

```
import numpy as np
import pandas as pd

import tensorflow
from tensorflow import keras
from tensorflow.keras import Sequential
from tensorflow.keras.layers import Dense, Flatten

(X_train,y_train),(X_test,y_test)=keras.datasets.mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/mnist.npz
11490434/11490434 [=====] - 1s 0us/step

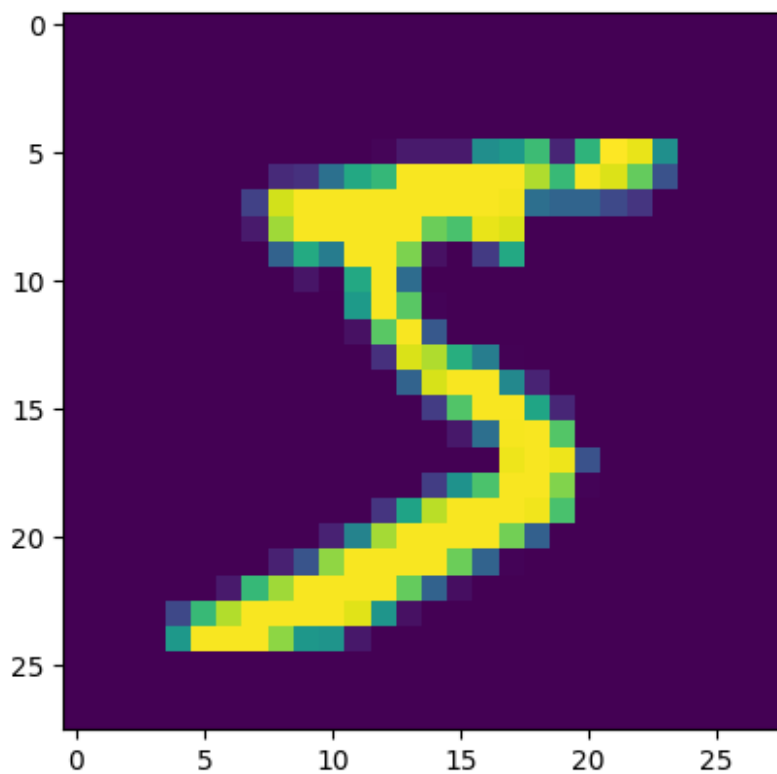
X_train.shape
(60000, 28, 28)

X_test.shape
(10000, 28, 28)

y_train
array([5, 0, 4, ..., 5, 6, 8], dtype=uint8)

import matplotlib.pyplot as plt
plt.imshow(X_train[0])

<matplotlib.image.AxesImage at 0x7eecf12177f0>
```



```
X_train[0].shape
```

```
(28, 28)
```

```
X_train = X_train / 255
```

```
X_test = X_test / 255
```

```
X_train[0]
```

```
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        0., 0., 0., 0., 0., 0.]])
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0.      , 0.      , 0.      , 11)

```

```

model = Sequential()
model.add(Flatten(input_shape=(28,28)))
model.add(Dense(32,activation='relu'))
model.add(Dense(128,activation='relu'))
model.add(Dense(10,activation='softmax'))

```

```
model.summary()
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
flatten (Flatten)	(None, 784)	0
dense (Dense)	(None, 32)	25120
dense_1 (Dense)	(None, 128)	4224
dense_2 (Dense)	(None, 10)	1290

```

=====
Total params: 30634 (119.66 KB)
Trainable params: 30634 (119.66 KB)
Non-trainable params: 0 (0.00 Byte)
=====

```

```

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

```

```
history = model.fit(X_train,y_train, epochs=10,validation_split=0.2)
```

```
Epoch 1/10
```

```
1500/1500 [=====] - 7s 3ms/step - loss:
0.3426 - accuracy: 0.8977 - val_loss: 0.1949 - val_accuracy: 0.9440
```

```
Epoch 2/10
```

```
1500/1500 [=====] - 7s 5ms/step - loss:
0.1662 - accuracy: 0.9496 - val_loss: 0.1402 - val_accuracy: 0.9578
```

```
Epoch 3/10
```

```
1500/1500 [=====] - 9s 6ms/step - loss:
0.1255 - accuracy: 0.9611 - val_loss: 0.1287 - val_accuracy: 0.9609
```

```
Epoch 4/10
```

```
1500/1500 [=====] - 7s 5ms/step - loss:
0.1051 - accuracy: 0.9671 - val_loss: 0.1225 - val_accuracy: 0.9620
```

```
Epoch 5/10
```

```
1500/1500 [=====] - 5s 3ms/step - loss:
0.0882 - accuracy: 0.9728 - val_loss: 0.1123 - val_accuracy: 0.9641
```

```
Epoch 6/10
1500/1500 [=====] - 6s 4ms/step - loss:
0.0763 - accuracy: 0.9759 - val_loss: 0.1122 - val_accuracy: 0.9676
Epoch 7/10
1500/1500 [=====] - 5s 3ms/step - loss:
0.0693 - accuracy: 0.9780 - val_loss: 0.1140 - val_accuracy: 0.9656
Epoch 8/10
1500/1500 [=====] - 5s 3ms/step - loss:
0.0588 - accuracy: 0.9811 - val_loss: 0.1151 - val_accuracy: 0.9672
Epoch 9/10
1500/1500 [=====] - 5s 3ms/step - loss:
0.0532 - accuracy: 0.9832 - val_loss: 0.1249 - val_accuracy: 0.9643
Epoch 10/10
1500/1500 [=====] - 5s 3ms/step - loss:
0.0448 - accuracy: 0.9851 - val_loss: 0.1161 - val_accuracy: 0.9672

y_prob = model.predict(X_test)

313/313 [=====] - 1s 2ms/step

y_pred = y_prob.argmax(axis=1)

from sklearn.metrics import accuracy_score
accuracy_score(y_test,y_pred)

0.968

plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])

[<matplotlib.lines.Line2D at 0x7eec5065d420>]
```

