```
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
(X_train, y_train), (X_test, y_test) = tf.keras.datasets.mnist.load_data()
Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist">https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist</a>.
      11490434/11490434 -
                                                  - 1s 0us/step
len(X_train)
→ 60000
len(X_test)
→ 10000
X_train[0].shape
→• (28, 28)
X_train[0]
```

ndarray (28, 28) hide data

array([[0, 0], 0], 0], 0], 0], 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 3, 18, 18, 126, 136, 175, 26, 166, 255, 247, 127, 18, 0, 0], 0, 0, 0, 0, 0, 0, 30, 36, 94, 154, 170, [0, 253, 253, 253, 253, 253, 225, 172, 253, 242, 195, 64, 0, 0], 0, 0, 0, 0, 0, 0, 49, 238, 253, 253, 253, 253, [0, 253, 253, 253, 253, 251, 93, 82, 82, 56, 39, 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, 80, 156, 107, 253, 253, [0, 205, 0. 43, 154, 0, 0, 0, 0, 0, 0, 0, 0, 11, 0], 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 14, 1, 154, 253, 0, 0, 0, 0, 0, 0, 0, 90, 0, 0, 0, 0, 0, 0], 0, 139, 253, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 190, 2, 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, 35, 0, 0, 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, 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, 45, 186, 253, 253, 150, 0, 27, 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, 0, 64, 0, 0, 0, 0, 0, 0, 0, 0], 0, 0, 0, 0,

0,

0,

0,

0,

0,

0,

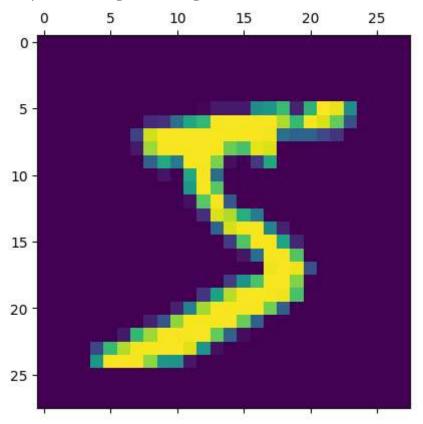
0,

0,

0,

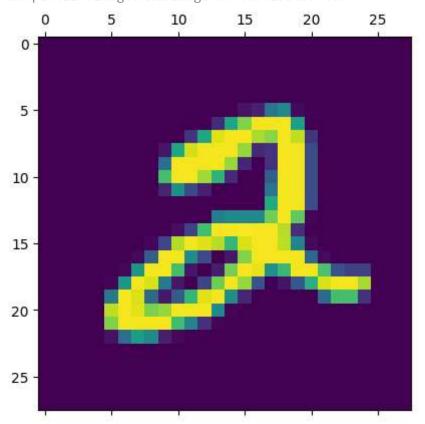
0, 46, 130, 183, 253, 253, 207, 2, 0, 0, 0, 0, 0, 0, 0], 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 39, [0, 148, 229, 253, 253, 253, 250, 182, 0, 0, 0, 0, 0, 0, 0, 0], 0, 0, 0, [0, 0, 0, 0, 0, 0, 0, 24, 114, 221, 0, 0, 253, 253, 253, 253, 201, 78, 0, 0, 0, 0, 0, 0], 0, [0, 0, 0, 0, 0, 0, 0, 0, 23, 66, 213, 253, 253, 253, 253, 198, 81, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0], 0, 0, 18, 171, 219, 253, 253, 253, 253, [0, 0, 0, 0, 195, 80, 9, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0], 0, 0, 0, 0, 55, 172, 226, 253, 253, 253, 253, 244, 133, [0, 0, 11, 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]], dtype=uint8)

<matplotlib.image.AxesImage at 0x78153d2d09a0>



plt.matshow(X_train[5])





y_train[2]

→ 4

y_train[:5]

⇒ array([5, 0, 4, 1, 9], dtype=uint8)

 $X_{train} = X_{train} / 255$ $X_{\text{test}} = X_{\text{test}} / 255$

X_train = X_train.reshape(60000, 784) $X_{\text{test}} = X_{\text{test.reshape}}(10000, 784)$

X_train[0].shape

→ (784,)

from tensorflow import keras from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Flatten, Dense

```
model = Sequential()
model.add(Dense(10, activation='softmax',input_shape=(784,)))
model.compile(optimizer='adam',loss='sparse categorical crossentropy',metrics=['accuracy'])
model.fit(X_train,y_train,epochs=5)
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
    Epoch 1/5
    1875/1875 -
                          3s 1ms/step - accuracy: 0.8072 - loss: 0.7267
    Epoch 2/5
    1875/1875 <del>-</del>
                 3s 2ms/step - accuracy: 0.9151 - loss: 0.3079
    Epoch 3/5
    1875/1875 -
                          ----- 3s 2ms/step - accuracy: 0.9199 - loss: 0.2894
    Epoch 4/5
    1875/1875 <del>-</del>
                       3s 1ms/step - accuracy: 0.9236 - loss: 0.2722
    Epoch 5/5
                       3s 1ms/step - accuracy: 0.9268 - loss: 0.2650
    1875/1875 <del>-</del>
    <keras.src.callbacks.history.History at 0x78153d464eb0>
model.evaluate(X_test,y_test)
→ 313/313 — Os 1ms/step - accuracy: 0.9143 - loss: 0.3023
    [0.26665347814559937, 0.9254000186920166]
model.predict(X test)
→ 313/313 ----
                     Os 1ms/step
    array([[5.6211493e-06, 7.9625834e-11, 2.0384223e-05, ..., 9.9146736e-01,
           3.1063890e-05, 4.6331936e-04],
           [2.3813271e-04, 2.6575958e-06, 9.8944682e-01, ..., 5.6023453e-16,
           1.2607417e-04, 1.1237867e-12],
           [2.2342490e-06, 9.7659731e-01, 1.0822067e-02, ..., 9.6586457e-04,
```

```
### 313/313
### array([[5.6211493e-06, 7.9625834e-11, 2.0384223e-05, ..., 9.9146736e-01, 3.1063890e-05, 4.6331936e-04],

[2.3813271e-04, 2.6575958e-06, 9.8944682e-01, ..., 5.6023453e-16, 1.2607417e-04, 1.1237867e-12],

[2.2342490e-06, 9.7659731e-01, 1.0822067e-02, ..., 9.6586457e-04, 4.8037283e-03, 3.8969322e-04],

...,

[1.1733340e-08, 1.7358657e-08, 6.3683083e-06, ..., 1.2374662e-03, 7.1569551e-03, 2.4908291e-02],

[4.9370118e-07, 5.9014070e-07, 9.2638209e-07, ..., 1.7624758e-07, 1.1376546e-02, 1.2455305e-06],

[1.1454872e-06, 5.3599926e-14, 4.9303831e-05, ..., 3.4453621e-12, 3.4974370e-08, 1.5631410e-10]], dtype=float32)
```

plt.matshow(X_test[0])

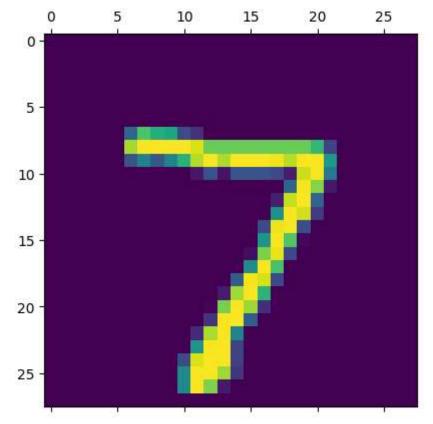
```
ValueError
                                           Traceback (most recent call last)
<ipython-input-26-3fb0095ace10> in <cell line: 1>()
----> 1 plt.matshow(X_test[0])
                                   🐧 1 frames
/usr/local/lib/python3.10/dist-packages/matplotlib/figure.py in figaspect(arg)
   3604
            # Extract the aspect ratio of the array
   3605
            if isarray:
-> 3606
                nr, nc = arg.shape[:2]
   3607
                arr_ratio = nr / nc
   3608
            else:
```

ValueError: not enough values to unpack (expected 2, got 1)

Next steps: Explain error

```
import matplotlib.pyplot as plt
# Reshape X_test[0] to a 2D array before plotting.
# Assuming X_test[0] represents a 28x28 image, reshape it accordingly.
plt.matshow(X_test[0].reshape(28, 28))
```

<matplotlib.image.AxesImage at 0x78153d92ec50>



y_predicted[0]

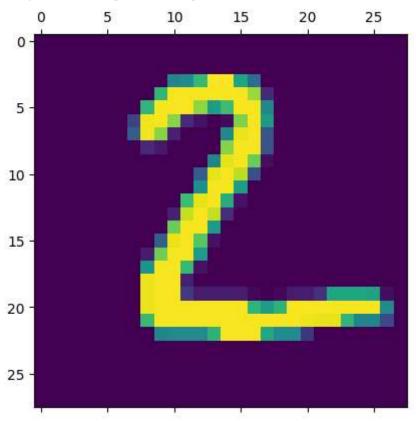
array([5.6211493e-06, 7.9625834e-11, 2.0384223e-05, 7.9663293e-03, 6.1376062e-07, 4.5238870e-05, 5.0766430e-10, 9.9146736e-01, 3.1063890e-05, 4.6331936e-04], dtype=float32)

np.argmax(y_predicted[0])

→ 7

plt.matshow(X_test[1].reshape(28, 28))

<matplotlib.image.AxesImage at 0x7815415d4280>



np.argmax(y_predicted[1])

→ 2

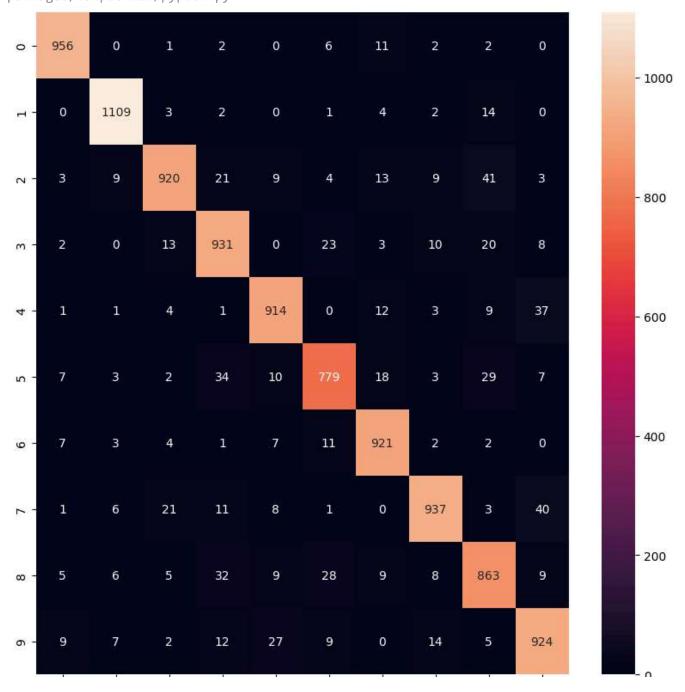
tf.math.confusion_matrix(y_test,np.argmax(y_predicted))

```
\rightarrow
```

```
InvalidArgumentError
                                            Traceback (most recent call last)
    <ipython-input-33-e807d09a9d81> in <cell line: 1>()
     ----> 1 tf.math.confusion_matrix(y_test,np.argmax(y_predicted))
                                      1 frames
    /usr/local/lib/python3.10/dist-packages/tensorflow/python/framework/ops.py in
    raise_from_not_ok_status(e, name)
       5981 def raise from not ok status(e, name) -> NoReturn:
              e.message += (" name: " + str(name if name is not None else ""))
    -> 5983
              raise core._status_to_exception(e) from None # pylint: disable=protected-
    access
       5984
       5985
    InvalidArgumentError: {{function node
     wrapped Pack N 2 device /job:localhost/replica:0/task:0/device:CPU:0}} Shapes of all
    inputs must match: values[0].shape = [10000] != values[1].shape = [] [Op:Pack] name:
    stack
 Next steps:
             Explain error
tf.math.confusion_matrix(y_test, np.argmax(y_predicted, axis=1))
<tf.Tensor: shape=(10, 10), dtype=int32, numpy=</pre>
    array([[ 956,
                     0,
                          1,
                              2,
                                      0,
                                                            2,
                                                                   0],
                                            6,
                                                 11,
                                                       2,
               0, 1109,
                           3,
                                2,
                                                 4,
                                                       2,
                                                            14,
                                                                   0],
           0,
                                            1,
               3,
                        920,
                               21,
                                     9,
                                           4,
                                                 13,
                                                       9,
                                                            41,
                                                                   3],
                     9,
                         13,
                             931,
               2,
                     0,
                                     0,
                                          23,
                                                  3,
                                                      10,
                                                            20,
                                                                   8],
                               1, 914,
               1,
                     1,
                         4,
                                            0,
                                                 12,
                                                       3,
                                                            9,
                                                                  37],
                                                       3,
                     3,
                         2,
                              34, 10, 779,
                                                18,
                                                            29,
                                                                  7],
                     3,
                                   7, 11, 921,
                          4,
                               1,
                                                       2,
                                                            2,
                                                                   0],
               1,
                   6,
                        21, 11,
                                     8,
                                           1,
                                                0, 937,
                                                            3,
                                                                  40],
                              32,
                                    9,
                                                9,
                                                      8, 863,
               5,
                        5,
                                           28,
                     6,
                                                                   9],
               9,
                    7,
                         2, 12, 27, 9,
                                                0,
                                                      14, 5,
                                                                 924]],
          dtype=int32)>
import seaborn as sn
plt.figure(figsize = (10,10))
sn.heatmap(tf.math.confusion_matrix(y_test, np.argmax(y_predicted, axis=1)), annot=True, fmt
plt
```



