

```
#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

mnist = tf.keras.datasets.mnist
(x_train, y_train), (x_test, y_test) = mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 [=====] - 4s 0us/step

#to see length of training dataset
len(x_train)

60000

##to see length of testing dataset
len(x_test)

10000

#shape of training dataset 60,000 images having 28*28 size
x_train.shape

(60000, 28, 28)

#shape of testing dataset 10,000 images having 28*28 size
x_test.shape

(10000, 28, 28)

x_train[0]

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]]
```

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0,
    0,  0],
  [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
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    0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
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0,
    0,  0],
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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,
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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,  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,  11,
190,
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0,	0, 0],
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35,	241, 225, 160, 108, 1, 0, 0, 0, 0, 0, 0, 0, 0,
0,	0, 0],
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0,	81, 240, 253, 253, 119, 25, 0, 0, 0, 0, 0, 0, 0,
0,	0, 0],
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0,	0, 45, 186, 253, 253, 150, 27, 0, 0, 0, 0, 0, 0,
0,	0, 0],
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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, 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, 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, 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, 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, 0, 0, 18, 171, 219, 253, 253, 253,

```

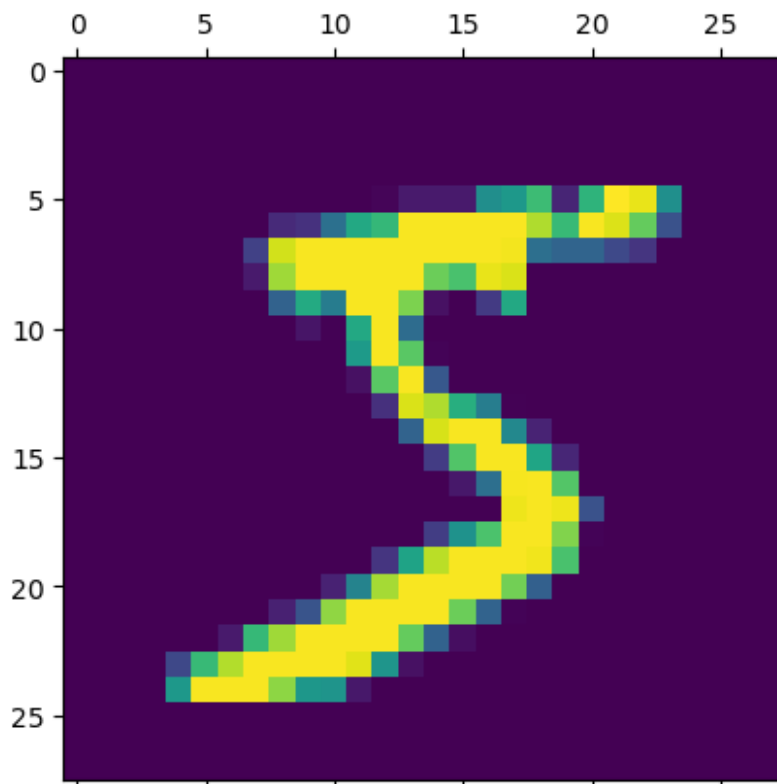
253,
    195, 80, 9, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
    0, 0],
[ 0, 0, 0, 0, 55, 172, 226, 253, 253, 253, 253, 244,
133,
    11, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
    0, 0],
[ 0, 0, 0, 0, 136, 253, 253, 253, 212, 135, 132, 16,
0,
    0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
    0, 0],
[ 0, 0, 0, 0, 0, 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, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
    0, 0]], dtype=uint8)

```

#to see how first image look

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

```
<matplotlib.image.AxesImage at 0x26590ab9790>
```



#normalize the images by scaling pixel intensities to the range 0,1

`x_train = x_train / 255`

`x_test = x_test / 255`

`x_train[0]`

```
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.
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0.	, 0.	, 0.	, 0.	, 0.	, 0.	, 0.
0.	, 0.	, 0.	, 0.	, 0.	, 0.	, 0.
[0.	, 0.	, 0.	, 0.	, 0.	, 0.	, 0.
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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.	, 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.		
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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.
0.	, 0.	, 0.07058824,	0.85882353,	0.99215686,		
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0.71372549,	0.96862745,	0.94509804,	0.	, 0.		
0.	, 0.	, 0.	, 0.	, 0.		
0.	, 0.	, 0.	, 0.			
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0.	, 0.	, 0.	, 0.			
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0.	, 0.	, 0.	, 0.	, 0.		
0.	, 0.	, 0.	, 0.			
[0.	, 0.	, 0.	, 0.	, 0.	, 0.	, 0.

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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.	, 0.	, 0.
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0.	, 0.	, 0.	, 0.	, 0.	, 0.
0.	, 0.	, 0.],		
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0.	, 0.	, 0.	, 0.	, 0.	, 0.
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0.	, 0.	, 0.],		
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0.	, 0.	, 0.	, 0.	, 0.	, 0.
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0.	, 0.	, 0.],		
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0.	, 0.	, 0.	, 0.	, 0.	, 0.
0.	, 0.	, 0.	, 0.	, 0.	, 0.
0.0627451	, 0.36470588,	0.98823529,	0.99215686,	0.73333333,	
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0.	, 0.	, 0.],		
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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.],		
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0.50980392,	0.71764706,	0.99215686,	0.99215686,	0.81176471,	
0.00784314,	0.	, 0.	, 0.	, 0.	, 0.
0.	, 0.	, 0.],		
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[0. , 0. , 0. , 0. , 0. ,
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```
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ]])
```

```
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 #
flatten (Flatten)	(None, 784)	0
dense (Dense)	(None, 128)	100480
dense_1 (Dense)	(None, 10)	1290

```
=====  
Total params: 101770 (397.54 KB)  
Trainable params: 101770 (397.54 KB)  
Non-trainable params: 0 (0.00 Byte)
```

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

```
history=model.fit(x_train,  
y_train,validation_data=(x_test,y_test),epochs=10)
```

```
Epoch 1/10
```

```
1875/1875 [=====] - 11s 6ms/step - loss:  
0.6417 - accuracy: 0.8362 - val_loss: 0.3575 - val_accuracy: 0.9035
```

```
Epoch 2/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:  
0.3356 - accuracy: 0.9055 - val_loss: 0.2922 - val_accuracy: 0.9166
```

```
Epoch 3/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:  
0.2861 - accuracy: 0.9195 - val_loss: 0.2585 - val_accuracy: 0.9270
```

```
Epoch 4/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:  
0.2555 - accuracy: 0.9285 - val_loss: 0.2337 - val_accuracy: 0.9322
```

```
Epoch 5/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:  
0.2324 - accuracy: 0.9352 - val_loss: 0.2163 - val_accuracy: 0.9379
```

```
Epoch 6/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:
0.2143 - accuracy: 0.9403 - val_loss: 0.2010 - val_accuracy: 0.9430
Epoch 7/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:
0.1994 - accuracy: 0.9440 - val_loss: 0.1893 - val_accuracy: 0.9458
Epoch 8/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:
0.1866 - accuracy: 0.9476 - val_loss: 0.1785 - val_accuracy: 0.9483
Epoch 9/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:
0.1754 - accuracy: 0.9508 - val_loss: 0.1698 - val_accuracy: 0.9512
Epoch 10/10
```

```
1875/1875 [=====] - 10s 5ms/step - loss:
0.1656 - accuracy: 0.9537 - val_loss: 0.1606 - val_accuracy: 0.9539
```

```
test_loss,test_acc=model.evaluate(x_test,y_test)
```

```
print("Loss=%.3f" %test_loss)
```

```
print("Accuracy=%.3f" %test_acc)
```

```
313/313 [=====] - 1s 4ms/step - loss: 0.1606
- accuracy: 0.9539
```

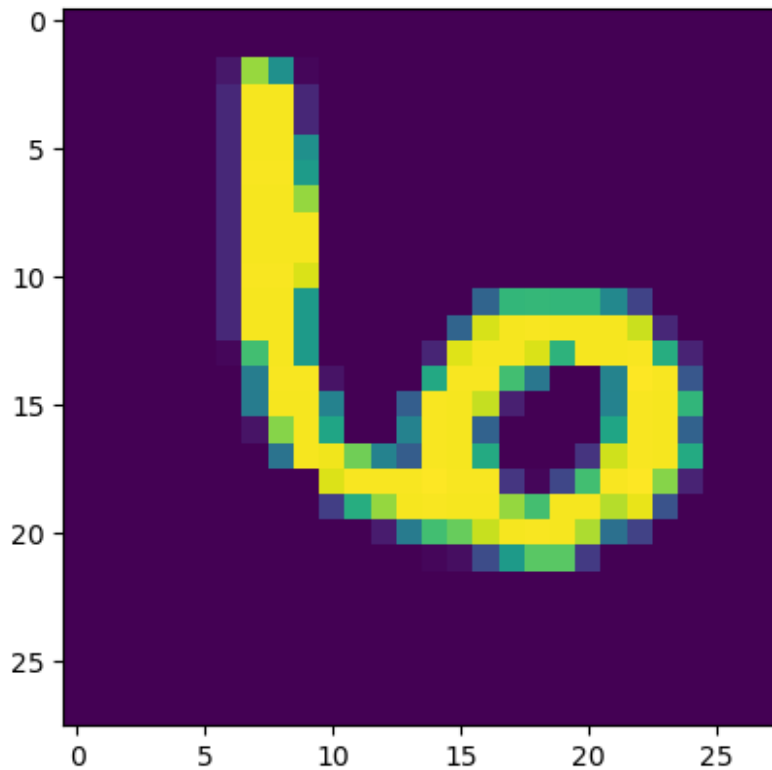
```
Loss=0.161
```

```
Accuracy=0.954
```

```
n=random.randint(0,9999)
```

```
plt.imshow(x_test[n])
```

```
plt.show()
```



```
#we use predict() on new data
predicted_value=model.predict(x_test)
print("Handwritten number in the image is= %d"
      %np.argmax(predicted_value[n]))

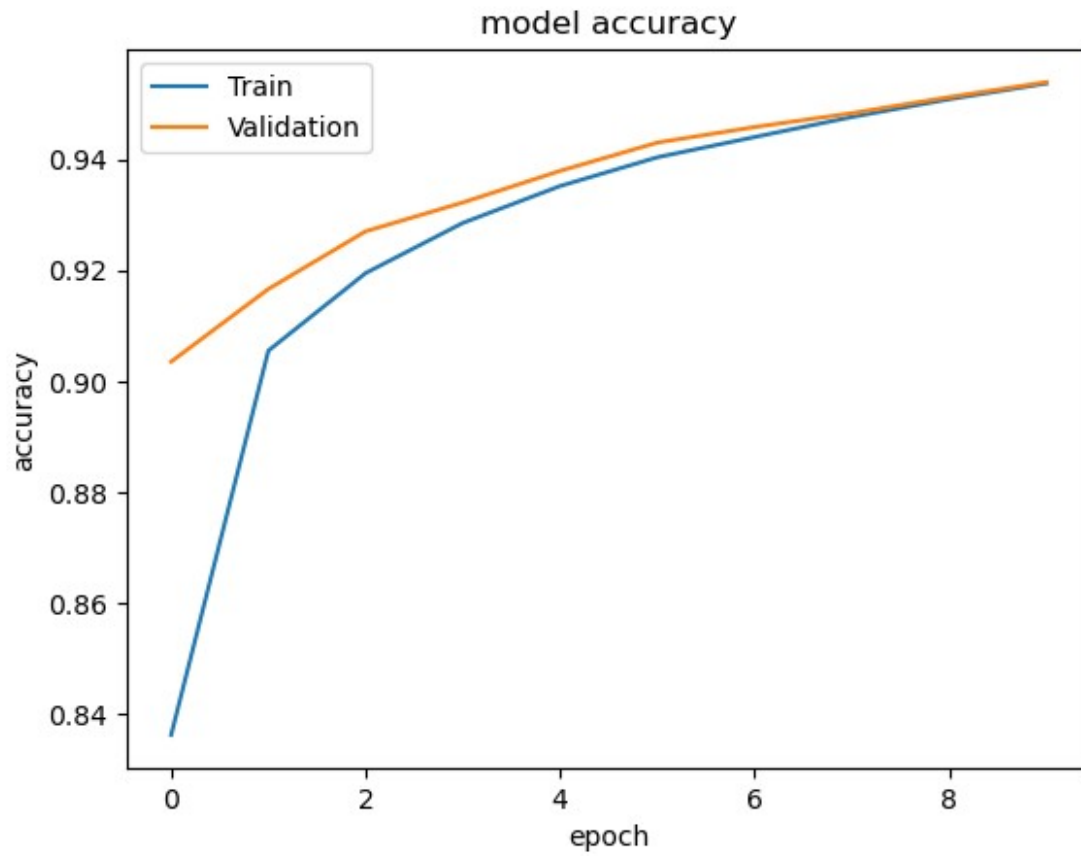
313/313 [=====] - 1s 3ms/step
Handwritten number in the image is= 6

history.history??

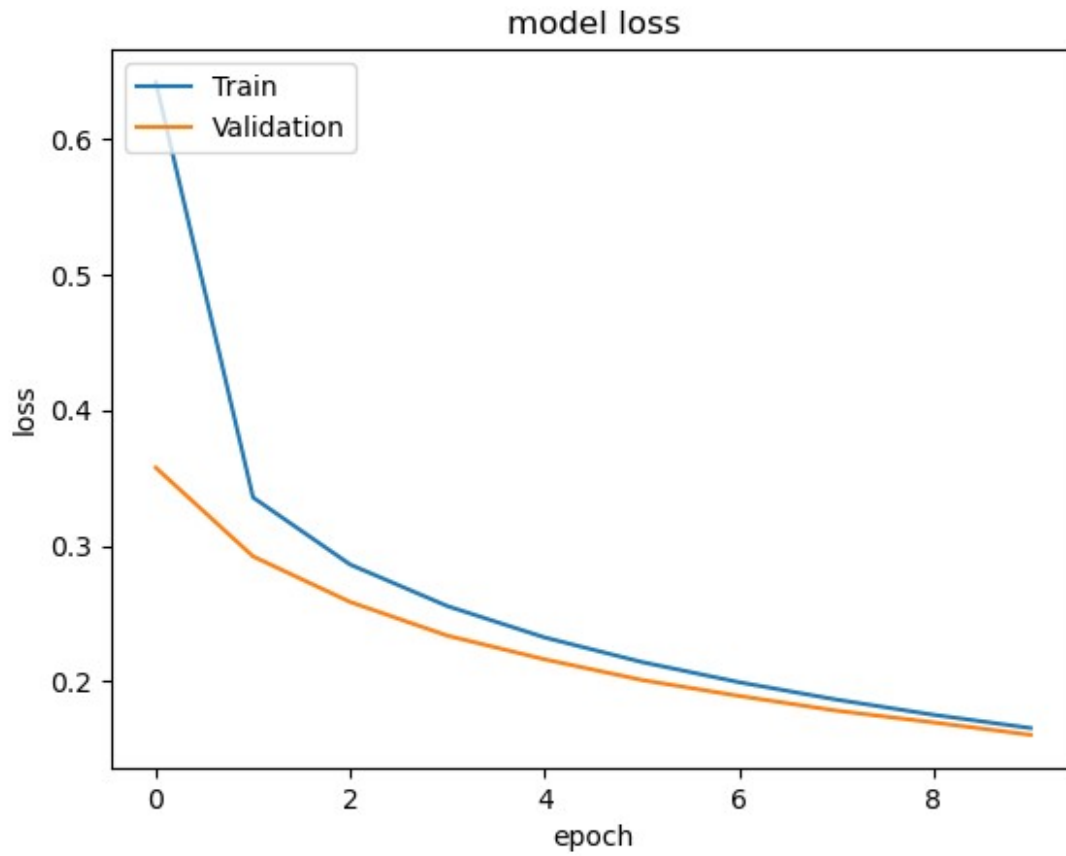
history.history.keys()

dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])

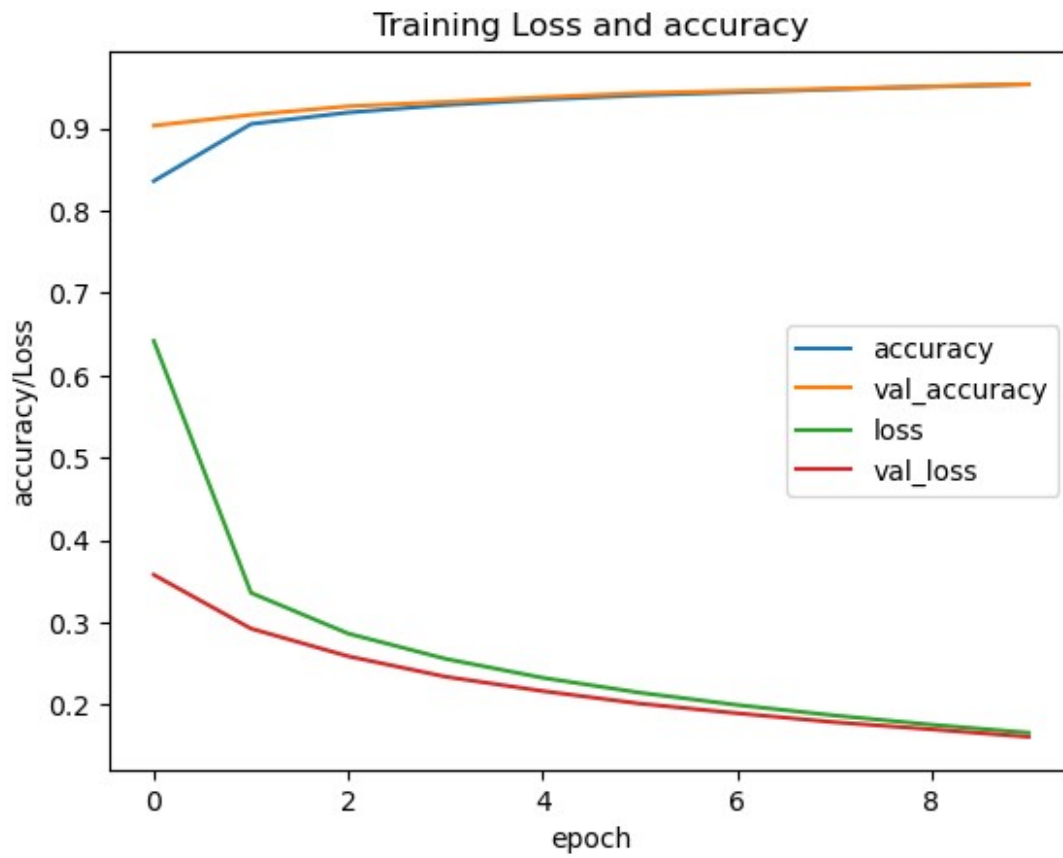
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
```



```
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()
```



```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('Training Loss and accuracy')
plt.ylabel('accuracy/Loss')
plt.xlabel('epoch')
plt.legend(['accuracy', 'val_accuracy', 'loss', 'val_loss'])
plt.show()
```



```
pwd
```

```
'C:\\Users\\ishwa'
```

```
keras_model_path='C:\\Users\\ishwa'  
model.save(keras_model_path)
```

```
INFO:tensorflow:Assets written to: C:\\Users\\ishwa\\assets
```

```
INFO:tensorflow:Assets written to: C:\\Users\\ishwa\\assets
```

```
#use the save model
```

```
restored_keras_model = tf.keras.models.load_model(keras_model_path)
```