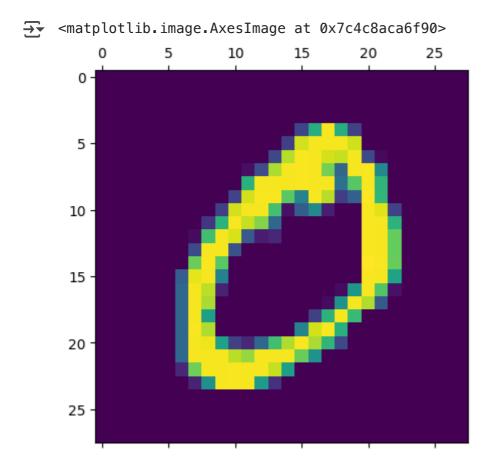
```
import tensorflow as tf
from tensorflow import keras
import matplotlib.pyplot as plt
import numpy as np
(x_train, y_train), (x_test, y_test) = keras.datasets.mnist.load_data()
Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-data:">https://storage.googleapis.com/tensorflow/tf-keras-data:</a>
     11490434/11490434 -
                                                – 0s 0us/step
len(x_train)
→ 60000
len(x_test)
→ 10000
x_train.shape
→ (60000, 28, 28)
x_train[0]
ndarray (28, 28) show data
plt.matshow(x_train[0])
```

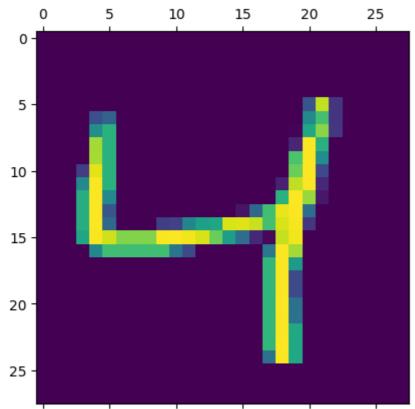
plt.matshow(x_train[1])

 $\overline{2}$



plt.matshow(x_train[2])





y_train[2]

y_train[:5]

$$\Rightarrow$$
 array([5, 0, 4, 1, 9], dtype=uint8)

x_train.shape

scalling the data

$$x_{train} = x_{train} / 255$$

 $x_{test} = x_{test} / 255$

x_train_flattened.shape

x_test_flattened.shape

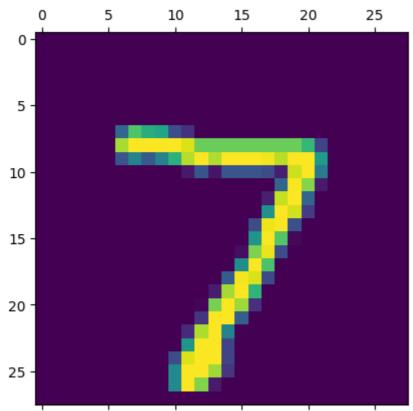
```
/usr/local/lib/python3.11/dist-packages/keras/src/layers/core/dense.py:93: Use
     super().__init__(activity_regularizer=activity_regularizer, **kwargs)
   Epoch 1/5
                          4s 2ms/step - accuracy: 0.8158 - loss: 0.7228
   1875/1875 -
   Epoch 2/5
                     4s 2ms/step - accuracy: 0.9136 - loss: 0.3107
   1875/1875 —
   Epoch 3/5
                       4s 2ms/step - accuracy: 0.9187 - loss: 0.2884
   1875/1875 -
   Epoch 4/5
                          5s 2ms/step - accuracy: 0.9227 - loss: 0.2747
   1875/1875 -
   Epoch 5/5
                         5s 2ms/step - accuracy: 0.9254 - loss: 0.2654
   1875/1875 -
   <keras.src.callbacks.history.History at 0x7c4c9093b950>
```

model.evaluate(x_test_flattened, y_test)

```
313/313 — 1s 2ms/step - accuracy: 0.9126 - loss: 0.3056 [0.2687021791934967, 0.925000011920929]
```

plt.matshow(x_test[0])





y_predicted = model.predict(x_test_flattened)
y_predicted[0]

