

Number Recognition



Handwritten digit recognition system not only detects scanned images of handwritten digits. Handwritten digit recognition using MNIST dataset is a major project made

```
In [1]: import tensorflow as tf
        from tensorflow import keras
        import matplotlib.pyplot as plt
        %matplotlib inline
        import numpy as np
```

```
In [2]: (X_train, y_train) , (X_test, y_test) = keras.datasets.mnist.load_data()
```

```
In [3]: len(X_train)
```

```
Out[3]: 60000
```

```
In [4]: len(X_test)
```

```
Out[4]: 10000
```

```
In [5]: X_train[0].shape
```

```
Out[5]: (28, 28)
```

```
In [6]: X_train[0]
```



```
Out[6]: 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,  0,  0,  0,
  0,  0],
 [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
 18, 18, 18, 126, 136, 175, 26, 166, 255, 247, 127,  0,  0,
  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,  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,  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,  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,
```

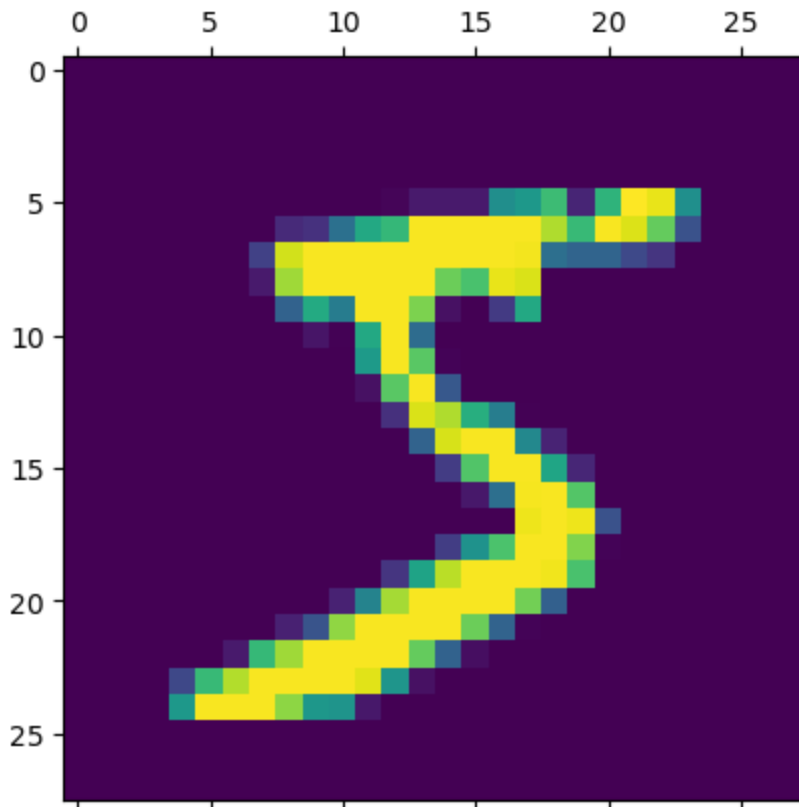
```

253, 253, 198, 81, 2, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 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, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0],
[ 0, 0, 0, 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)

```

In [7]: `plt.matshow(X_train[0])`

Out[7]: `<matplotlib.image.AxesImage at 0x2227af98f90>`



In [8]: `y_train[0]`

Out[8]: `5`

In [9]: `X_train = X_train / 255`
`X_test = X_test / 255`

In [10]: `X_train[0]`



0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.],
[0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0.54509804, 0.99215686, 0.74509804, 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. , 0.04313725, 0.74509804, 0.99215686, 0.2745098 ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.],
[0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.1372549 , 0.94509804, 0.88235294,
0.62745098, 0.42352941, 0.00392157, 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.],
[0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0.31764706, 0.94117647,
0.99215686, 0.99215686, 0.46666667, 0.09803922, 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.],
[0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0.17647059,
0.72941176, 0.99215686, 0.99215686, 0.58823529, 0.10588235,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.],
[0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0.0627451 , 0.36470588, 0.98823529, 0.99215686, 0.73333333,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.],
[0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0.97647059, 0.99215686, 0.97647059,
0.25098039, 0. , 0. , 0. , 0. ,
0. , 0. , 0.],
[0. , 0. , 0. , 0. , 0. ,
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. , 0.15294118, 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. ,
0.09411765, 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. , 0.09019608, 0.25882353,



```

0.83529412, 0.99215686, 0.99215686, 0.99215686, 0.99215686,
0.77647059, 0.31764706, 0.00784314, 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.          , 0.21568627,
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.          , 0.53333333,
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.          ,
0.          , 0.          , 0.          , 0.          , 0.          ,
0.          , 0.          , 0.          , 0.          , 0.          ,
0.          , 0.          , 0.          , 0.          , 0.          ,
0.          , 0.          , 0.          ],
[0.          , 0.          , 0.          , 0.          , 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]: X_train_flattened = X_train.reshape(len(X_train), 28*28)
X_test_flattened = X_test.reshape(len(X_test), 28*28)
```

```
In [12]: X_train_flattened.shape
```

```
Out[12]: (60000, 784)
```

```
In [13]: X_train_flattened[0]
```



0.99215686, 0.74509804, 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. , 0.04313725, 0.74509804, 0.99215686,
0.2745098 , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0.1372549 , 0.94509804, 0.88235294, 0.62745098,
0.42352941, 0.00392157, 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0.31764706, 0.94117647, 0.99215686, 0.99215686, 0.46666667,
0.09803922, 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0.17647059,
0.72941176, 0.99215686, 0.99215686, 0.58823529, 0.10588235,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0.0627451 , 0.36470588,
0.98823529, 0.99215686, 0.73333333, 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0.97647059, 0.99215686,
0.97647059, 0.25098039, 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 0. ,
0. , 0. , 0. , 0. , 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. , 0.15294118,
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. ,
0.09411765, 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. , 0.09019608, 0.25882353, 0.83529412, 0.99215686,
0.99215686, 0.99215686, 0.99215686, 0.77647059, 0.31764706,
0.00784314, 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. ,



[illegible]

```
In [14]: model = keras.Sequential([
            keras.layers.Dense(10, input_shape=(784,)), activation='sigmoid'
        ])

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

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

Epoch 1/10
1875/1875 [=====] - 7s 2ms/step - loss: 0.4678 - accuracy: 0.8777
Epoch 2/10
1875/1875 [=====] - 5s 3ms/step - loss: 0.3036 - accuracy: 0.9155
Epoch 3/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.2835 - accuracy: 0.9207
Epoch 4/10
1875/1875 [=====] - 4s 2ms/step - loss: 0.2734 - accuracy: 0.9237
Epoch 5/10
1875/1875 [=====] - 4s 2ms/step - loss: 0.2666 - accuracy: 0.9254
Epoch 6/10
1875/1875 [=====] - 5s 2ms/step - loss: 0.2618 - accuracy: 0.9275
Epoch 7/10
1875/1875 [=====] - 4s 2ms/step - loss: 0.2584 - accuracy: 0.9284
Epoch 8/10
1875/1875 [=====] - 4s 2ms/step - loss: 0.2552 - accuracy: 0.9295
Epoch 9/10
1875/1875 [=====] - 4s 2ms/step - loss: 0.2532 - accuracy: 0.9302
Epoch 10/10
1875/1875 [=====] - 5s 3ms/step - loss: 0.2509 - accuracy: 0.9302



Out[14]: <keras.src.callbacks.History at 0x2227af2add0>

In [15]: `model.evaluate(X_test_flattened, y_test)`

313/313 [=====] - 1s 2ms/step - loss: 0.2652 - accuracy: 0.9264

Out[15]: [0.26522335410118103, 0.9264000058174133]

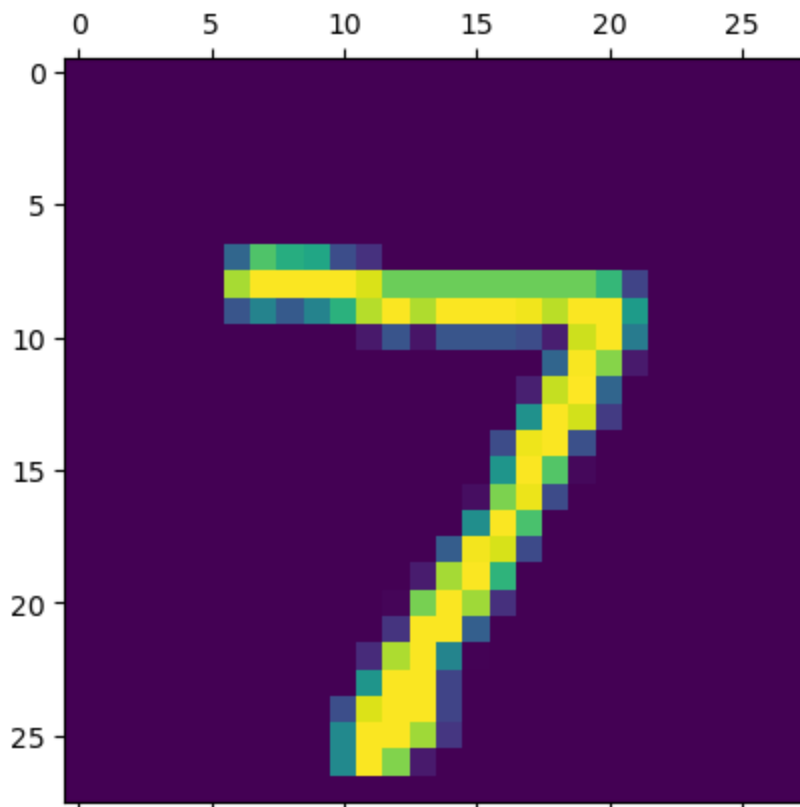
In [16]: `y_predicted = model.predict(X_test_flattened)`
`y_predicted[0]`

313/313 [=====] - 1s 2ms/step

Out[16]: array([3.6841894e-03, 1.7399653e-08, 2.3899361e-02, 9.7188163e-01,
1.4701609e-03, 1.1469388e-01, 2.0632944e-08, 9.9987030e-01,
9.5993996e-02, 7.0904833e-01], dtype=float32)

In [17]: `plt.matshow(X_test[0])`

Out[17]: <matplotlib.image.AxesImage at 0x2221c9e8cd0>



```
In [18]: np.argmax(y_predicted[0])
```

```
Out[18]: 7
```

```
In [19]: y_predicted_labels = [np.argmax(i) for i in y_predicted]
```

```
In [20]: y_predicted_labels[:5]
```

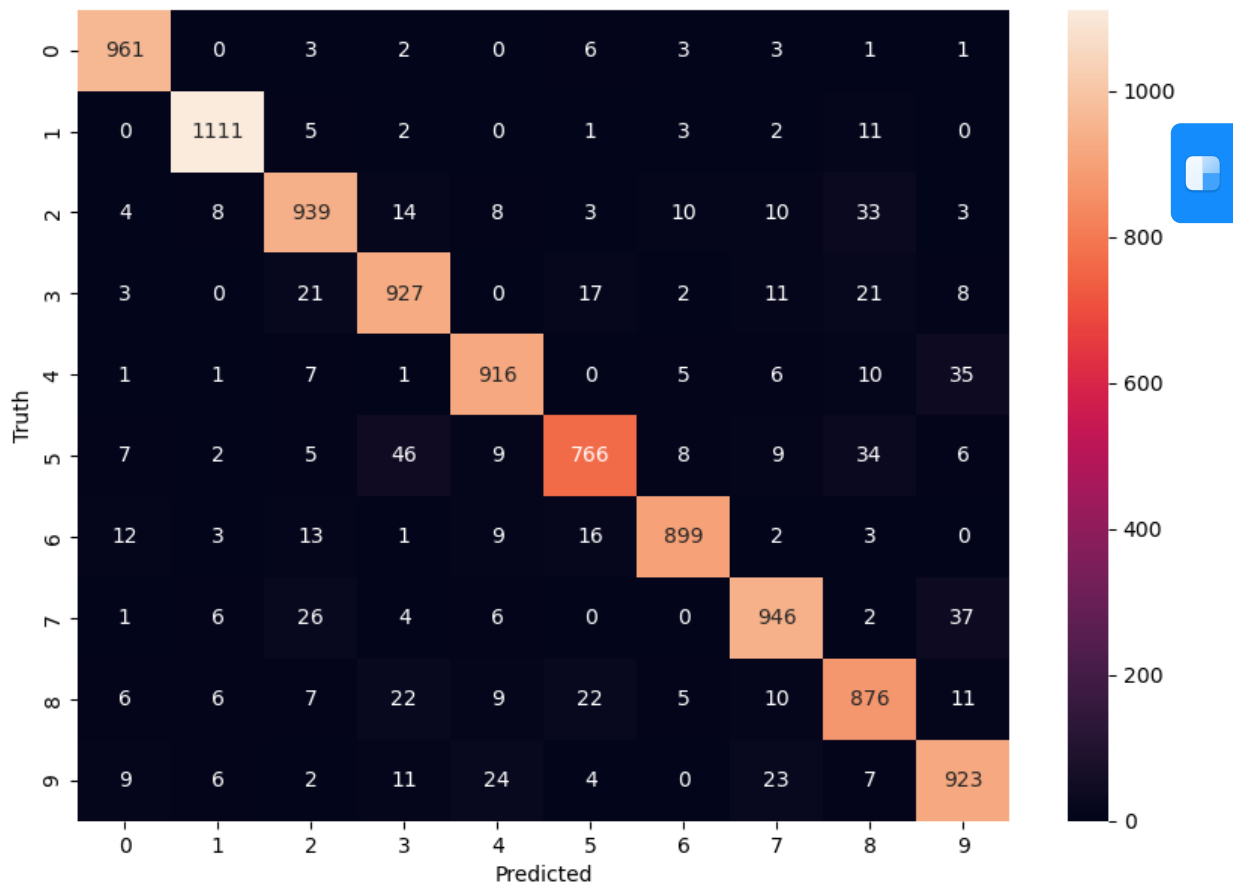
```
Out[20]: [7, 2, 1, 0, 4]
```

```
In [21]: cm = tf.math.confusion_matrix(labels=y_test, predictions=y_predicted_labels)
cm
```

```
Out[21]: <tf.Tensor: shape=(10, 10), dtype=int32, numpy=
array([[ 961,    0,    3,    2,    0,    6,    3,    3,    1,    1],
       [    0, 1111,    5,    2,    0,    1,    3,    2,   11,    0],
       [    4,    8,  939,   14,    8,    3,   10,   10,   33,    3],
       [    3,    0,   21,  927,    0,   17,    2,   11,   21,    8],
       [    1,    1,    7,    1,  916,    0,    5,    6,   10,   35],
       [    7,    2,    5,   46,    9,  766,    8,    9,   34,    6],
       [   12,    3,   13,    1,    9,   16,  899,    2,    3,    0],
       [    1,    6,   26,    4,    6,    0,    0,  946,    2,   37],
       [    6,    6,    7,   22,    9,   22,    5,   10,  876,   11],
       [    9,    6,    2,   11,   24,    4,    0,   23,    7,  923]])>
```

```
In [22]: import seaborn as sn
plt.figure(figsize = (10,7))
sn.heatmap(cm, annot=True, fmt='d')
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

```
Out[22]: Text(95.72222222222221, 0.5, 'Truth')
```



Using hidden layer

```
In [23]: 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=10)
```

Epoch 1/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.2688 - accuracy: 0.9241
Epoch 2/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.1212 - accuracy: 0.9640
Epoch 3/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0844 - accuracy: 0.9747
Epoch 4/10
1875/1875 [=====] - 5s 3ms/step - loss: 0.0655 - accuracy: 0.9798
Epoch 5/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0506 - accuracy: 0.9844
Epoch 6/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0409 - accuracy: 0.9875
Epoch 7/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0335 - accuracy: 0.9897
Epoch 8/10
1875/1875 [=====] - 7s 4ms/step - loss: 0.0275 - accuracy: 0.9917
Epoch 9/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0231 - accuracy: 0.9926
Epoch 10/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0193 - accuracy: 0.9942



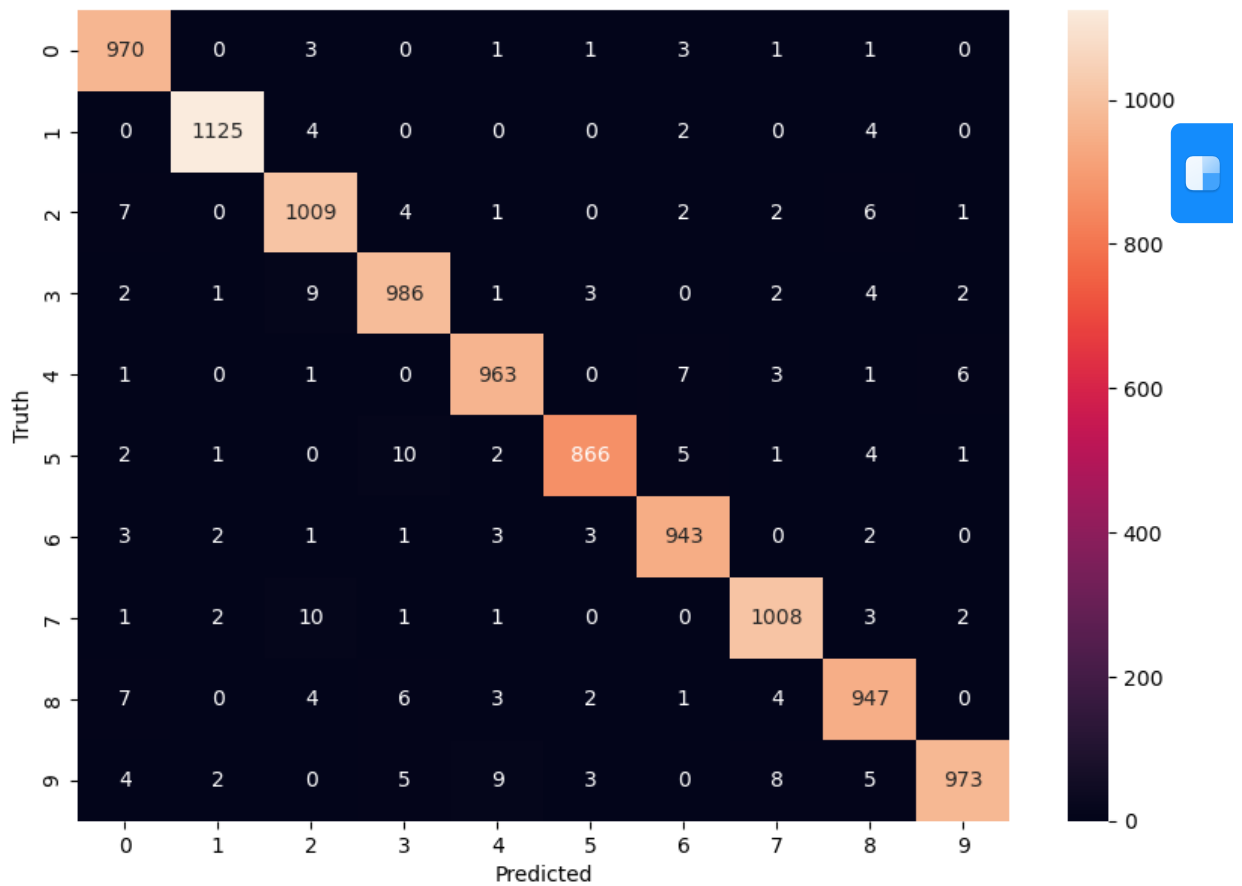
Out[23]: <keras.src.callbacks.History at 0x2221ed56ad0>

```
In [24]: 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')
```

313/313 [=====] - 1s 2ms/step
Text(95.7222222222221, 0.5, 'Truth')

Out[24]:



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

```
In [25]: 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 [=====] - 6s 3ms/step - loss: 0.2721 - accuracy:
0.9230
Epoch 2/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.1216 - accuracy:
0.9643
Epoch 3/10
1875/1875 [=====] - 7s 4ms/step - loss: 0.0849 - accuracy:
0.9750
Epoch 4/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0645 - accuracy:
0.9805
Epoch 5/10
1875/1875 [=====] - 7s 4ms/step - loss: 0.0507 - accuracy:
0.9844
Epoch 6/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0424 - accuracy:
0.9865
Epoch 7/10
1875/1875 [=====] - 7s 3ms/step - loss: 0.0327 - accuracy:
0.9899
Epoch 8/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0270 - accuracy:
0.9916
Epoch 9/10
1875/1875 [=====] - 6s 3ms/step - loss: 0.0231 - accuracy:
0.9928
Epoch 10/10
1875/1875 [=====] - 7s 4ms/step - loss: 0.0189 - accuracy:
0.9943



Out[25]: <keras.src.callbacks.History at 0x2221f4384d0>

In [26]: `model.evaluate(X_test,y_test)`

313/313 [=====] - 1s 3ms/step - loss: 0.0865 - accuracy: 0.
9767

Out[26]: [0.08650301396846771, 0.9767000079154968]