```
In [1]:
             1 # Ignore warnings for cleaner output
               import warnings
             2
             3
               warnings.filterwarnings('ignore')
             4
               # Importing libraries for machine learning and deep learning
             5
               from sklearn.preprocessing import MinMaxScaler # For data normaliza
             7
               from keras.models import Sequential # For creating a sequential new
             8 from keras.layers import Dense, Dropout, LSTM, Bidirectional # For
In [2]:
             1 import tensorflow as tf
In [3]:
         H
               from keras.datasets import mnist
             1
               from keras.utils import to_categorical
In [4]:
             1 # Load the MNIST dataset
               (x_train, y_train), (x_test, y_test) = mnist.load_data()
            Downloading data from https://storage.googleapis.com/tensorflow/tf-ker
            as-datasets/mnist.npz (https://storage.googleapis.com/tensorflow/tf-ke
            ras-datasets/mnist.npz)
            11490434/11490434 [============== ] - 6s lus/step
In [9]:
             1 print(x_train.shape)
             2 print(y_train.shape)
             3 print(x_test.shape)
               print(y_test.shape)
             5
            (60000, 28, 28)
            (60000, 10)
            (10000, 28, 28)
```

(10000, 10)

```
In [10]:
                 print(x_train)
             [[[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. 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.]]]

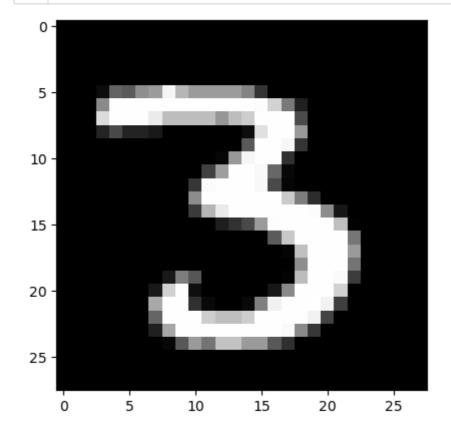
```
In [11]:
                 print(x_test)
             [[[0. 0. 0. ... 0. 0. 0.]]
               [0. 0. 0. ... 0. 0. 0.]
               [0. 0. 0. ... 0. 0. 0.]
               [0. 0. 0. ... 0. 0. 0.]
               [0. 0. 0. ... 0. 0. 0.]
               [0. 0. 0. ... 0. 0. 0.]]
              [[0. 0. 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. 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.]]

```
In [12]:
                           H
                                        1
                                              # Reshaping the input Data which is used as a input in CNN in Tense
                                        2
                                             # CNN takes the input Data in 4D Format with the shape (num_samples
                                        3 # Here (num_channels) is set to 1 which means input image is Graysc
                                        5
                                              x_train = x_train.reshape((x_train.shape[0] , x_train.shape[1] , x_
                                        6 x_test = x_test.reshape((x_test.shape[0] , x_test.shape[1] , x_test
                                        7
                                              print(x_train.shape)
                                             print(x_test.shape)
                                        8
                                        9 print(x_train.dtype)
                                     10 print(x test.dtype)
                                    (60000, 28, 28, 1)
                                    (10000, 28, 28, 1)
                                   float64
                                   float64
In [13]:
                           H
                                        1
                                               # Normalizing Pixel Values
                                        2
                                        3
                                             x train = x train.astype('float32')/255.0
                                        4 x_test = x_test.astype('float32')/255.0
                                        5 print(x train.dtype)
                                              print(x_test.dtype)
                                   float32
                                   float32
In [14]:
                                        1
                                              fig=plt.figure(figsize=(5,3))
                                        2
                                              for i in range(20):
                                                          ax =fig.add_subplot(2,10,i+1, xticks=[], yticks=[])
                                        3
                                        4
                                                          ax.imshow(np.squeeze(x_train[i]), cmap='Blues')
                                        5
                                                          ax.set_title(y_train[i])
                                       [O. O[O. OLO. 00]. (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40). (40).
In [15]:
                                             # showing shape of single image
                            M
                                        1
                                               img_shape= x_train.shape[1:]
                                        3
                                               img_shape
         Out[15]: (28, 28, 1)
```

```
In [48]:
          H
              1
                 model = tf.keras.models.Sequential([
                     tf.keras.layers.Flatten(input_shape=(28,28,1)),
               2
                     tf.keras.layers.Dense(128, activation='relu'),
               3
              4
                     tf.keras.layers.Dropout(0.2),
               5
                     tf.keras.layers.Dense(10)
               6
                 ])
In [49]:
                 model.summary()
             Model: "sequential_5"
              Layer (type)
                                          Output Shape
                                                                     Param #
              flatten_5 (Flatten)
                                          (None, 784)
              dense_10 (Dense)
                                          (None, 128)
                                                                     100480
              dropout 5 (Dropout)
                                          (None, 128)
                                                                     0
              dense_11 (Dense)
                                          (None, 10)
                                                                     1290
             Total params: 101770 (397.54 KB)
             Trainable params: 101770 (397.54 KB)
             Non-trainable params: 0 (0.00 Byte)
In [50]:
              1 # Displaying Neural Network Model
               2 from tensorflow.keras.utils import plot_model
               3 plot_model(model, 'model.jpg', show_shapes = True)
             You must install pydot (`pip install pydot`) and install graphviz (see
             instructions at https://graphviz.gitlab.io/download/) (https://graphvi
             z.gitlab.io/download/)) for plot_model to work.
In [ ]:
               1
In [ ]:
              1
In [51]:
                 # Applying Softmax() Function to prediction array
          M
              1
                 # This convert an output vector of real numbers into a probability
                 tf.nn.softmax(prediction).numpy()
   Out[51]: array([[0.09981512, 0.09993055, 0.09990993, 0.1002295 , 0.10028599,
                     0.09991155, 0.10023038, 0.09992979, 0.09978843, 0.09996876]],
                   dtype=float32)
In [57]:
          M
              1
               2
              3
                 # Compile the model with this loss
              4
                 model.compile(optimizer='adam', loss='categorical_crossentropy', me
               5
               6
```

```
In [58]:
              model.fit(x train, y train, epochs=5,batch size=32)
            2
           Epoch 1/5
           3 - accuracy: 0.1020
           Epoch 2/5
           3 - accuracy: 0.1026
           Epoch 3/5
           1875/1875 [============= ] - 3s 1ms/step - loss: 9.627
           3 - accuracy: 0.1154
           Epoch 4/5
           2 - accuracy: 0.1138
           Epoch 5/5
           1875/1875 [============= ] - 3s 2ms/step - loss: 9.594
           6 - accuracy: 0.1132
   Out[58]: <keras.src.callbacks.History at 0x20897f708b0>
In [59]:
        H
            1 # Evaluating the Model
            2 model.evaluate(x test, y test, verbose=2)
           313/313 - 0s - loss: 11.1021 - accuracy: 0.1437 - 367ms/epoch - 1ms/st
           ер
   Out[59]: [11.10214900970459, 0.1437000036239624]
        H
            1 # Creating a new sequential model which includes both previously tr
In [60]:
            probability model = tf.keras.Sequential([ model,tf.keras.layers.Sof
              probability_model(x_test[:5])
   Out[60]: <tf.Tensor: shape=(5, 10), dtype=float32, numpy=</pre>
           array([[0.09602615, 0.09623323, 0.10268615, 0.09590027, 0.10280494,
                  0.10257605, 0.10288402, 0.10250293, 0.10222689, 0.09615933],
                 [0.09596132, 0.09618185, 0.10310774, 0.09572108, 0.10281739,
                  0.10295112, 0.10280609, 0.10200837, 0.10237346, 0.09607158],
                 [0.09649432, 0.09598102, 0.10274434, 0.09603737, 0.10268992,
                  0.10249823, 0.10279985, 0.10238555, 0.10209569, 0.09627371],
                 [0.09496093, 0.09643601, 0.10280189, 0.0956667, 0.10293792,
                  0.10328193, 0.10299419, 0.10222725, 0.10274 , 0.09595316],
                 [0.0960004 , 0.09665856, 0.10272933, 0.096179 , 0.10258953,
                  0.10252755, 0.10248248, 0.10221114, 0.10235294, 0.09626911],
                dtype=float32)>
```



```
In []: H 1
```

```
In []: 🔰 1
```