

## practical-12-character

May 9, 2024

```
[ ]: 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 [=====] - 0s 0us/step
```

```
[ ]: X_test.shape
```

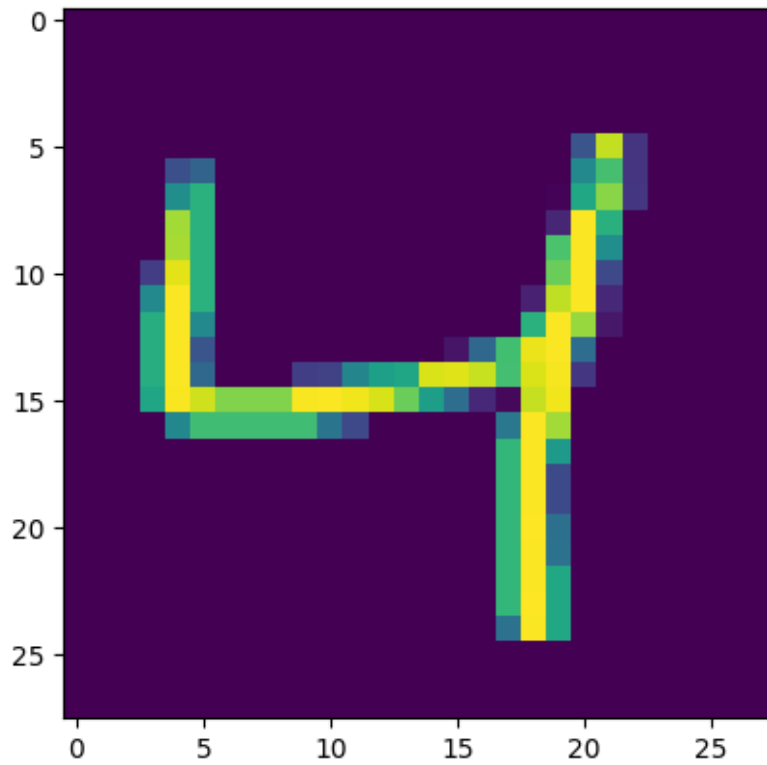
```
[ ]: (10000, 28, 28)
```

```
[ ]: y_train
```

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

```
[ ]: import matplotlib.pyplot as plt
      plt.imshow(X_train[2])
```

```
[ ]: <matplotlib.image.AxesImage at 0x7d33567e8160>
```



```
[ ]: X_train = X_train/255
      X_test = X_test/255
```

```
[ ]: X_train[0]
```

```
[ ]: array([[0., 0., 0., 0., 0., 0.,
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```

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```

```
[ ]: model = Sequential()

model.add(Flatten(input_shape=(28,28)))
model.add(Dense(128,activation='relu'))
model.add(Dense(32,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, 128)	100480
dense_1 (Dense)	(None, 32)	4128
dense_2 (Dense)	(None, 10)	330

```

=====
Total params: 104938 (409.91 KB)
Trainable params: 104938 (409.91 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=25,validation_split=0.2)
```

```

Epoch 1/25
1500/1500 [=====] - 8s 5ms/step - loss: 0.2850 -
accuracy: 0.9177 - val_loss: 0.1462 - val_accuracy: 0.9571
Epoch 2/25

```

1500/1500 [=====] - 6s 4ms/step - loss: 0.1228 - accuracy: 0.9634 - val\_loss: 0.1319 - val\_accuracy: 0.9593  
Epoch 3/25  
1500/1500 [=====] - 7s 4ms/step - loss: 0.0862 - accuracy: 0.9745 - val\_loss: 0.1081 - val\_accuracy: 0.9688  
Epoch 4/25  
1500/1500 [=====] - 5s 4ms/step - loss: 0.0643 - accuracy: 0.9794 - val\_loss: 0.1023 - val\_accuracy: 0.9728  
Epoch 5/25  
1500/1500 [=====] - 7s 5ms/step - loss: 0.0507 - accuracy: 0.9841 - val\_loss: 0.0941 - val\_accuracy: 0.9737  
Epoch 6/25  
1500/1500 [=====] - 5s 4ms/step - loss: 0.0395 - accuracy: 0.9876 - val\_loss: 0.1084 - val\_accuracy: 0.9706  
Epoch 7/25  
1500/1500 [=====] - 7s 4ms/step - loss: 0.0344 - accuracy: 0.9891 - val\_loss: 0.1202 - val\_accuracy: 0.9697  
Epoch 8/25  
1500/1500 [=====] - 6s 4ms/step - loss: 0.0266 - accuracy: 0.9912 - val\_loss: 0.1071 - val\_accuracy: 0.9748  
Epoch 9/25  
1500/1500 [=====] - 7s 5ms/step - loss: 0.0232 - accuracy: 0.9925 - val\_loss: 0.1168 - val\_accuracy: 0.9720  
Epoch 10/25  
1500/1500 [=====] - 5s 4ms/step - loss: 0.0219 - accuracy: 0.9926 - val\_loss: 0.1360 - val\_accuracy: 0.9711  
Epoch 11/25  
1500/1500 [=====] - 6s 4ms/step - loss: 0.0181 - accuracy: 0.9940 - val\_loss: 0.1127 - val\_accuracy: 0.9740  
Epoch 12/25  
1500/1500 [=====] - 5s 4ms/step - loss: 0.0157 - accuracy: 0.9950 - val\_loss: 0.1206 - val\_accuracy: 0.9753  
Epoch 13/25  
1500/1500 [=====] - 6s 4ms/step - loss: 0.0137 - accuracy: 0.9957 - val\_loss: 0.1224 - val\_accuracy: 0.9755  
Epoch 14/25  
1500/1500 [=====] - 6s 4ms/step - loss: 0.0134 - accuracy: 0.9953 - val\_loss: 0.1414 - val\_accuracy: 0.9732  
Epoch 15/25  
1500/1500 [=====] - 6s 4ms/step - loss: 0.0131 - accuracy: 0.9956 - val\_loss: 0.1207 - val\_accuracy: 0.9771  
Epoch 16/25  
1500/1500 [=====] - 7s 4ms/step - loss: 0.0092 - accuracy: 0.9971 - val\_loss: 0.1608 - val\_accuracy: 0.9729  
Epoch 17/25  
1500/1500 [=====] - 6s 4ms/step - loss: 0.0144 - accuracy: 0.9950 - val\_loss: 0.1378 - val\_accuracy: 0.9741  
Epoch 18/25

```

1500/1500 [=====] - 7s 4ms/step - loss: 0.0102 -
accuracy: 0.9965 - val_loss: 0.1451 - val_accuracy: 0.9754
Epoch 19/25
1500/1500 [=====] - 5s 4ms/step - loss: 0.0106 -
accuracy: 0.9958 - val_loss: 0.1571 - val_accuracy: 0.9737
Epoch 20/25
1500/1500 [=====] - 7s 4ms/step - loss: 0.0112 -
accuracy: 0.9960 - val_loss: 0.1470 - val_accuracy: 0.9747
Epoch 21/25
1500/1500 [=====] - 5s 4ms/step - loss: 0.0068 -
accuracy: 0.9979 - val_loss: 0.1562 - val_accuracy: 0.9753
Epoch 22/25
1500/1500 [=====] - 7s 4ms/step - loss: 0.0091 -
accuracy: 0.9971 - val_loss: 0.1507 - val_accuracy: 0.9773
Epoch 23/25
1500/1500 [=====] - 5s 4ms/step - loss: 0.0066 -
accuracy: 0.9978 - val_loss: 0.1569 - val_accuracy: 0.9783
Epoch 24/25
1500/1500 [=====] - 7s 5ms/step - loss: 0.0108 -
accuracy: 0.9963 - val_loss: 0.1622 - val_accuracy: 0.9758
Epoch 25/25
1500/1500 [=====] - 5s 4ms/step - loss: 0.0077 -
accuracy: 0.9974 - val_loss: 0.1702 - val_accuracy: 0.9745

```

```
[ ]: y_prob = model.predict(X_test)
```

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

```
[ ]: y_pred = y_prob.argmax(axis=1)
```

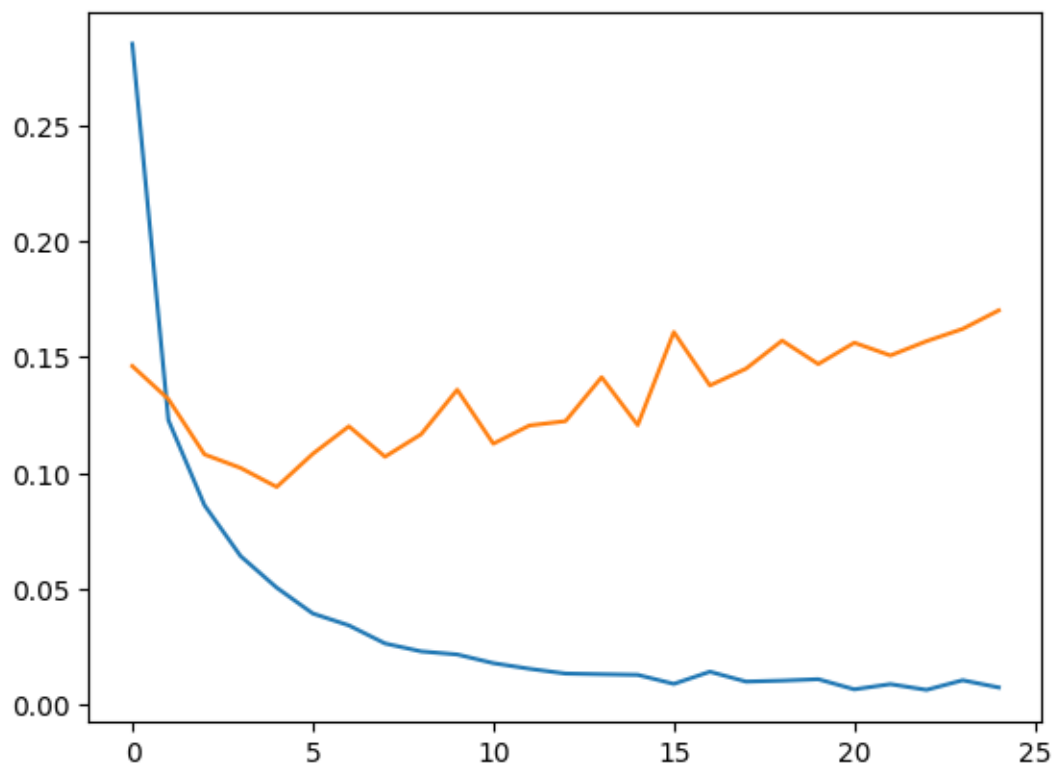
```
[ ]: from sklearn.metrics import accuracy_score
accuracy_score(y_test, y_pred)
```

```
[ ]: 0.9768
```

```
[ ]: plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
```

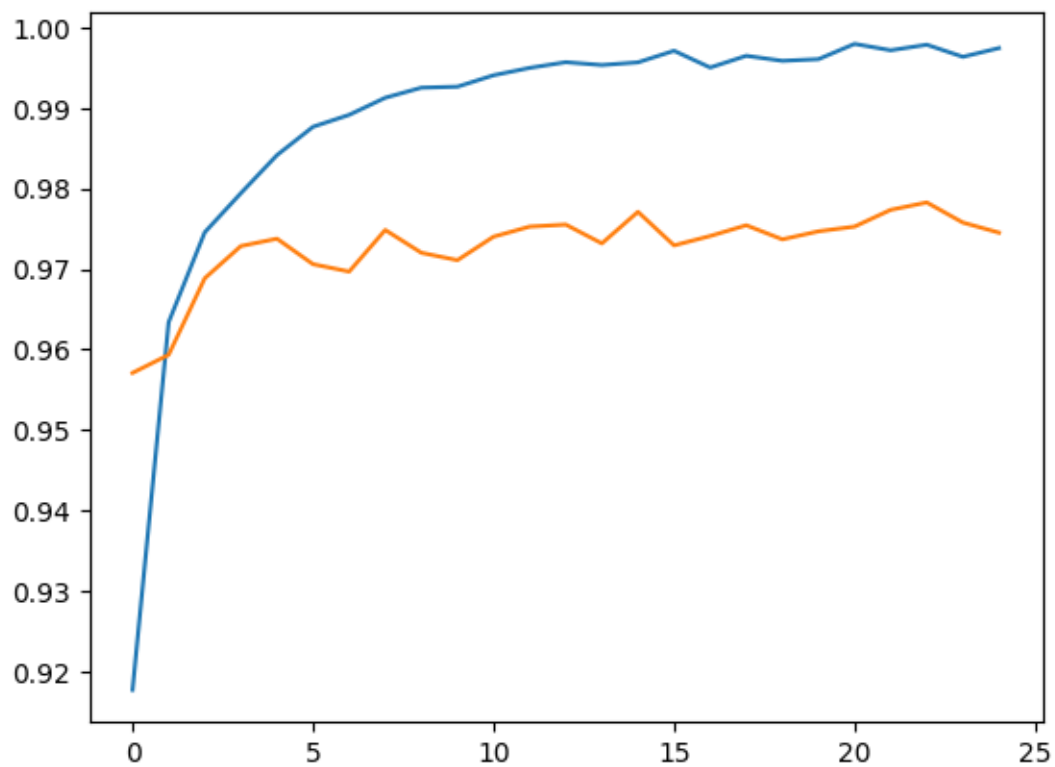
```
[ ]: [<matplotlib.lines.Line2D at 0x7d33507c1570>]
```





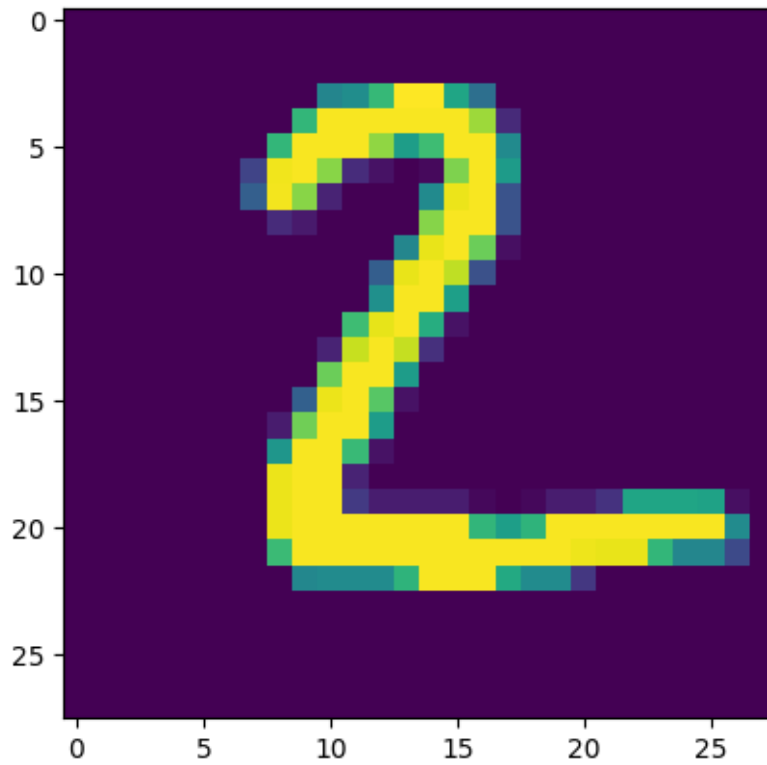
```
[ ]: plt.plot(history.history['accuracy'])  
      plt.plot(history.history['val_accuracy'])
```

```
[ ]: [<matplotlib.lines.Line2D at 0x7d3330a6d5d0>]
```



```
[ ]: plt.imshow(X_test[1])
```

```
[ ]: <matplotlib.image.AxesImage at 0x7d33303b3850>
```



```
[ ]: model.predict(X_test[1].reshape(1,28,28)).argmax(axis=1)
```

```
1/1 [=====] - 0s 89ms/step
```

```
[ ]: array([2])
```