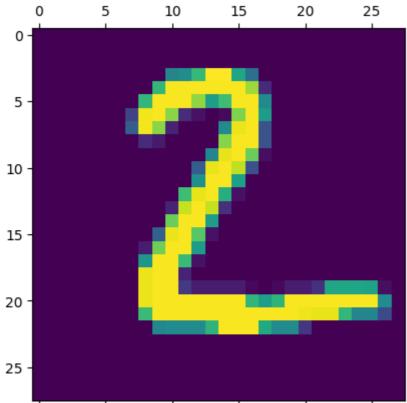
```
313/313 — Os 1ms/step
array([3.7691776e-02, 3.6071768e-07, 6.0681131e-02, 9.6168643e-01,
1.6371487e-03, 8.0357373e-02, 2.2784391e-06, 9.9973071e-01,
8.6109348e-02, 5.7833678e-01], dtype=float32)
```

np.argmax(y_predicted[0])

→ np.int64(7)

plt.matshow(x_test[1])

<matplotlib.image.AxesImage at 0x7c4c8569d610>



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

np.argmax(y_predicted[1])

→ np.int64(2)

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

 \rightarrow [np.int64(7), np.int64(2), np.int64(1), np.int64(0), np.int64(4)]

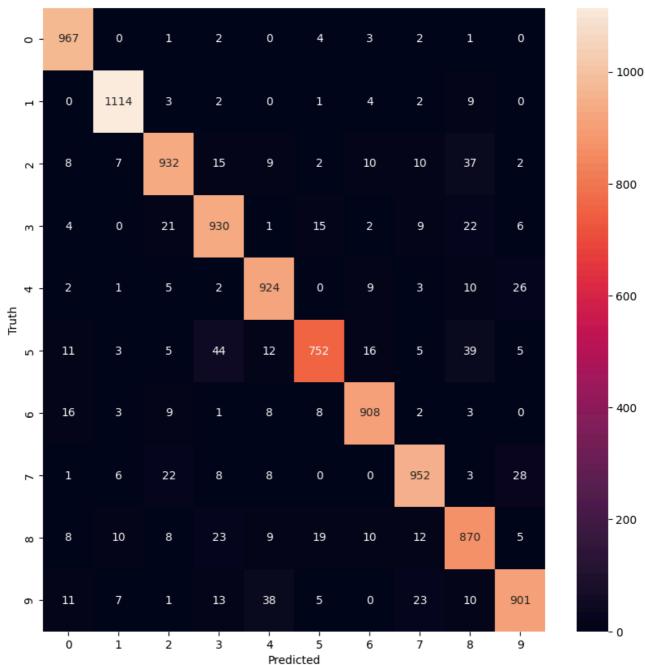
cm = tf.math.confusion_matrix(labels = y_test, predictions = y_predicted_lbels)
cm

```
<tf.Tensor: shape=(10, 10), dtype=int32, numpy=
\rightarrow
                                                                    3,
                                                                                            0],
     array([[ 967,
                            0,
                                    1,
                                            2,
                                                    0,
                                                            4,
                                                                            2,
                                                                                    1,
                                    3,
                                            2,
                                                                                            0],
                        1114,
                                                    0,
                                                                            2,
                                                                                    9,
                    8,
                                 932,
                                           15,
                                                    9,
                                                                                   37,
                            7,
                                                            2,
                                                                   10,
                                                                           10,
                                                                                            2],
                            0,
                                                           15,
                                  21,
                                         930,
                                                    1,
                                                                    2,
                                                                            9,
                                                                                   22,
                                                                                            6],
                            1,
                                                 924,
                                                                                           26],
                                    5,
                                            2,
                                                            0,
                                                                    9,
                                                                            3,
                                                                                   10,
                  11,
                            3,
                                    5,
                                           44,
                                                         752,
                                                                  16,
                                                                            5,
                                                                                   39,
                                                                                            5],
                                                  12,
                                    9,
                                                            8,
                                                                 908,
                                                                                            0],
                  16,
                            3,
                                            1,
                                                    8,
                                                                            2,
                                                                                    3,
                                                    8,
                                                                         952,
                    1,
                            6,
                                  22,
                                            8,
                                                                                    3,
                                                                                           28],
```

5], 10, 23, 19, 8, 8, 9, 10, 12, 870, 11, 23, 10, 901]], [7, 1, 13, 38, 5, 0, dtype=int32)>

