

Proxenix Session 2

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In [1]:
         import numpy as np
In [3]: import pandas as pd
         from matplotlib import pyplot as plt
In [6]: data= pd.read csv('/content/train.csv.zip')
In [9]: data= np.array(data)
         #m- total no. of images , n- total no. of columns(how many px)
         \#m= total no. of images , n= 784px + 1(label)=785
         #this is divided into testing/dev and train
         m, n=data.shape
         np.random.shuffle(data)
In [11]: data dev=data[0:1000].T
         Y dev=data dev[0]
         X_dev=data_dev[1:n]
         X \text{ dev=} X \text{ dev/} 255
In [13]: data train=data[1000:m].T
         Y_train=data_train[0]
         X_train=data_train[1:n]
         X train=X train/255
         _,m_train=X_train.shape
In [14]: Y train
Out[14]: array([1, 0, 3, ..., 1, 5, 4])
In [31]: def init params():
           #it will create a matrix
           W1=np.random.rand(10,784)-0.5
           b1=np.random.rand(10,1)-0.5
           W2=np.random.rand(10,10)-0.5
           b2=np.random.rand(10,1)-0.5
           return W1,b1,W2,b2
         def ReLU(Z):
           return np.maximum(Z,0)
         def softmax(Z):
           A=np.exp(Z)/sum(np.exp(Z))
           return A
         def forward_prop(W1,b1,W2,b2,X):
           Z1=W1.dot(X)+b1
           A1=ReLU(Z1)
           Z2=W2.dot(A1)+b2
           A2=softmax(Z2)
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return Z1,A1,Z2,A2
         def ReLU deriv(Z):
           return Z>0
         def one hot(Y):
           one hot Y=np.zeros((Y.size,Y.max()+1))
           #first i will crate a matrix of zeros
           one hot Y[np.arange(Y.size),Y]=1
           one hot Y=one hot Y.T
           return one hot Y
         def backward prop(Z1,A1,Z2,A2,W1,W2,X,Y):
           one hot Y=one hot(Y)
           dZ2=A2-one hot Y
           dW2=1/m*dZ2.dot(A1.T)
           db2=1/m*np.sum(dZ2)
           dZ1=W2.T.dot(dZ2)*ReLU deriv(Z1)
           dW1=1/m*dZ1.dot(X.T)
           db1=1/m*np.sum(dZ1)
           return dW1,db1,dW2,db2
         def update params(W1,b1,W2,b2,dW1,db1,dW2,db2,alpha):
           W1=W1-alpha*dW1
           b1=b1-alpha*db1
           W2=W2-alpha*dW2
           b2=b2-alpha*db2
           return W1,b1,W2,b2
In [35]: def get predictions(A2):
           return np.argmax(A2,0)
         def get accuracy(predictions, Y):
           print(predictions, Y)
           return np.sum(predictions == Y) / Y.size
In [36]: def gradient descent(X,Y,alpha,iterations):
           W1,b1,W2,b2=init params()
           for i in range(iterations):
             Z1,A1,Z2,A2=forward prop(W1,b1,W2,b2,X)
             dW1,db1,dW2,db2=backward prop(Z1,A1,Z2,A2,W1,W2,X,Y)
             W1,b1,W2,b2=update params(W1,b1,W2,b2,dW1,db1,dW2,db2,alpha)
             if i%10==0:
               print("Iteration: ",i)
               predictions=get predictions(A2)
               print(get accuracy(predictions,Y))
           return W1,b1,W2,b2
In [37]: W1,b1,W2,b2 = gradient_descent(X_train,Y_train,0.10,500)
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Iteration: 0
[6 6 6 ... 6 6 6] [1 0 3 ... 1 5 4]
0.11139024390243903
Iteration: 10
[6 0 6 ... 6 6 6] [1 0 3 ... 1 5 4]
0.20341463414634145
Iteration: 20
[6 0 6 ... 6 6 6] [1 0 3 ... 1 5 4]
0.2873170731707317
Iteration: 30
[6\ 0\ 6\ \dots\ 1\ 6\ 6]\ [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.35324390243902437
Iteration: 40
[6 0 6 ... 1 0 9] [1 0 3 ... 1 5 4]
0.40187804878048783
Iteration: 50
[6 0 6 ... 1 0 9] [1 0 3 ... 1 5 4]
0.44521951219512196
Iteration: 60
[6\ 0\ 6\ \dots\ 1\ 0\ 4]\ [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.48741463414634145
Iteration: 70
[6 0 3 ... 1 0 4] [1 0 3 ... 1 5 4]
0.5262926829268293
Iteration: 80
[6\ 0\ 3\ \dots\ 1\ 0\ 4]\ [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.5559512195121952
Iteration: 90
[1\ 0\ 6\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.5802926829268292
Iteration: 100
[1\ 0\ 6\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.6029024390243902
Iteration: 110
[1\ 0\ 6\ \dots\ 1\ 5\ 4]\ [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.6231951219512195
Iteration: 120
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.6422682926829268
Iteration: 130
[1 0 5 ... 1 5 4] [1 0 3 ... 1 5 4]
0.6594146341463415
Iteration: 140
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.6754634146341464
Iteration: 150
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.6905365853658536
Iteration: 160
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.7036585365853658
Iteration: 170
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.7166585365853658
```

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Iteration: 180
[1 0 5 ... 1 5 4] [1 0 3 ... 1 5 4]
0.7291219512195122
Iteration: 190
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.7396341463414634
Iteration: 200
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.7494878048780488
Iteration: 210
[1\ 0\ 5\ \dots\ 1\ 5\ 4]\ [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.7574390243902439
Iteration: 220
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.7650487804878049
Iteration: 230
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.7714878048780488
Iteration: 240
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.7779024390243903
Iteration: 250
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.7824146341463415
Iteration: 260
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.7880243902439025
Iteration: 270
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.7934146341463415
Iteration: 280
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.7976341463414635
Iteration: 290
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.8020243902439025
Iteration: 300
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.8058536585365854
Iteration: 310
[1 0 5 ... 1 5 4] [1 0 3 ... 1 5 4]
0.8089756097560976
Iteration: 320
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.8123414634146342
Iteration: 330
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.8155365853658536
Iteration: 340
[1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
0.8188292682926829
Iteration: 350
[1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
