In []: import tensorflow from tensorflow import keras from tensorflow.keras import Sequential from tensorflow.keras.layers import Dense,Flatten In []: (X_train, y_train), (X_test, y_test) = keras.datasets.mnist.load_data() In []: X_test.shape Out[]: (10000, 28, 28) In []: y_train Out[]: array([5, 0, 4, ..., 5, 6, 8], dtype=uint8) In []: import matplotlib.pyplot as plt plt.imshow(X_train[2]) Out[]: <matplotlib.image.AxesImage at 0x7f573ed9f790> 10 15 · 20 · 25 10 15 20 In []: X_train = X_train/255 $X_{test} = X_{test/255}$ In []: X_train[0] , 0. , 0. , 0. Out[]: array([[0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. Θ. , 0. , 0. , 0. , 0. Θ. , 0. , 0. , 0. Θ. , 0. , 0. , 0. Θ.], , 0. , 0. , 0. , 0. , 0. , 0. Θ. , 0. , 0. , 0. , 0. 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. Θ.], , 0. , 0. , 0. , 0. [0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. Θ. , 0. , 0. , 0. , 0. Θ. , 0.], , 0. , 0. , 0. , 0. [0. , 0. , 0. , 0. , 0. Θ. , 0. , 0. , 0. , 0. Θ. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. Θ.], , 0. [0. Θ. Θ. , 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. 0.07058824, 0.49411765, 0.53333333, 0.68627451, 0.10196078, 0.65098039, 1. , 0.96862745, 0.49803922, 0. 0. , 0. , 0.], , 0. 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.

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Out[]: array([2])

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

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