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

Text(95.72222222221, 0.5, 'Truth')



adding a hidden layer

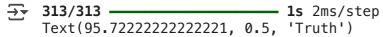
```
model = keras.Sequential([
         keras.layers.Dense(100, input_shape=(784,), activation='relu'),
         keras.layers.Dense(10, activation='sigmoid')
])
model.compile(optimizer='adam',
```

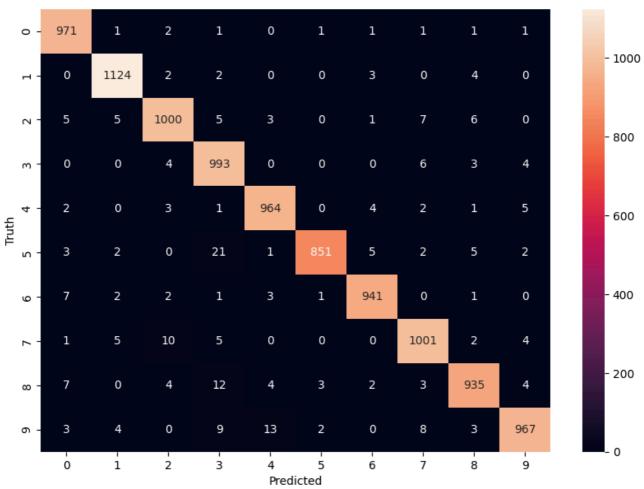
```
/usr/local/lib/python3.11/dist-packages/keras/src/layers/core/dense.py:93: Use
     super().__init__(activity_regularizer=activity_regularizer, **kwargs)
   Epoch 1/5
                      6s 3ms/step - accuracy: 0.8716 - loss: 0.4453
   1875/1875 —
   Epoch 2/5
                              — 7s 3ms/step - accuracy: 0.9614 - loss: 0.1309
   1875/1875 -
   Epoch 3/5
   1875/1875 -
                        9s 3ms/step - accuracy: 0.9729 - loss: 0.0894
   Epoch 4/5
                       6s 3ms/step - accuracy: 0.9798 - loss: 0.0671
   1875/1875 —
   Epoch 5/5
                    5s 3ms/step - accuracy: 0.9851 - loss: 0.0483
   1875/1875 ———
   <keras.src.callbacks.history.History at 0x7c4c8aca6150>
```

model.evaluate(x_test_flattened, y_test)

```
313/313 — 2s 4ms/step - accuracy: 0.9699 - loss: 0.0967 [0.08165591210126877, 0.9746999740600586]
```

```
y_predicted = model.predict(x_test_flattened)
y_predicted_lbels = [np.argmax(i) for i in y_predicted]
cm = tf.math.confusion_matrix(labels = y_test, predictions = y_predicted_lbels)
plt.figure(figsize = (10,7))
sn.heatmap(cm, annot=True, fmt='d')
plt.xlabel('Predicted')
plt.ylabel('Truth')
```





to avoid flattern arraay

```
model = keras.Sequential([
    keras.layers.Flatten(input_shape=(28, 28)),
    keras.layers.Dense(100, activation='relu'),#hidden layer
    keras.layers.Dense(10, activation='sigmoid')
])
model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])
model.fit(x_train, y_train, epochs=5)
\rightarrow
    /usr/local/lib/python3.11/dist-packages/keras/src/layers/reshaping/flatten.py
      super().__init__(**kwargs)
    Epoch 1/5
    1875/1875
                                  - 7s 3ms/step - accuracy: 0.8756 - loss: 0.4463
    Epoch 2/5
                                   - 5s 3ms/step - accuracy: 0.9641 - loss: 0.1265
    1875/1875
    Epoch 3/5
```

1875/1875 — 6s 3ms/step - accuracy: 0.9761 - loss: 0.0828
Epoch 4/5
1875/1875 — 5s 3ms/step - accuracy: 0.9816 - loss: 0.0608
Epoch 5/5
1875/1875 — 12s 4ms/step - accuracy: 0.9868 - loss: 0.0457
<keras.src.callbacks.history.History at 0x7c4c6546ef50>

Start coding or generate with AI.

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