_output.T)*sigmoid_derivative(hidden_output)

w_hidden_output+=hidden_output.T.dot(output_error)*learning_rate

b_output+=np.sum(output_error)*learning_rate

w_input_hidden+=x.T.dot(hidden_error)*learning_rate

b_hidden+=np.sum(hidden_error)*learning_rate

if epoch % 1000==0:

    print(f"Epoch: {epoch},Error: {np.mean(np.abs(output_error))}")

print("\nTesting the trained network:")

for i in range(3):

    hidden_input = np.dot(x[i],w_input_hidden)+b_hidden

    hidden_output = sigmoid(hidden_input)

    output = np.dot(hidden_output,w_hidden_output)+b_output

    output = sigmoid(output)

    print(f"input: {x[i]},Predicted Output: {output[0][0]},Target Output: {t[i][0]}")

```

OUTPUT:

Epoch:0,Error:0.9612435747420139
Epoch:1000,Error:0.5040833500390071
Epoch:2000,Error:0.5017794479227115
Epoch:3000,Error:0.5011257955648463
Epoch:4000,Error:0.50081982834948
Epoch:5000,Error:0.5006431517612333
Epoch:6000,Error:0.5005283743537952
Epoch:7000,Error:0.500447932412263
Epoch:8000,Error:0.5003884828374154
Epoch:9000,Error:0.5003427898178361

Testing the trained network:

input: [-1 -1], Predicted Output: -0.9996290483699555, Target Output: -1
input: [-1 1], Predicted Output: -0.00022888265647258432, Target Output: 1
input: [1 -1], Predicted Output: 0.999615535881984, Target Outp

6) Write a program to load an image and apply the Convolution layer, Activation layer and pooling layer Operation to extract the inside feature

PROGRAM:

```
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from itertools import product

# set the param
plt.rc('figure', autolayout=True)
plt.rc('image', cmap='magma')

kernel = tf.constant([[ -1, -1, -1],
                      [ -1,  8, -1],
                      [ -1, -1, -1],
                      ])

#Load the image

image_path = r"C:\Users\STUDENT.HCAS-PGLAB-4\Pictures\Saved Pictures\ga
nesh.jpg"
image = tf.io.read_file(image_path)
image = tf.io.decode_jpeg(image, channels=1)
image = tf.image.resize(image, size=[300, 300])

#Load the image

image_path = r"C:\Users\STUDENT.HCAS-PGLAB-4\Pictures\Saved Pictures\ga
nesh.jpg"
image = tf.io.read_file(image_path)
image = tf.io.decode_jpeg(image, channels=1)
image = tf.image.resize(image, size=[300, 300])

#plot the image
```

```
img = tf.squeeze(image).numpy()
plt.figure(figsize=(5, 5))
plt.imshow(img, cmap='gray')
plt.axis('off')
plt.title('Original Gray Scale image')
plt.show();
```

#Reformat

```
image = tf.image.convert_image_dtype(image, dtype=tf.float32)
image = tf.expand_dims (image, axis=0)
kernel = tf.reshape(kernel, [*kernel.shape, 1, 1])
kernel = tf.cast (kernel, dtype=tf.float32)
```

conv_fn = tf.nn.conv2d#Reformat

```
image = tf.image.convert_image_dtype(image, dtype=tf.float32)
image = tf.expand_dims (image, axis=0)
kernel = tf.reshape(kernel, [*kernel.shape, 1, 1])
kernel = tf.cast (kernel, dtype=tf.float32)
```

```
image_filter = conv_fn(
    input = image,
    filters = kernel,
    strides = 1,
    padding = 'SAME',
)
plt.figure(figsize=(15,5))
```

Plot the convolved image

```
plt.subplot(1, 1, 1)
```

```
plt.imshow(
    tf.squeeze(image_filter)
)
```

```
plt.axis('off')
plt.title('Convolution')
```

```
#activation layer
tanh_fn = tf.nn.tanh

# Image detection
image_detect = relu_fn(image_filter)

plt.subplot(1, 1, 1)
plt.imshow(
    # Reformat for plotting
    tf.squeeze(image_detect)
)

plt.axis('off')
plt.title('Activation')

pool = tf.nn.pool

image_condense = pool(input=image_detect,
                      window_shape=(2,2),
                      pooling_type='MAX',
                      strides=(2,2),
                      padding='SAME',
                      )

plt.subplot(1,1,1)
plt.imshow(tf.squeeze(image_condense))
plt.axis('off')
plt.title('pooling')
plt.show()
```

OUTPUT:

Original Gray Scale image



Convolution



Activation



pooling

