```
import numpy as np
import tensorflow as tf
from keras.models import Sequential
from keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout, BatchNormalization, GlobalAveragePooling2D
from tensorflow.keras.preprocessing.image import ImageDataGenerator
# Initialize the sequential model
model custom = Sequential()
# First convolutional layer
model_custom.add(Conv2D(32, (3, 3), input_shape=(256, 256, 3), activation='relu'))
model_custom.add(BatchNormalization())
model custom.add(MaxPooling2D(pool_size=(2, 2)))
model_custom.add(Dropout(0.25))
# Second convolutional layer
model_custom.add(Conv2D(64, (3, 3), activation='relu'))
model custom.add(BatchNormalization())
model_custom.add(MaxPooling2D(pool_size=(2, 2)))
model_custom.add(Dropout(0.25))
# Third convolutional layer
model_custom.add(Conv2D(128, (3, 3), activation='relu'))
model_custom.add(BatchNormalization())
model_custom.add(MaxPooling2D(pool_size=(2, 2)))
model custom.add(Dropout(0.25))
# Fourth convolutional layer
model_custom.add(Conv2D(128, (3, 3), activation='relu'))
model_custom.add(BatchNormalization())
model_custom.add(MaxPooling2D(pool_size=(2, 2)))
model_custom.add(Dropout(0.25))
# Fifth convolutional laver
model_custom.add(Conv2D(256, (3, 3), activation='relu'))
model custom.add(BatchNormalization())
model_custom.add(MaxPooling2D(pool_size=(2, 2)))
model_custom.add(Dropout(0.25))
# Global average pooling
model_custom.add(GlobalAveragePooling2D())
# Dense layers
model_custom.add(Dense(256, activation='relu'))
model_custom.add(Dropout(0.5))
model_custom.add(Dense(128, activation='relu'))
model custom.add(Dropout(0.5))
model_custom.add(Dense(3, activation='softmax'))
# Compile the model
compilation = model\_custom.compile(optimizer='adam', loss='categorical\_crossentropy', metrics=['accuracy','confusion\_metics'])
# Summary of the model
model custom.summary()
# Data augmentation and normalization
train_datagen = ImageDataGenerator(rescale=1./255,
   shear_range=0.2,
   zoom range=0.2,
   horizontal_flip=True,
   rotation_range=40,
   width_shift_range=0.2,
   height_shift_range=0.2,
   brightness_range=[0.5, 1.3],
    fill_mode='nearest')
val_datagen = ImageDataGenerator(rescale=1./255)
test_datagen = ImageDataGenerator(rescale=1./255)
# Load the training, validation, and test datasets
validation_set = val_datagen.flow_from_directory('/content/drive/MyDrive/potato_leaf_disease_classification/Validation', target_size=(256
test_set = test_datagen.flow_from_directory('/content/drive/MyDrive/potato_leaf_disease_classification/Test', target_size=(256, 256), bat
# Train the model
history = model_custom.fit(training_set, steps_per_epoch=20, epochs=50, validation_data=validation_set, validation_steps=20)
# Evaluate the model
# Evaluate the model on the test set
```

```
# Print the results
#print("\n\n")
#print("Test Loss: \t", test_loss, "\n")
#print("Test Accuracy: \t", test_acc, "\n")
#print("Test Precision: \t", test_precision, "\n")
#print("Test Recall: \t", test_recall, "\n")
#print("Test F1 Score: \t", test_f1, "\n")
test_loss, test_acc = model_custom.evaluate(test_set, verbose=2)
print("\n\n")
print("Test Loss: \t", test_loss, "\n")
print("Test Accuracy: \t", test_acc, "\n")
```

/usr/local/lib/python3.10/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape super().__init__(activity_regularizer=activity_regularizer, **kwargs)

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 254, 254, 32)	896
batch_normalization (BatchNormalization)	(None, 254, 254, 32)	128
max_pooling2d (MaxPooling2D)	(None, 127, 127, 32)	0
dropout (Dropout)	(None, 127, 127, 32)	0
conv2d_1 (Conv2D)	(None, 125, 125, 64)	18,496
batch_normalization_1 (BatchNormalization)	(None, 125, 125, 64)	256
max_pooling2d_1 (MaxPooling2D)	(None, 62, 62, 64)	0
dropout_1 (Dropout)	(None, 62, 62, 64)	0
conv2d_2 (Conv2D)	(None, 60, 60, 128)	73,856
batch_normalization_2 (BatchNormalization)	(None, 60, 60, 128)	512
max_pooling2d_2 (MaxPooling2D)	(None, 30, 30, 128)	0
dropout_2 (Dropout)	(None, 30, 30, 128)	0
conv2d_3 (Conv2D)	(None, 28, 28, 128)	147,584
batch_normalization_3 (BatchNormalization)	(None, 28, 28, 128)	512
max_pooling2d_3 (MaxPooling2D)	(None, 14, 14, 128)	0
dropout_3 (Dropout)	(None, 14, 14, 128)	0
conv2d_4 (Conv2D)	(None, 12, 12, 256)	295,168
batch_normalization_4 (BatchNormalization)	(None, 12, 12, 256)	1,024
max_pooling2d_4 (MaxPooling2D)	(None, 6, 6, 256)	0
dropout_4 (Dropout)	(None, 6, 6, 256)	0
global_average_pooling2d (GlobalAveragePooling2D)	(None, 256)	0
dense (Dense)	(None, 256)	65,792
dropout_5 (Dropout)	(None, 256)	0
dense_1 (Dense)	(None, 128)	32,896

```
y true = test set.classes
cm = confusion_matrix(y_true, y_pred_classes)
print("Confusion Matrix:\n", cm)
report = classification_report(y_true, y_pred_classes, target_names=test_set.class_indices.keys())
print("Classification Report:\n", report)
    27/27 -
                                 - 2s 66ms/step
     Confusion Matrix:
       [[109 70 21]
       [113 80 7]
       [ 18 12 0]]
     Classification Report:
                               precision
                                            recall f1-score support
     Potato___Early_blight
                                   0.45
                                              0.55
                                                         0.50
                                                                     200
      Potato___Late_blight
                                   0.49
                                              9.49
                                                         0.44
                                                                     200
          Potato___healthy
                                   0.00
                                              0.00
                                                         0.00
                                                                      30
                                                         0.44
                   accuracy
                                                                     430
                  macro avg
                                    0.32
                                              0.32
                                                         0.31
                                                                      430
               weighted avg
                                    0.44
                                              0.44
                                                         0.44
                                                                     430
                                - 6s 328ms/step - accuracv: 0.8329 - loss: 0.4219 - val accuracv: 0.5455 - val loss: 2.7436
results = model_custom.evaluate(test_set, verbose=2)
# Extract loss and accuracy
test loss = results[0]
test_acc = results[1]
# Predict probabilities for the test set
y_pred_proba = model_custom.predict(test_set)
# Get predicted classes
y_pred = np.argmax(y_pred_proba, axis=1)
# Get true classes
y true = test set.classes
# Calculate precision, recall, and F1-score
# Note: 'macro' average calculates metrics globally by considering each class independently
from sklearn.metrics import precision_score, recall_score, f1_score # Import necessary metrics
test_precision = precision_score(y_true, y_pred, average='macro')
test_recall = recall_score(y_true, y_pred, average='macro')
test_f1 = f1_score(y_true, y_pred, average='macro')
# Print the results
print("\n\n")
print("Test Loss: \t", test_loss, "\n")
print("Test Accuracy: \t", test_acc, "\n")
print("Test Precision: \t", test_precision, "\n")
print("Test Recall: \t", test_recall, "\n")
print("Test F1 Score: \t", test_f1, "\n")
    27/27 - 2s - 68ms/step - accuracy: 0.8744 - loss: 0.3023
                                -- 2s 65ms/step
                       0.30227696895599365
     Test Loss:
     Test Accuracy: 0.8744186162948608
                                0.33628747795414465
     Test Precision:
     Test Recall:
                       0.337777777777778
     Test F1 Score: 0.33392854050122106
     Enach 40/F0
pip install tensorflow
     Requirement already satisfied: tensorflow in /usr/local/lib/python3.10/dist-packages (2.17.0)
     Requirement already satisfied: absl-py>=1.0.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.4.0)
     Requirement already satisfied: astunparse>=1.6.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.6.3)
     Requirement already satisfied: flatbuffers>=24.3.25 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (24.3.25)
     Requirement already satisfied: gast!=0.5.0,!=0.5.1,!=0.5.2,>=0.2.1 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0.6
     Requirement already satisfied: google-pasta>=0.1.1 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0.2.0)
     Requirement already satisfied: h5py>=3.10.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (3.11.0)
     Requirement already satisfied: libclang>=13.0.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (18.1.1)
     Requirement already satisfied: ml-dtypes<0.5.0,>=0.3.1 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0.4.0)
     Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (3.3.0)
     Requirement already satisfied: packaging in /usr/local/lib/python3.10/dist-packages (from tensorflow) (24.1)

Requirement already satisfied: protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.3,!=4.21.4,!=4.21.5,<5.0.0dev,>=3.20.3 in /usr/local/lib/py

Requirement already satisfied: requests<3,>=2.21.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (2.32.3)
```

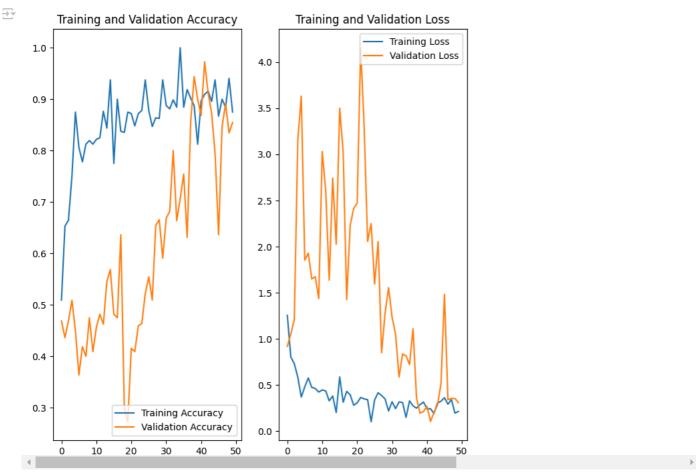
```
Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-packages (from tensorflow) (71.0.4)
Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.16.0)
Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (2.4.0)
Requirement already satisfied: typing-extensions>=3.6.6 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (4.12.2)
Requirement already satisfied: wrapt>=1.11.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.16.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.64.1)
Requirement already satisfied: tensorboard<2.18,>=2.17 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (2.17.0)
Requirement already satisfied: keras>=3.2.0 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (3.4.1)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (0
Requirement already satisfied: numpy<2.0.0,>=1.23.5 in /usr/local/lib/python3.10/dist-packages (from tensorflow) (1.26.4)
Requirement already satisfied: wheel<1.0,>=0.23.0 in /usr/local/lib/python3.10/dist-packages (from astunparse>=1.6.0->tensorflow) (@
Requirement already satisfied: rich in /usr/local/lib/python3.10/dist-packages (from keras>=3.2.0->tensorflow) (13.7.1)
Requirement already \ satisfied: \ namex in \ /usr/local/lib/python 3.10/dist-packages \ (from \ keras>= 3.2.0-) tensorflow) \ (0.0.8)
Requirement already satisfied: optree in /usr/local/lib/python3.10/dist-packages (from keras>=3.2.0->tensorflow) (0.12.1)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensor
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensorflow) (3.7)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensorflow)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.21.0->tensorflow)
Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.10/dist-packages (from tensorboard<2.18,>=2.17->tensorflow
Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0 in /usr/local/lib/python3.10/dist-packages (from tensorboard<2 Requirement already satisfied: werkzeug>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from tensorboard<2.18,>=2.17->tensorflow
Requirement already satisfied: MarkupSafe>=2.1.1 in /usr/local/lib/python3.10/dist-packages (from werkzeug>=1.0.1->tensorboard<2.18,
Requirement already satisfied: markdown-it-py>=2.2.0 in /usr/local/lib/python3.10/dist-packages (from rich->keras>=3.2.0->tensorflow
Requirement already satisfied: pygments < 3.0.0, >= 2.13.0 in /usr/local/lib/python 3.10/dist-packages (from rich->keras>= 3.2.0->tensorflue for the control of the contr
Requirement already satisfied: mdurl~=0.1 in /usr/local/lib/python3.10/dist-packages (from markdown-it-py>=2.2.0->rich->keras>=3.2.0
```

```
from google.colab import drive
drive.mount('/content/drive')

>>> Drive already mounted at /
```

From Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force remount=True).

```
from matplotlib import pyplot as plt
FPOCHS = 50
acc = history.history['accuracy']
val acc = history.history['val accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
plt.figure(figsize=(8, 8))
plt.subplot(1, 2, 1)
plt.plot(range(EPOCHS), acc, label='Training Accuracy')
plt.plot(range(EPOCHS), val_acc, label='Validation Accuracy')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')
plt.subplot(1, 2, 2)
plt.plot(range(EPOCHS), loss, label='Training Loss')
plt.plot(range(EPOCHS), val_loss, label='Validation Loss')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')
plt.show()
```



model custom.save('final model.h5') 🚌 WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is c Start coding or generate with AI. import numpy as np from keras.preprocessing import image import keras.utils as image # Load and preprocess the test image test_image = image.load_img('/content/drive/MyDrive/Potato_Healthy.jpg', target_size=(256, 256)) test_image = image.img_to_array(test_image) test_image = np.expand_dims(test_image, axis=0) # Predict the class of the test image result = model_custom.predict(test_image) print(result) # Assuming the class indices mapping # training_set.class_indices will return a dictionary like {'class1': 0, 'class2': 1, 'class3': 2} class_indices = training_set.class_indices $class_labels = \{v: \ k \ for \ k, \ v \ in \ class_indices. items()\} \\ \# \ Reverse \ the \ dictionary \ to \ get \ labels \ from \ indices$ # Get the predicted class index predicted_class_index = np.argmax(result, axis=1)[0] prediction = class_labels[predicted_class_index] print(prediction) 1/1 -- 0s 17ms/sten [[0.0000000e+00 1.0000000e+00 1.2121233e-18]] Potato___Late_blight

test_image = image.load_img('/content/drive/MyDrive/test_potato_early_blight.jpg', target_size=(256, 256))

import numpy as np

import keras.utils as image

from keras.preprocessing import image

Load and preprocess the test image

test_image = image.img_to_array(test_image)
test_image = np.expand_dims(test_image, axis=0)

```
# Predict the class of the test image
result = model_custom.predict(test_image)
print(result)

# Assuming the class indices mapping
# training_set.class_indices will return a dictionary like {'class1': 0, 'class2': 1, 'class3': 2}
class indices = training set.class indices
```