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Roll No: TETB19

Sub: Soft Computitng

Batch: B2

```
import tensorflow
```

```
from tensorflow import keras
```

```
import tensorflow as tf
from tensorflow import keras
import matplotlib.pyplot as plt
%matplotlib inline
import numpy as np
```

```
(X_train, y_train) , (X_test, y_test) = keras.datasets.mnist.load_data()
```

```
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11493376/11490434 [=====] - 0s 0us/step
11501568/11490434 [=====] - 0s 0us/step
```



```
len(X_train)
```

```
60000
```

```
len(X_test)
```

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```
X_train[0].shape
```

```
(28, 28)
```

```
X_train[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,
```



```

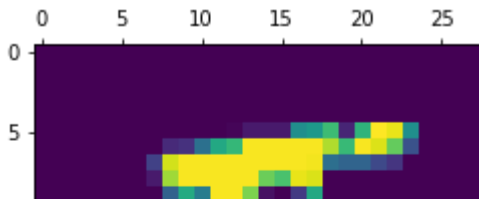
    0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 35,
241, 225, 160, 108, 1, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
81, 240, 253, 253, 119, 25, 0, 0, 0, 0, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 45, 186, 253, 253, 150, 27, 0, 0, 0, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 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, 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,
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, 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, 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, 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, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0]], dtype=uint8)

```

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```
plt.matshow(X_train[0])
```

<matplotlib.image.AxesImage at 0x7efc976e77d0>



y\_train[0]

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# Scaling Technique

X\_train = X\_train / 255

X\_test = X\_test / 255

X\_train[0]

```
0.0, 0.0, 0.0, 0.0, 0.18039216,
0.50980392, 0.71764706, 0.99215686, 0.99215686, 0.81176471,
0.00784314, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, ],
[0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.15294118, 0.58039216, 0.89803922,
0.99215686, 0.99215686, 0.99215686, 0.98039216, 0.71372549,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, ],
[0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.09411765, 0.44705882, 0.86666667, 0.99215686, 0.99215686,
0.99215686, 0.99215686, 0.78823529, 0.30588235, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, ],
[0.0, 0.0, 0.0, 0.0, 0.0,
0.77047059, 0.51764706, 0.00784314, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, ],
[0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.07058824, 0.67058824, 0.85882353, 0.99215686,
0.99215686, 0.99215686, 0.99215686, 0.76470588, 0.31372549,
0.03529412, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, ],
[0.0, 0.0, 0.0, 0.0, 0.21568627,
0.6745098, 0.88627451, 0.99215686, 0.99215686, 0.99215686,
0.99215686, 0.95686275, 0.52156863, 0.04313725, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, ],
[0.0, 0.0, 0.0, 0.0, 0.53333333,
0.99215686, 0.99215686, 0.99215686, 0.83137255, 0.52941176,
0.51764706, 0.0627451, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0]
```

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Downloaded from <http://ajph.org/> on November 10, 2015

model: Linear Convolutional / F

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```
<keras.callbacks.History at 0x7efc93077f50>
```

```
model.evaluate(X_test_flattened, y_test)
```

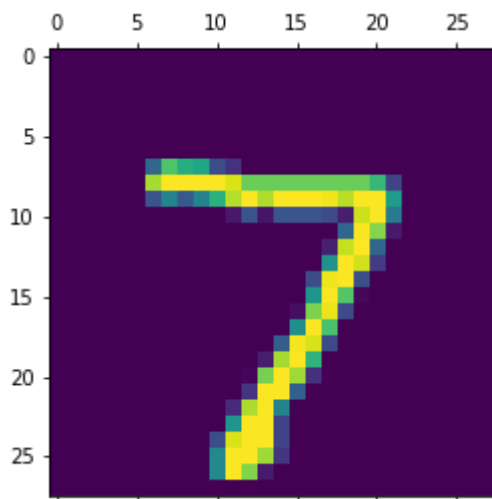
```
313/313 [=====] - 1s 2ms/step - loss: 0.2695 - accuracy: 0.9  
[0.2695145010948181, 0.9254000186920166]
```

```
y_predicted = model.predict(X_test_flattened)  
y_predicted[0]
```

```
array([1.5027165e-02, 3.9325224e-07, 6.3410342e-02, 9.5975685e-01,  
       3.2548308e-03, 1.0176152e-01, 1.0720740e-06, 9.9978119e-01,  
       7.0441395e-02, 6.0749489e-01], dtype=float32)
```

```
plt.matshow(X_test[0])
```

```
<matplotlib.image.AxesImage at 0x7efc920031d0>
```



```
# np.argmax finds a maximum element from an array and returns the index of it
```

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```
y_predicted_labels = [np.argmax(i) for i in y_predicted]
```

```
y_predicted_labels[:5]
```

```
[7, 2, 1, 0, 4]
```

```
cm = tf.math.confusion_matrix(labels=y_test, predictions=y_predicted_labels)  
cm
```

```
<tf.Tensor: shape=(10, 10), dtype=int32, numpy=  
array([[ 965,    0,    1,    2,    0,    4,    5,    2,    1,    0],  
       [   0, 1117,    3,    2,    0,    1,    4,    2,    6,    0],  
       [   6,   10,  923,   15,    9,    6,   12,   10,   38,    3],  
       [   4,    0,   19,  914,    1,   34,    2,   11,   19,    6],
```

```

[ 2, 2, 4, 2, 929, 0, 9, 4, 9, 21],
[ 8, 3, 6, 24, 12, 794, 9, 6, 25, 5],
[ 12, 3, 9, 1, 8, 20, 901, 2, 2, 0],
[ 1, 8, 22, 6, 9, 0, 0, 957, 1, 24],
[ 6, 11, 5, 23, 9, 32, 8, 12, 862, 6],
[ 11, 7, 1, 11, 45, 8, 0, 29, 5, 892]],
dtype=int32)>

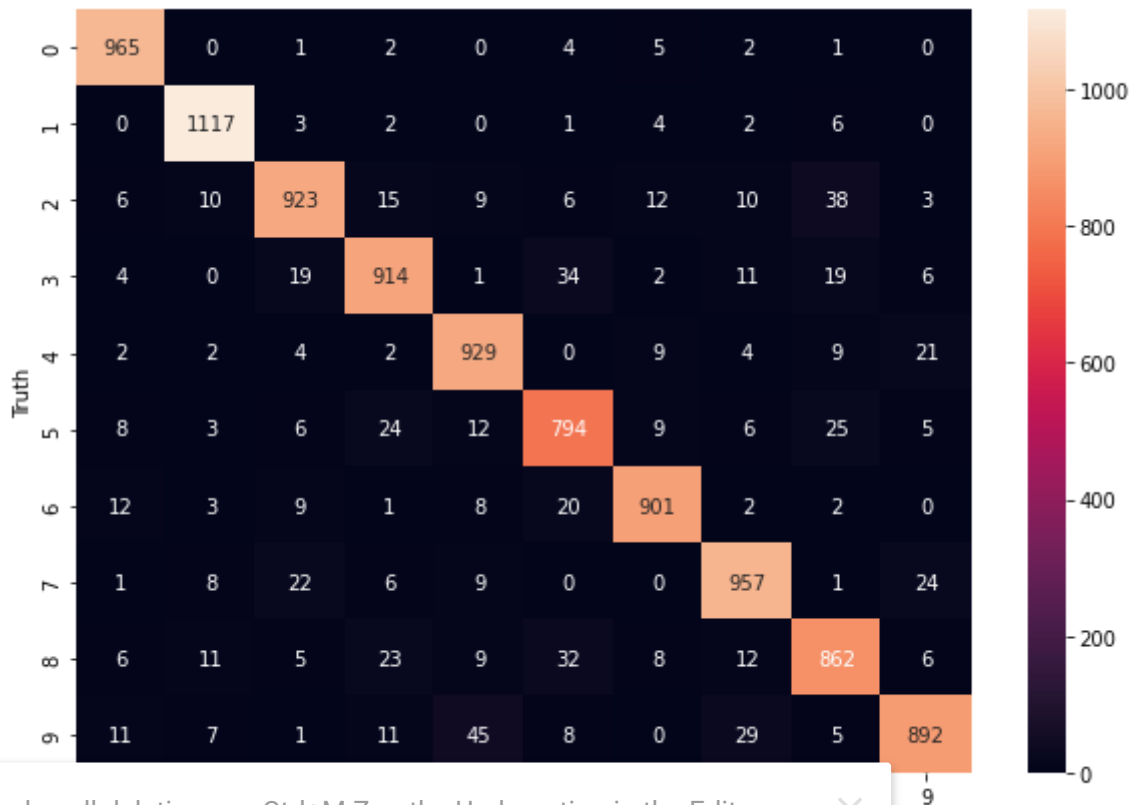
```

```

import seaborn as sn
plt.figure(figsize = (10,7))
sn.heatmap(cm, annot=True, fmt='d')
plt.xlabel('Predicted')
plt.ylabel('Truth')

```

Text(69.0, 0.5, 'Truth')



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## Using hidden layer

```

model = keras.Sequential([
    keras.layers.Dense(100, input_shape=(784,), activation='relu'),
    keras.layers.Dense(10, activation='sigmoid')
])

```

```

model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])

```

```

model.fit(X_train_flattened, y_train, epochs=5)

```

Epoch 1/5

1875/1875 [=====] - 3s 2ms/step - loss: 0.2726 - accuracy: 0.92

```
Epoch 2/5
1875/1875 [=====] - 3s 2ms/step - loss: 0.1230 - accuracy: 0.61
Epoch 3/5
1875/1875 [=====] - 3s 2ms/step - loss: 0.0852 - accuracy: 0.71
Epoch 4/5
1875/1875 [=====] - 3s 2ms/step - loss: 0.0660 - accuracy: 0.78
Epoch 5/5
1875/1875 [=====] - 3s 1ms/step - loss: 0.0521 - accuracy: 0.83
<keras.callbacks.History at 0x7efc8ed501d0>
```

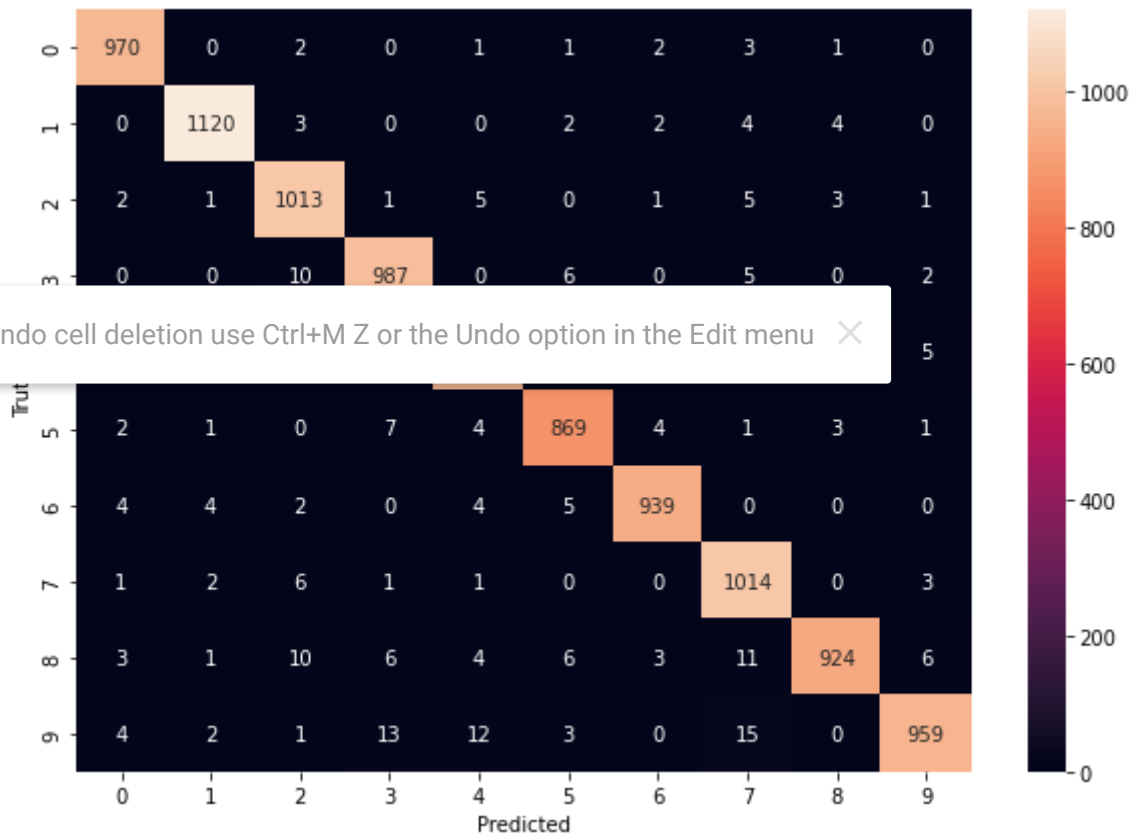
```
model.evaluate(X_test_flattened,y_test)
```

```
313/313 [=====] - 0s 1ms/step - loss: 0.0786 - accuracy: 0.95
[0.07863084971904755, 0.9763000011444092]
```

```
y_predicted = model.predict(X_test_flattened)
y_predicted_labels = [np.argmax(i) for i in y_predicted]
cm = tf.math.confusion_matrix(labels=y_test,predictions=y_predicted_labels)
```

```
plt.figure(figsize = (10,7))
sn.heatmap(cm, annot=True, fmt='d')
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

```
Text(69.0, 0.5, 'Truth')
```





Using Flatten layer so that we don't have to call .reshape on input dataset

```
model = keras.Sequential([
    keras.layers.Flatten(input_shape=(28, 28)),
    keras.layers.Dense(100, activation='relu'),
    keras.layers.Dense(10, activation='sigmoid')
])

model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])

model.fit(X_train, y_train, epochs=10)
```

```
Epoch 1/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.2693 - accuracy: 0.0000
Epoch 2/10
1875/1875 [=====] - 3s 1ms/step - loss: 0.1207 - accuracy: 0.0000
Epoch 3/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.0839 - accuracy: 0.0000
Epoch 4/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.0657 - accuracy: 0.0000
Epoch 5/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0490 - accuracy: 0.0000
Epoch 6/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.0408 - accuracy: 0.0000
Epoch 7/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.0332 - accuracy: 0.0000
Epoch 8/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.0281 - accuracy: 0.0000
Epoch 9/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.0225 - accuracy: 0.0000
Epoch 10/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.0196 - accuracy: 0.0000
<keras.callbacks.History at 0x7efc92f47c90>
```

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```
313/313 [=====] - 0s 1ms/step - loss: 0.0913 - accuracy: 0.9740
[0.0912703275680542, 0.9740999937057495]
```

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