

```
from tensorflow import keras
import numpy as np
import matplotlib.pyplot as plt
from keras.layers import Dense, Flatten
from keras.models import Sequential
from keras.utils import to_categorical
from keras.datasets import mnist
import tensorflow as tf
```

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

(60000, 28, 28)

(10000, 28, 28)

60000

(60000,)

(10000,)

```
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, 3,
18, 18, 18, 126, 136, 175, 26, 166, 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, 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, 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,
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, 23, 66, 213, 253, 253,
253, 253, 198, 81, 2, 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, 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, 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]], dtype=uint8)

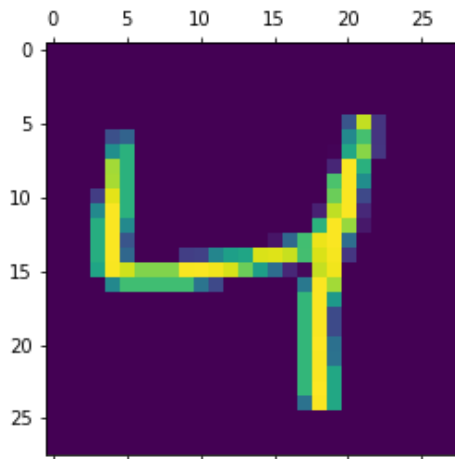
```

In [10]:

```
plt.matshow(X_train[2])
```

Out[10]:

<matplotlib.image.AxesImage at 0x7f63ef8a7a20>



In [11]:

```
y_train[2]
```

Out[11]:

4

In [12]:

```
y_train[:5]
```

Out[12]:

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

In [13]:

```
X_train=X_train/255
X_test=X_test/255
```

In [14]:

```
X_train[0]
```

Out[14]:

```
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. , 0.01176471, 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,
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. , 0.04313725, 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. ,
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.	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.	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.	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.	0.15294118	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.
0.09411765	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.	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.	0.21568627
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.	0.53333333
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.	0.
0.	0.	0.	0.	0.
[0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
[0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
[0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
[0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
[0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
0.	0.	0.	0.	0.
[0.	0.	0.	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 [15]:

```
X_train_flattend=X_train.reshape(len(X_train),28*28)
X_test_flattend=X_test.reshape(len(X_test),28*28)
```

In [16]:

```
X_train_flattend.shape
```

Out[16]:

```
(60000, 784)
```

In [17]:

```
X_test_flattend.shape
```

Out[17]:

```
(10000, 784)
```

X_train_flattend[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']
)
```

In [19]:

```
model.fit(X_train_flattend,y_train,epochs=5)
```

```
Epoch 1/5
1875/1875 [=====] - 9s 5ms/step - loss: 0.4858 - accuracy: 0.879
1
Epoch 2/5
1875/1875 [=====] - 9s 5ms/step - loss: 0.3065 - accuracy: 0.915
3
Epoch 3/5
1875/1875 [=====] - 9s 5ms/step - loss: 0.2853 - accuracy: 0.921
0
Epoch 4/5
1875/1875 [=====] - 9s 5ms/step - loss: 0.2747 - accuracy: 0.924
3
Epoch 5/5
1875/1875 [=====] - 9s 5ms/step - loss: 0.2674 - accuracy: 0.926
0
```

Out[19]:

```
<tensorflow.python.keras.callbacks.History at 0x7f63f005ff28>
```

In [20]:

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

```
313/313 [=====] - 1s 4ms/step - loss: 0.2657 - accuracy: 0.9261
```

Out[20]:

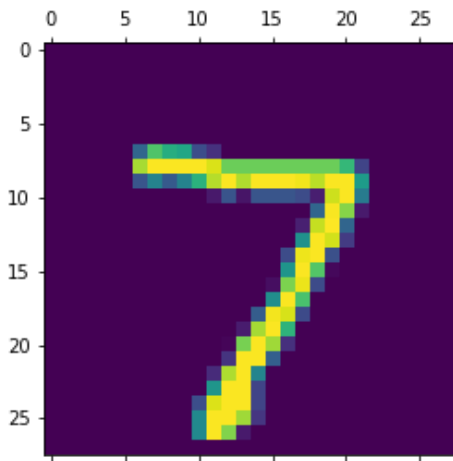
```
[0.26566585898399353, 0.9261000156402588]
```

In [21]:

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

Out[21]:

<matplotlib.image.AxesImage at 0x7f63eff2ac50>



In [22]:

```
y_predicted=model.predict(X_test_flattend)
```

In [23]:

```
y_predicted[0]
```

Out[23]:

```
array([1.2869386e-05, 9.6633576e-11, 3.9769660e-05, 8.1052184e-03,  
       1.2041669e-06, 8.2391372e-05, 7.4705303e-10, 7.8519523e-01,  
       8.3045808e-05, 6.4727664e-04], dtype=float32)
```

In [24]:

```
np.argmax(y_predicted[0])
```

Out[24]:

7

In [25]:

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

Out[25]:

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

In [26]:

```
y_test[:5]
```

Out[26]:

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

In [27]:

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

Out[27]:

```
<tf.Tensor: shape=(10, 10), dtype=int32, numpy=  
array([[ 962,    0,    0,    2,    0,    5,    6,    3,    2,    0],  
       [   0, 1112,    3,    2,    0,    1,    4,    2,   11,    0],  
       [   5,   11,  911,   19,   11,    6,   12,   13,   42,    2],  
       [   3,    0,   16,  916,    1,   24,    1,   12,   31,    6],  
       [   1,    1,    2,    1,  920,    0,    0,    4,   11,   25],  
       [   0,    0,    0,    0,    0,  900,    0,    0,    0,    0],  
       [   0,    0,    0,    0,    0,    0,  900,    0,    0,    0],  
       [   0,    0,    0,    0,    0,    0,    0,  900,    0,    0],  
       [   0,    0,    0,    0,    0,    0,    0,    0,  900,    0],  
       [   0,    0,    0,    0,    0,    0,    0,    0,    0,  900]])
```

```
[ 1, 1, 2, 1, 328, 0, 9, 4, 11, 25],  
[ 7, 3, 2, 29, 11, 783, 12, 7, 32, 6],  
[ 9, 3, 7, 0, 8, 15, 910, 2, 4, 0],  
[ 1, 7, 23, 4, 10, 0, 0, 957, 1, 25],  
[ 5, 9, 5, 15, 9, 21, 9, 13, 884, 4],  
[ 10, 7, 1, 9, 34, 6, 0, 33, 11, 898]],  
dtype=int32)>
```

In []:

In []: