

```
!pip install jupyter-dash
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/pypi/simple
Requirement already satisfied: jupyter-dash in /usr/local/lib/python3.7/dist-packages
Requirement already satisfied: dash in /usr/local/lib/python3.7/dist-packages (from jupyter-dash)
Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: nest-asyncio in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: flask in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: ipykernel in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: retrying in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: ipython in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: ansi2html in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: importlib-metadata in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: plotly>=5.0.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: dash-core-components==2.0.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: dash-table==5.0.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: flask-compress in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: dash-html-components==2.0.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: Jinja2<3.0,>=2.10.1 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: itsdangerous<2.0,>=0.24 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: Werkzeug<2.0,>=0.15 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: click<8.0,>=5.1 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: MarkupSafe>=0.23 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: tenacity>=6.2.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: brotli in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: typing-extensions>=3.6.4 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: traitlets>=4.1.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: tornado>=4.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: jupyter-client in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: pickleshare in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: pygments in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: setuptools>=18.5 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: pexpect in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: decorator in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: prompt-toolkit<2.0.0,>=1.0.4 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: simplegeneric>0.8 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: wcwidth in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: pyzmq>=13 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: pyzmq-dateutil>=2.1 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: jupyter-core>=4.6.0 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: ptyprocess>=0.5 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from dash)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from dash)
```

```
import plotly.express as px
from jupyter_dash import JupyterDash
import dash_core_components as dcc
import dash_html_components as html
```

✓ 0s completed at 2:14 AM



webpage

Google Drive

```
# Connecting to the Google Drive for the permanent storage
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.r

```
# Checking if the correct version of tensorflow is installed or not
import tensorflow as tf
print("TensorFlow version:", tf.__version__)
```

TensorFlow version: 2.8.2

```
from numpy.random import seed
seed(786)
# Tensor flow for the fast calculation or computation
import tensorflow
tensorflow.random.set_seed(786)
```

```
# Other imports like os for directory management and numpy for mathematical operation on a
import os
```

```
# Numpy for performing mathematical operations on arrays
import numpy as np
```

```
# Importing pandas for data visualization in the forms of tables or data frames
import pandas as pd
```

```
# Time for dealing with real time and the pillow for dealing with the image for prediction:
from time import strftime
```

```
# Pillow for the image used for prediction
from PIL import Image
```

```
# Loading all the files from the directory into variables
```

```
X_TRAINING_PATH = '/content/drive/MyDrive/MNIST/digit_xtrain.csv'
X_TESTING_PATH = '/content/drive/MyDrive/MNIST/digit_xtest.csv'
Y_TRAINING_PATH = '/content/drive/MyDrive/MNIST/digit_ytrain.csv'
Y_TESTING_PATH = '/content/drive/MyDrive/MNIST/digit_ytest.csv'
```

```
%%time
```

```
# Getting the wall time required for reading the y-training csv file
```

```
y_training_all = np.loadtxt(Y_TRAINING_PATH, delimiter=',', dtype=int)
```

```
CPU times: user 271 ms, sys: 12 ms, total: 283 ms
```

```
Wall time: 324 ms
```

```
y_training_all.shape
```

```
(60000,)
```

```
%%time
```

```
y_testing = np.loadtxt(Y_TESTING_PATH, delimiter=',', dtype=int)
```

```
CPU times: user 39.6 ms, sys: 1.98 ms, total: 41.6 ms
```

```
Wall time: 44 ms
```

```
%%time
```

```
x_training_all = np.loadtxt(X_TRAINING_PATH, delimiter=',', dtype=int)
```

```
CPU times: user 35.5 s, sys: 4.48 s, total: 40 s
```

```
Wall time: 48.4 s
```

```
%%time
```

```
x_testing = np.loadtxt(X_TESTING_PATH, delimiter=',', dtype=int)
```

```
CPU times: user 4.22 s, sys: 218 ms, total: 4.43 s
```

```
Wall time: 4.44 s
```

Visualizing the Dataset

```
x_training_all.shape
```

```
(60000, 784)
```

```
x_training_all[0]
```

```
array([ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  3, 18, 18, 18,
126, 136, 175,  26, 166, 255, 247, 127,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  30,  36,  94, 154, 170, 253,
253, 253, 253, 253, 225, 172, 253, 242, 195,  64,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  49, 238, 253, 253, 253,
253, 253, 253, 253, 253, 251,  93,  82,  82,  56,  39,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 18, 219, 253,
253, 253, 253, 253, 198, 182, 247, 241,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        80, 156, 107, 253, 253, 205,  11,  0,  43, 154,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0, 14,  1, 154, 253,  90,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0, 139, 253, 190,  2,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  11, 190, 253,  70,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  35,
241, 225, 160, 108,  1,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  81, 240, 253, 253, 119,  25,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  45, 186, 253, 253, 150,  27,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0, 16,  93, 252, 253, 187,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 249,
253, 249,  64,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  46, 130,
183, 253, 253, 207,  2,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  39, 148,
229, 253, 253, 253, 250, 182,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  24, 114,
221, 253, 253, 253, 253, 201,  78,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  23,  66,
213, 253, 253, 253, 253, 198,  81,  2,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 18, 171,
219, 253, 253, 253, 253, 195,  80,  9,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  55, 172,
226, 253, 253, 253, 253, 244, 133, 11,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
136, 253, 253, 253, 212, 135, 132, 16,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
```

[illegible]

```
y_training_all.shape
```

(60000,)

```
x_testing.shape
```

 $(10000, 784)$

```
# First 10 labels from training dataset
```

```
y_training_all[:10]
```

```
array([5, 0, 4, 1, 9, 2, 1, 3, 1, 4])
```

Our Features are between 0 and 255 which is a large range. So, we have to rescale our tra

```
# After rescaling our data is between 0 and 1
```

```
x_training_all = x_training_all / 255.0
```

```
x_testing = x_testing / 255.0
```

```
x_training_all[0]
```

[illegible]

```

0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.01176471, 0.07058824, 0.07058824,
0.07058824, 0.49411765, 0.53333333, 0.68627451, 0.10196078,
0.65098039, 1.      , 0.96862745, 0.49803922, 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.11764706, 0.14117647, 0.36862745, 0.60392157,
0.66666667, 0.99215686, 0.99215686, 0.99215686, 0.99215686,
0.99215686, 0.88235294, 0.6745098 , 0.99215686, 0.94901961,
0.76470588, 0.25098039, 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.19215686, 0.93333333,
0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.99215686,
0.99215686, 0.99215686, 0.99215686, 0.98431373, 0.36470588,
0.32156863, 0.32156863, 0.21960784, 0.15294118, 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.07058824, 0.85882353, 0.99215686, 0.99215686,
0.99215686, 0.99215686, 0.99215686, 0.77647059, 0.71372549,
0.96862745, 0.94509804, 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.31372549, 0.61176471, 0.41960784, 0.99215686, 0.99215686,
0.80392157, 0.04313725, 0.      , 0.16862745, 0.60392157,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.05490196,

```

```
# Our labels are 10 in total from 0 to 9
```

```
NR_CLASSES = 10
```

```
y_training_all = np.eye(NR_CLASSES)[y_training_all]
```

```
y_training_all.shape
```

```
(60000, 10)
```

```
y_testing = np.eye(NR_CLASSES)[y_testing]
```

```
y_testing.shape
```

```
(10000, 10)
```

```
LOGGING_PATH = '/content/sample_data/MNISTtensorboard_mnist_digit_logs/'
```

```
VALIDATION_SIZE = 10000
```

```
IMAGE_WIDTH = 28
```

```
IMAGE_HEIGHT = 28
CHANNELS = 1
# Total number of features
TOTAL_INPUTS = IMAGE_WIDTH*IMAGE_HEIGHT*CHANNELS

# Now we have to divide our training dataset into smaller training and validation dataset
# Training dataset contains 50000 and avalidation contains 10000

# From start to the validation size
x_val = x_training_all[:VALIDATION_SIZE]
y_val = y_training_all[:VALIDATION_SIZE]

x_val.shape

(10000, 784)

# From validation size till the end
x_training = x_training_all[VALIDATION_SIZE:]
y_training = y_training_all[VALIDATION_SIZE:]

x_training.shape

(50000, 784)

# Creating tensors
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()

# tf.placeholder is used for creating tensors 2 parameters (datatype, shape of tensor)
X = tf.placeholder(tf.float32, shape=[None, TOTAL_INPUTS], name='X')
Y = tf.placeholder(tf.float32, shape=[None, NR_CLASSES], name='labels')

# Hyperparameters
# nr_epochs = 25
# learning_rate = 1e-3
nr_epochs = 50
learning_rate = 1e-3

n_hidden1 = 512
n_hidden2 = 64

1e-3

0.001
```

```

def setup_layer(input, weight_dim, bias_dim, name):

    with tf.name_scope(name):
        """
        Iniatiating the weights
        tf.truncated_normal generates random values except for extreme values
        Shape = no of inputs and neurons in the layer
        Standard Deviation tells how far appart weights should from each others
        """

        initial_w = tf.truncated_normal(shape=weight_dim, stddev=0.1, seed=42)
        # Calculations for weights
        w = tf.Variable(initial_value=initial_w, name='W')

        """
        Initializing the biases
        All biases start from same value that is 0
        Shape is no of neurons in layer
        """

        initial_b = tf.constant(value=0.0, shape=bias_dim)
        # Calculation of the biases
        b = tf.Variable(initial_value=initial_b, name='B')

        """
        Input layer of next hidden layer
        MatrixMultiplication
        """

        layer_in = tf.matmul(input, w) + b

        # Checking for the the last hidden layer to apply activation accordingly
        if name=='out':
            # Apply softmax for the last layer
            layer_out = tf.nn.softmax(layer_in)
        else:
            # Apply relu for the remaining layers
            layer_out = tf.nn.relu(layer_in)

        tf.summary.histogram('weights', w)
        tf.summary.histogram('biases', b)

    return layer_out

layer_1 = setup_layer(X, weight_dim=[TOTAL_INPUTS, n_hidden1],
                      bias_dim=[n_hidden1], name='layer_1')

layer_drop = tf.nn.dropout(layer_1, keep_prob=0.8, name='dropout_layer')

layer_2 = setup_layer(layer_drop, weight_dim=[n_hidden1, n_hidden2],
                      bias_dim=[n_hidden2], name='layer_2')

output = setup_layer(layer_2, weight_dim=[n_hidden2, NR_CLASSES],

```



```

output = setup_layer(layer_2, weight_dim=[n_hidden1, n_hidden2, nr_classes],
                    bias_dim=[nr_classes], name='out')

model_name = f'{n_hidden1}-D0-{n_hidden2} LR{learning_rate} E{nr_epochs}'

# Folder for Tensorboard

folder_name = f'{model_name} at {strftime("%H:%M")}'
directory = os.path.join(LOGGING_PATH, folder_name)

try:
    os.makedirs(directory)
except OSError as exception:
    print(exception.strerror)
else:
    print('Successfully created directories!')

    Successfully created directories!

```

Defining Loss Function

```

with tf.name_scope('loss_calc'):
    loss = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits_v2(labels=Y, logits=output))

```

Defining Optimizer

```

with tf.name_scope('optimizer'):
    optimizer = tf.train.AdamOptimizer(learning_rate)
    train_step = optimizer.minimize(loss)

```

Accuracy Metric

```

with tf.name_scope('accuracy_calc'):
    correct_pred = tf.equal(tf.argmax(output, axis=1), tf.argmax(Y, axis=1))
    accuracy = tf.reduce_mean(tf.cast(correct_pred, tf.float32))

with tf.name_scope('performance'):
    tf.summary.scalar('accuracy', accuracy)
    tf.summary.scalar('cost', loss)

```

Check Input Images in Tensorboard

```
with tf.name_scope('show_image'):  
    x_image = tf.reshape(X, [-1, 28, 28, 1])  
    tf.summary.image('image_input', x_image, max_outputs=4)
```

Run Session

```
sess = tf.Session()
```

Setup Filewriter and Merge Summaries

```
merged_summary = tf.summary.merge_all()  
  
train_writer = tf.summary.FileWriter(directory + '/train')  
train_writer.add_graph(sess.graph)  
  
validation_writer = tf.summary.FileWriter(directory + '/validation')
```

Initialise all the variables

```
init = tf.global_variables_initializer()  
sess.run(init)
```

Batching the Data

```
size_of_batch = 1000  
# size_of_batch = 500  
# size_of_batch = 2000  
  
num_examples = y_training.shape[0]  
nr_iterations = int(num_examples/size_of_batch)  
  
index_in_epoch = 0
```

this function is to go to the next batch

```
def next_batch(batch_size, data, labels):

    global num_examples
    global index_in_epoch

    start = index_in_epoch
    index_in_epoch += batch_size

    if index_in_epoch > num_examples:
        start = 0
        index_in_epoch = batch_size

    end = index_in_epoch

    return data[start:end], labels[start:end]
```

Training Loop

```
import array as arr
accuracy_array = []

%%time
for epoch in range(nr_epochs):

    # ===== Training Dataset =====
    for i in range(nr_iterations):

        batch_x, batch_y = next_batch(batch_size=size_of_batch, data=x_training, labels=y_1

        feed_dictionary = {X:batch_x, Y:batch_y}

        sess.run(train_step, feed_dict=feed_dictionary)

    s, batch_accuracy = sess.run(fetches=[merged_summary, accuracy], feed_dict=feed_diction

    train_writer.add_summary(s, epoch)

    print(f'Epoch {epoch} \t| Training Accuracy = {batch_accuracy}')
```

```
# ===== Validation =====

summary = sess.run(fetches=merged_summary, feed_dict={X:x_val, Y:y_val})
validation_writer.add_summary(summary, epoch)

accuracy_array.append(batch_accuracy)
```

```
print('Done training!')
```

Epoch 0	Training Accuracy = 0.8450000286102295
Epoch 1	Training Accuracy = 0.859000027179718
Epoch 2	Training Accuracy = 0.8659999966621399
Epoch 3	Training Accuracy = 0.8730000257492065
Epoch 4	Training Accuracy = 0.9739999771118164
Epoch 5	Training Accuracy = 0.9769999980926514
Epoch 6	Training Accuracy = 0.9810000061988831
Epoch 7	Training Accuracy = 0.9819999933242798
Epoch 8	Training Accuracy = 0.9819999933242798
Epoch 9	Training Accuracy = 0.9860000014305115
Epoch 10	Training Accuracy = 0.9850000143051147
Epoch 11	Training Accuracy = 0.9829999804496765
Epoch 12	Training Accuracy = 0.9829999804496765
Epoch 13	Training Accuracy = 0.9890000224113464
Epoch 14	Training Accuracy = 0.9900000095367432
Epoch 15	Training Accuracy = 0.9909999966621399
Epoch 16	Training Accuracy = 0.9879999756813049
Epoch 17	Training Accuracy = 0.9909999966621399
Epoch 18	Training Accuracy = 0.9890000224113464
Epoch 19	Training Accuracy = 0.9879999756813049
Epoch 20	Training Accuracy = 0.9900000095367432
Epoch 21	Training Accuracy = 0.9879999756813049
Epoch 22	Training Accuracy = 0.9909999966621399
Epoch 23	Training Accuracy = 0.9900000095367432
Epoch 24	Training Accuracy = 0.9919999837875366
Epoch 25	Training Accuracy = 0.9919999837875366
Epoch 26	Training Accuracy = 0.9909999966621399
Epoch 27	Training Accuracy = 0.9890000224113464
Epoch 28	Training Accuracy = 0.9909999966621399
Epoch 29	Training Accuracy = 0.9919999837875366
Epoch 30	Training Accuracy = 0.9919999837875366
Epoch 31	Training Accuracy = 0.9919999837875366
Epoch 32	Training Accuracy = 0.9919999837875366
Epoch 33	Training Accuracy = 0.9919999837875366
Epoch 34	Training Accuracy = 0.9919999837875366
Epoch 35	Training Accuracy = 0.9900000095367432
Epoch 36	Training Accuracy = 0.9909999966621399
Epoch 37	Training Accuracy = 0.9900000095367432
Epoch 38	Training Accuracy = 0.9909999966621399
Epoch 39	Training Accuracy = 0.9909999966621399
Epoch 40	Training Accuracy = 0.9919999837875366
Epoch 41	Training Accuracy = 0.9909999966621399
Epoch 42	Training Accuracy = 0.9919999837875366
Epoch 43	Training Accuracy = 0.9919999837875366
Epoch 44	Training Accuracy = 0.9919999837875366
Epoch 45	Training Accuracy = 0.9919999837875366
Epoch 46	Training Accuracy = 0.9919999837875366
Epoch 47	Training Accuracy = 0.9919999837875366
Epoch 48	Training Accuracy = 0.9919999837875366
Epoch 49	Training Accuracy = 0.9919999837875366

Done training!

CPU times: user 3min 20s, sys: 4.59 s, total: 3min 24s

Wall time: 1min 53s

```
# accuracy of the whole training model
accuracy_of_model = np.mean(accuracy_array)
print("Accuracy of model = {}".format(accuracy_of_model*100))
```

Accuracy of model = 97.88000583648682%

Make a Prediction

```
im = Image.open('/content/drive/MyDrive/MNIST/2.png')
im
```



```
img = im.resize((28,28))
```

```
bw = img.convert('L')
```

```
img_array = np.invert(bw)
```

```
img_array.shape
```

(28, 28)

```
test_img = img_array.ravel()
```

```
test_img.shape
```

(784,)

img_array

```

array([[ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 38, 111, 160, 170,
        149, 78, 2,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  3, 134, 199, 136, 82, 63,
        95, 180, 185, 39,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  2, 157, 174, 19,  0,  0,  0,
        0,  0, 101, 208, 20,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0, 111, 184, 6,  0,  0,  0,  0,
        0,  0,  0, 143, 142,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0, 10, 206, 40,  0,  0,  0,  0,  0,
        0,  0,  0, 24, 206, 20,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0, 67, 186,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0, 186, 61,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0, 153, 109,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0, 157, 96,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  3, 199, 53,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0, 154, 98,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  4, 154, 24,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  1, 189, 56,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  1,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0, 32, 204, 12,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0, 142, 139,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0, 73, 209, 22,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0, 38, 216, 55,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0],
       r  a  a  a  a  a  a  a  a  a  a  a  a  a

```

```

[[ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  14, 199, 97,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
   0,  0],
 [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 30,
 189, 134,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
   0,  0],
 [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 63, 211,

```

```
prediction = sess.run(fetches=tf.argmax(output, axis=1), feed_dict={X:[test_img]})
```

```
print(f'Prediction for test image is {prediction}')
```

```
Prediction for test image is [2]
```

Testing and Evaluation

```
test_accuracy = sess.run(fetches=accuracy, feed_dict={X:x_testing, Y:y_testing})
```

```
print(f'Accuracy on test set is {test_accuracy:0.2%}')
```

```
Accuracy on test set is 97.66%
```

```
def predict_test_image(Img_url):
    im = Image.open(Img_url)
    img = im.resize((28, 28))
    bw = img.convert('L')
    img_array = np.invert(bw)
    test_img = img_array.ravel()
    prediction = sess.run(fetches=tf.argmax(output, axis=1), feed_dict={X:[test_img]})
    print(f'Prediction for test image is {prediction}')
```

```
def predict_test_image1(Img_url):
    im = Image.open(Img_url)
    img = im.resize((28, 28))
    bw = img.convert('L')
    img_array = np.invert(bw)
    test_img = img_array.ravel()
    prediction = sess.run(fetches=tf.argmax(output, axis=1), feed_dict={X:[test_img]})
    print(f'Prediction for test image is {prediction}')
```

Prediction 1

```
img = Image.open('/content/drive/MyDrive/MNIST/1.png')
display(img)
predict_test_image('/content/drive/MyDrive/MNIST/1.png');
```

A handwritten digit '1' in black ink on a white background.

Prediction for test image is [1]

Prediction 2

```
img = Image.open('/content/drive/MyDrive/MNIST/2.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/2.png');
```

A handwritten digit '2' in black ink on a white background.

Prediction for test image is [2]

Prediction 3

```
img = Image.open('/content/drive/MyDrive/MNIST/3.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/3.png');
```



A large, handwritten digit '3' in black ink on a white background.

Prediction for test image is [3]

Prediction 4

```
img = Image.open('/content/drive/MyDrive/MNIST/4.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/4.png');
```

A large, handwritten digit '4' in black ink on a white background.

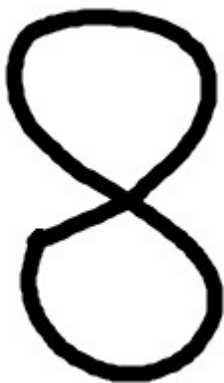
Prediction for test image is [4]

```
img = Image.open('/content/drive/MyDrive/MNIST/ii_4.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/ii_4.png');
```

A large, handwritten digit '4' in black ink on a white background.

Prediction for test image is [4]

```
img = Image.open('/content/drive/MyDrive/MNIST/8.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/8.png');
```



Prediction for test image is [8]

```
img = Image.open('/content/drive/MyDrive/MNIST/7777.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/7777.png');
```



Prediction for test image is [7]

```
img = Image.open('/content/drive/MyDrive/MNIST/22222.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/22222.png');
```



Prediction for test image is [2]

```
img = Image.open('/content/drive/MyDrive/MNIST/555555.png')  
display(img)  
predict_test_image('/content/drive/MyDrive/MNIST/555555.png');
```



Prediction for test image is [8]

Webpage

```
temp = test_accuracy*100
model_accuracy = round(temp,2)
df = px.data.tips()# Build App
app = JupyterDash(__name__)
app.layout = html.Div([
    html.H1("ML SEMESTER PROJECT"),
    html.H1("DIGIT RECOGNITION"),
    html.H2("Model Accuracy: {}".format(model_accuracy)),
    html.H2("Img:"),
    #html.Img(src = "https://drive.google.com/uc?export=view&id=1UcMUe3ifP9ijRuQtHWavWsvVnF"),
    html.Img(src = "https://drive.google.com/uc?export=view&id=1_L1CCgQ-x1c5FCeExEPmXoXmBpr"),

    html.H2("Prediction: {}".format(prediction)),
])

# app.run_server(mode='external')
app.run_server(host="127.0.0.1", port="8000")
```

Dash app running on:

<http://127.0.0.1:8000/>

Reset for the Next Run

```
train_writer.close()
validation_writer.close()
sess.close()
tf.reset_default_graph()
```

