

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,  3,
        18, 18, 18, 126, 136, 175, 26, 1
 66, 255, 247, 127,  0,  0,
        0,  0],
      [ 0,  0,  0,  0,  0,  0,  0,
 0, 30, 36, 94, 154, 170,
        253, 253, 253, 253, 253, 225, 172, 2
 53, 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,
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, 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, 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, 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,
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, 0, 18, 1
71, 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, 2
53, 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, 2
53, 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, 0]], dtype=uint8)

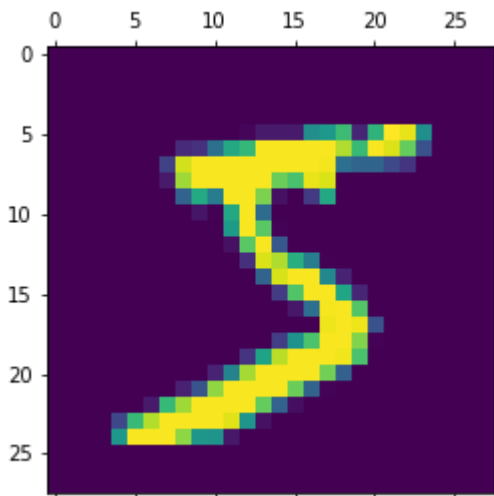
```

In [7]:

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

Out[7]:

<matplotlib.image.AxesImage at 0x21d738dc508
>



In [8]:

```
y_train[2]
```

Out[8]:

4

In [9]:

```
y_train[:5]
```

Out[9]:

```
array([5, 0, 4, 1, 9], dtype=uint8)
```

In [10]:

```
X_train.shape
```

Out[10]:

```
(60000, 28, 28)
```

In [11]:

```
X_train = X_train / 255
```

In [12]:

```
X_train_flattened = X_train.reshape(len(X_train), 28 * 28)  
X_train_flattened.shape
```

Out[12]:

```
(60000, 784)
```

In [13]:

```
X_test.shape
```

Out[13]:

```
(10000, 28, 28)
```

In [14]:

```
X_test = X_test / 255
```


In [15]:

```
X_train[0]
```

Out[15]:

[illegible]

```

0.      , 0.      ,
      0.      , 0.      , 0.      ,
0.      , 0.      ,
      0.      , 0.      , 0.      ],
[0.      , 0.      , 0.      ,
0.      , 0.      ,
      0.      , 0.      ,
0.      , 0.      ,
      0.      , 0.      ,
0.      , 0.      ,
      0.      , 0.      ,
0.      , 0.      ,
      0.      , 0.      , 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.    , 0.05490196,
    0.00392157, 0.60392157, 0.99215686,
0.35294118, 0.    ,
    0.    , 0.    , 0.    ,
0.    , 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.      ,
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.      ,
0.31764706, 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.      ,
      0.0627451 , 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.      , 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.      , 0.18039216,
      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.      , 0.15294118,
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.      ,
0.09019608, 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.      , 0.07058824, 0.67058824,
0.85882353, 0.99215686,
0.99215686, 0.99215686, 0.99215686,
0.76470588, 0.31372549,
0.03529412, 0.      , 0.      ,
0.      , 0.      ,

```


In [16]:

```
X_test_flattened = X_test.reshape(len(X_test), 28 * 28)  
X_test_flattened.shape
```

Out[16]:

(10000, 784)

In [17]:

```
X_train_flattened[0]
```

Out[17]:

[illegible]

```

0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 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.      ,
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.      , 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. ,
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. ,
0.31372549, 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. ,
0. , 0.05490196,
0.00392157, 0.60392157, 0.99215686,
0.35294118, 0. ,
0. , 0. , 0. ,
0. , 0. ,
0. , 0. , 0. ,
0. , 0. ,
0. , 0. , 0. ,
0. , 0. , 0. ,
0. , 0. ,

localhost:8888/nbconvert/html/Handwritten Digit Recognition using Tensorflow and keras.ipyn... 23/39

```

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.      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.      ,
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      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. , 0.15294118,
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. ,
0.09411765, 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. , 0.09019608, 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. , 0.07058824,
0.67058824, 0.85882353,


```
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 0.      ,
0.      0.      , 0.      , 0.      ,
0.      , 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 [18]:

```
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)
```

Train on 60000 samples

Epoch 1/10

60000/60000 [=====]

- 5s 79us/sample - loss: 0.4883 - accuracy:
0.8758

Epoch 2/10

60000/60000 [=====]

- 4s 71us/sample - loss: 0.3059 - accuracy:
0.9147

Epoch 3/10

60000/60000 [=====]

- 4s 71us/sample - loss: 0.2854 - accuracy:
0.9213

Epoch 4/10

60000/60000 [=====]

- 4s 72us/sample - loss: 0.2742 - accuracy:
0.9238

Epoch 5/10

60000/60000 [=====]

- 4s 71us/sample - loss: 0.2679 - accuracy:
0.9262

Epoch 6/10

60000/60000 [=====]

- 4s 71us/sample - loss: 0.2626 - accuracy:
0.9270

Epoch 7/10

60000/60000 [=====]

- 4s 71us/sample - loss: 0.2588 - accuracy:
0.9286

Epoch 8/10

60000/60000 [=====]

- 4s 71us/sample - loss: 0.2551 - accuracy:
0.9300

Epoch 9/10

60000/60000 [=====]

- 4s 71us/sample - loss: 0.2529 - accuracy:
0.9306

Epoch 10/10

60000/60000 [=====]

- 4s 73us/sample - loss: 0.2501 - accuracy:
0.9317

Out[18]:

```
<tensorflow.python.keras.callbacks.History at 0x21d73379f88>
```

In [19]:

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

```
10000/10000 [=====]  
- 1s 60us/sample - loss: 0.2622 - accuracy:  
0.9286
```

Out[19]:

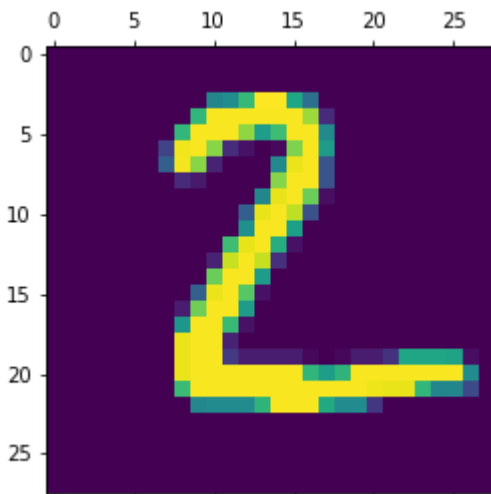
```
[0.2622289415150881, 0.9286]
```

In [20]:

```
plt.matshow(X_test[1])
```

Out[20]:

```
<matplotlib.image.AxesImage at 0x21d7a010788  
>
```



In [21]:

```
y_predicted = model.predict(X_test_flattened)
y_predicted[1]
```

Out[21]:

```
array([1.7646527e-04, 7.3083322e-07, 8.26081
16e-01, 6.3831882e-05,
        1.6834984e-15, 1.9083178e-03, 2.06563
53e-03, 8.8439915e-20,
        2.4465753e-05, 2.1175482e-16], dtype=
float32)
```

In [22]:

```
np.argmax(y_predicted[1])
```

Out[22]:

2

In [23]:

```
y_predicted_labels = [np.argmax(i) for i in y_predicted]
y_predicted_labels[:5]
```

Out[23]:

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

In [24]:

```
y_test[:5]
```

Out[24]:

```
array([7, 2, 1, 0, 4], dtype=uint8)
```

In [25]:

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

Out[25]:

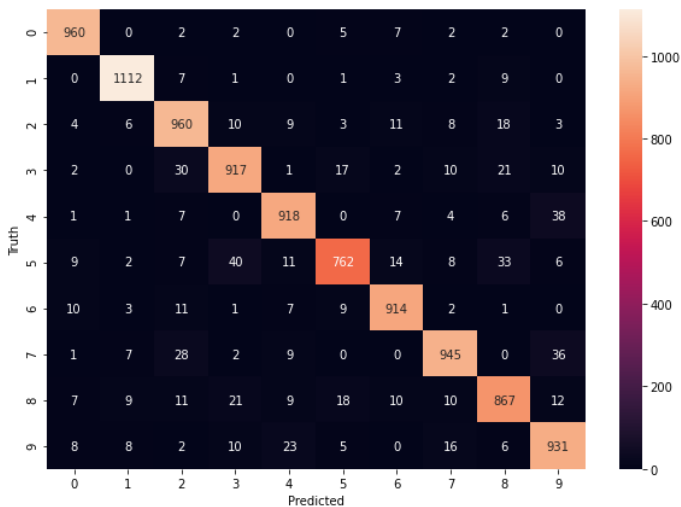
```
<tf.Tensor: shape=(10, 10), dtype=int32, num
py=
array([[ 960,    0,    2,    2,    0,    5,
 7,    2,    2,    0],
 [    0, 1112,    7,    1,    0,    1,
 3,    2,    9,    0],
 [    4,    6, 960,   10,    9,    3,
11,    8,   18,    3],
 [    2,    0,   30, 917,    1,   17,
 2,   10,   21,   10],
 [    1,    1,    7,    0, 918,    0,
 7,    4,    6,   38],
 [    9,    2,    7,   40,   11, 762,
14,    8,   33,    6],
 [   10,    3,   11,    1,    7,    9,
914,    2,    1,    0],
 [    1,    7,   28,    2,    9,    0,
 0, 945,    0,   36],
 [    7,    9,   11,   21,    9,   18,
10,   10, 867,   12],
 [    8,    8,    2,   10,   23,    5,
 0,   16,    6, 931]])>
```


In [26]:

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

Out[26]:

Text(69.0, 0.5, 'Truth')



In [27]:

```
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)
```

Train on 60000 samples

Epoch 1/5

60000/60000 [=====]

- 6s 97us/sample - loss: 0.2928 - accuracy: 0.9178

Epoch 2/5

60000/60000 [=====]

- 5s 89us/sample - loss: 0.1388 - accuracy: 0.9587

Epoch 3/5

60000/60000 [=====]

- 5s 90us/sample - loss: 0.0983 - accuracy: 0.9704

Epoch 4/5

60000/60000 [=====]

- 5s 91us/sample - loss: 0.0765 - accuracy: 0.9770

Epoch 5/5

60000/60000 [=====]

- 5s 89us/sample - loss: 0.0619 - accuracy: 0.9813

Out[27]:

<tensorflow.python.keras.callbacks.History at 0x21d7bd3dc48>

In [28]:

```
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)
```

Train on 60000 samples

Epoch 1/10

60000/60000 [=====]

- 6s 97us/sample - loss: 0.2891 - accuracy:
0.9204

Epoch 2/10

60000/60000 [=====]

- 5s 88us/sample - loss: 0.1307 - accuracy:
0.9621

Epoch 3/10

60000/60000 [=====]

- 5s 87us/sample - loss: 0.0929 - accuracy:
0.9727

Epoch 4/10

60000/60000 [=====]

- 5s 87us/sample - loss: 0.0728 - accuracy:
0.9782

Epoch 5/10

60000/60000 [=====]

- 5s 90us/sample - loss: 0.0599 - accuracy:
0.9817

Epoch 6/10

60000/60000 [=====]

- 5s 88us/sample - loss: 0.0492 - accuracy:
0.9855

Epoch 7/10

60000/60000 [=====]

- 5s 89us/sample - loss: 0.0407 - accuracy:
0.9880

Epoch 8/10

60000/60000 [=====]

- 5s 89us/sample - loss: 0.0344 - accuracy:
0.9896

Epoch 9/10

60000/60000 [=====]

- 5s 89us/sample - loss: 0.0290 - accuracy:
0.9915

Epoch 10/10

60000/60000 [=====]

- 5s 89us/sample - loss: 0.0252 - accuracy:
0.9923

Out[28]:

```
<tensorflow.python.keras.callbacks.History a  
t 0x21d7e3ca608>
```

In [29]:

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

```
10000/10000 [=====]  
- 1s 63us/sample - loss: 0.0858 - accuracy:  
0.9756
```

Out[29]:

```
[0.08582128985947929, 0.9756]
```

In [30]:

```

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, prediction
s = y_predicted_labels)
cm

```

Out[30]:

```

<tf.Tensor: shape=(10, 10), dtype=int32, num
py=
array([[ 973,    1,    0,    0,    0,    2,
 1,    1,    1,    1],
 [    0, 1122,    4,    0,    0,    0,
 2,    1,    6,    0],
 [    2,    0, 1013,    0,    4,    0,
 2,    7,    4,    0],
 [    1,    1,    9, 950,    0,   38,
 0,    2,    7,    2],
 [    1,    0,    1,    0, 965,    0,
 3,    1,    3,    8],
 [    3,    0,    0,    4,    0, 879,
 3,    1,    2,    0],
 [    8,    2,    1,    0,    1,    6,
 938,    0,    2,    0],
 [    2,    4,    7,    1,    1,    0,
 0, 1001,    4,    8],
 [    7,    0,    6,    0,    2,   18,
 1,    4, 934,    2],
 [    2,    2,    1,    3,    7,    7,
 1,    2,    3, 981]])>

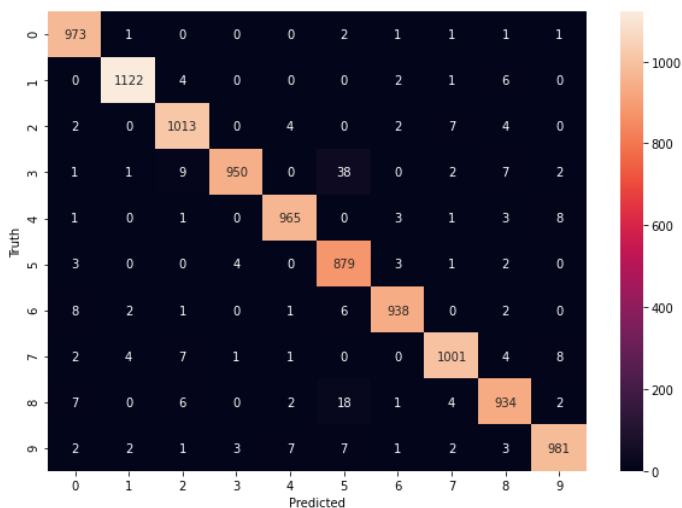
```

In [31]:

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

Out[31]:

Text(69.0, 0.5, 'Truth')



In []: