In [21]: import numpy as np import matplotlib.pyplot as plt

init the perceptron

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In [22]: import numpy as np
          import matplotlib.pyplot as plt
          # create a new class for a perceptron with a hidden layer
          class Perceptron:
               rng = np.random.default_rng()
               errArr = []
               def __init__(self, input_size, hidden_size, output_size):
                   self.hiddenWeights = np.random.normal(0, 1/2, (hidden_size, input_size))
                   self.hiddenThreshold = np.zeros([hidden_size,1])
                   self.outputWeights = np.random.normal(0, 1/hidden_size, (hidden_size))
                   self.outputThreshold = np.zeros(output_size)
                   if(0):
                        #shapes of the initializations
                        print("hiddenWeights.shape: ", self.hiddenWeights.shape)
print("hiddenThreshold.shape: ", self.hiddenThreshold.shape)
                        print("outputWeights.shape: ", self.outputWeights.shape)
print("outputThreshold.shape: ", self.outputThreshold.shape)
                        # print the values
                        print("hiddenWeights: ", self.hiddenWeights)
print("hiddenThreshold: ", self.hiddenThreshold)
print("outputWeights: ", self.outputWeights)
print("outputThreshold: ", self.outputThreshold)
               def tanh(self, x):
                   return np.tanh(x)
               def tanhDerivative(self, x):
                   return 1 - np.power(np.tanh(x), 2)
               def forward(self, input):
                   self.hidden_b = np.dot(self.hiddenWeights, input.T) - self.hiddenThreshold
                   self.hiddenLayer = self.tanh(self.hidden b)
                   self.output_B = np.dot(self.outputWeights.T, self.hiddenLayer) - self.outputThreshold
                   self.outputLayer = self.tanh(self.output_B)
                   return self.outputLayer
               def backward(self, X, target, eta, output):
                   error = target - output
                                                                                                                  #(batch si
                   error_2 = error * self.tanhDerivative(self.output_B)
                                                                                                                        #(bat
                   error_2_reshaped = error_2[:, np.newaxis]
                   output_weights_reshaped = self.outputWeights[np.newaxis, :]
                   tmp = error_2_reshaped * output_weights_reshaped
                   gprime = self.tanhDerivative(self.hidden b)
                   error_1 = (tmp * gprime.T)
                   self.outputWeights += error_2 @ self.hiddenLayer.T *eta
self.outputThreshold -= eta * np.sum(error_2, axis=0, keepdims=True)
                   self.hiddenWeights += error 1.T @ X * eta
                   self.hiddenThreshold -= eta * np.sum(error_1, axis=0, keepdims=True).T
               def classificationError(self, input, target):
                   prediction = np.sign(self.forward(input))
                    return 0.5 * np.mean(np.abs(prediction-target))
               def trainMiniBatches(self, xTrain, yTrain, eta, epochs, mB):
                   for in range(epochs):
                        mu = self.rng.choice(xTrain.shape[0],size=mB, replace=True)
                        xTrain_rand = xTrain[mu]
                        yTrain_rand = yTrain[mu]
                        output = self.forward(xTrain rand)
                        self.backward(xTrain_rand, yTrain_rand, eta=eta)
                   return
               def trainMiniBatchesWhile(self, xTrain, yTrain, xValid, yValid, eta, epochs, mB):
                   err = 1
                   lowestErr = 1
                   while err > 0.115:
                        if i > 1000:
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eta = 0.001
                      oldErr = err
                      mu = self.rng.choice(xTrain.shape[0],size=mB, replace=False)
                      xTrain_rand = xTrain[mu]
                      yTrain_rand = yTrain[mu]
                      output = self.forward(xTrain rand,)
                      self.backward(xTrain_rand, yTrain_rand, eta=eta, output=output)
                      # calculate the error
                      err = self.classificationError(xValid, yValid)
                      self.errArr.append(err)
                      if err<lowestErr:</pre>
                          lowestErr = err
                      if (i % 100 == 0):
                          print("epoch: ", i, "error: ", err, "lowestErr: ", lowestErr)
                      i += 1
                      # if new lowest error, and error < 0.13 save the weights and thresholds
                      if (err < 0.12):</pre>
                          np.savetxt("csv/wl.csv", self.hiddenWeights, delimiter=",")
                          np.savetxt("csv/w2.csv", self.outputWeights, delimiter=",")
np.savetxt("csv/t1.csv", self.hiddenThreshold, delimiter=",")
                           np.savetxt("csv/t2.csv", self.outputThreshold, delimiter=",")
                  print("epoch: ", i, "error: ", err, "lowestErr: ", lowestErr)
                   ""print("hidden Weights: ", self.hiddenWeights)
                  print("hidden Threshold: ", self.hiddenThreshold)
                  print("output Weights: ", self.outputWeights)
print("output Threshold: ", self.outputThreshold)"""
                  return
         perceptron = Perceptron(2, 2, 1)
In [23]: def main():
              # read the training and validation data
             training_set = np.loadtxt('training_set.csv', delimiter=',')
             validation_set = np.loadtxt('validation_set.csv', delimiter=',')
             xTraining = training_set[:, :2]
             yTraining = training_set[:, 2]
             xValidation = validation_set[:, :2]
             yValidation = validation_set[:, 2]
             xTraining mean = xTraining.mean(axis=0)
             xTraining_std = xTraining.std(axis=0)
             xTraining normalised = (xTraining-xTraining mean)/xTraining std
             xValidation_normalised = (xValidation-xTraining_mean)/xTraining_std
             perceptron = Perceptron(input size=2, hidden size=50, output size=1)
              # train the perceptron
             #perceptron.trainMiniBatches(xTraining normalised, yTraining, eta=0.001, epochs=100000, mB=2**8)
             perceptron.trainMiniBatchesWhile(xTraining_normalised, yTraining, xValidation_normalised, yValid
              # classify the validation data
             valErr = perceptron.classificationError(xValidation_normalised, yValidation)
             print("Validation error: ", valErr)
```

epoch: 1800 error: 0.1354 lowestErr: 0.1342