

```
In [1]: import tensorflow as tf  
from tensorflow import keras
```



```
In [2]: mnist_dataset = tf.keras.datasets.mnist
```



```
In [3]: (x_train, y_train), (x_test, y_test) = mnist_dataset.load_data()  
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz  
11490434/11490434 ━━━━━━━━ 3s 0us/step
```

```
In [4]: len(x_train)
```

```
Out[4]: 60000
```

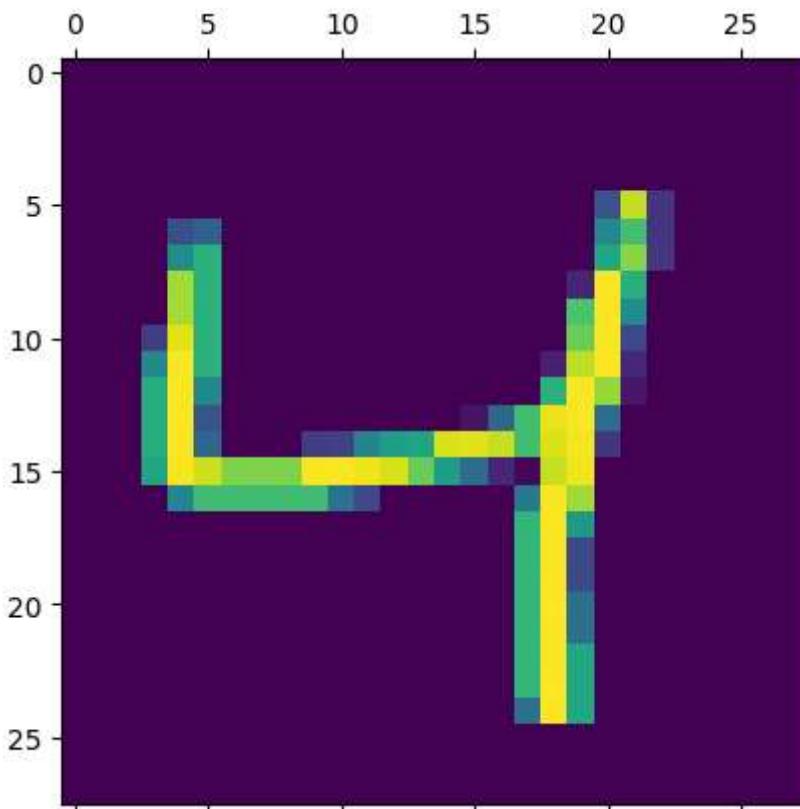
```
In [5]: x_train.shape
```

```
Out[5]: (60000, 28, 28)
```

```
In [6]: import matplotlib.pyplot as plt
```

```
In [7]: plt.matshow(x_train[2])
```

```
Out[7]: <matplotlib.image.AxesImage at 0x2074ac58590>
```



```
In [8]: x_train = x_train/255  
x_test = x_test/255
```

In [9]: `x_train[0]`

0.41960784, 0.99215686, 0.99215686, 0.80392157, 0.04313725,
 0., 0.16862745, 0.60392157, 0., 0., 0.,
 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0.,
 [0., 0., 0., 0., 0., 0., 0.],
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.05490196,
 0.00392157, 0.60392157, 0.99215686, 0.35294118, 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 [0., 0., 0., 0., 0., 0., 0.],
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0.54509804, 0.99215686, 0.74509804, 0.00784314,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 [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.],
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0.1372549, 0.94509804, 0.88235294,
 0.62745098, 0.42352941, 0.00392157, 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 [0., 0., 0., 0., 0., 0., 0.],
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0.31764706, 0.94117647,
 0.99215686, 0.99215686, 0.46666667, 0.09803922, 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 [0., 0., 0., 0., 0., 0., 0.],
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0.17647059,
 0.72941176, 0.99215686, 0.99215686, 0.58823529, 0.10588235,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 [0., 0., 0., 0., 0., 0., 0.],
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0.0627451, 0.36470588, 0.98823529, 0.99215686, 0.73333333,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 [0., 0., 0., 0., 0., 0., 0.],
 0., 0., 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., 0., 0., 0., 0., 0.,
 [0., 0., 0., 0., 0., 0., 0.],
 0., 0., 0., 0., 0., 0., 0.,
 0., 0., 0., 0., 0., 0., 0.,
 0., 0.18039216,
 0.50980392, 0.71764706, 0.99215686, 0.99215686, 0.81176471,


```
In [10]: model = keras.Sequential([
    keras.layers.Flatten(input_shape=(28, 28)),
    keras.layers.Dense(128, activation='relu'),
    keras.layers.Dense(10, activation='softmax')
])
```

```
C:\Users\SMIT DESHMUKH\AppData\Local\Programs\Python\Python313\Lib\site-packages\keras\src\layers\reshaping\flatten.py:37: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.
    super().__init__(**kwargs)
```

```
In [11]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape
flatten (Flatten)	(None, 784)
dense (Dense)	(None, 128)
dense_1 (Dense)	(None, 10)



Total params: 101,770 (397.54 KB)

Trainable params: 101,770 (397.54 KB)

Non-trainable params: 0 (0.00 B)

```
In [12]: model.compile(optimizer='sgd',
                      loss='sparse_categorical_crossentropy',
                      metrics=['accuracy'])
```

```
In [13]: history = model.fit(x_train, y_train, validation_data=(x_test, y_test), epochs=10)
```

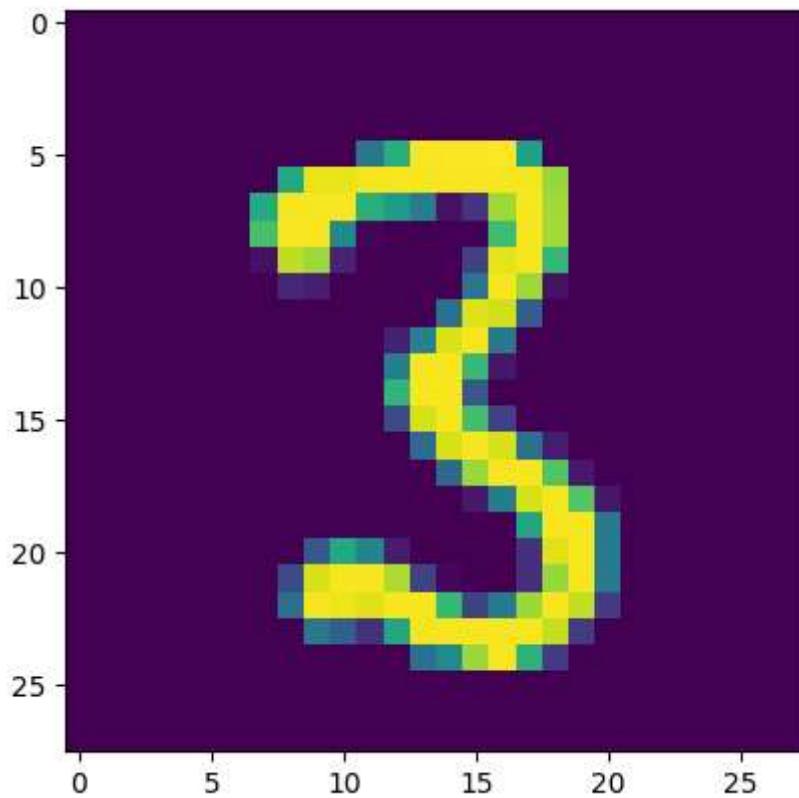
```
Epoch 1/10
1875/1875 24s 12ms/step - accuracy: 0.8414 - loss: 0.6359 - val
_accuracy: 0.9037 - val_loss: 0.3593
Epoch 2/10
1875/1875 40s 11ms/step - accuracy: 0.9065 - loss: 0.3390 - val
_accuracy: 0.9195 - val_loss: 0.2944
Epoch 3/10
1875/1875 21s 11ms/step - accuracy: 0.9179 - loss: 0.2910 - val
_accuracy: 0.9281 - val_loss: 0.2626
Epoch 4/10
1875/1875 21s 11ms/step - accuracy: 0.9270 - loss: 0.2610 - val
_accuracy: 0.9344 - val_loss: 0.2411
Epoch 5/10
1875/1875 21s 11ms/step - accuracy: 0.9339 - loss: 0.2381 - val
_accuracy: 0.9395 - val_loss: 0.2226
Epoch 6/10
1875/1875 24s 12ms/step - accuracy: 0.9386 - loss: 0.2195 - val
_accuracy: 0.9425 - val_loss: 0.2083
Epoch 7/10
1875/1875 38s 11ms/step - accuracy: 0.9431 - loss: 0.2041 - val
_accuracy: 0.9449 - val_loss: 0.1948
Epoch 8/10
1875/1875 18s 10ms/step - accuracy: 0.9469 - loss: 0.1904 - val
_accuracy: 0.9492 - val_loss: 0.1840
Epoch 9/10
1875/1875 22s 11ms/step - accuracy: 0.9499 - loss: 0.1790 - val
_accuracy: 0.9511 - val_loss: 0.1745
Epoch 10/10
1875/1875 22s 12ms/step - accuracy: 0.9524 - loss: 0.1689 - val
_accuracy: 0.9531 - val_loss: 0.1658
```

```
In [14]: test_loss, test_acc = model.evaluate(x_test, y_test)
print("Loss=%.3f"%test_loss)
print("Accuracy=%.3f"%test_acc)
```

```
313/313 3s 9ms/step - accuracy: 0.9531 - loss: 0.1658
Loss=0.166
Accuracy=0.953
```

```
In [15]: import random
```

```
In [16]: n = random.randint(0,9999)
plt.imshow(x_test[n])
plt.show()
```



```
In [17]: import numpy as np  
predicted_value=model.predict(x_test)  
print("Handwritten number in the image is= %d" %np.argmax(predicted_value[0]))
```

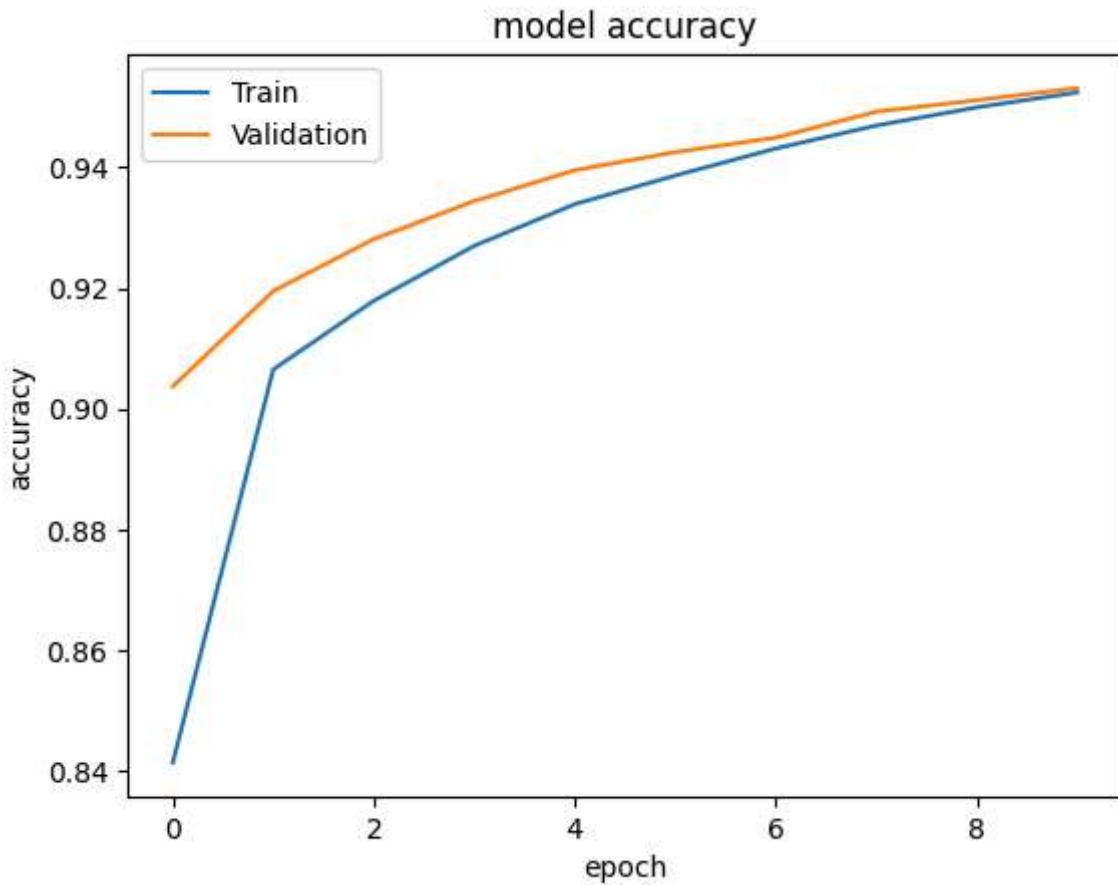
313/313 ————— 2s 7ms/step
Handwritten number in the image is= 3

```
In [18]: history.history.keys()
```

Out[18]: dict_keys(['accuracy', 'loss', 'val_accuracy', 'val_loss'])

Training Accuracy

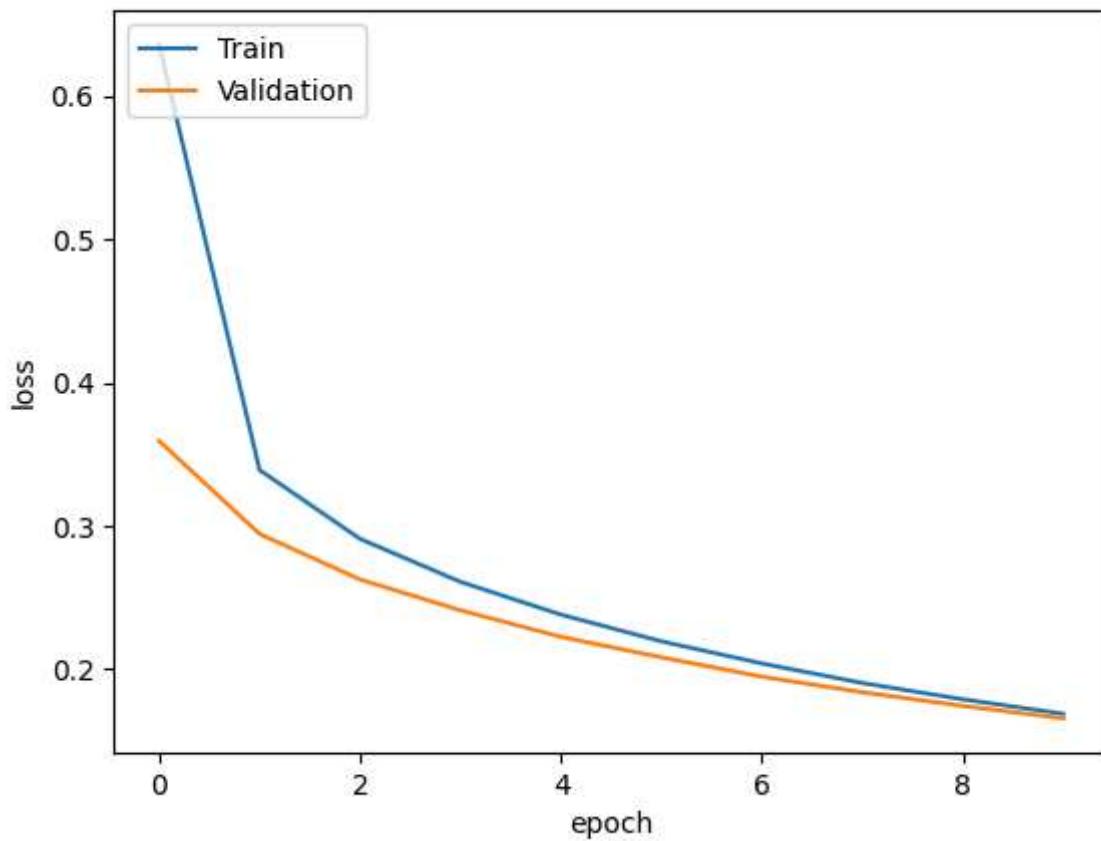
```
In [19]: plt.plot(history.history['accuracy'])  
plt.plot(history.history['val_accuracy'])  
plt.title('model accuracy')  
plt.ylabel('accuracy')  
plt.xlabel('epoch')  
plt.legend(['Train', 'Validation'], loc='upper left')  
plt.show()
```



Training Loss

```
In [20]: 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()
```

model loss



In []: