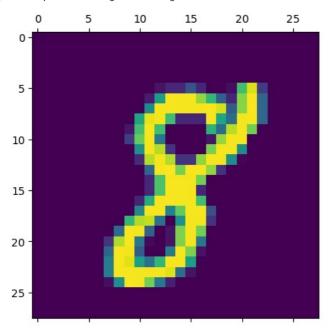
```
In [17]: #importing necessary libraries
import tensorflow as tf
from tensorflow import keras
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import random
%matplotlib inline
```

```
In [18]: #import dataset and split into train and test data
mnist = tf.keras.datasets.mnist
  (x_train, y_train), (x_test, y_test) = mnist.load_data()
```

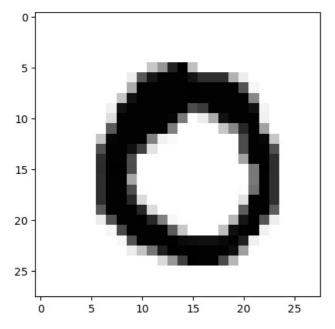
In [55]: plt.matshow(x_train[97])

Out[55]: <matplotlib.image.AxesImage at 0x288a21b5390>



In [56]: plt.imshow(-x_train[56], cmap="gray")

Out[56]: <matplotlib.image.AxesImage at 0x288a2365390>

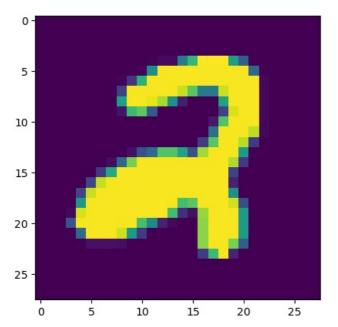


```
In [58]:
    x_train = x_train / 255
    x_test = x_test / 255
    # to bring the array in a specific range
    print(x_train.min())
    print(x_train.max())
```

0.0 2.365044290627876e-10

```
In [31]: model = keras.Sequential([
```

```
keras.layers.Flatten(input_shape=(28, 28)),
     keras.layers.Dense(128, activation="relu"),
     keras.layers.Dense(10, activation="softmax")
     model.summary()
     Model: "sequential_1"
     Layer (type)
                                       Param #
                       Output Shape
     flatten 1 (Flatten)
                       (None, 784)
                                       0
     dense 2 (Dense)
                       (None, 128)
                                       100480
                                       1290
     dense_3 (Dense)
                       (None, 10)
     Total params: 101770 (397.54 KB)
     Trainable params: 101770 (397.54 KB)
     Non-trainable params: 0 (0.00 Byte)
In [32]: model.compile(optimizer="sgd",
     loss="sparse_categorical_crossentropy",
     metrics=['accuracy'])
In [33]: history=model.fit(x train,
     y_train,validation_data=(x_test,y_test),epochs=10)
     Fnoch 1/10
     val_accuracy: 0.1135
     Epoch 2/10
     val accuracy: 0.1135
     Epoch 3/10
     1875/1875 [==
                     ========] - 9s 5ms/step - loss: 2.2992 - accuracy: 0.1124 - val loss: 2.2987 -
     val accuracy: 0.1135
     Epoch 4/10
     1875/1875 [==:
                     =========] - 9s 5ms/step - loss: 2.2988 - accuracy: 0.1124 - val loss: 2.2983 -
     val accuracy: 0.1135
     Epoch 5/10
     val_accuracy: 0.1135
     Epoch 6/10
     val accuracy: 0.1135
     Epoch 7/10
     val accuracy: 0.1135
     Epoch 8/10
     val_accuracy: 0.1135
     Epoch 9/10
                    ==========] - 9s 5ms/step - loss: 2.2965 - accuracy: 0.1124 - val_loss: 2.2959 -
     1875/1875 [========
     val accuracy: 0.1135
     Epoch 10/10
               1875/1875 [==
     val_accuracy: 0.1135
In [43]: test_loss,test_acc=model.evaluate(x_test,y_test)
     print("Loss=%.3f" %test_loss)
     print("Accuracy=%.3f" %test_acc)
     Loss=2.301
     Accuracy=0.113
In [44]: n=random.randint(0,9999)
     plt.imshow(x_test[n])
     plt.show()
```



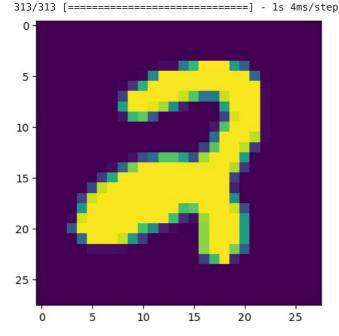
```
In [48]: x_train[:5]
Out[48]: array([[[0., 0., 0., ..., 0., 0., 0.],
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]
                           [0., 0., 0., \dots, 0., 0., 0.]
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]],
                          \hbox{\tt [[0., 0., 0., ..., 0., 0., 0.],}
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
                           [0., 0., 0., \ldots, 0., 0., 0.]
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]],
                          \hbox{\tt [[0., 0., 0., ..., 0., 0., 0.],}
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]
                           [0., 0., 0., \ldots, 0., 0., 0.]
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]],
                          \hbox{\tt [[0., 0., 0., ..., 0., 0., 0.],}
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
                           [0., 0., 0., \dots, 0., 0., 0.]
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]],
                          [[0., 0., 0., ..., 0., 0., 0.],
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]
                           [0., 0., 0., \ldots, 0., 0., 0.]
                           [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]]])
```

```
In [50]: x_test[:5]
```

```
Out[50]: array([[[0., 0., 0., ..., 0., 0., 0.],
                       [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]
                       [0., 0., 0., \dots, 0., 0., 0.]
                       [0., 0., 0., \ldots, 0., 0., 0.]
                       [0., 0., 0., \dots, 0., 0., 0.]
                      [[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
                       [0., 0., 0., \dots, 0., 0., 0.],
                       [0., 0., 0., \ldots, 0., 0., 0.]
                       [0., 0., 0., ..., 0., 0., 0.]
                      [[0., 0., 0., \ldots, 0., 0., 0.],
                       [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
                       [0.,\ 0.,\ 0.,\ \dots,\ 0.,\ 0.,\ 0.],
                       [0., 0., 0., \dots, 0., 0., 0.]
                       [0., 0., 0., \ldots, 0., 0., 0.]
                      \hbox{\tt [[0., 0., 0., ..., 0., 0., 0.],}
                       [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]
                       [0., 0., 0., \dots, 0., 0., 0.]
                       [0., 0., 0., \ldots, 0., 0., 0.]
                       [0., 0., 0., ..., 0., 0., 0.]
                      [[0., 0., 0., \ldots, 0., 0., 0.],
                       [0., 0., 0., \ldots, 0., 0., 0.]
                       [0., 0., 0., ..., 0., 0., 0.],
                       [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
                       [0., 0., 0., \dots, 0., 0., 0.]])
```

```
In [53]: predicted_value=model.predict(x_test)
    plt.imshow(x_test[n])
    plt.show()

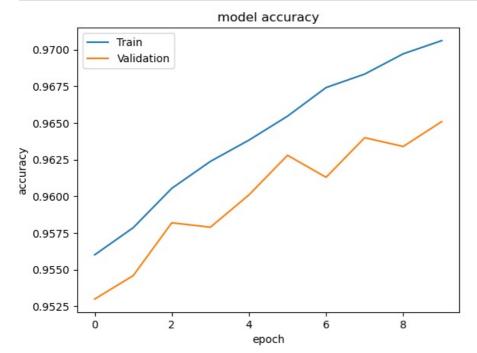
print(predicted_value[n])
```



```
In [28]: # history.history()
history.history.keys()
# dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])

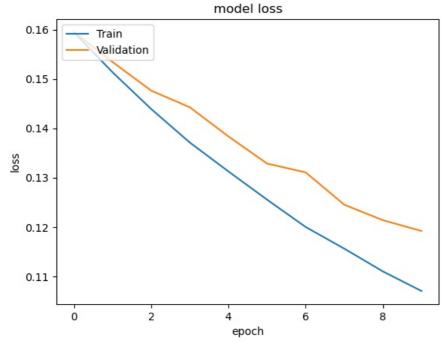
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
```

```
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
```



```
In [29]: # history.history()
history.history.keys()
# dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])

plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
```



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