Feed Forward Neural Networks - ML Lab assignment

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Section: A

Code includes implementation of:

- XOR gate: Adaline Backpropagation
- Hand-written digits classification
- Hand-written character classification

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
```

Functions required for Neural Networks

Initializing parameters for all layers

Forward propagation

· Activation function: sigmoid

Forward propagation equations:

```
Z[1] = W[1].X + b[1]
A[1] = g( Z[1] )
Where,
    Z = weighted sum of input and bias
    A = activations of particular layer
    1 = layer
```

Backward propagation

Backward propagation equations:

Implementation

```
def sigmoid(X):
    return 1/(1 + np.exp(-1*X))

def computation(X, y, parameters, eta, num_iters, batch = False):
    W1_storage = []
    W2_storage = []
    b1_storage = []
    b2_storage = []
    m = X.shape[0]  # number of training examples
```

```
for itr in range(num_iters):
 # iterate for each training example
 for i in range(m):
    # forward pass for each example
    hidden_output = sigmoid(np.dot(X[i], parameters["W1"].T) + parameters["b1"])
    final output = sigmoid(np.dot(hidden output, parameters["W2"].T) + parameters["b2"])
    # backward pass for each example
    dOutput = final output*(1 - final output)*(y[i] - final output)
    dHidden = hidden_output*(1 - hidden_output)*np.dot(dOutput, parameters["W2"])
    # weight changes
    dW2 = eta*dOutput.reshape(-1, 1)*hidden output
    dW1 = eta*dHidden.reshape(-1, 1)*X[i]
    # bias changes
    db2 = eta*dOutput
    db1 = eta*dHidden
    if batch == True:
     W1 storage.append(dW1)
     W2_storage.append(dW2)
      b1 storage.append(db1)
     b2_storage.append(db2)
    else:
      parameters["W2"] += dW2
      parameters["W1"] += dW1
      parameters["b2"] += db2
      parameters["b1"] += db1
 # for batch update, parameters updated here
 if batch == True:
    parameters["W2"] += sum(W2 storage)
    parameters["W1"] += sum(W1_storage)
    parameters["b2"] += sum(b2_storage)
    parameters["b1"] += sum(b1 storage)
parameters["W2"] = np.squeeze(parameters["W2"])
parameters["W1"] = np.squeeze(parameters["W1"])
parameters["b2"] = np.squeeze(parameters["b2"])
parameters["b1"] = np.squeeze(parameters["b1"])
return parameters
```

Training and testing model

```
def train(X, y, parameters, alpha, num_iters, batch=True):
 parameters = computation(X, y, parameters, alpha, num_iters, batch)
 return parameters
def test(X, y_test, parameters):
 y pred = []
 counter = 0
 for i in range(X.shape[0]):
   hidden_output = sigmoid(np.dot(X[i], parameters["W1"].T) + parameters["b1"])
   final_output = sigmoid(np.dot(hidden_output, parameters["W2"].T) + parameters["b2"])
   y_pred.append(final_output)
 y_pred = np.asarray(y_pred)
 #print(y_pred)
 y_pred[y_pred < 0.5] = 0
 y_pred[y_pred >= 0.5] = 1
 #print(y_pred)
 #print(y_test)
 accuracy = np.mean(np.asarray(y_pred) == y_test)
  print("Accuracy : {} %".format(accuracy*100))
```

Hand-written digits: Loading + Formatting + Training + **Testing**

```
def modify_label(y, n):
     new_y = []
     for i in range(y.shape[0]):
        row = np.zeros(n)
        row[y[i, 0]] = 1.
        new_y.append(row)
     return np.asarray(new_y)
   data = pd.read_csv("/content/sample_data/mnist_train_small.csv", header=None)
   data = data.to numpy()
   x_train = data[:1000, 1:]
   y_train = data[:1000, 0]
   y train = np.expand dims(y train, axis=1)
   y_train = modify_label(y_train, 10)
   x_{train} = x_{train} / 255.0
   print("Features : \n{}".format(x_train.shape))
https://colab.research.google.com/drive/1lNuHpi1TjBWEeo1XmXEOCdgupjq8-ws-#scrollTo=74qC-17OwpFx&printMode=true
```

```
print("Labels : \n{}".format(y_train.shape))
print("\nDataset description : ")
print("Digits : 0-9")
print("Image size : 28x28 = 784 pixels")
print("Pixel values range : 0-255")
print("Total number of images : {}".format(x_train.shape[0]))
     Features:
     (1000, 784)
     Labels :
     (1000, 10)
     Dataset description :
     Digits: 0-9
     Image size : 28x28 = 784 pixels
     Pixel values range : 0-255
     Total number of images: 1000
parameters = initialize_parameters([784, 50, 10])
print("Length of parameters dictionary : {}".format(len(parameters)))
     Length of parameters dictionary: 4
print("Training model...")
parameters = train(x_train, y_train, parameters, 0.01, 10)
     Training model...
print("Testing model..")
data_2 = pd.read_csv("/content/sample_data/mnist_test.csv", header=None)
data_2 = data_2.to_numpy()
x_test = data_2[:100, 1:]
y_test = data_2[:100, 0]
y_test = np.expand_dims(y_test, axis=1)
y_test = modify_label(y_test, 10)
x_{test} = x_{test} / 255.0
test(x_test, y_test, parameters)
     Testing model..
     Accuracy : 90.0 %
```

Character recognition: Loading + Formatting + Training + Testing

```
data_3 = pd.read_csv("/content/drive/My Drive/A_Z Handwritten Data.csv", header=None)
print(data_3.describe())
```

```
data 3 = data 3.to numpy()
X = data 3[:3000, 1:]/255.0
y = data_3[:3000, 0]
y = np.expand_dims(y, axis=1)
y = modify label(y, 26)
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
print("Features : \n{}".format(x_train.shape))
print("Labels : \n{}".format(y_train.shape))
print("\nDataset description : ")
print("Alphabets : A-Z")
print("Image size : 28x28 = 784 pixels")
print("Pixel values range : 0-255")
print("Total number of images : {}".format(x_train.shape[0]))
print()
print("x_train shape : {}".format(x_train.shape))
print("x_test shape : {}".format(x_test.shape))
print("y_train shape : {}".format(y_train.shape))
print("y_test shape : {}".format(y_test.shape))
                                                    783
                                1
                                                                    784
С⇒
     count 372451.000000 372451.0
                                          372451.000000 372451.000000
                                0.0 ...
                13.523454
                                               0.000239
                                                              0.000011
     mean
     std
                 6.740852
                                0.0
                                               0.134852
                                                              0.006554
     min
                0.000000
                                0.0
                                               0.000000
                                                              0.000000
                                     . . .
     25%
               10.000000
                                0.0 ...
                                               0.000000
                                                              0.000000
     50%
                14.000000
                                0.0
                                               0.000000
                                                              0.000000
     75%
                18.000000
                                               0.000000
                                                              0.000000
                                0.0 ...
     max
                25.000000
                                0.0 ...
                                              82.000000
                                                              4.000000
     [8 rows x 785 columns]
     Features:
     (2400, 784)
     Labels :
     (2400, 26)
     Dataset description :
     Alphabets : A-Z
     Image size : 28x28 = 784 pixels
     Pixel values range: 0-255
     Total number of images : 2400
     x_train shape : (2400, 784)
     x_test shape : (600, 784)
     y train shape : (2400, 26)
     y_test shape : (600, 26)
parameters = initialize_parameters([784, 50, 26])
print("Length of parameters dictionary : {}".format(len(parameters)))
     Length of parameters dictionary: 4
```

```
print("Training model...")
parameters = train(x_train, y_train, parameters, 0.01, 10)

Training model...

print("Testing model...")
test(x_test, y_test, parameters)

Testing model...
   Accuracy : 100.0 %
```

XOR: Loading + Formatting + Training + Testing

```
x_train = np.array([[0.1, 0.1], [0.1, 0.9], [0.9, 0.1], [0.9, 0.9]])
x_test = np.array([[0.1, 0.1], [0.1, 0.9], [0.9, 0.1], [0.9, 0.9]])
y_train = np.array([[0.1], [0.9], [0.9], [0.1]])
y_test = np.array([0, 1, 1, 0])

print("x_train shape : {}".format(x_train.shape))
print("x_test shape : {}".format(x_test.shape))
print("y_train shape : {}".format(y_train.shape))
print("y_test shape : {}".format(y_test.shape))

x_train shape : (4, 2)
x_test shape : (4, 2)
y_train shape : (4, 1)
y_test shape : (4, 1)
```

XOR

```
print("Testing model..")
test(x_test, y_test, parameters)
```

Testing model..
Accuracy : 75.0 %