

Deep Learning: fully connected MNIST network

1. The goal is to train a fully connected network for the popular MNIST dataset. You will use the code fragments below as a guide to perform this task. Use the MIT code libraries found at: www.deeplearning.ai

```
1 %tensorflow_version 2.x
2 import tensorflow as tf
3
4 !pip install mitdeeplearning
5 import mitdeeplearning as mdl
6
7 import matplotlib.pyplot as plt
8 import numpy as np
9 import random
10 from tqdm import tqdm
11
12 #download and load the MNIST dataset
13 #The training set is made up of 28x28 grayscale images of handwritten digits.
14
15 mnist = tf.keras.datasets.mnist
16 (train_images, train_labels), (test_images, test_labels) = mnist.load_data()
17 train_images = (np.expand_dims(train_images, axis=-1)/255.).astype(np.float32)
18 train_labels = (train_labels).astype(np.int64)
19 test_images = (np.expand_dims(test_images, axis=-1)/255.).astype(np.float32)
20 test_labels = (test_labels).astype(np.int64)
21
22 #use the Keras API and define the model using the Sequential class.
23 #First use a Flatten layer, which flattens the input so that it can be fed
24 #into the model.
25
26 #In this next block, define the fully connected layers of this simple work.
27 #Use a sigmoid activation
28
29 def build_fc_model():
30     fc_model = tf.keras.Sequential([
31         # First define a Flatten layer
32         tf.keras.layers.Flatten(),
33
34         # '''TODO: Define the activation function for the first fully connected
35         # (Dense) layer.'''
36         tf.keras.layers.Dense(128, activation= '''TODO''' ),
37
38         # '''TODO: Define the second Dense layer to output the classification
39         # probabilities'''
```

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```
37     '''TODO: Dense layer to output classification probabilities'''
38 ]
39 return fc_model
40
41 model = build_fc_model()
42
43 #####%
44 #Compile the model with appropriate optimizer and loss function
45
46 model.compile(optimizer=tf.keras.optimizers.SGD(learning_rate=1e-1),
47                 loss='sparse_categorical_crossentropy',
48                 metrics=['accuracy'])
49
50
51 #train the fully connected model
52
53
54 # Define the batch size and the number of epochs to use during training
55 BATCH_SIZE = 64
56 EPOCHS = 5
57
58 model.fit(train_images, train_labels, batch_size=BATCH_SIZE, epochs=EPOCHS)
59
60
61 #Use the evaluate method to evaluate the model on the test dataset!
62
63
64 '''TODO: Use the evaluate method to test the model!'''
65 test_loss, test_acc = # TODO
66
67 print('Test accuracy:', test_acc)
```

2. Compare the performance of other loss functions:

```
1 mse = tf.keras.losses.MeanSquaredError()
2
3 mape = tf.keras.losses.MeanAbsolutePercentageError()
```