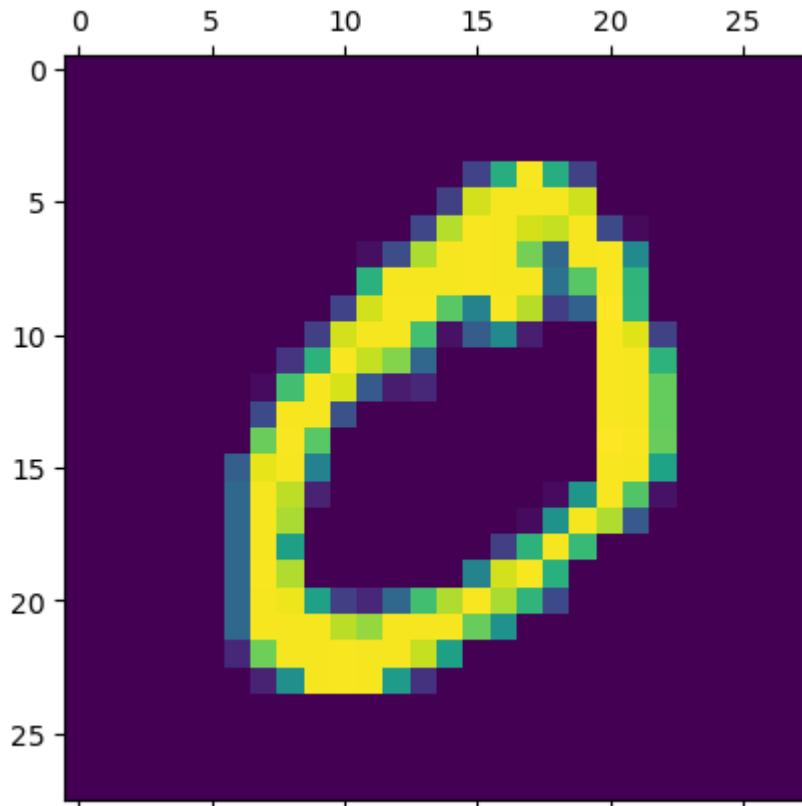


```
In [2]: #importing necessary libraries
import tensorflow as tf
from tensorflow import keras
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import random
%matplotlib inline
```

```
In [3]: #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()
```

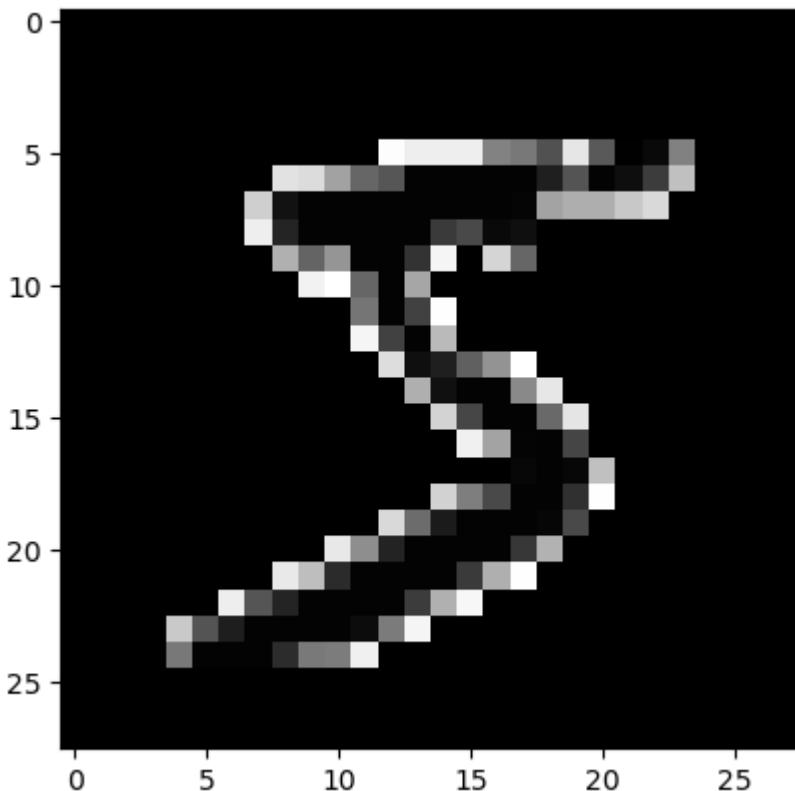
```
In [4]: plt.matshow(x_train[1])
```

```
Out[4]: <matplotlib.image.AxesImage at 0x2318f9c2788>
```



```
In [5]: plt.imshow(-x_train[0], cmap="gray")
```

```
Out[5]: <matplotlib.image.AxesImage at 0x2318f7af388>
```



```
In [5]: x_train = x_train / 255
x_test = x_test / 255
```

```
In [6]: 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"

Layer (type)	Output Shape	Param #
<hr/>		
flatten (Flatten)	(None, 784)	0
dense (Dense)	(None, 128)	100480
dense_1 (Dense)	(None, 10)	1290
<hr/>		
Total params: 101,770		
Trainable params: 101,770		
Non-trainable params: 0		

```
In [7]: model.compile(optimizer="sgd",
                    loss="sparse_categorical_crossentropy",
                    metrics=['accuracy'])
```

```
In [8]: history=model.fit(x_train,
                        y_train,validation_data=(x_test,y_test),epochs=10)
```

```

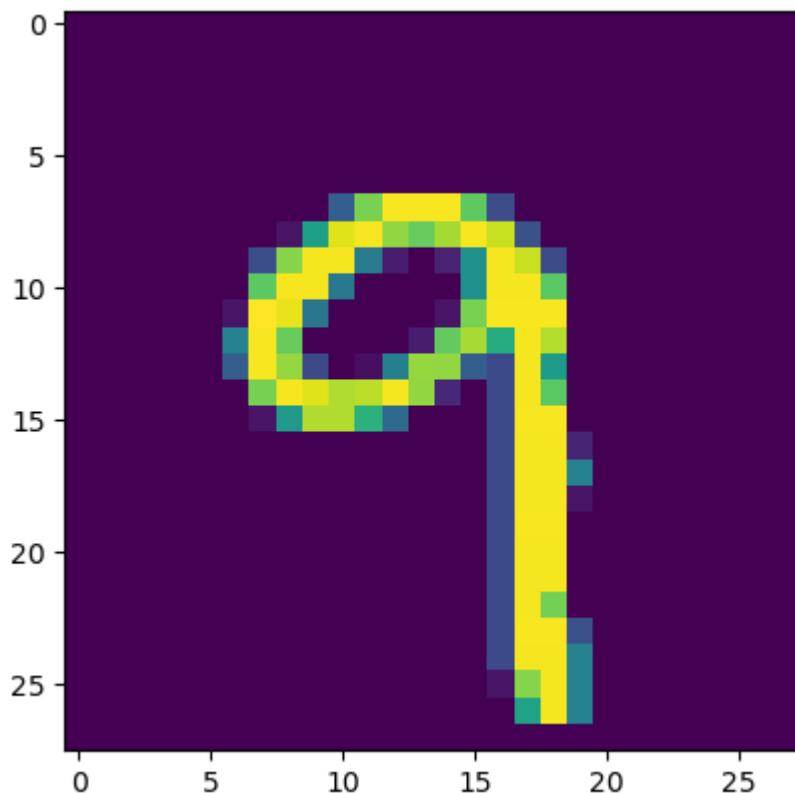
Epoch 1/10
1875/1875 [=====] - 1s 619us/step - loss: 0.6385
- accuracy: 0.8383 - val_loss: 0.3554 - val_accuracy: 0.9044
Epoch 2/10
1875/1875 [=====] - 1s 552us/step - loss: 0.3383
- accuracy: 0.9051 - val_loss: 0.2977 - val_accuracy: 0.9176
Epoch 3/10
1875/1875 [=====] - 1s 546us/step - loss: 0.2920
- accuracy: 0.9176 - val_loss: 0.2663 - val_accuracy: 0.9275
Epoch 4/10
1875/1875 [=====] - 1s 545us/step - loss: 0.2623
- accuracy: 0.9269 - val_loss: 0.2438 - val_accuracy: 0.9319
Epoch 5/10
1875/1875 [=====] - 1s 547us/step - loss: 0.2399
- accuracy: 0.9331 - val_loss: 0.2238 - val_accuracy: 0.9377
Epoch 6/10
1875/1875 [=====] - 1s 658us/step - loss: 0.2211
- accuracy: 0.9385 - val_loss: 0.2086 - val_accuracy: 0.9399
Epoch 7/10
1875/1875 [=====] - 1s 639us/step - loss: 0.2054
- accuracy: 0.9432 - val_loss: 0.1947 - val_accuracy: 0.9440
Epoch 8/10
1875/1875 [=====] - 1s 627us/step - loss: 0.1919
- accuracy: 0.9467 - val_loss: 0.1847 - val_accuracy: 0.9461
Epoch 9/10
1875/1875 [=====] - 1s 556us/step - loss: 0.1801
- accuracy: 0.9500 - val_loss: 0.1736 - val_accuracy: 0.9500
Epoch 10/10
1875/1875 [=====] - 1s 552us/step - loss: 0.1695
- accuracy: 0.9535 - val_loss: 0.1656 - val_accuracy: 0.9521

```

```
In [9]: test_loss,test_acc=model.evaluate(x_test,y_test)
print("Loss=%f" %test_loss)
print("Accuracy=%f" %test_acc)
```

```
313/313 [=====] - 0s 424us/step - loss: 0.1656 -
accuracy: 0.9521
Loss=0.166
Accuracy=0.952
```

```
In [10]: n=random.randint(0,9999)
plt.imshow(x_test[n])
plt.show()
```



```
In [11]: x_train
```

```
Out[11]: 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., 0., 0., ..., 0., 0., 0.],
   ...,
   [0., 0., 0., ..., 0., 0., 0.],
   [0., 0., 0., ..., 0., 0., 0.],
   [0., 0., 0., ..., 0., 0., 0.]],

   [[0., 0., 0., ..., 0., 0., 0.],
   [0., 0., 0., ..., 0., 0., 0.],
   [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 [12]: x_test
```

```
Out[12]: 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.]],

 ...,

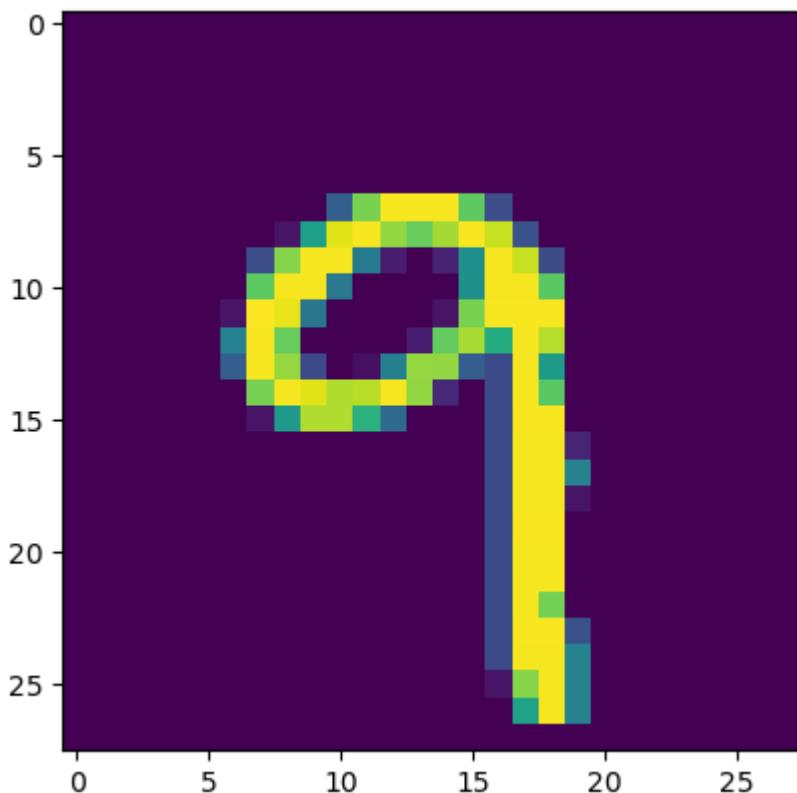
 [[[0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 ...,
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.]],

 [[[0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 ...,
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.]],

 [[[0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 ...,
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.],
 [0., 0., 0., ..., 0., 0., 0.]]])
```

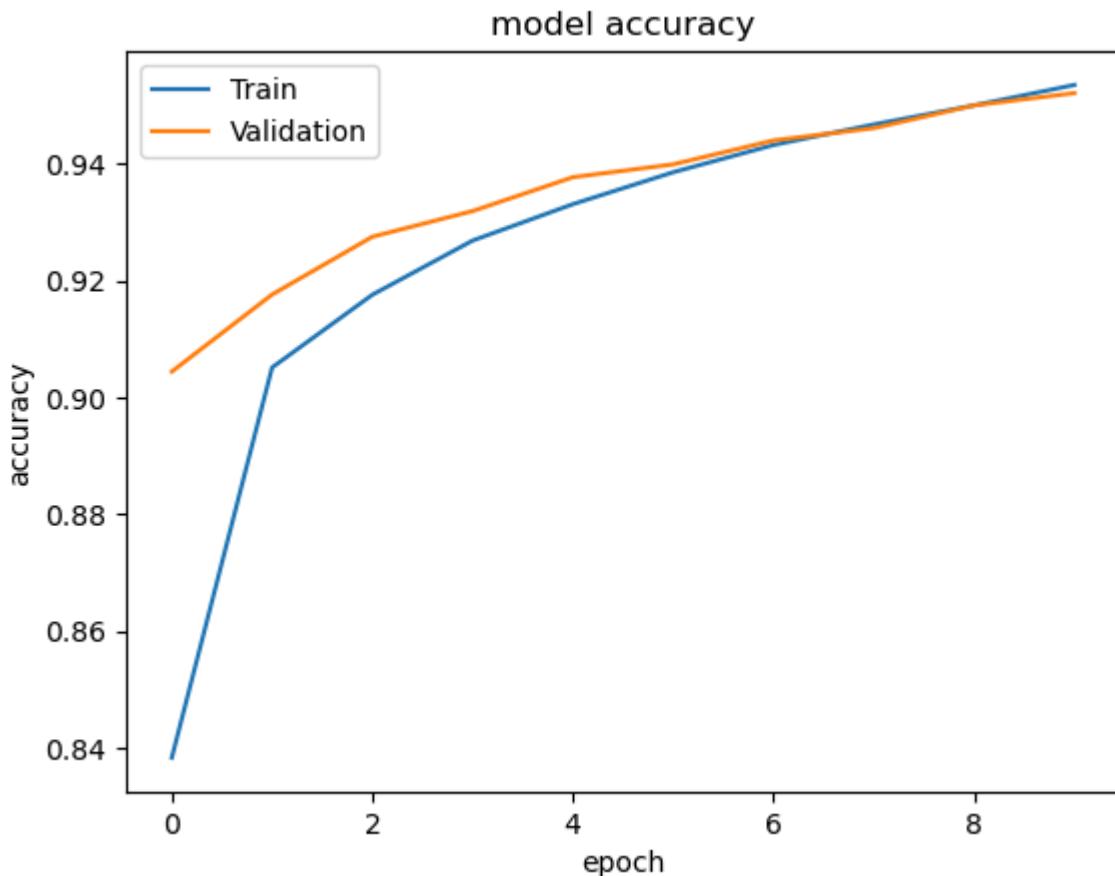
```
In [13]: predicted_value=model.predict(x_test)
plt.imshow(x_test[n])
plt.show()

print(predicted_value[n])
```



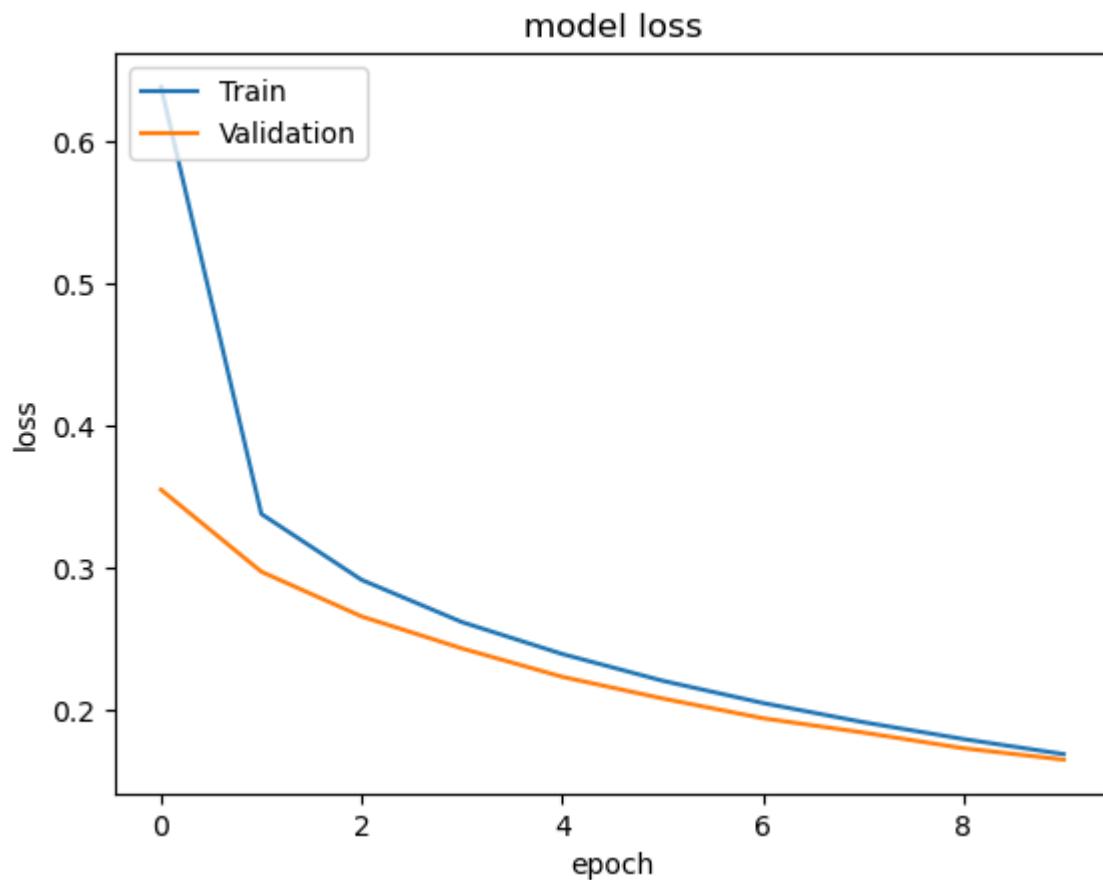
```
[7.2151641e-05 3.4558122e-06 3.3534434e-05 4.9630189e-03 2.8357599e-03  
1.7946344e-03 4.8126030e-06 1.5905222e-02 6.7955634e-04 9.7370785e-01]
```

```
In [14]: # 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 [15]: # 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.xlabel('epoch')
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
```



In []: