

```
import tensorflow as tf
from tensorflow import keras
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import random
```

```
from tensorflow.keras.datasets import mnist
```

```
(trainX, trainY), (testX, testY) = mnist.load_data()
```

↗ Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist11490434/11490434> — 0s 0us/step



```
len(trainX)
```

↗ 60000

```
len(testX)
```

↗ 10000

```
trainX.shape
```

↗ (60000, 28, 28)

```
plt.matshow(trainX[500])
```



```
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.01176471, 0.07058824, 0.07058824,
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.      , 0.11764706, 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.      ,
0.      , 0.      , 0.      ],
[0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.19215686, 0.93333333, 0.99215686,
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.      , 0.07058824, 0.85882353, 0.99215686,
0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.77647059,
0.71372549, 0.96862745, 0.94509804, 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.31372549, 0.61176471,
0.41960784, 0.99215686, 0.99215686, 0.80392157, 0.04313725,
0.      , 0.16862745, 0.60392157, 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_1 (Flatten)	(None, 784)	0
dense_1 (Dense)	(None, 128)	100480
dense_2 (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(trainX, trainY, validation_data=(testX, testY), epochs=10)
```



```
-----
NameError                                Traceback (most recent call last)
<ipython-input-1-7a960ee5a744> in <cell line: 1>()
----> 1 history=model.fit(trainX, trainY, validation_data=(testX, testY), epochs=10)

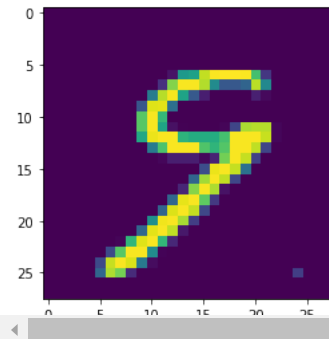
NameError: name 'model' is not defined
```

```
test_loss, test_acc=model.evaluate(testX, testY)
print("Loss=%.3f" %test_loss)
print("Accuracy=%.3f" %test_acc)
```



```
258/313 [=====>.....] - ETA: 0s - loss: 0.1751 - accuracy: 0.948826
313/313 [=====] - 0s 590us/step - loss: 0.1653 - accuracy: 0.
Loss=0.165
Accuracy=0.952
```

```
n=random.randint(0,9999)
plt.imshow(testX[n])
plt.show()
```

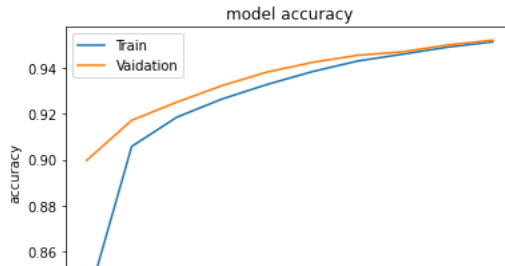


```
predicted_value=model.predict(testX)
print("Handwritten number in the image is=%d" %np.argmax(predicted_value[n]))
```



```
313/313 [=====] - 0s 510us/step
Handwritten number in the image is=9
```

```
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['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()
```

