```
In [1]: import numpy as np
         import matplotlib.pyplot as plt
         import pandas as pd
         from numpy import random
         import time
         from sklearn import *
In [2]: from mnist import MNIST
         mndata = MNIST(r"C:\Users\kalya\OneDrive - University of Illinois at Chicago\!UIC\!Semesters\2nd Sem\Courses
         \CS 559 NN\Homeworks\HW2\Q2\data\t")
         xtrain, ytrain = mndata.load training()
         xtest, ytest = mndata.load testing()
         xtrain = np.reshape(xtrain,(60000,784))
         xtest = np.reshape(xtest,(10000,784))
         ytrain = np.reshape(ytrain,(60000,1))
         ytest = np.reshape(ytest,(10000,1))
         xtrain = (xtrain)/255
         xtest = (xtest)/255
         xtrain.shape
Out[2]: (60000, 784)
In [3]: x train = []
         x test = []
         for i in range(len(xtrain)):
             x train.append(np.append(xtrain[i],1))
         for i in range(len(xtest)):
             x test.append(np.append(xtest[i],1))
         x train = np.asarray(x train)
         x test = np.asarray(x test)
In [4]: \# x \text{ test1} = pd.DataFrame(data = x \text{ test.reshape(len}(x \text{ test}), 785))
         # ytest1 = pd.DataFrame(data = ytest.reshape(len(ytest),1))
In [5]: | # x test1.head()
```

```
In [6]: \# x = [1, 1]
          # np.linalq.norm(x)
 In [7]: | ytrain
 Out[7]: array([[5],
                  [0],
                  [4],
                  . . . ,
                  [5],
                  [6],
                  [8]], dtype=uint8)
In [20]: \# x = [[0,1, 1], [1,0, 0], [1,1, 0], [0,0, 0]]
          # y=[4,1,2,3]
          \# x = np.asarray(x)
          \# y = np.asarray(y)
          d = np.zeros(shape = (len(xtrain),10))
          for i in range(len(ytrain)):
              d[i][ytrain[i]] = 1
          dt = np.zeros(shape = (len(xtest),10))
          for i in range(len(ytest)):
              dt[i][ytest[i]] = 1
          \# d = np.zeros(shape = (len(x), 10))
          # for i in range(len(y)):
                d\lceil i\rceil\lceil y\lceil i\rceil\rceil = 1
          # xtrain = x
          # xtrain
          d[0]
Out[20]: array([0., 0., 0., 0., 0., 1., 0., 0., 0., 0.])
```

```
In [21]: | # # # d = np.zeros(shape = (len(xtrain), 10))
         # # # for i in range(len(ytrain)):
                   d[i][ytrain[i]] = 1
         # # #
         # # # dt = np.zeros(shape = (len(xtest),10))
         # # # for i in range(len(ytest)):
                dt[i][ytest[i]] = 1
         # # #
         # # #w = random.normal(size = (10,784))
         # d = np.zeros(shape = (len(xtrain)))
         # for i in range(len(ytrain)):
               d[i] = ytrain[i][0]
         # dt = np.zeros(shape = (len(xtest)))
         # for i in range(len(ytest)):
               dt[i] = ytest[i][0]
In [22]: d
Out[22]: array([[0., 0., 0., ..., 0., 0., 0.],
                [1., 0., 0., ..., 0., 0., 0.]
                [0., 0., 0., ..., 0., 0., 0.]
                [0., 0., 0., ..., 0., 0., 0.]
                [0., 0., 0., \ldots, 0., 0., 0.]
                [0., 0., 0., \ldots, 0., 1., 0.]
In [23]: a = [[1.5, 2.5], [2.6, 3.3]]
         np.around(a)
Out[23]: array([[2., 2.],
                [3., 3.]])
In [24]: # weights = np.random.uniform(low =-2, high = 2, size = (784, 60000))
         # np.matmul(weights, x train)
```

```
In [25]: d
 Out[25]: array([[0., 0., 0., ..., 0., 0., 0.],
                 [1., 0., 0., ..., 0., 0., 0.]
                 [0., 0., 0., ..., 0., 0., 0.]
                 [0., 0., 0., ..., 0., 0., 0.]
                 [0., 0., 0., ..., 0., 0., 0.]
                 [0., 0., 0., \ldots, 0., 1., 0.]
 In [26]: \# t = np.asarray([1,2,3])*(np.asarray([4,5,6]))
          # t
In [232]: a=np.asarray([[1,2,3], [4,5,6]])
          print(a[:-1])
          c = np.asarray([[7,8,9]])
          b =np.append(a,c, axis=0)
          [[1 2 3]]
Out[232]: array([[1, 2, 3],
                 [4, 5, 6],
                 [7, 8, 9]])
```

```
In [427]: class Neural Network():
              def __init__(self, x=[[]], y=[], p=[[]], q=[], nHiddenLayers = 0, nHiddenNodes =0, numOutputs = 0, eta =
          1, iter = 0, prec = 0):
                  self.data = x
                  self.labels = v
                  self.test = p
                  self.testlabels = q
                  self.nInputNodes = x.shape[1]
                  self.nHiddenLayers = nHiddenLayers
                  self.nHiddenNodes = nHiddenNodes
                  self.numOutputs = numOutputs
                  self.eta = eta
                  self.maxIt = iter
                  self.prec = prec
                  self.weights=[np.random.uniform(low = -2, high = 2, size = (self.nHiddenNodes, self.nInputNodes))]
                  for i in range(self.nHiddenLayers-1):
                      self.weights.append(np.random.uniform(low =-2, high = 2, size =(self.nHiddenNodes, self.nHiddenNo
          des+1)))
                  self.weights.append(np.random.uniform(low =-2, high = 2, size = (self.numOutputs, self.nHiddenNodes+1
          )))
              # Tanh activation function for all layers except the output layer
              def tanh(self, s):
                  return np.tanh(s)
              # Derivative of tanh for all layers
              def tanhPrime(self, s):
                  return (1 - s**2)
              # Activation function for all the input data
              def af predict(self,t):
                    t = np.tanh(t)
                  p = []
                  for i in range(len(t)):
                      max = np.argmax(t[i])
                      y = np.zeros(len(t[i]))
                      y[max] = 1
                      p.append(y)
                  return np.asarray(p)
              # Activation function for the output layer of the feedforward graph
```

```
def af(self,t):
#
         t = np.tanh(t)
       max = np.argmax(t[0])
       y = np.zeros(self.numOutputs)
       y[max] = 1
         temp = np.asarray(y).reshape(1,self.numOutputs)
       return np.asarray(y)
     # Derivative of sigmoid for the output layer
     def sigmoidPrime(self, s):
         return s * (1 - s)
     def decision A(self,x):
#
          out =[]
# #
           print(x.shape)
           print(f)
# #
         for i in range(x.shape[0]):
              s = np.zeros(x.shape[1])
             max = np.arqmax(x[i])
             #print(max)
              s[max] = 1
              out.append(s)
         out = np.asarray(out)
         print(out.shape)
         print(f)
          return out
   def misclassifications(self, x,y):
        count = 0
       for i in range(y.shape[0]):
            if np.any(x[i]-y[i]):
                count += 1
       return count
   def predict(self, data=[]):
       prev = data.T
       for i in range(len(self.weights)-1):
            temp = (np.matmul(self.weights[i], prev))
            temp2 = self.tanh(temp)
            y=[np.ones(temp2.shape[1])]
            prev = np.append(y, temp2, axis = 0)
             print(prev.shape)
```

```
print(f)
       temp_f = np.matmul(self.weights[self.nHiddenLayers], prev)
       temp4 = self.af predict(np.transpose(temp f))
       return temp4
   def feedforward(self, x):
         print(x)
       self.r = []
       self.r.append(x)
         print(self.r)
         print(f)
       prev = x.T
       for i in range(len(self.weights)-1):
             print(self.weights[i])
#
             print(prev)
           temp = np.matmul(self.weights[i], prev)
             print(temp)
#
           temp2 = self.tanh(temp)
           prev = np.append([[1]], temp2).reshape(temp2.shape[0]+1,1)
             print(prev.shape)
             print(f)
           self.r.append(prev)
             print(self.r)
#
             print('\n')
         print(self.weights[self.nHiddenLayers])
         print(prev)
       temp f = np.matmul(self.weights[self.nHiddenLayers], prev)
       temp4 =self.af(temp f.T)
       return temp4
   def backward(self, x, y):
       self.gradient=[]
         print(x)
         print(y)
         print(f)
       o = self.feedforward(x)
         print(o)
       self.out error = y - o
         print(self.out error)
         print(f)
       for i in reversed(range(len(self.weights))):
```

```
if i == 0 and i != len(self.weights)-1:
        temp2 = temp.T.dot(self.weights[i+1]).T
          print(self.sigmoidPrime(self.r[i+1]))
        temp3 = temp2*self.tanhPrime(self.r[i+1])
        temp4 = (temp3.dot(self.r[i]))*2/len(self.data)
          print('w')
          print(self.weights[i])
          print((self.eta*temp4*2)/len(self.data))
        self.gradient.append(self.eta*2*temp4[1:])
    elif i > 0 and i<len(self.weights)-1:</pre>
        temp2 = temp.T.dot(self.weights[i+1]).T
          print(self.sigmoidPrime(self.r[i+1]))
        temp3 = temp2*self.tanhPrime(self.r[i+1])
          print(temp3.shape)
        temp4 = (temp3.dot(self.r[i].T))*2/len(self.data)
          print('w')
          print(self.weights[i])
          print((self.eta*temp4*2)/len(self.data))
        self.gradient.append(self.eta*2*temp4[1:])
        temp = temp3[1:]
    elif i == len(self.weights)-1:
          print(self.weights)
          print(self.out error)
          print(o)
          print(self.af derivative(o).T)
        temp = self.out error.reshape(self.numOutputs,1)
          temp = (self.out error*self.tanhPrime(o)).reshape(self.numOutputs,1)
          print(temp.shape)
          print(f)
          print(self.weights[i])
        temp2 = self.r[i].dot(temp.T)*2/len(self.data)
        self.gradient.append(self.eta*temp2.T)
          print(self.weights[i])
          print('\n')
self.gradient = self.gradient[::-1]
for i in range(len(self.weights)):
    self.weights[i] += self.gradient[i]
  print(self.weights)
```

```
#
         print(f)
   def train (self):
       #print(self.eta)
         print(self.weights[0].shape)
         print(f)
       e = 0
       obj training =[]
       obj testing = []
       epoch = []
       mis training = []
       mis testing =[]
       epoch.append(e)
       pred train = self.predict(self.data)
       pred test = self.predict(self.test)
       obj training.append(((np.linalg.norm(self.labels - pred train))**2)/len(self.data))
       obj testing.append(((np.linalg.norm(self.testlabels - pred test))**2)/len(self.test))
       mis training.append(self.misclassifications(self.labels, pred train))
       mis testing.append(self.misclassifications(self.testlabels, pred test))
       mse = 100000000000
         while e <= self.maxIt and mse >= self.prec:
       while e <= self.maxIt:</pre>
            prev = mse
             print(prev)
             print(f)
            e += 1
           for i in range(len(self.data)):
                self.backward(self.data[i].reshape(1,self.nInputNodes), self.labels[i])
                  print(self.delta)
             print(self.weights)
             print(f)
             print('after')
             print(self.weights)
           pred train = self.predict(self.data)
           pred test = self.predict(self.test)
           mse = ((np.linalg.norm(self.labels - pred train))**2)/len(self.data)
           mse test = ((np.linalg.norm(self.testlabels - pred test))**2)/len(self.test)
             print(mse)
             print(f)
           obj training.append(mse)
           obj testing.append(mse test)
            epoch.append(e)
           mis training.append(self.misclassifications(self.labels, pred train))
```

```
mis_testing.append(self.misclassifications(self.testlabels, pred_test))
if mse >= prev:
    self.eta = 0.7*self.eta
    prev = mse
    return self.weights, epoch, obj_training, mis_training, obj_testing, mis_testing
```

## **Hyperparameter Selection**

```
In [417]: # nHiddenLayers = 1
          # nHiddenNodes =2
          # numOutputs = 10
          # eta = 1
          # iter = 100
          # prec = 1.6
          #print(xtrain.shape)
          hid = [1,2]
          hidnodes= [16,128]
          lrate = [10, 15]
          a =[]
          b =[]
          c =[]
          q = []
          weights =[]
          u =[]
          iter = 30
          numOutputs = 10
          prec = 0.2
          for i in range(len(hid)):
              for j in range(len(hidnodes)):
                  for k in range(len(lrate)):
                      u.append([hid[i],hidnodes[j],lrate[k]])
                      print([hid[i],hidnodes[j],lrate[k]])
                      NN = Neural_Network(x_train, d, x_test, dt, hid[i], hidnodes[j], numOutputs, lrate[k], iter, prec
                      w, epoch, obj_training, mis_training, obj_testing, mis_testing = NN.train()
                      weights.append(w)
                      a.append(obj_training)
                      b.append(mis_training)
                      c.append(obj testing)
                      q.append(mis testing)
```

[1, 16, 10]

[1, 16, 15]

[1, 128, 10]

[1, 128, 15]

[2, 16, 10]

[2, 16, 15]

[2, 128, 10]

[2, 128, 15]

```
In [418]: fig, ax = plt.subplots(2,4,figsize = (20,10))
               # for i in range(len(hid)):
                       for j in range(len(hidnodes)):
                             for k in range(len(lta)):
               k = 0
               for i in range(2):
                    for j in range(4):
                          ax[i,j].scatter(epoch,a[k], s = 7)
                          ax[i,j].plot(epoch,a[k], label='Training')
                          ax[i,j].scatter(epoch,c[k], s = 7, c = 'r')
                          ax[i,j].plot(epoch, c[k], c='r',label='Testing')
                          ax[i,j].legend()
                          ax[i,j].title.set text('Plot for the combination ' + str(u[k]))
                          k += 1
                      Plot for the combination [1, 16, 10]
                                                             Plot for the combination [1, 16, 15]
                                                                                                    Plot for the combination [1, 128, 10]
                                                                                                                                            Plot for the combination [1, 128, 15]
                                                        1.9
                1.80
                                                                                               1.8
                                                                                                                                       1.8

    Training

    Training

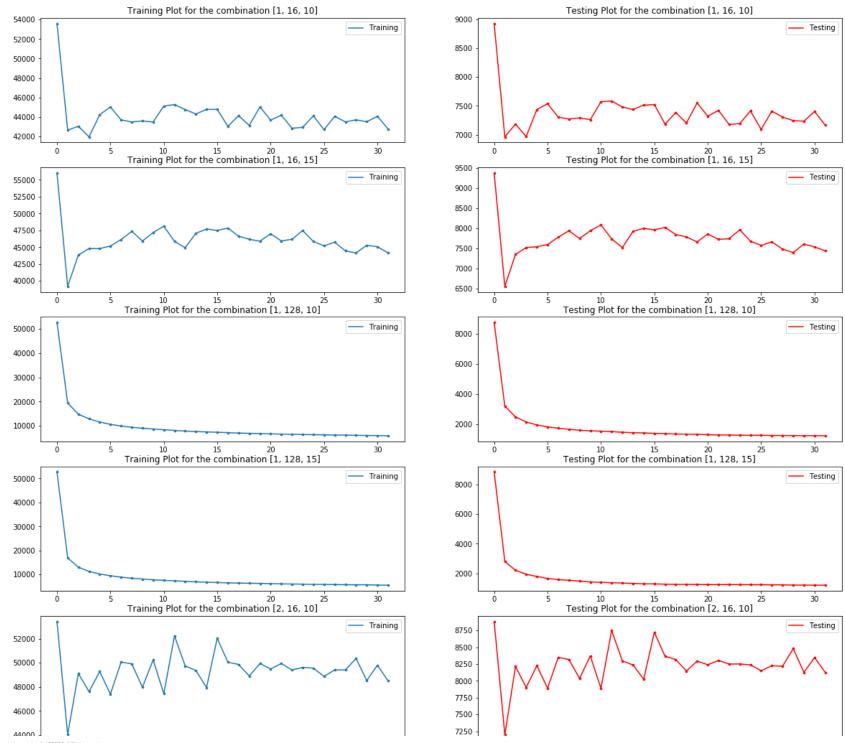
                                                                                                                         — Training

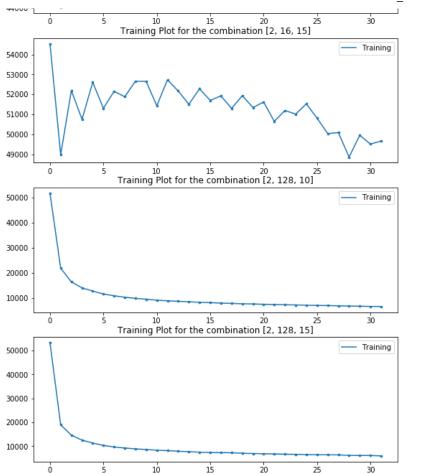
    Training

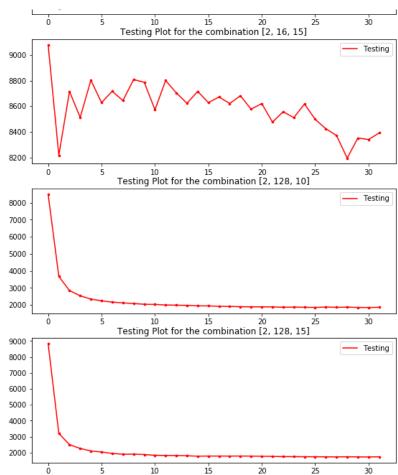
                                              Testing
                                                                                     Testing
                                                                                                                             Testing
                                                                                                                                                                    Testing
                                                                                               1.6
               1.75
                                                                                                                                       1.6
                                                        1.8
                1.70
                                                                                               1.4
                                                                                                                                       1.4
                                                        1.7
               1.65
                                                                                               1.2
                                                                                                                                       1.2
               1.60
                                                        1.6
                                                                                               1.0
                                                                                                                                       1.0
               1.55
                                                                                                0.8
                                                                                                                                       0.8
                                                        1.5
               1.50
                                                                                                0.6
                                                                                                                                       0.6
                                                        1.4
                1.45
                                                                                                                                       0.4
                                                                                                0.4
                1.40
                                                                                                                                       0.2
                                                                                                0.2
                                                        1.3
                              10
                                   15
                                                                     10
                                                                          15
                                                                                                             10
                                                                                                                  15
                                                                                                                                                    10
                                                                                                                                                         15
                                                                                    25
                                                                                                                           25
                                                                                                    Plot for the combination [2, 128, 10]
                                                                                                                                            Plot for the combination [2, 128, 15]
                      Plot for the combination [2, 16, 10]
                                                             Plot for the combination [2, 16, 15]
                1.80
                                                      1.825
                                                                                                                                       1.8
                                                                                  Training
                                                                                                                         - Training
                                              Training

    Training

                                              Testing
                                                                                     Testing
                                                                                               1.6
                                                                                                                            Testing
                                                                                                                                                                    Testing
               1.75
                                                      1.800
                                                                                                                                       1.6
                                                                                               14
                                                      1.775
               1.70
                                                                                                                                       1.4
                                                                                               1.2
                                                      1.750
                                                                                                                                       1.2
                1.65
                                                      1.725
                                                                                                1.0
                                                                                                                                       1.0
                1.60
                                                      1.700
                                                                                                0.8
                                                                                                                                       0.8
               1.55
                                                      1.675
                                                                                                0.6
                                                                                                                                       0.6
               1.50
                                                      1.650
                                                                                                                                       0.4
                                                                                                0.4
                1.45
                                                      1.625
                                                                                                0.2
                                            25
```







```
In [420]: # Tanh activation function for all layers except the output layer
          def tanh(s):
              return np.tanh(s)
          # Derivative of tanh for all layers except the output layer
          def tanhPrime(s):
              return (1 - s**2)
          # Activation function for all the input data
          def af predict(t):
             t = np.tanh(t)
              p = []
              for i in range(len(t)):
                  max = np.argmax(t[i])
                  y = np.zeros(len(t[i]))
                  y[max] = 1
                  p.append(y)
              return np.asarray(p)
          # Activation function for the output layer of the feedforward graph
          def af(t):
                t = np.tanh(t)
              max = np.argmax(t[0])
              y = np.zeros(numOutputs)
              y[max] = 1
                    temp = np.asarray(y).reshape(1,self.numOutputs)
              return np.asarray(y)
          def misclassifications(x,y):
              count = 0
              for i in range(y.shape[0]):
                  if np.any(x[i]-y[i]):
                      count += 1
              return count
          def predict(data, weights):
              prev = data.T
              for i in range(len(weights)-1):
                  temp = (np.matmul(weights[i], prev))
                  temp2 = tanh(temp)
```

```
y=[np.ones(temp2.shape[1])]
                  prev = np.append(y, temp2, axis = 0)
                        print(prev.shape)
                        print(f)
              temp f = np.matmul(weights[nHiddenLayers], prev)
              temp4 = af predict(np.transpose(temp f))
              return temp4
In [425]: nHiddenLayers = 1
          nHiddenNodes =128
          numOutputs = 10
           # eta = 10
           # iter =50
          y = predict(x_test, weights[3])
          misclassifications(dt, y)
Out[425]: 1206
In [435]: nHiddenLayers = 1
          nHiddenNodes =128
          numOutputs = 10
          eta = 15
          iter =50
          NN = Neural_Network(x_train, d, x_test, dt, nHiddenLayers, nHiddenNodes, numOutputs, eta, iter, prec)
          w, epoch, obj_training, mis_training, obj_testing, mis_testing = NN.train()
In [436]: y = predict(x_test, w)
          misclassifications(dt, y)
```

file:///C:/Users/kalya/Downloads/ TEMP.html

Out[436]: 1097