## mlp

## December 3, 2021

## 0.0.1 MT21MCS013 Jay - MLP

```
[1]: import numpy as np
  import pandas as pd
  from matplotlib import pyplot as plt

data = pd.read_csv('train.csv')
  from sklearn.metrics import classification_report

[3]: data = np.array(data)
```

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[3]: data = np.array(data)
m, n = data.shape
np.random.shuffle(data) # shuffle before splitting into dev and training sets
print(data.shape)
data_dev = data[0:1000].T
Y_dev = data_dev[0]
X_dev = data_dev[1:n]
X_dev = X_dev / 255.

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
```

(42000, 785)

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[3]: def init_params():
    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))
```

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def forward_prop(W1, b1, W2, b2, X):
          Z1 = W1.dot(X) + b1
          A1 = ReLU(Z1)
          Z2 = W2.dot(A1) + b2
          A2 = softmax(Z2)
          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))
          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
[14]: def get_predictions(A2):
          return np.argmax(A2, 0)
      def get_accuracy(predictions, Y):
          print(predictions, Y)
          return np.sum(predictions == Y) / Y.size
      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)
```

return A

```
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))

print(classification_report(Y,get_predictions(A2)))

return W1, b1, W2, b2
```

[16]: W1, b1, W2, b2 = gradient\_descent(X\_train, Y\_train, 0.10, 1000)

	precision	recall	f1-score	support
0	0.93	0.95	0.94	4042
1	0.94	0.97	0.95	4567
2	0.89	0.86	0.87	4082
3	0.85	0.85	0.85	4265
4	0.88	0.88	0.88	3969
5	0.84	0.82	0.83	3711
6	0.90	0.92	0.91	4033
7	0.90	0.89	0.90	4298
8	0.85	0.83	0.84	3965
9	0.82	0.84	0.83	4068
accuracy			0.88	41000
macro avg	0.88	0.88	0.88	41000
weighted avg	0.88	0.88	0.88	41000

[]: