```
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_train
\rightarrow array([[[0, 0, 0, ..., 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, ..., 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0]],
             [[0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
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              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0]],
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              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
              . . . ,
              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, \ldots, 0, 0, 0],
              [0, 0, 0, ..., 0, 0, 0]]], dtype=uint8)
```

X_train.shape

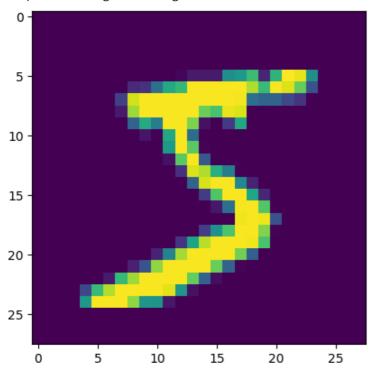
→ (60000, 28, 28)

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 0x78b687310990>



#scaling
X_train = X_train/255

 $X_{\text{test}} = X_{\text{test}}/255$

X_train[0]

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```
#Input features are a total of 28*28 = 784, but they're stored in a 2D manner before sending them
#We need to flatten the dataset (28,28) -> (1,784)
model = Sequential()
model.add(Flatten(input_shape = (28,28))) # Added the input Layer (layer0)
model.add(Dense(128, activation = 'relu')) #Layer1 = Hidden Layer
model.add(Dense(32, activation = 'relu')) #Layer2 = Hidden Layer
model.add(Dense(10, activation = 'softmax')) #Layer3 = Output Layer = 10 labels(0 to 9)
```

/usr/local/lib/python3.11/dist-packages/keras/src/layers/reshaping/flatten.py:37: UserWarning: super().__init__(**kwargs)

model.summary()

→ Model: "sequential_3"

Layer (type)	Output Shape	Param #
flatten_3 (Flatten)	(None, 784)	0
dense_6 (Dense)	(None, 128)	100,480
dense_7 (Dense)	(None, 32)	4,128
dense_8 (Dense)	(None, 10)	330

Total params: 104,938 (409.91 KB)
Trainable params: 104,938 (409.91 KB)
Non-trainable params: 0 (0.00 B)



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

history = model.fit(X_train, y_train, epochs = 15, validation_split = 0.2)

```
→ Epoch 1/15
    1500/1500
                                - 8s 5ms/step - accuracy: 0.9959 - loss: 0.0111 - val accuracy: 0.
    Epoch 2/15
                                - 10s 5ms/step - accuracy: 0.9969 - loss: 0.0100 - val_accuracy: 0
    1500/1500
    Epoch 3/15
    1500/1500 ·
                                - 7s 5ms/step - accuracy: 0.9969 - loss: 0.0097 - val_accuracy: 0.
    Epoch 4/15
    1500/1500
                                - 11s 5ms/step - accuracy: 0.9971 - loss: 0.0090 - val_accuracy: 0
    Epoch 5/15
                                 - 6s 4ms/step - accuracy: 0.9974 - loss: 0.0074 - val_accuracy: 0.
    1500/1500
    Epoch 6/15
                                - 8s 5ms/step - accuracy: 0.9976 - loss: 0.0079 - val_accuracy: 0.
    1500/1500
    Epoch 7/15
    1500/1500
                                 - 7s 4ms/step - accuracy: 0.9975 - loss: 0.0073 - val_accuracy: 0.
    Epoch 8/15
    1500/1500
                                - 10s 4ms/step - accuracy: 0.9965 - loss: 0.0106 - val_accuracy: 0
    Epoch 9/15
    1500/1500 -
                                - 13s 8ms/step - accuracy: 0.9975 - loss: 0.0070 - val_accuracy: 0
    Epoch 10/15
    1500/1500 -
                                - 10s 7ms/step - accuracy: 0.9981 - loss: 0.0060 - val_accuracy: 0
    Epoch 11/15
    1500/1500 -
                                - 10s 7ms/step - accuracy: 0.9970 - loss: 0.0080 - val_accuracy: 0
    Epoch 12/15
                              1500/1500 -
    Epoch 13/15
    1500/1500 ·
                                - 22s 9ms/step - accuracy: 0.9980 - loss: 0.0065 - val_accuracy: 0
    Epoch 14/15
    1500/1500 -
                                - 12s 8ms/step - accuracy: 0.9984 - loss: 0.0052 - val_accuracy: 0
    Epoch 15/15
    1500/1500 ·
                                - 15s    4ms/step - accuracy: 0.9972 - loss: 0.0090 - val_accuracy: 0
```

 $y_prob = model.predict(X_test)$ # returns a 2D Matrix, where each row coltains 10columns, each entit $Y_pred = y_prob.argmax(axis= 1)$ # return ns the index with the most prob for each row #Coincidentally each index == value of output too

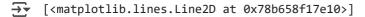
→ 313/313 — 1s 2ms/step

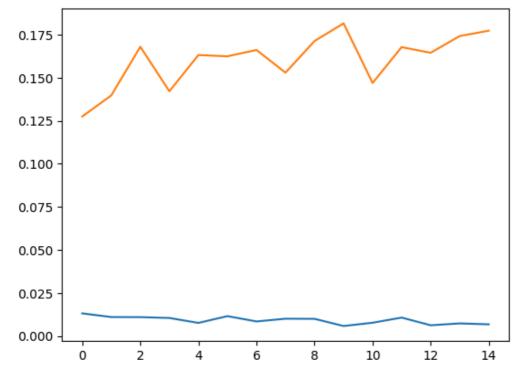
from sklearn.metrics import accuracy_score
accuracy_score(y_test, Y_pred)

→ 0.9772

```
import matplotlib.pyplot as plt
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
```

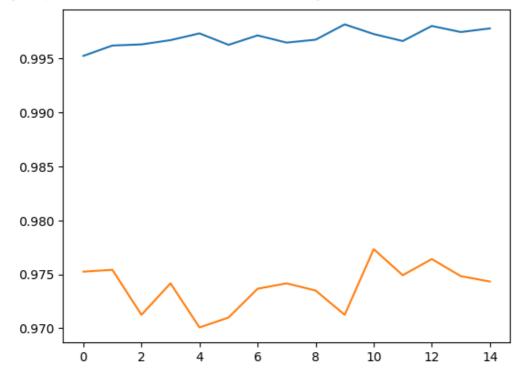
D



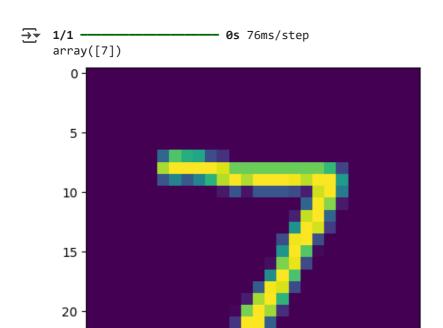


```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
```

[<matplotlib.lines.Line2D at 0x78b6fc5ed850>]



```
plt.imshow(X_test[0])
model.predict(X_test[0].reshape(1,28,28)).argmax(axis = 1)
```



Start coding or $\underline{\text{generate}}$ with AI.

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