0.8222926829268292
```

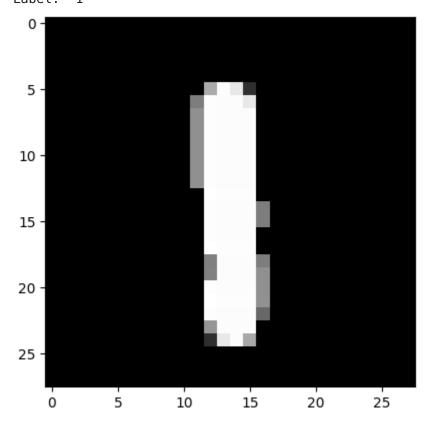
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[1 0 5 ... 1 5 4] [1 0 3 ... 1 5 4]
         0.8248292682926829
         Iteration: 370
         [1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
         0.8274878048780487
         Iteration: 380
         [1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
         0.8294390243902439
         Iteration: 390
         [1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
         0.8317073170731707
         Iteration: 400
         [1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
         0.8335853658536585
         Iteration: 410
         [1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
         0.8358048780487805
         Iteration: 420
         [1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
         0.8374146341463414
         Iteration: 430
         [1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
         0.8393414634146341
         Iteration: 440
         [1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
         0.8408536585365853
         Iteration: 450
         [1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
         0.8427317073170731
         Iteration: 460
         [1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
         0.8442195121951219
         Iteration: 470
         [1\ 0\ 5\ \dots\ 1\ 5\ 4] [1\ 0\ 3\ \dots\ 1\ 5\ 4]
         0.8457317073170731
         Iteration: 480
         [1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
         0.8469268292682927
         Iteration: 490
         [1 \ 0 \ 5 \ \dots \ 1 \ 5 \ 4] \ [1 \ 0 \ 3 \ \dots \ 1 \ 5 \ 4]
         0.8483414634146341
In [39]: def make predictions(X,W1,b1,W2,b2):
             _{,},_{,},A2 = forward_prop(W1,b1,W2,b2,X)
             predictions = get predictions(A2)
             return predictions
           def test prediction(index,W1,b1,W2,b2):
              current_image = X_train[:,index,None]
             prediction = make predictions(X train[:, index, None], W1, b1, W2, b2)
             label = Y train[index]
             print("Prediction: ", prediction)
             print("Label: ", label)
```

Iteration: 360

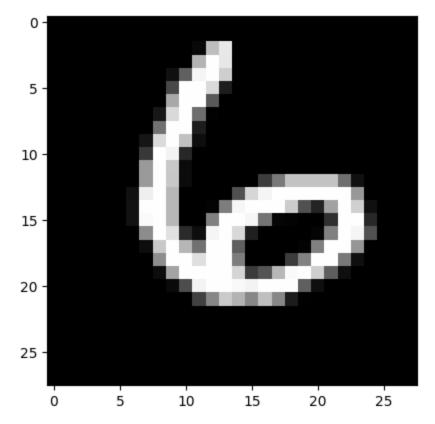
```
current_image = current_image.reshape((28, 28)) * 255
plt.gray()
plt.imshow(current_image, interpolation='nearest')
plt.show()
```

In [41]: test_prediction(0,W1,b1,W2,b2)
 test_prediction(3,W1,b1,W2,b2)
 test_prediction(5,W1,b1,W2,b2)
 test_prediction(6,W1,b1,W2,b2)
 test_prediction(22,W1,b1,W2,b2)

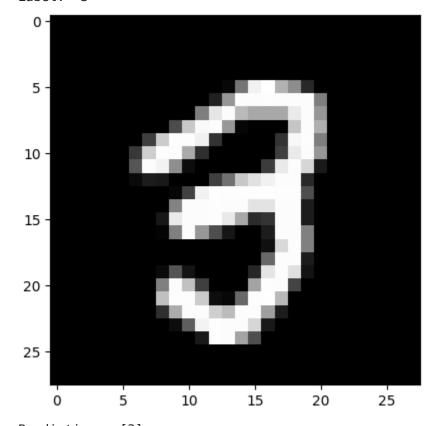
Prediction: [1] Label: 1



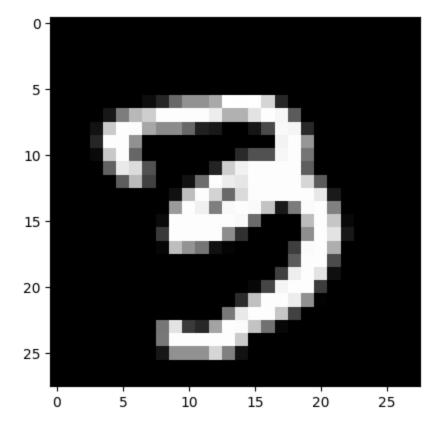
Prediction: [6] Label: 6



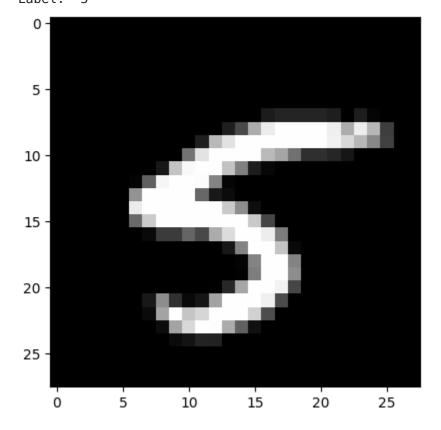
Prediction: Label: 3 [8]



Prediction: Label: 3 [3]

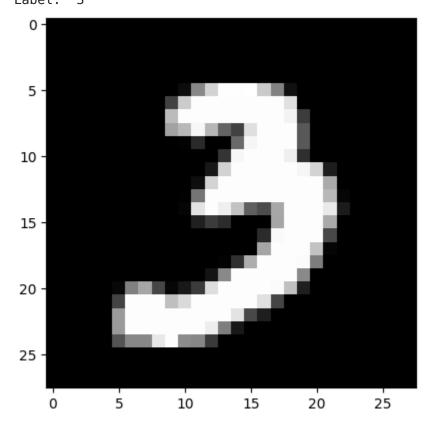


Prediction: [5] Label: 5



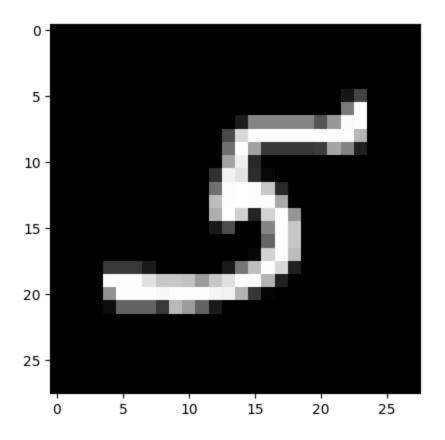
In [43]: test_prediction(33,W1,b1,W2,b2)

Prediction: [3] Label: 3



In [44]: test_prediction(12,W1,b1,W2,b2)

Prediction: [3] Label: 5



In [45]: test_prediction(34,W1,b1,W2,b2)

Prediction: [3] Label: 3

