```
1 import tensorflow as tf
2 from tensorflow import keras
 3 from tensorflow.keras import layers
4 import numpy as np
1 (x_train, y_train), (x_test, y_test) = keras.datasets.mnist.load_data()
1 x_train = x_train.astype("float32") / 255.0
2 \times test = x_test.astype("float32") / 255.0
4 \times train = x train.reshape((-1, 28, 28, 1))
5 x_{test} = x_{test.reshape}((-1, 28, 28, 1))
7 y_train = keras.utils.to_categorical(y_train, num_classes=10)
 8 y_test = keras.utils.to_categorical(y_test, num_classes=10)
1 model = keras.Sequential(
2
      Γ
3
          layers.Conv2D(32, (3, 3), activation="relu", input_shape=(28, 28, 1)),
4
          layers.MaxPooling2D((2, 2)),
5
          layers.Conv2D(64, (3, 3), activation="relu"),
6
          layers.MaxPooling2D((2, 2)),
7
          layers.Flatten(),
8
           layers.Dense(128, activation="relu"),
9
          layers.Dense(10, activation="softmax"),
10
      ]
11 )
1 model.compile(
2
      optimizer="adam", loss="categorical crossentropy", metrics=["accuracy"]
3)
1 model.summary()
```

## → Model: "sequential\_1"

Layer (type)	Output Shape	Param #
conv2d_2 (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d_2 (MaxPooling2D)	(None, 13, 13, 32)	0
conv2d_3 (Conv2D)	(None, 11, 11, 64)	18,496
max_pooling2d_3 (MaxPooling2D)	(None, 5, 5, 64)	0
flatten_1 (Flatten)	(None, 1600)	0
dense_2 (Dense)	(None, 128)	204,928
dense_3 (Dense)	(None, 10)	1,290

Total params: 225,034 (879.04 KB)

1 history = model.fit(x train, y train, epochs=10, validation split=0.1)

```
→ Epoch 1/10

    1688/1688
                                 - 57s 32ms/step - accuracy: 0.9075 - loss: 0.3084 - val_accuracy: 0.9830 - val_loss: 0.0571
    Epoch 2/10
    1688/1688
                                 - 79s 31ms/step - accuracy: 0.9858 - loss: 0.0472 - val_accuracy: 0.9883 - val_loss: 0.0398
    Epoch 3/10
                                  - 50s 29ms/step - accuracy: 0.9907 - loss: 0.0282 - val_accuracy: 0.9885 - val_loss: 0.0418
    1688/1688
    Epoch 4/10
    1688/1688
                                  - 80s 28ms/step - accuracy: 0.9930 - loss: 0.0203 - val_accuracy: 0.9892 - val_loss: 0.0410
    Epoch 5/10
                                  - 84s 30ms/step - accuracy: 0.9956 - loss: 0.0142 - val_accuracy: 0.9908 - val_loss: 0.0357
    1688/1688
    Epoch 6/10
    1688/1688
                                  - 50s 30ms/step - accuracy: 0.9962 - loss: 0.0120 - val_accuracy: 0.9877 - val_loss: 0.0432
    Epoch 7/10
    1688/1688
                                  - 82s 30ms/step - accuracy: 0.9976 - loss: 0.0076 - val_accuracy: 0.9882 - val_loss: 0.0534
    Epoch 8/10
    1688/1688
                                 - 82s 30ms/step - accuracy: 0.9980 - loss: 0.0067 - val_accuracy: 0.9915 - val_loss: 0.0422
    Epoch 9/10
                                 — 83s 30ms/step - accuracy: 0.9984 - loss: 0.0049 - val_accuracy: 0.9907 - val_loss: 0.0402
```

```
Epoch 10/10
                                  – 80s 29ms/step - accuracy: 0.9981 - loss: 0.0055 - val_accuracy: 0.9915 - val_loss: 0.0438
    1688/1688
 1 loss, accuracy = model.evaluate(x_test, y_test)
 2 print(f"Test loss: {loss:.4f}")
 3 print(f"Test accuracy: {accuracy:.4f}")
<del>→</del>▼ 313/313 -
                               — 3s 9ms/step - accuracy: 0.9903 - loss: 0.0408
    Test loss: 0.0317
    Test accuracy: 0.9919
 1 predictions = model.predict(x_test)
 2 print(predictions)
<del>___</del> 313/313 -
                                - 4s 12ms/step
    [[2.8182216e-15 1.6171050e-13 3.4985261e-12 ... 9.9999994e-01
      1.6728505e-17 4.6382453e-11]
     [2.3480274e-13 1.3916217e-10 9.9999994e-01 ... 4.6401180e-16
      3.2465846e-17 1.1698152e-21]
     [2.4907302e-11 9.9999911e-01 5.7571814e-10 ... 5.6047583e-07
      1.9578926e-07 4.4505510e-11]
     [1.1599857e-21 1.1680350e-15 2.2768531e-18 ... 2.6072008e-14
      2.6681270e-11 3.2994381e-15]
     [4.4203076e-16 3.5325142e-20 2.6083705e-21 ... 9.2329536e-21
      4.9647345e-11 4.7253544e-18]
     [1.2013277e-12 3.6340803e-14 2.0375839e-12 ... 2.0544706e-19
      3.3551479e-13 1.8980761e-18]]
 1 from IPython.display import display
 2 from google.colab import files
 3 import numpy as np
 4 from tensorflow.keras.preprocessing import image
 1 def predict_image_class(model):
     """Predicts the class of an uploaded image using the given model."""
 2
     uploaded = files.upload()
 4
 6
    for fn in uploaded.keys():
 7
       # Preprocess the image
 8
       img path = fn
       img = image.load_img(img_path, target_size=(28, 28), color_mode='grayscale')
 9
10
       img_array = image.img_to_array(img)
11
       img_array = img_array / 255.0 # Normalize
12
       img_array = img_array.reshape((-1, 28, 28, 1)) # Reshape
13
14
       # Make the prediction
15
       prediction = model.predict(img array)
16
       predicted_class = np.argmax(prediction)
17
18
       # Get the class name
       class_names = ['T-shirt/top', 'Trouser', 'Pullover', 'Dress', 'Coat',
19
20
                       'Sandal', 'Shirt', 'Sneaker', 'Bag', 'Ankle boot']
21
       predicted_class_name = class_names[predicted_class]
22
23
       # Display the result
       print(f"The predicted class for {fn} is: {predicted_class_name}")
24
 1 # Call the function with your trained model
 2 predict_image_class(model)
    Choose Files images.jpg
    • images.jpg(image/jpeg) - 7477 bytes, last modified: 4/4/2025 - 100% done
    Saving images.jpg to images (1).jpg
    1/1 -
                            - 0s 38ms/step
    The predicted class for images (1).jpg is: T-shirt/top
```