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```
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()
```

```
[ ] X_test.shape
```

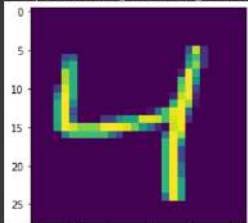
```
(10000, 28, 28)
```

```
[ ] 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 0x7f573ed9f790>
```



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

• `x_train[0]`

[illegible]

[illegible]

```
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\_5"

Layer (type)	Output Shape	Param #
flatten_4 (Flatten)	(None, 784)	0
dense_11 (Dense)	(None, 128)	100480
dense_12 (Dense)	(None, 32)	4128
dense_13 (Dense)	(None, 10)	330

=====  
Total params: 104,938  
Trainable params: 104,938  
Non-trainable params: 0

```
[ ] 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 [=====] - 5s 3ms/step - loss: 0.2845 - accuracy: 0.9168 - val_loss: 0.1425 - val_accuracy: 0.9570
Epoch 2/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.1164 - accuracy: 0.9638 - val_loss: 0.1258 - val_accuracy: 0.9605
Epoch 3/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0800 - accuracy: 0.9754 - val_loss: 0.0912 - val_accuracy: 0.9728
Epoch 4/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0594 - accuracy: 0.9817 - val_loss: 0.1014 - val_accuracy: 0.9697
Epoch 5/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0462 - accuracy: 0.9851 - val_loss: 0.1024 - val_accuracy: 0.9687
Epoch 6/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0377 - accuracy: 0.9881 - val_loss: 0.1060 - val_accuracy: 0.9712
Epoch 7/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0313 - accuracy: 0.9898 - val_loss: 0.0960 - val_accuracy: 0.9758
Epoch 8/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0283 - accuracy: 0.9902 - val_loss: 0.1061 - val_accuracy: 0.9732
Epoch 9/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0230 - accuracy: 0.9923 - val_loss: 0.1160 - val_accuracy: 0.9713
Epoch 10/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0179 - accuracy: 0.9939 - val_loss: 0.1134 - val_accuracy: 0.9761
Epoch 11/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0177 - accuracy: 0.9937 - val_loss: 0.1204 - val_accuracy: 0.9742
Epoch 12/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0156 - accuracy: 0.9947 - val_loss: 0.1103 - val_accuracy: 0.9761
Epoch 13/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0156 - accuracy: 0.9949 - val_loss: 0.1221 - val_accuracy: 0.9743
Epoch 14/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0121 - accuracy: 0.9960 - val_loss: 0.1246 - val_accuracy: 0.9764
Epoch 15/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0122 - accuracy: 0.9958 - val_loss: 0.1362 - val_accuracy: 0.9753
Epoch 16/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0100 - accuracy: 0.9964 - val_loss: 0.1340 - val_accuracy: 0.9772
Epoch 17/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0131 - accuracy: 0.9955 - val_loss: 0.1371 - val_accuracy: 0.9752
Epoch 18/25
1500/1500 [=====] - 4s 3ms/step - loss: 0.0099 - accuracy: 0.9969 - val_loss: 0.1330 - val_accuracy: 0.9760
```

```
[ ] y_prob = model.predict(x_test)

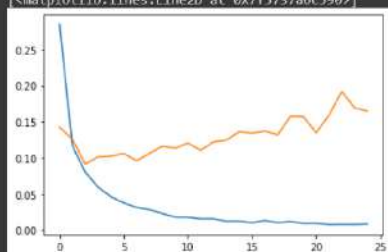
[ ] y_pred = y_prob.argmax(axis=1)

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

0.9755

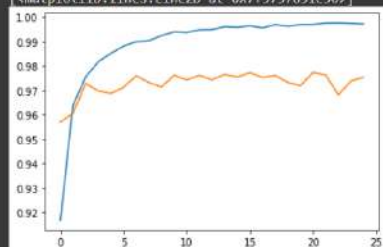
```
[ ] plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
```

[<matplotlib.lines.Line2D at 0x7f5737a6c590>]



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

<matplotlib.lines.Line2D at 0x7f5737831e50>



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

<matplotlib.image.AxesImage at 0x7f5736844c50>

