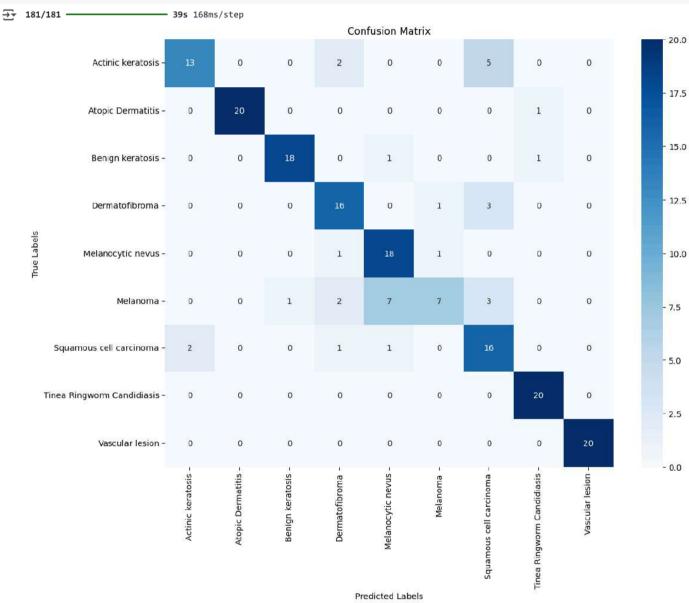
```
# -*- coding: utf-8 -*-
from google.colab import drive
drive.mount('/content/drive')

→ Mounted at /content/drive

import numpy as np
import tensorflow as tf
from tensorflow.keras.applications import DenseNet121
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Dense, GlobalAveragePooling2D, Dropout
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.callbacks import EarlyStopping, ReduceLROnPlateau
from sklearn.utils import class_weight
import os
# Set random seed
tf.random.set seed(42)
np.random.seed(42)
# === Data Paths ===
train_dir = '/content/drive/MyDrive/train'
val_dir = '/content/drive/MyDrive/val'
test_dir = '/content/drive/MyDrive/val'
img_size = (224, 224)
batch_size = 32
# === Data Augmentation ===
train_datagen = ImageDataGenerator(
    rescale=1./255,
    rotation_range=25,
    width shift range=0.2,
    height shift range=0.2,
    zoom_range=0.2,
    horizontal flip=True,
    vertical_flip=True
val_datagen = ImageDataGenerator(rescale=1./255)
test_datagen = ImageDataGenerator(rescale=1./255)
train_gen = train_datagen.flow_from_directory(
    train dir,
    target_size=img_size,
    batch_size=batch_size,
    class_mode='categorical'
val_gen = val_datagen.flow_from_directory(
    val_dir,
    target_size=img_size,
    batch_size=batch_size,
    class_mode='categorical'
)
test_gen = test_datagen.flow_from_directory(
    test_dir,
    target_size=img_size,
    batch_size=1,
    class_mode='categorical',
    shuffle=False
Found 697 images belonging to 9 classes.
     Found 181 images belonging to 9 classes.
     Found 181 images belonging to 9 classes.
# === Class Weights ===
labels = train_gen.classes
class_weights = class_weight.compute_class_weight(
    class_weight='balanced',
    classes=np.unique(labels),
class_weights = dict(enumerate(class_weights))
# === Build Model ===
base_model = DenseNet121(include_top=False, weights='imagenet', input_shape=(224, 224, 3))
```

```
base model.trainable = True # Unfreeze all layers
x = base_model.output
x = GlobalAveragePooling2D()(x)
x = Dropout(0.4)(x)
predictions = Dense(train_gen.num_classes, activation='softmax')(x)
model = Model(inputs=base_model.input, outputs=predictions)
model.compile(
    optimizer=tf.keras.optimizers.Adam(learning_rate=1e-4),
    loss='categorical crossentropy',
    metrics=['accuracy']
)
<del>J</del>₹
    Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/densenet/densenet121 weights tf dim ordering tf |
     29084464/29084464
                                            - 0s Ous/sten
# === Callbacks ===
early_stop = EarlyStopping(monitor='val_loss', patience=5, restore_best_weights=True)
reduce_lr = ReduceLROnPlateau(monitor='val_loss', patience=3, factor=0.2, min_lr=1e-6)
# === Train ===
history = model.fit(
   train_gen,
    epochs=30,
    validation_data=val_gen,
    class weight=class weights.
    callbacks=[early_stop, reduce_lr]
)
/usr/local/lib/python3.11/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` ci
       self._warn_if_super_not_called()
     Epoch 1/30
     22/22 -
                               - 720s 28s/step - accuracy: 0.2113 - loss: 2.4315 - val_accuracy: 0.4144 - val_loss: 1.6041 - learning_rate
     Epoch 2/30
     22/22
                               - 546s 25s/step - accuracy: 0.6306 - loss: 1.1169 - val_accuracy: 0.5249 - val_loss: 1.3003 - learning_rate
     Epoch 3/30
                               - 571s 26s/step - accuracy: 0.7561 - loss: 0.6954 - val accuracy: 0.6077 - val loss: 1.1616 - learning rate
     22/22
     Epoch 4/30
                               – 537s 25s/step - accuracy: 0.7746 - loss: 0.6048 - val_accuracy: 0.6409 - val_loss: 1.0738 - learning_rat։
     22/22 -
     Epoch 5/30
     22/22 -
                               – 535s 24s/step - accuracy: 0.8429 - loss: 0.4714 - val_accuracy: 0.6740 - val_loss: 0.9532 - learning_rat։
     Epoch 6/30
     22/22 ·
                              <mark>– 533s</mark> 24s/step - accuracy: 0.8786 - loss: 0.3715 - val_accuracy: 0.7182 - val_loss: 0.8194 - learning_rat@
     Epoch 7/30
     22/22 -
                               – 564s 25s/step - accuracy: 0.8820 - loss: 0.3274 - val_accuracy: 0.7514 - val_loss: 0.6590 - learning_rat։
     Epoch 8/30
     22/22 -
                              − 536s 24s/step - accuracy: 0.9233 - loss: 0.2161 - val accuracy: 0.7956 - val loss: 0.5665 - learning rate
     Epoch 9/30
                              — 538s 25s/step - accuracy: 0.9063 - loss: 0.2668 - val_accuracy: 0.8177 - val_loss: 0.5197 - learning_rat։
     22/22 .
     Epoch 10/30
     22/22 -
                               - 536s 24s/step - accuracy: 0.9441 - loss: 0.2051 - val accuracy: 0.8066 - val loss: 0.6335 - learning rate
     Epoch 11/30
     22/22 -
                              — 584s 25s/step - accuracy: 0.9294 - loss: 0.1917 - val_accuracy: 0.8066 - val_loss: 0.6492 - learning_rat։
     Epoch 12/30
     22/22 -
                               - 542s 25s/step - accuracy: 0.9618 - loss: 0.1271 - val_accuracy: 0.7790 - val_loss: 0.6988 - learning_rate
     Epoch 13/30
     22/22
                               - 528s 24s/step - accuracy: 0.9377 - loss: 0.1798 - val_accuracy: 0.7956 - val_loss: 0.5992 - learning_rat։
     Epoch 14/30
                               - 542s 25s/step - accuracy: 0.9642 - loss: 0.1104 - val_accuracy: 0.7956 - val_loss: 0.5550 - learning_rato
     22/22 -
# === Evaluate ===
loss, acc = model.evaluate(test gen)
print(f"Final Test Accuracy: {acc * 100:.2f}%")
→ 181/181 -
                                 - 30s 167ms/step - accuracy: 0.7827 - loss: 0.6292
     Final Test Accuracy: 81.77%
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import classification_report, confusion_matrix
# Step 1: Get predictions
Y_pred = model.predict(test_gen, steps=test_gen.samples)
y_pred = np.argmax(Y_pred, axis=1)
# Step 2: True labels
y_true = test_gen.classes
```

```
# Step 3: Confusion Matrix
cm = confusion_matrix(y_true, y_pred)
class_names = list(test_gen.class_indices.keys())
# Step 4: Plot confusion matrix
plt.figure(figsize=(12, 10))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues',
            xticklabels=class_names,
            yticklabels=class_names)
plt.xlabel('Predicted Labels')
plt.ylabel('True Labels')
plt.title('Confusion Matrix')
plt.tight_layout()
plt.show()
# Optional: Classification Report
print("Classification Report:\n")
\verb|print(classification_report(y_true, y_pred, target_names=class_names))| \\
```



Classification Report:

precision recall f1-score support

Actinic keratosis 0.87 0.65 0.74 20

=== Save Model ===
model.save('/content/drive/MyDrive/final_project/DenseNet_81PercentageAccuracy.h5')

 accuracy
 0.82
 181

 macro avg
 0.83
 0.82
 0.81
 181

 weighted avg
 0.83
 0.82
 0.81
 181