

## ✓ Digit Number



```
1 import matplotlib.pyplot as plt
2 from sklearn.metrics import accuracy_score
3 from sklearn.preprocessing import OneHotEncoder
4 from sklearn.model_selection import train_test_split
5 from sklearn.preprocessing import StandardScaler
6 from sklearn.datasets import fetch_openml
7 import numpy as np
8
9 np.random.seed(4)
```

## ✓ Dataset

```
1
2 mnist = fetch_openml('mnist_784')
3 X, y = mnist['data'].to_numpy(), mnist['target'].to_numpy().astype(int)
4
5
6 1 # show
7 num_imgs = 5
8 fig, axes = plt.subplots(1, num_imgs)
9 for i in range(num_imgs):
10     axes[i].imshow(X[i].reshape(28,28))
11     axes[i].set_title(f"Label: {y[i]}")
12     axes[i].axis('off')
13 plt.show()
```



## ✓ Pre-Processing

```
1 y_ohe = OneHotEncoder().fit_transform(y.reshape(-1,1)).toarray()
2 X_train, X_test, y_train, y_test = train_test_split(X, y_ohe, test_size=0.2, random_state=12)
3
4 scaler = StandardScaler()
5 X_train = scaler.fit_transform(X_train)
6 X_test = scaler.transform(X_test)
7
8
9 1 def sigmoid(x):
10
11     x = np.clip(x, -100, 100)
12     return 1 / (1 + np.exp(-x))
13
14
15 6 def sigmoid_prime(x):
16     return x * (1 - x)
17
18
19 1 def forward_prop(w1, w2, b1, b2, x):
20     z1 = x @ w1 + b1
21     h1 = sigmoid(z1)
22
23     z2 = h1 @ w2 + b2
24     y_hat = sigmoid(z2)
25
26     return z1, h1, z2, y_hat
27
28
29 10 def back_prop(m,w1,w2,z1,h1,z2,y_hat,x,y):
30     dz2 = y_hat - y
31     dw2 = h1.T @ dz2
32     db2 = np.ones((1,m)) @ dz2 / m
33
34     dz1 = dz2 @ w2.T * sigmoid_prime(h1)
35     dw1 = x.T @ dz1
36
37     db1 = np.ones((1,m)) @ dz1 / m
```

```

19
20     return dw1, db1, dw2, db2
21     pass

```

## Initialization



```

1 ## Initialize weights
2 n_x = X_train.shape[1]
3 n_y = y_train.shape[1]
4 n_h = 100
5
6 w1 = np.random.rand(n_x, n_h) - 0.5
7 w2 = np.random.rand(n_h, n_y) - 0.5
8 b1 = np.random.rand(1, n_h) - 0.5
9 b2 = np.random.rand(1, n_y) - 0.5

```

## Main Loop

```

1 epoch = 30
2 losses = []
3 m = y_train.shape[0]          # of data set
4 lr = 0.01                     # Learning rate
5
6 y_train_true = np.argmax(y_train, axis=1)
7 y_test_true = np.argmax(y_test, axis=1)
8 for i in range(epoch):
9     z1, a1, z2, y_hat = forward_prop(w1, w2, b1, b2, X_train)
10    loss = -(1/m) * np.sum(y_train * np.log(y_hat + 1e-10) + (1-y_train)*np.log(1-y_hat+1e-10))
11
12    losses.append(loss)
13
14    dw1, db1, dw2, db2 = back_prop(m,w1,w2,z1,a1,z2,y_hat,X_train,y_train)
15    w2 = w2 - lr * dw2
16    w1 = w1 - lr * dw1
17
18    b2 = b2 - lr * db2
19    b1 = b1 - lr * db1
20
21    print(f'loss: {loss}')

```



```

loss: 7.862787070527112
loss: 41.457222246718935
loss: 41.21939602001462
loss: 44.322213370817344
loss: 24.244484037920216
loss: 25.085660769440214
loss: 40.98787024898897
loss: 45.98855818444205
loss: 26.217540929525434
loss: 28.090226106233935
loss: 20.38737869311946
loss: 18.679223100004922
loss: 22.300543158052015
loss: 17.261754708416124
loss: 14.234768289358929
loss: 13.304205942274837
loss: 12.53498234024039
loss: 11.758791319884178
loss: 14.258909790888914
loss: 17.03586375639763
loss: 12.854193823814963
loss: 14.303639151047744
loss: 13.47702477792577
loss: 11.1418756273927
loss: 11.779950605992529
loss: 10.865521317878816
loss: 10.12921926027023
loss: 8.434598714105855
loss: 9.043357339978035
loss: 7.942675569074206

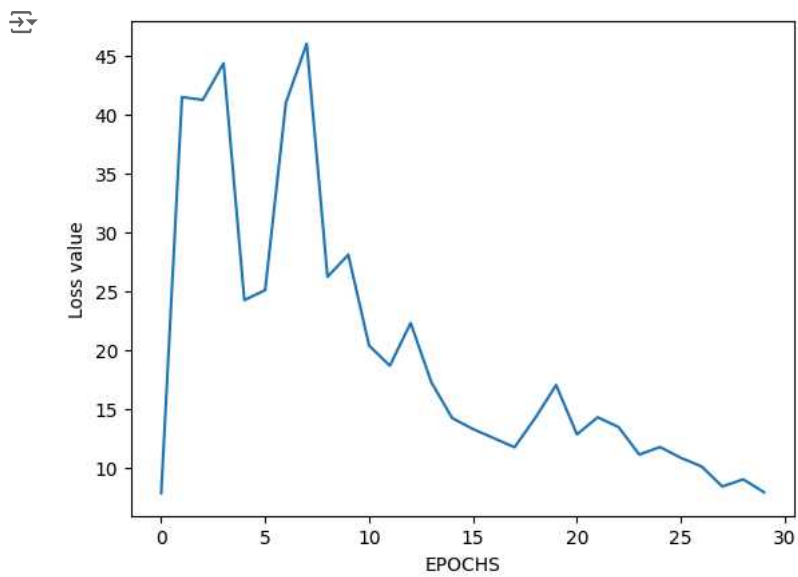
```

## Loss

```

1 plt.plot(losses)
2 plt.xlabel("EPOCHS")
3 plt.ylabel("Loss value")
4 plt.show()

```



## Accuracy

```
1 _, _, _, y_test_hat = forward_prop(w1, w2, b1, b2, X_test)
2 y_test_hat_true = np.argmax(y_test_hat, axis=1)
3 accuracy = accuracy_score(y_test_true, y_test_hat_true)
4 print(f'loss: {loss:.2f}, acc: {accuracy:.2f}')
```

loss: 7.94, acc: 0.83