```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout
from tensorflow.keras.optimizers import RMSprop
from tensorflow.keras.callbacks import EarlyStopping
import matplotlib.pyplot as plt
from tensorflow.keras.datasets import mnist
from tensorflow.keras.utils import to_categorical
(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_{train} = x_{train.reshape}(60000, 784).astype("float32") / 255
x_{\text{test}} = x_{\text{test.reshape}}(10000, 784).astype("float32") / 255
y_train = to_categorical(y_train, 10)
y_test = to_categorical(y_test, 10)
model_4 = Sequential()
model_4.add(Dense(400, activation="relu", input_shape=(784,)))
model_4.add(Dropout(0.4))
model 4.add(Dense(300, activation="relu"))
model_4.add(Dropout(0.4))
model_4.add(Dense(10, activation="softmax"))
Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz">https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz</a>
                                            - 0s Ous/step
     11490434/11490434
     /usr/local/lib/python3.11/dist-packages/keras/src/layers/core/dense.py:87: UserWarning: Do not pass an `input_shape`/`input_dim` arg
       super().__init__(activity_regularizer=activity_regularizer, **kwargs)
learning_rate = 0.003
model 4.compile(
    loss="categorical_crossentropy",
    optimizer=RMSprop(learning_rate=learning_rate),
    metrics=["accuracy"],
)
batch_size = 128
epochs = 20
history_4 = model_4.fit(
    x train.
    v train,
    batch_size=batch_size,
    epochs=epochs,
    verbose=1,
    validation_data=(x_test, y_test),
→ Epoch 1/20
     469/469
                                 — 8s 14ms/step - accuracy: 0.8302 - loss: 0.5392 - val_accuracy: 0.9635 - val_loss: 0.1245
     Enoch 2/20
     469/469
                                 - 9s 11ms/step - accuracy: 0.9505 - loss: 0.1723 - val_accuracy: 0.9684 - val_loss: 0.1111
     Epoch 3/20
     469/469
                                 — 11s 12ms/step - accuracy: 0.9595 - loss: 0.1414 - val_accuracy: 0.9746 - val_loss: 0.1005
     Epoch 4/20
     469/469
                                  - 6s 13ms/step - accuracy: 0.9628 - loss: 0.1318 - val_accuracy: 0.9752 - val_loss: 0.0894
     Epoch 5/20
     469/469
                                 - 9s 12ms/step - accuracy: 0.9692 - loss: 0.1179 - val_accuracy: 0.9763 - val_loss: 0.0955
     Epoch 6/20
     469/469
                                 - 10s 12ms/step - accuracy: 0.9694 - loss: 0.1130 - val_accuracy: 0.9777 - val_loss: 0.0952
     Epoch 7/20
     469/469
                                 - 6s 13ms/step - accuracy: 0.9717 - loss: 0.1079 - val accuracy: 0.9784 - val loss: 0.0934
     Fnoch 8/20
     469/469
                                 - 5s 11ms/step - accuracy: 0.9744 - loss: 0.0959 - val_accuracy: 0.9772 - val_loss: 0.0961
     Epoch 9/20
                                 - 6s 14ms/step - accuracy: 0.9767 - loss: 0.0911 - val_accuracy: 0.9814 - val_loss: 0.0863
     469/469
     Epoch 10/20
     469/469
                                 - 6s 12ms/step - accuracy: 0.9757 - loss: 0.0943 - val_accuracy: 0.9789 - val_loss: 0.1011
     Epoch 11/20
                                 – 10s 12ms/step - accuracy: 0.9755 - loss: 0.0958 - val_accuracy: 0.9793 - val_loss: 0.1007
     469/469
     Epoch 12/20
     469/469
                                 – 11s 14ms/step - accuracy: 0.9788 - loss: 0.0882 - val_accuracy: 0.9797 - val_loss: 0.1047
     Epoch 13/20
                                 – 10s 13ms/step - accuracy: 0.9773 - loss: 0.0910 - val_accuracy: 0.9811 - val_loss: 0.0969
     469/469 -
     Epoch 14/20
     469/469
                                 - 6s 12ms/step - accuracy: 0.9786 - loss: 0.0845 - val_accuracy: 0.9800 - val_loss: 0.1016
     Epoch 15/20
     469/469
                                 – 10s 12ms/step - accuracy: 0.9796 - loss: 0.0877 - val_accuracy: 0.9831 - val_loss: 0.0930
     Epoch 16/20
     469/469
                                 - 11s 13ms/step - accuracy: 0.9800 - loss: 0.0875 - val_accuracy: 0.9820 - val_loss: 0.0956
     Epoch 17/20
     469/469
                                 - 11s 14ms/step - accuracy: 0.9823 - loss: 0.0782 - val accuracy: 0.9805 - val loss: 0.1092
     Epoch 18/20
     469/469
                                 - 10s 13ms/step - accuracy: 0.9820 - loss: 0.0772 - val_accuracy: 0.9805 - val_loss: 0.1067
     Epoch 19/20
```

```
— 10s 12ms/step - accuracy: 0.9807 - loss: 0.0826 - val_accuracy: 0.9827 - val_loss: 0.0954
     469/469
     Epoch 20/20
     469/469
                                — 7s 14ms/step - accuracy: 0.9814 - loss: 0.0824 - val_accuracy: 0.9808 - val_loss: 0.1200
model_4.summary()
score = model 4.evaluate(x test, y test, verbose=0)
```

```
print("Model 4 - Test loss:", score[0])
print("Model 4 - Test accuracy:", score[1])
```

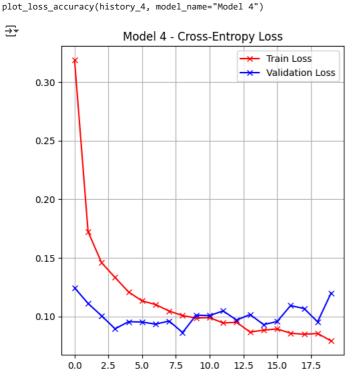
ax.set_title(f"{model_name} - Accuracy")

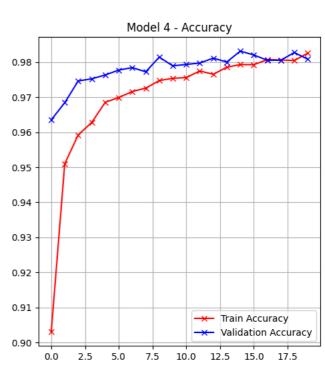
ax.grid(True)

→ Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 400)	314,000
dropout (Dropout)	(None, 400)	0
dense_1 (Dense)	(None, 300)	120,300
dropout_1 (Dropout)	(None, 300)	0
dense_2 (Dense)	(None, 10)	3,010

```
Total params: 874,622 (3.34 MB)
      Trainable params: 437,310 (1.67 MB)
      Non-trainable params: 0 (0.00 B)
      Optimizer params: 437,312 (1.67 MB)
     Model 4 - Test loss: 0.1200367659330368
     Model 4 - Test accuracy: 0.9807999730110168
def plot_loss_accuracy(history, model_name="Model"):
    fig = plt.figure(figsize=(12, 6))
    ax = fig.add_subplot(1, 2, 1)
    ax.plot(history.history["loss"], "r-x", label="Train Loss")
    ax.plot(history.history["val_loss"], "b-x", label="Validation Loss")
    ax.legend()
    ax.set_title(f"{model_name} - Cross-Entropy Loss")
    ax.grid(True)
    ax = fig.add_subplot(1, 2, 2)
acc_key = "accuracy" if "accuracy" in history.history else "acc"
    val acc key = "val accuracy" if "val accuracy" in history.history else "val acc"
    ax.plot(history.history[acc_key], "r-x", label="Train Accuracy")
    ax.plot(history.history[val\_acc\_key], \ "b-x", \ label="Validation \ Accuracy")\\
    ax.legend()
```





Start coding or generate with AI.