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import tensorflow as tf
from tensorflow.keras.datasets import mnist
from sklearn.model_selection import KFold

# Load MNIST dataset
(x_train, y_train), (x_test, y_test) = mnist.load_data()

# Preprocess data
x_train = x_train.reshape(x_train.shape[0], 28, 28, 1).astype('float32') / 255.0
x_test = x_test.reshape(x_test.shape[0], 28, 28, 1).astype('float32') / 255.0

# Define CNN architecture
def create_model():
    model = tf.keras.Sequential([
        tf.keras.layers.Conv2D(32, kernel_size=(3, 3), activation='relu', input_shape=(28, 28, 1)),
        tf.keras.layers.MaxPooling2D(pool_size=(2, 2)),
        tf.keras.layers.Conv2D(64, kernel_size=(3, 3), activation='relu'),
        tf.keras.layers.MaxPooling2D(pool_size=(2, 2)),
        tf.keras.layers.Conv2D(128, kernel_size=(3, 3), activation='relu'),
        tf.keras.layers.Flatten(),
        tf.keras.layers.Dense(128, activation='relu'),
        tf.keras.layers.Dense(10, activation='softmax')
    ])
    return model

# K-Fold Cross-Validation
kfold = KFold(n_splits=5, shuffle=True)
accuracy, loss = [], []

for train_index, val_index in kfold.split(x_train):
    x_train_fold, x_val_fold = x_train[train_index], x_train[val_index]
    y_train_fold, y_val_fold = y_train[train_index], y_train[val_index]

    model = create_model()
    model.compile(loss='sparse_categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    model.fit(x_train_fold, y_train_fold, epochs=3, validation_data=(x_val_fold, y_val_fold))

    # Evaluate model
    val_loss, val_acc = model.evaluate(x_val_fold, y_val_fold)
    accuracy.append(val_acc)
    loss.append(val_loss)

# Print K-Fold results
print("Average Accuracy:", sum(accuracy) / len(accuracy))
print("Average Loss:", sum(loss) / len(loss))
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Epoch 1/3  
1500/1500 [=====] - 59s 38ms/step - loss: 0.1464 - accuracy: 0.9540 - val\_loss: 0.0731 - val\_accuracy: 0.9778  
Epoch 2/3  
1500/1500 [=====] - 47s 31ms/step - loss: 0.0456 - accuracy: 0.9857 - val\_loss: 0.0449 - val\_accuracy: 0.9861  
Epoch 3/3  
1500/1500 [=====] - 49s 33ms/step - loss: 0.0329 - accuracy: 0.9895 - val\_loss: 0.0344 - val\_accuracy: 0.9902  
375/375 [=====] - 4s 11ms/step - loss: 0.0344 - accuracy: 0.9902  
Epoch 1/3  
1500/1500 [=====] - 49s 32ms/step - loss: 0.1444 - accuracy: 0.9561 - val\_loss: 0.0713 - val\_accuracy: 0.9756  
Epoch 2/3  
1500/1500 [=====] - 49s 32ms/step - loss: 0.0455 - accuracy: 0.9860 - val\_loss: 0.0358 - val\_accuracy: 0.9892  
Epoch 3/3  
1500/1500 [=====] - 45s 30ms/step - loss: 0.0333 - accuracy: 0.9896 - val\_loss: 0.0315 - val\_accuracy: 0.9908  
375/375 [=====] - 4s 12ms/step - loss: 0.0315 - accuracy: 0.9908  
Epoch 1/3  
1500/1500 [=====] - 50s 32ms/step - loss: 0.1463 - accuracy: 0.9544 - val\_loss: 0.0540 - val\_accuracy: 0.9835  
Epoch 2/3  
1500/1500 [=====] - 51s 34ms/step - loss: 0.0454 - accuracy: 0.9856 - val\_loss: 0.0526 - val\_accuracy: 0.9844  
Epoch 3/3  
1500/1500 [=====] - 49s 33ms/step - loss: 0.0307 - accuracy: 0.9898 - val\_loss: 0.0468 - val\_accuracy: 0.9859  
375/375 [=====] - 3s 8ms/step - loss: 0.0468 - accuracy: 0.9859  
Epoch 1/3  
1500/1500 [=====] - 50s 33ms/step - loss: 0.1487 - accuracy: 0.9534 - val\_loss: 0.0586 - val\_accuracy: 0.9822  
Epoch 2/3  
1500/1500 [=====] - 48s 32ms/step - loss: 0.0460 - accuracy: 0.9858 - val\_loss: 0.0388 - val\_accuracy: 0.9885  
Epoch 3/3  
1500/1500 [=====] - 49s 33ms/step - loss: 0.0331 - accuracy: 0.9894 - val\_loss: 0.0363 - val\_accuracy: 0.9893  
375/375 [=====] - 3s 8ms/step - loss: 0.0363 - accuracy: 0.9893  
Epoch 1/3  
1500/1500 [=====] - 48s 31ms/step - loss: 0.1465 - accuracy: 0.9551 - val\_loss: 0.0517 - val\_accuracy: 0.9837  
Epoch 2/3  
1500/1500 [=====] - 48s 32ms/step - loss: 0.0465 - accuracy: 0.9853 - val\_loss: 0.0381 - val\_accuracy: 0.9887  
Epoch 3/3  
1500/1500 [=====] - 48s 32ms/step - loss: 0.0320 - accuracy: 0.9896 - val\_loss: 0.0424 - val\_accuracy: 0.9859  
375/375 [=====] - 3s 9ms/step - loss: 0.0424 - accuracy: 0.9859  
Average Accuracy: 0.9884166717529297  
Average Loss: 0.0382773831486702