model = Sequential()



```
model.add(Dense(100, activation='relu',input_shape=(784,)))
model.add(Dense(10,activation='relu'))
model.add(Dense(10, activation='sigmoid'))
model.compile(optimizer='adam',loss='sparse_categorical_crossentropy',metrics=['accuracy'])
model.fit(X_train,y_train,epochs=10)
```

/usr/local/lib/python3.10/dist-packages/keras/src/layers/core/dense.py:87: UserWarning: super().__init__(activity_regularizer=activity_regularizer, **kwargs) Epoch 1/10 1875/1875 • 6s 3ms/step - accuracy: 0.7852 - loss: 0.6714 Epoch 2/10

1875/1875	6s	3ms/step	-	accuracy:	0.9606	-	loss:	0.1334
Epoch 3/10 1875/1875	5s	3ms/step	-	accuracy:	0.9714	-	loss:	0.0936
Epoch 4/10 1875/1875	6s	3ms/step	-	accuracy:	0.9785	-	loss:	0.0690
Epoch 5/10 1875/1875 ————————————————————————————————————	6s	3ms/step	-	accuracy:	0.9820	-	loss:	0.0591
Epoch 6/10 1875/1875 ————————————————————————————————————	5s	3ms/step	-	accuracy:	0.9867	-	loss:	0.0443
	7s	3ms/step	-	accuracy:	0.9878	-	loss:	0.0372
1875/1875	5s	3ms/step	-	accuracy:	0.9898	-	loss:	0.0323
Epoch 9/10 1875/1875	6s	3ms/step	-	accuracy:	0.9921	-	loss:	0.0247
Epoch 10/10 1875/1875				accuracy:		-	loss:	0.0205
<pre><keras.src.callbacks.history.history 0x781524ff2980="" at=""></keras.src.callbacks.history.history></pre>								

Start coding or generate with AI.

model.evaluate(X_test,y_test)

313/313 ______ 1s 3ms/step - accuracy: 0.9708 - loss: 0.1061 [0.09192464500665665, 0.9742000102996826]

y_predicted = model.predict(X_test)
tf.math.confusion_matrix(y_test, np.argmax(y_predicted, axis=1))

plt.figure(figsize = (10,10))
sn.heatmap(tf.math.confusion_matrix(y_test, np.argmax(y_predicted, axis=1)), annot
plt

