

---

# **A deep learning pipeline for medical image classification on mobile phones**

***Release 1.0***

**Muhammad Muneeb and Samuel F. Feng**

**Mar 14, 2022**



# CONTENTS

<b>1</b>	<b>Content</b>	<b>1</b>
1.1	Find a dataset . . . . .	1
1.2	Cancer classification . . . . .	1
1.3	Covid classification . . . . .	13
1.4	Tree classification . . . . .	33



## CONTENT

### 1.1 Find a dataset

There are many websites from which medical/plants/objects data can be downloaded.

See (Medical) [This Link](#).

See (Medical) [This Link](#).

See (Plants) [This Link](#).

Sample datasets used to demonstrate the working of this project are mentioned in the paper.

#### 1.1.1 Directory form.

Name	Date modified	Type	Size
ADI	3/21/2018 3:25 ...	File folder	
BACK	3/21/2018 3:27 ...	File folder	
DEB	3/21/2018 3:27 ...	File folder	
LYM	3/21/2018 3:09 ...	File folder	
MUC	3/21/2018 3:10 ...	File folder	
MUS	3/21/2018 3:29 ...	File folder	
NORM	3/21/2018 3:11 ...	File folder	
STR	3/21/2018 3:11 ...	File folder	
TUM	3/21/2018 3:12 ...	File folder	

This directory shows 9 folders; each contains images for a specific class. Any number of images can be removed for real testing, or another dataset can be used for this purpose. We removed about 5 images from each category for real testing, which is the last step in the pipeline.

### 1.2 Cancer classification

We included only cross-validation + image augmentation with variation in picture sizes in a simple train machine learning model.

## 1.2.1 python imports

```
# This section contains the list of modules and imports require to execute the code.

# We ignored some warnings.
import warnings
warnings.filterwarnings("ignore")
warnings.filterwarnings("ignore", message="libpng warning: iCCP: known incorrect sRGB_
↳profile")
warnings.filterwarnings("ignore", message="'rm' is not recognized as an")

# We run the code on the CPU, so the following 2 lines disable GPU-based execution.
import os
os.environ["CUDA_VISIBLE_DEVICES"] = "-1"

from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
import seaborn as sn
import matplotlib.pyplot as plt
from pylab import rcParams

from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
import seaborn as sn
import matplotlib.pyplot as plt
from pylab import rcParams
from PIL import Image
import tensorflow as tf
import os
seed_value= 0
import os
os.environ['PYTHONHASHSEED']=str(seed_value)
import random
random.seed(seed_value)
import numpy as np
np.random.seed(seed_value)
import numpy as np
import tensorflow as tf
tf.random.set_seed(seed_value)
from tensorflow.keras.utils import to_categorical
from sklearn.model_selection import StratifiedKFold
from PIL import Image
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Conv2D, Dropout, Flatten, MaxPooling2D
from tensorflow.keras import layers
from tensorflow.keras import optimizers
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from tensorflow.keras.utils import to_categorical
import tensorflow as tf

# CV2 module is required to read PNG images.
import cv2
from sklearn.preprocessing import LabelEncoder
```

(continues on next page)

(continued from previous page)

```

from sklearn.preprocessing import OneHotEncoder
import sys
tf.compat.v1.logging.set_verbosity(tf.compat.v1.logging.ERROR)

```

## 1.2.2 Helper functions

```

# This section contains helper functions.

# Plot the confusion matrix.
def textplotting(model,x_test,y_test):
    y_pred = model.predict(x_test)
    y_pred = np.argmax(y_pred, axis=1)
    y_test = np.argmax(y_test, axis=1)
    y_pred = y_pred.reshape(-1,1)
    y_test = y_test.reshape(-1,1)
    print(confusion_matrix(y_test, y_pred))
    print(accuracy_score(y_test, y_pred))

# Plot the confusion matrix.
def metric(best_model,x_train,y_train,x_test,y_test):

    y_pred = best_model.predict(x_test)
    sn.set(font_scale=2)
    rcParams['figure.figsize'] = 7, 7
    confusion_matrix = pd.crosstab(y_test.argmax(axis=1), y_pred.argmax(axis=1),
    ↪rownames=['Actual'], colnames=['Predicted'])
    sn.heatmap(confusion_matrix, annot=True)

    plt.savefig("Test.png")
    plt.clf()
    confusion_matrix = pd.crosstab(y_train.argmax(axis=1), best_model.predict(x_train).
    ↪argmax(axis=1), rownames=['Actual'], colnames=['Predicted'])
    sn.heatmap(confusion_matrix, annot=True)

    plt.savefig("Train.png")
    plt.clf()

# Plot the parameter reduction.
def simplot(parameterslist):
    plt.clf()
    sn.set(font_scale=1)
    rcParams['figure.figsize'] = 10, 10
    plt.rcParams["axes.edgecolor"] = "black"
    plt.rcParams["axes.linewidth"] = 1
    lists = sorted(parameterslist.items())
    x, y = zip(*lists)
    plt.minorticks_on()
    fig,ax=plt.subplots()
    ax.plot(x, y, marker="o")
    ax.grid(True)
    ax.patch.set_edgecolor('black')
    ax.patch.set_linewidth('1')
    ax.set_xscale('log')
    ax.set_xlabel("Number of Parameters")
    ax.set_ylabel("Accuracy")

```

(continues on next page)

(continued from previous page)

```

plt.savefig("Parameters.png")
plt.show()

# Save parameter reduction file in a particular format.
def saveLogfile(NeuronsInLayer1,NeuronsInLayer2,IntermediateAccuracy,
↪IntermediateParameters):
    data = np.column_stack([NeuronsInLayer1,NeuronsInLayer2,IntermediateAccuracy,
↪IntermediateParameters])
    datafile_path = "./Logfile.txt"
    np.savetxt(datafile_path , data, fmt=['%f','%d','%f','%d'])

# This segment is used to place images in specific directories.
def makedirectory(name,x_train,y_train,x_test, y_test,x_val, y_val):
    base_dir = './fold'+str(name)
    if not os.path.isdir(base_dir):
        os.mkdir(base_dir)

    train_dir = os.path.join(base_dir, 'train')
    if not os.path.isdir(train_dir):
        os.mkdir(train_dir)

    validation_dir = os.path.join(base_dir, 'validation')
    if not os.path.isdir(validation_dir):
        os.mkdir(validation_dir)

    test_dir = os.path.join(base_dir, 'test')
    if not os.path.isdir(test_dir):
        os.mkdir(test_dir)

# Make a folder for train,test, and validation set for each category.
for loop in folders:
    temptrain = os.path.join(train_dir, loop)
    if not os.path.isdir(temptrain):
        os.mkdir(temptrain)
    tempvalid = os.path.join(validation_dir, loop)
    if not os.path.isdir(tempvalid):
        os.mkdir(tempvalid)
    temptest = os.path.join(test_dir, loop)

    if not os.path.isdir(temptest):
        os.mkdir(temptest)

    for loop2 in range(0,len(y_train)):
        if y_train[loop2]==loop:
            im = Image.fromarray(x_train[loop2])
            im.save(temptrain+os.sep+str(loop2)+".jpeg")
    for loop2 in range(0,len(y_test)):
        if y_test[loop2]==loop:
            im = Image.fromarray(x_test[loop2])
            im.save(temptest+os.sep+str(loop2)+".jpeg")
    for loop2 in range(0,len(y_val)):

```

(continues on next page)



(continued from previous page)

```
        if y_val[loop2]==loop:
            im = Image.fromarray(x_val[loop2])
            im.save(tempvalid+os.sep+str(loop2)+".jpeg")

# Read images
def load_images_from_folder(folder,shape):
    images = []
    for filename in os.listdir(folder):
        img = cv2.imread(os.path.join(folder, filename))
        if img is not None:
            img = cv2.resize(img, (shape, shape))
            images.append(img)

    return images
```

### 1.2.3 Step 1 - Train the model.

```
# The folders in which images are placed.
folders = [
    'ADI',
    'BACK',
    'DEB',
    'LYM',
    'MUC',
    'MUS',
    'NORM',
    'STR',
    'TUM',
]

label_encoder = LabelEncoder()
enc = OneHotEncoder(handle_unknown='ignore')

# Any number of image generators can be specified and used, but we used 4 different
↳ image augmentation architectures.
gen=[1,2,3,4]

# Images can be reshaped to various sizes to reduce the training time and neural
↳ network size.
shapes = [50,100,200]

# For each shape
for shape in shapes:
    print("Shape:", shape)
    X = load_images_from_folder(folders[0],shape)
    Y = [folders[0]]*len(X)
    cat = 1
    for loop in range(1,len(folders)):
        print("Processing : ",folders[loop])
        tempX = load_images_from_folder(folders[loop],shape)
        tempY = [folders[loop]]*len(tempX)
        Y = Y + tempY
```

(continues on next page)

(continued from previous page)

```

X = X + tempX

X = np.array(X)
Y= np.array(Y)
skf = StratifiedKFold(n_splits=5)
skf.get_n_splits(X, Y)
fold_no = 1
acc_per_fold = {}
loss_per_fold = {}
input_shape = (shape , shape , 3)

# For cross validation
for train_index, test_index in skf.split(X, Y):
    x_train, x_test = X[train_index], X[test_index]
    y_train, y_test = Y[train_index], Y[test_index]
    x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size=0.
↪1, random_state=1)
    makedirs(str(fold_no)+"_"+str(shape),x_train,y_train,x_test, y_test,x_val,
↪y_val)
    train_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'train')
    validation_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'validation')
    test_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'test')
    y_train = to_categorical(label_encoder.fit_transform(y_train))
    y_test = to_categorical(label_encoder.fit_transform(y_test))
    y_val = to_categorical(label_encoder.fit_transform(y_val))

    model = Sequential()
    model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    #model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
    #model.add(MaxPooling2D(pool_size = (2, 2)))
    # Use the best number of neurons for the first layer in this layer obtained
↪from the previous step.
    #model.add(Conv2D(32, kernel_size=(3,3), input_shape=input_shape))
    #model.add(Conv2D(64, kernel_size=(3,3)))
    # Use the best number of neurons for the second layer in this layer obtained
↪from the previous step.
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Flatten())
    model.add(Dense(128, activation=tf.nn.relu))
    model.add(Dropout(0.2))
    model.add(Dense(50, activation=tf.nn.relu))
    model.add(Dropout(0.2))
    model.add(Dense(20, activation=tf.nn.relu))
    model.add(Dropout(0.2))
    model.add(Dense(len(folders), activation=tf.nn.softmax))

    param = model.count_params()

    # For each generator
    for selectgen in gen:

```

(continues on next page)

(continued from previous page)

```
if selectgen==1:
    train_datagen = ImageDataGenerator(
        rescale=1./255)
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==2:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==3:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        featurewise_center=True,
        featurewise_std_normalization=True,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
else:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        width_shift_range=0.2,
        height_shift_range=0.2,
        shear_range=0.2,
        featurewise_center=True,
        featurewise_std_normalization=True,
        zoom_range=0.2,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

validation_generator = test_datagen.flow_from_directory(
    validation_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

model.compile(optimizer='adam',
              loss='categorical_crossentropy',
              metrics=['accuracy'])
```

(continues on next page)

(continued from previous page)

```

history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=20,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,
    verbose=1)

#print("DONE")
test_generator = test_datagen.flow_from_directory(
    test_dir,
    target_size=(shape , shape),
    batch_size=1 ,
    class_mode='categorical')

test_loss, test_acc = model.evaluate_generator(test_generator, steps=50)
print('test acc:', test_acc)
if selectgen not in acc_per_fold:
    acc_per_fold[selectgen] = [test_acc * 100]
    loss_per_fold[selectgen] = [test_loss]
else:
    acc_per_fold[selectgen].append(test_acc * 100)
    loss_per_fold[selectgen].append(test_loss)
#os.system("rm -rf fold"+str(fold_no)+"_"+str(shape))
fold_no = fold_no + 1

# For each generator, print the results for each fold.
for g in gen:
    print("Generator",g)
    acc_per_foldt = acc_per_fold[g]
    loss_per_foldt = loss_per_fold[g]
    print('-----')
    print('Score per fold')
    for i in range(0, len(acc_per_foldt)):
        print('-----')
        print(f'> Fold {i+1} - Loss: {loss_per_foldt[i]} - Accuracy: {acc_per_')
        print(f'foldt[i]}%')
        print('-----')
    print('Average scores for all folds:')
    print(f'> Accuracy: {np.mean(acc_per_foldt)} (+- {np.std(acc_per_foldt)})')
    print(f'> Loss: {np.mean(loss_per_foldt)}')
    print('-----')

```

## 1.2.4 Step 2 - Convnet Model to TFlite And AddDetadata

```

gen = [1]
shapes = [150]
for shape in shapes:
    X = load_images_from_folder(folders[0],shape)
    Y = [folders[0]]*len(X)
    cat = 1

```

(continues on next page)

(continued from previous page)

```

for loop in range(1,len(folders)):
    print("Processing : ",folders[loop])
    tempX = load_images_from_folder(folders[loop],shape)
    tempY = [folders[loop]]*len(tempX)
    Y = Y + tempY
    X = X + tempX
X = np.array(X)
Y= np.array(Y)
skf = StratifiedKFold(n_splits=5)
input_shape = (shape , shape , 3)
for selectgen in gen:
    batch_size=100
    acc_per_fold = []
    loss_per_fold = []
    fold_no = 0
    for train_index, test_index in skf.split(X, Y):
        x_train, x_test = X[train_index], X[test_index]
        y_train, y_test = Y[train_index], Y[test_index]
        x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_
↪size=0.3, random_state=1)
        makedirectory(str(fold_no)+"_"+str(shape),x_train,y_train,x_test, y_test,
↪x_val, y_val)
        train_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'train')
        validation_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape),
↪'validation')
        test_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'test')
        y_train = to_categorical(label_encoder.fit_transform(y_train))
        y_test = to_categorical(label_encoder.fit_transform(y_test))
        y_val = to_categorical(label_encoder.fit_transform(y_val))
        mobile = tf.keras.applications.mobilenet.MobileNet()

        #print("Shape ",str(shape))
        model = Sequential()
        model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        #model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
        #model.add(MaxPooling2D(pool_size = (2, 2)))
        # Use the best number of neurons for the first layer in this layer_
↪obtained from the previous step.
        #model.add(Conv2D(32, kernel_size=(3,3), input_shape=input_shape))
        #model.add(Conv2D(64, kernel_size=(3,3)))
        # Use the best number of neurons for the second layer in this layer_
↪obtained from the previous step.
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Flatten())
        model.add(Dense(128, activation=tf.nn.relu))
        model.add(Dropout(0.2))
        model.add(Dense(50, activation=tf.nn.relu))
        model.add(Dropout(0.2))
        model.add(Dense(20, activation=tf.nn.relu))
        model.add(Dropout(0.2))

```

(continues on next page)

(continued from previous page)

```

model.add(Dense(len(folders), activation=tf.nn.softmax))

param = model.count_params()

if selectgen==1:
    train_datagen = ImageDataGenerator(
        rescale=1./255)
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==2:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==3:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        featurewise_center=True,
        featurewise_std_normalization=True,
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
else:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        width_shift_range=0.2,
        height_shift_range=0.2,
        shear_range=0.2,
        featurewise_center=True,
        featurewise_std_normalization=True,
        zoom_range=0.2,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

validation_generator = test_datagen.flow_from_directory(
    validation_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

```

(continues on next page)

(continued from previous page)

```

model.compile(optimizer='adam',
              loss='categorical_crossentropy',
              metrics=['accuracy'])

history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=20,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,
    verbose=1)

#print("DONE")
test_generator = test_datagen.flow_from_directory(
    test_dir,
    target_size=(shape , shape),
    batch_size=1 ,
    class_mode='categorical')

textplotting(model,x_test/255,y_test)
textplotting(model,x_train/255,y_train)

test_loss, test_acc = model.evaluate_generator(test_generator, steps=test_
→generator.n // batch_size)
print(test_acc)
acc_per_fold.append(test_acc * 100)
loss_per_fold.append(test_loss)
fold_no = fold_no + 1

def representative_dataset_gen():
    for _ in range(100):
        img = train_generator.next()
        yield [img[0]]

model.save("alsome.hd5")
### In this step we will simply save the model
model_save_path = "."
tf.saved_model.save(model, model_save_path)
converter = tf.lite.TFLiteConverter.from_saved_model(model_save_path)
converter.optimizations = [tf.lite.Optimize.DEFAULT]
converter.representative_dataset = representative_dataset_gen
→INT8] #converter.target_spec.supported_ops = [tf.lite.OpsSet.TFLITE_BUILTINS_

converter.experimental_new_converter = True

#converter.target_spec.supported_types = [tf.uint8]
#converter.inference_input_type = tf.int8 # or tf.uint8
#converter.inference_output_type = tf.uint8 # or tf.uint8

tflite_model = converter.convert()
open("model.tflite", "wb").write(tflite_model)
import tensorflow_model_optimization as tfmot

```

(continues on next page)

(continued from previous page)

```

quantize_model = tfmot.quantization.keras.quantize_model

# q_aware stands for quantization aware.
q_aware_model = quantize_model(model)

# `quantize_model` requires a recompile.
q_aware_model.compile(optimizer='adam',
                      loss='categorical_crossentropy',
                      metrics=['accuracy'])
history = q_aware_model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=20,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,
    verbose=1)
q_aware_model.summary()
_, baseline_model_accuracy = model.evaluate(x_test/255, y_test, verbose=0)

_, q_aware_model_accuracy = q_aware_model.evaluate(x_test/255, y_test,
→verbose=0)
print('Baseline test accuracy:', baseline_model_accuracy)
print('Quant test accuracy:', q_aware_model_accuracy)
textplotting(q_aware_model, x_test/255, y_test)
textplotting(q_aware_model, x_train/255, y_train)
q_aware_model.save("qalsome.hd5")
### In this step we will simply save the model
model_save_path = "."
tf.saved_model.save(q_aware_model, model_save_path)
converter = tf.lite.TFLiteConverter.from_saved_model(model_save_path)
tflite_model = converter.convert()
open("qmodel.tflite", "wb").write(tflite_model)
# We can save any model for the 5 fold validation, so we exit the code,
→after the training of the first fold.

```

### 1.2.5 Step 3 - Convnet Model to TFlite And AddMetadata

```

import os
# Change values in metadata_writer_for_image_classifier.py
_MODEL_INFO = {
    "model.tflite":
        ModelSpecificInfo(
            name="model",
            version="v1",
            image_width=100,
            image_height=100,
            image_min=0,
            image_max=1,
            mean=[0],
            std=[255],
            num_classes=9,
            author="Alpha")
}

```

(continues on next page)



(continued from previous page)

```
# Change labels in labels.txt file

os.mkdir("Finalmodel")
os.system("python ./metadata_writer_for_image_classifier.py --model_file=model.tflite_
↪--label_file=labels.txt --export_directory=./Finalmodel")
```

## 1.2.6 Step 4 - Make Android Application

Make Android Application - TensorFlow Lite

The following picture shows the android studio, and we have to change the assets.

Replace

G:/Application/Cancer Application/Tensorflow Application/models/src/main/assets

With

Finalmodel/model and rename model with efficientnet-lite0-int8

Open Android studio and open Image-Classificaion-master, build the application.

Make Android Application - Flutter

The following picture shows the android studio, and we have to change the assets.

Replace

G:/Application/Cancer Application/Cat-Dog-Classifer-main/assets

With

Finalmodel/qmodel and rename model with model\_unquant

Open Android studio and open Cat-Dog-Classifer-main, build the application.

## 1.3 Covid classification

We included only cross-validation + image augmentation with variation in picture sizes in a simple train machine learning model.

### 1.3.1 python imports

```
# This section contains the list of modules and imports require to execute the code.

# We ignored some warnings.
import warnings
warnings.filterwarnings("ignore")
warnings.filterwarnings("ignore", message="libpng warning: iCCP: known incorrect sRGB_
↪profile")
warnings.filterwarnings("ignore", message="'rm' is not recognized as an")

# We run the code on the CPU, so the following 2 lines disable GPU-based execution.
import os
os.environ["CUDA_VISIBLE_DEVICES"] = "-1"
```

(continues on next page)

(continued from previous page)

```
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
import seaborn as sn
import matplotlib.pyplot as plt
from pylab import rcParams

from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
import seaborn as sn
import matplotlib.pyplot as plt
from pylab import rcParams
from PIL import Image
import tensorflow as tf
import os
seed_value= 0
import os
os.environ['PYTHONHASHSEED']=str(seed_value)
import random
random.seed(seed_value)
import numpy as np
np.random.seed(seed_value)
import numpy as np
import tensorflow as tf
tf.random.set_seed(seed_value)
from tensorflow.keras.utils import to_categorical
from sklearn.model_selection import StratifiedKFold
from PIL import Image
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Conv2D, Dropout, Flatten, MaxPooling2D
from tensorflow.keras import layers
from tensorflow.keras import optimizers
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from tensorflow.keras.utils import to_categorical
import tensorflow as tf

# CV2 module is required to read PNG images.
import cv2
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import OneHotEncoder
import sys
tf.compat.v1.logging.set_verbosity(tf.compat.v1.logging.ERROR)
```

### 1.3.2 Helper functions

```
# This section contains helper functions.

# Plot the confusion matrix.
def textplotting(model,x_test,y_test):
    y_pred = model.predict(x_test)
    y_pred = np.argmax(y_pred, axis=1)
```

(continues on next page)

(continued from previous page)

```

y_test = np.argmax(y_test, axis=1)
y_pred = y_pred.reshape(-1,1)
y_test = y_test.reshape(-1,1)
print(confusion_matrix(y_test, y_pred))
print(accuracy_score(y_test, y_pred))

# Plot the confusion matrix.
def metric(best_model,x_train,y_train,x_test,y_test):

    y_pred = best_model.predict(x_test)
    sn.set(font_scale=2)
    rcParams['figure.figsize'] = 7, 7
    confusion_matrix = pd.crosstab(y_test.argmax(axis=1), y_pred.argmax(axis=1),
↪rownames=['Actual'], colnames=['Predicted'])
    sn.heatmap(confusion_matrix, annot=True)

    plt.savefig("Test.png")
    plt.clf()
    confusion_matrix = pd.crosstab(y_train.argmax(axis=1), best_model.predict(x_train).
↪argmax(axis=1), rownames=['Actual'], colnames=['Predicted'])
    sn.heatmap(confusion_matrix, annot=True)

    plt.savefig("Train.png")
    plt.clf()

# Plot the parameter reduction.
def simplot(parameterslist):
    plt.clf()
    sn.set(font_scale=1)
    rcParams['figure.figsize'] = 10, 10
    plt.rcParams["axes.edgecolor"] = "black"
    plt.rcParams["axes.linewidth"] = 1
    lists = sorted(parameterslist.items())
    x, y = zip(*lists)
    plt.minorticks_on()
    fig,ax=plt.subplots()
    ax.plot(x, y, marker="o")
    ax.grid(True)
    ax.patch.set_edgecolor('black')
    ax.patch.set_linewidth('1')
    ax.set_xscale('log')
    ax.set_xlabel("Number of Parameters")
    ax.set_ylabel("Accuracy")
    plt.savefig("Parameters.png")
    plt.show()

# Save parameter reduction file in a particular format.
def saveLogfile(NeuronsInLayer1,NeuronsInLayer2,IntermediateAccuracy,
↪IntermediateParameters):
    data = np.column_stack([NeuronsInLayer1,NeuronsInLayer2,IntermediateAccuracy,
↪IntermediateParameters])
    datafile_path = "./Logfile.txt"
    np.savetxt(datafile_path , data, fmt=['%f','%d','%f','%d'])

# This segment is used to place images in specific directories.
def makedirectory(name,x_train,y_train,x_test, y_test,x_val, y_val):

```

(continues on next page)

(continued from previous page)

```

base_dir = './fold'+str(name)
if not os.path.isdir(base_dir):
    os.mkdir(base_dir)

train_dir = os.path.join(base_dir, 'train')
if not os.path.isdir(train_dir):
    os.mkdir(train_dir)

validation_dir = os.path.join(base_dir, 'validation')
if not os.path.isdir(validation_dir):
    os.mkdir(validation_dir)

test_dir = os.path.join(base_dir, 'test')
if not os.path.isdir(test_dir):
    os.mkdir(test_dir)

# Make a folder for train, test, and validation set for each category.
for loop in folders:
    temptrain = os.path.join(train_dir, loop)
    if not os.path.isdir(temptrain):
        os.mkdir(temptrain)
    tempvalid = os.path.join(validation_dir, loop)
    if not os.path.isdir(tempvalid):
        os.mkdir(tempvalid)
    temptest = os.path.join(test_dir, loop)

    if not os.path.isdir(temptest):
        os.mkdir(temptest)

    for loop2 in range(0, len(y_train)):
        if y_train[loop2] == loop:
            im = Image.fromarray(x_train[loop2])
            im.save(temptrain+os.sep+str(loop2)+".jpeg")
    for loop2 in range(0, len(y_test)):
        if y_test[loop2] == loop:
            im = Image.fromarray(x_test[loop2])
            im.save(temptest+os.sep+str(loop2)+".jpeg")
    for loop2 in range(0, len(y_val)):
        if y_val[loop2] == loop:
            im = Image.fromarray(x_val[loop2])
            im.save(tempvalid+os.sep+str(loop2)+".jpeg")

# Read images
def load_images_from_folder(folder, shape):
    images = []
    for filename in os.listdir(folder):
        img = cv2.imread(os.path.join(folder, filename))
        if img is not None:
            img = cv2.resize(img, (shape, shape))
            images.append(img)

    return images

```

### 1.3.3 Step 1 - Train the model.

```
# The folders in which images are placed.
folders = [
    'CT_COVID',
    'CT_NonCOVID',
]

label_encoder = LabelEncoder()
enc = OneHotEncoder(handle_unknown='ignore')

# Any number of image generators can be specified and used, but we used 4 different
↳ image augmentation architectures.
gen=[1,2,3,4]

# Images can be reshaped to various sizes to reduce the training time and neural
↳ network size.
shapes = [50,100,200]

# For each shape
for shape in shapes:
    print("Shape:", shape)
    X = load_images_from_folder(folders[0], shape)
    Y = [folders[0]]*len(X)
    cat = 1
    for loop in range(1, len(folders)):
        print("Processing : ", folders[loop])
        tempX = load_images_from_folder(folders[loop], shape)
        tempY = [folders[loop]]*len(tempX)
        Y = Y + tempY
        X = X + tempX

    X = np.array(X)
    Y = np.array(Y)
    skf = StratifiedKFold(n_splits=5)
    skf.get_n_splits(X, Y)
    fold_no = 1
    acc_per_fold = {}
    loss_per_fold = {}
    input_shape = (shape, shape, 3)

    # For cross validation
    for train_index, test_index in skf.split(X, Y):
        x_train, x_test = X[train_index], X[test_index]
        y_train, y_test = Y[train_index], Y[test_index]
        x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size=0.
↳ 1, random_state=1)
        makedirs(str(fold_no)+"_"+str(shape), x_train, y_train, x_test, y_test, x_val,
↳ y_val)
        train_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'train')
        validation_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'validation')
        test_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'test')
        y_train = to_categorical(label_encoder.fit_transform(y_train))
        y_test = to_categorical(label_encoder.fit_transform(y_test))
```

(continues on next page)

(continued from previous page)

```

y_val = to_categorical(label_encoder.fit_transform(y_val))

model = Sequential()
model.add(Conv2D(32, kernel_size=(3,3), input_shape=input_shape))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Flatten())
model.add(Dense(128, activation=tf.nn.relu))
model.add(Dropout(0.2))
model.add(Dense(len(folders), activation=tf.nn.softmax))

# For each generator
for selectgen in gen:
    if selectgen==1:
        train_datagen = ImageDataGenerator(
            rescale=1./255)
        test_datagen = ImageDataGenerator(rescale=1./255)
    elif selectgen==2:
        train_datagen = ImageDataGenerator(
            rescale=1./255,
            rotation_range=40,
            brightness_range=[0.2,1.0],
            horizontal_flip=True,
            vertical_flip=True,
            fill_mode='nearest')
        test_datagen = ImageDataGenerator(rescale=1./255)
    elif selectgen==3:
        train_datagen = ImageDataGenerator(
            rescale=1./255,
            rotation_range=40,
            featurewise_center=True,
            featurewise_std_normalization=True,
            brightness_range=[0.2,1.0],
            horizontal_flip=True,
            vertical_flip=True,
            fill_mode='nearest')
        test_datagen = ImageDataGenerator(rescale=1./255)
    else:
        train_datagen = ImageDataGenerator(
            rescale=1./255,
            rotation_range=40,
            width_shift_range=0.2,
            height_shift_range=0.2,
            shear_range=0.2,
            featurewise_center=True,
            featurewise_std_normalization=True,
            zoom_range=0.2,
            brightness_range=[0.2,1.0],
            horizontal_flip=True,
            vertical_flip=True,
            fill_mode='nearest')
        test_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,

```

(continues on next page)

(continued from previous page)

```

class_mode='categorical')

validation_generator = test_datagen.flow_from_directory(
    validation_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

model.compile(optimizer='adam',
              loss='categorical_crossentropy',
              metrics=['accuracy'])

history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=15,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,
    verbose=0)

test_generator = test_datagen.flow_from_directory(
    test_dir,
    target_size=(shape , shape),
    batch_size=batch_size ,
    class_mode='categorical')

test_loss, test_acc = model.evaluate_generator(test_generator, steps=50)
print('test acc:', test_acc)
if selectgen not in acc_per_fold:
    acc_per_fold[selectgen] = [test_acc * 100]
    loss_per_fold[selectgen] = [test_loss]
else:
    acc_per_fold[selectgen].append(test_acc * 100)
    loss_per_fold[selectgen].append(test_loss)
#os.system("rm -rf fold"+str(fold_no)+"_"+str(shape))
fold_no = fold_no + 1

# For each generator, print the results for each fold.
for g in gen:
    print("Generator",g)
    acc_per_foldt = acc_per_fold[g]
    loss_per_foldt = loss_per_fold[g]
    print('-----')
    print('Score per fold')
    for i in range(0, len(acc_per_foldt)):
        print('-----')
        print(f'> Fold {i+1} - Loss: {loss_per_foldt[i]} - Accuracy: {acc_per_
        foldt[i]}%')
        print('-----')
    print('Average scores for all folds:')
    print(f'> Accuracy: {np.mean(acc_per_foldt)} (+- {np.std(acc_per_foldt)})')
    print(f'> Loss: {np.mean(loss_per_foldt)}')
    print('-----')

```

### 1.3.4 Step 2 - Find the Best Model And reduce the model's size

```

# From the previous step, use the best generator and shape depending on the
↳ deployment.
# For example, generator 1 is not feasible for real images because images can be
↳ various rotations when captured through a phone.
# For medical images, if we reduce the shape size, then many different classes may
↳ converge to the same class.

gen = [1]
shapes = [50]
for shape in shapes:
    X = load_images_from_folder(folders[0], shape)
    Y = [folders[0]]*len(X)
    cat = 1
    for loop in range(1, len(folders)):
        print("Processing : ", folders[loop])
        tempX = load_images_from_folder(folders[loop], shape)
        tempY = [folders[loop]]*len(tempX)
        Y = Y + tempY
        X = X + tempX

X = np.array(X)
Y = np.array(Y)
skf = StratifiedKFold(n_splits=5)
input_shape = (shape, shape, 3)
for selectgen in gen:
    batch_size=5
    NeuronsInLayer1 = []
    NeuronsInLayer2 = []
    IntermediateAccuracy = []
    IntermediateParameters = []
    parameterslist = {}
    # Change the number of neurons in the first and the second layer.
    # If you use an existing architecture, then there is no need to reduce the
↳ number of neurons in each layer.
    for loop in range(1, 20):
        for loop2 in range(1, 20):
            acc_per_fold = []
            loss_per_fold = []
            fold_no = 0
            for train_index, test_index in skf.split(X, Y):
                x_train, x_test = X[train_index], X[test_index]
                y_train, y_test = Y[train_index], Y[test_index]
                x_train, x_val, y_train, y_val = train_test_split(x_train, y_train,
↳ test_size=0.1, random_state=1)
                makedirs(directory(str(fold_no)+"_"+str(shape), x_train, y_train, x_test,
↳ y_test, x_val, y_val)
                train_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'train
↳ ')
                validation_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape),
↳ 'validation')
                test_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'test')
                y_train = to_categorical(label_encoder.fit_transform(y_train))
                y_test = to_categorical(label_encoder.fit_transform(y_test))
                y_val = to_categorical(label_encoder.fit_transform(y_val))
                #print("Shape ", str(shape))

```

(continues on next page)



(continued from previous page)

```

model = Sequential()
model.add(Conv2D(loop, kernel_size=(3,3), input_shape=input_shape))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Flatten())
model.add(Dense(loop2, activation=tf.nn.relu))
model.add(Dropout(0.2))
model.add(Dense(len(folders), activation=tf.nn.softmax))
param = model.count_params()

if selectgen==1:
    train_datagen = ImageDataGenerator(
        rescale=1./255)
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==2:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==3:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        featurewise_center=True,
        featurewise_std_normalization=True,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
else:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        width_shift_range=0.2,
        height_shift_range=0.2,
        shear_range=0.2,
        featurewise_center=True,
        featurewise_std_normalization=True,
        zoom_range=0.2,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

```

(continues on next page)

(continued from previous page)

```

validation_generator = test_datagen.flow_from_directory(
    validation_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

model.compile(optimizer='adam',
              loss='categorical_crossentropy',
              metrics=['accuracy'])

history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=15,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,
    verbose=0)

#print("DONE")
test_generator = test_datagen.flow_from_directory(
    test_dir,
    target_size=(shape , shape),
    batch_size=batch_size ,
    class_mode='categorical')

test_loss, test_acc = model.evaluate_generator(test_generator, ↵
↵steps=test_generator.n // batch_size)
#print('test acc:', test_acc)
acc_per_fold.append(test_acc * 100)
loss_per_fold.append(test_loss)
fold_no = fold_no + 1
print(fold_no)
print(str(int(np.mean(acc_per_fold))), loop, loop2, param)
NeuronsInLayer1.append(loop)
NeuronsInLayer2.append(loop2)
IntermediateAccuracy.append(int(np.mean(acc_per_fold)))
IntermediateParameters.append(param)
parameterslist[str(model.count_params())] = int(np.mean(acc_per_fold))

saveLogfile(NeuronsInLayer1, NeuronsInLayer2, IntermediateAccuracy,
↵IntermediateParameters)
parameterslist = dict(sorted(parameterslist.items()))
simplot(parameterslist)

# We can save any model for the 5 fold validation, so we exit the code after ↵
↵the training of the first fold.

exit(0)

```

### 1.3.5 Step 2 - Plot parameter reduction

```

import pandas as pd
import seaborn as sns
import numpy as np
import math

```

(continues on next page)

(continued from previous page)

```
import matplotlib.pyplot as plt
data = pd.read_csv("LogFile.txt", sep="\s+", header=None)
print(data)
data2 = data[2].values
data2 = data2.reshape(int(math.sqrt(len(data))), int(math.sqrt(len(data))))
ax = sns.heatmap(data2, cmap="YlGnBu")
ax.set(xlabel="Neurons in the second layer", ylabel = "Filters in the first layer")
ax.set_xlim(1, int(math.sqrt(len(data)))+1)
ax.set_ylim(1, int(math.sqrt(len(data)))+1)
plt.show()
```

### 1.3.6 Step 3 - Convnet Model to TFlite And AddDetaddata

```
gen = [1]
shapes = [50]
for shape in shapes:
X = load_images_from_folder(folders[0], shape)
Y = [folders[0]]*len(X)
cat = 1
for loop in range(1, len(folders)):
    print("Processing : ", folders[loop])
    tempX = load_images_from_folder(folders[loop], shape)
    tempY = [folders[loop]]*len(tempX)
    Y = Y + tempY
    X = X + tempX

X = np.array(X)
Y = np.array(Y)
skf = StratifiedKFold(n_splits=5)
input_shape = (shape, shape, 3)
for selectgen in gen:
    batch_size=5
    acc_per_fold = []
    loss_per_fold = []
    fold_no = 0
    for train_index, test_index in skf.split(X, Y):
        x_train, x_test = X[train_index], X[test_index]
        y_train, y_test = Y[train_index], Y[test_index]
        x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_
↪size=0.1, random_state=1)
        makedirectory(str(fold_no)+"_"+str(shape), x_train, y_train, x_test, y_test, x_
↪val, y_val)
        train_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'train')
        validation_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'validation
↪')

        test_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'test')
        y_train = to_categorical(label_encoder.fit_transform(y_train))
        y_test = to_categorical(label_encoder.fit_transform(y_test))
        y_val = to_categorical(label_encoder.fit_transform(y_val))
        #print("Shape ", str(shape))
        model = Sequential()

        # Use the best number of neurons for the first layer in this layer obtained_
↪from the previous step.
        model.add(Conv2D(2, kernel_size=(3,3), input_shape=input_shape))
```

(continues on next page)

(continued from previous page)

```

# Use the best number of neurons for the second layer in this layer obtained
from the previous step.
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Flatten())
model.add(Dense(17, activation=tf.nn.relu))
model.add(Dropout(0.2))
model.add(Dense(2, activation=tf.nn.softmax))
param = model.count_params()

if selectgen==1:
    train_datagen = ImageDataGenerator(
        rescale=1./255)
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==2:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==3:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        featurewise_center=True,
        featurewise_std_normalization=True,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
else:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        width_shift_range=0.2,
        height_shift_range=0.2,
        shear_range=0.2,
        featurewise_center=True,
        featurewise_std_normalization=True,
        zoom_range=0.2,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

```

(continues on next page)

(continued from previous page)

```

validation_generator = test_datagen.flow_from_directory(
    validation_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

model.compile(optimizer='adam',
              loss='categorical_crossentropy',
              metrics=['accuracy'])

history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=15,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,
    verbose=0)

test_generator = test_datagen.flow_from_directory(
    test_dir,
    target_size=(shape , shape),
    batch_size=batch_size ,
    class_mode='categorical')

test_loss, test_acc = model.evaluate_generator(test_generator, steps=test_
↪generator.n // batch_size)

acc_per_fold.append(test_acc * 100)
loss_per_fold.append(test_loss)
fold_no = fold_no + 1

### In this step we will simply save the model
model_save_path = "."
tf.saved_model.save(model, model_save_path)
converter = tf.lite.TFLiteConverter.from_saved_model(model_save_path)
tflite_model = converter.convert()

# Save the simple model.
open("model.tflite", "wb").write(tflite_model)

# Generate Quantized Model
import tensorflow_model_optimization as tfmot
quantize_model = tfmot.quantization.keras.quantize_model

# q_aware stands for for quantization aware.
q_aware_model = quantize_model(model)
# `quantize_model` requires a recompile.
q_aware_model.compile(optimizer='adam',
                      loss='categorical_crossentropy',
                      metrics=['accuracy'])
history = q_aware_model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=30,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,

```

(continues on next page)

(continued from previous page)

```

        verbose=1)
    q_aware_model.summary()

    _, baseline_model_accuracy = model.evaluate(x_test/255, y_test, verbose=0)
    _, q_aware_model_accuracy = q_aware_model.evaluate(x_test/255, y_test,
↳ verbose=0)

    print('Baseline test accuracy:', baseline_model_accuracy)
    print('Quant test accuracy:', q_aware_model_accuracy)

    textplotting(model,x_train/255,y_train)
    textplotting(model,x_test/255,y_test)
    textplotting(q_aware_model,x_train/255,y_train)
    textplotting(q_aware_model,x_test/255,y_test)

    #q_aware_model.save("qalsome.hd5")
    ### In this step we will simply save the model
    model_save_path = "."
    tf.saved_model.save(q_aware_model, model_save_path)
    converter = tf.lite.TFLiteConverter.from_saved_model(model_save_path)
    tflite_model = converter.convert()
    # Save the quantized model.
    open("qmodel.tflite", "wb").write(tflite_model)

    # We can save any model for the 5 fold validation, so we exit the code after
↳ the training of the first fold.

```

### 1.3.7 Step 3 - Convet Model to TFlite And AddDetadata

```

import os
# Change values in metadata_writer_for_image_classifier.py
#_MODEL_INFO = {
#     "model.tflite":
#         ModelSpecificInfo(
#             name="model",
#             version="v1",
#             image_width=100,
#             image_height=100,
#             image_min=0,
#             image_max=1,
#             mean=[0],
#             std=[255],
#             num_classes=2,
#             author="Alpha")
#}

# Change labels in labels.txt file

os.mkdir("Finalmodel")
os.system("python ./metadata_writer_for_image_classifier.py --model_file=model.tflite,
↳ --label_file=labels.txt --export_directory=./Finalmodel")

```

### 1.3.8 Step 4 - Make Android Application

Make Android Application - TensorFlow Lite

The following picture shows the android studio, and we have to change the assets.

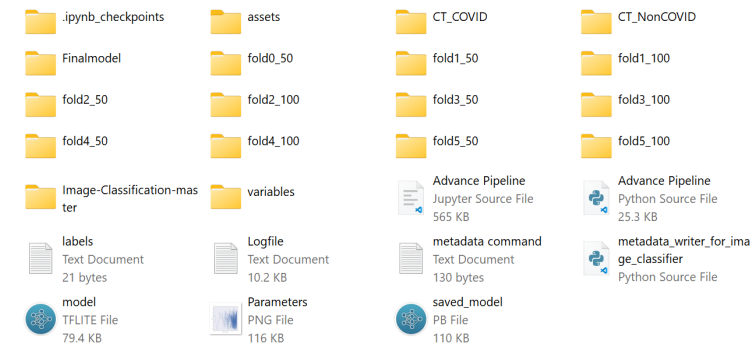
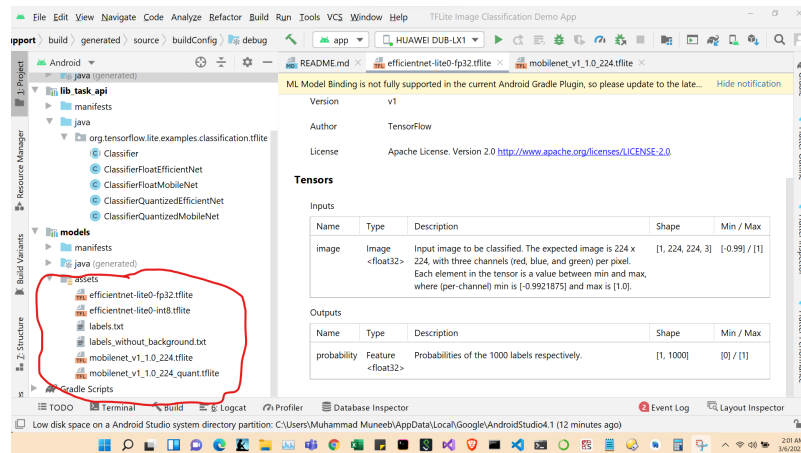
Replace

G:/Application/Covid Application/Tensorflow Application/models/src/main/assets

With

Finalmodel/model and rename model with efficientnet-lite0-int8

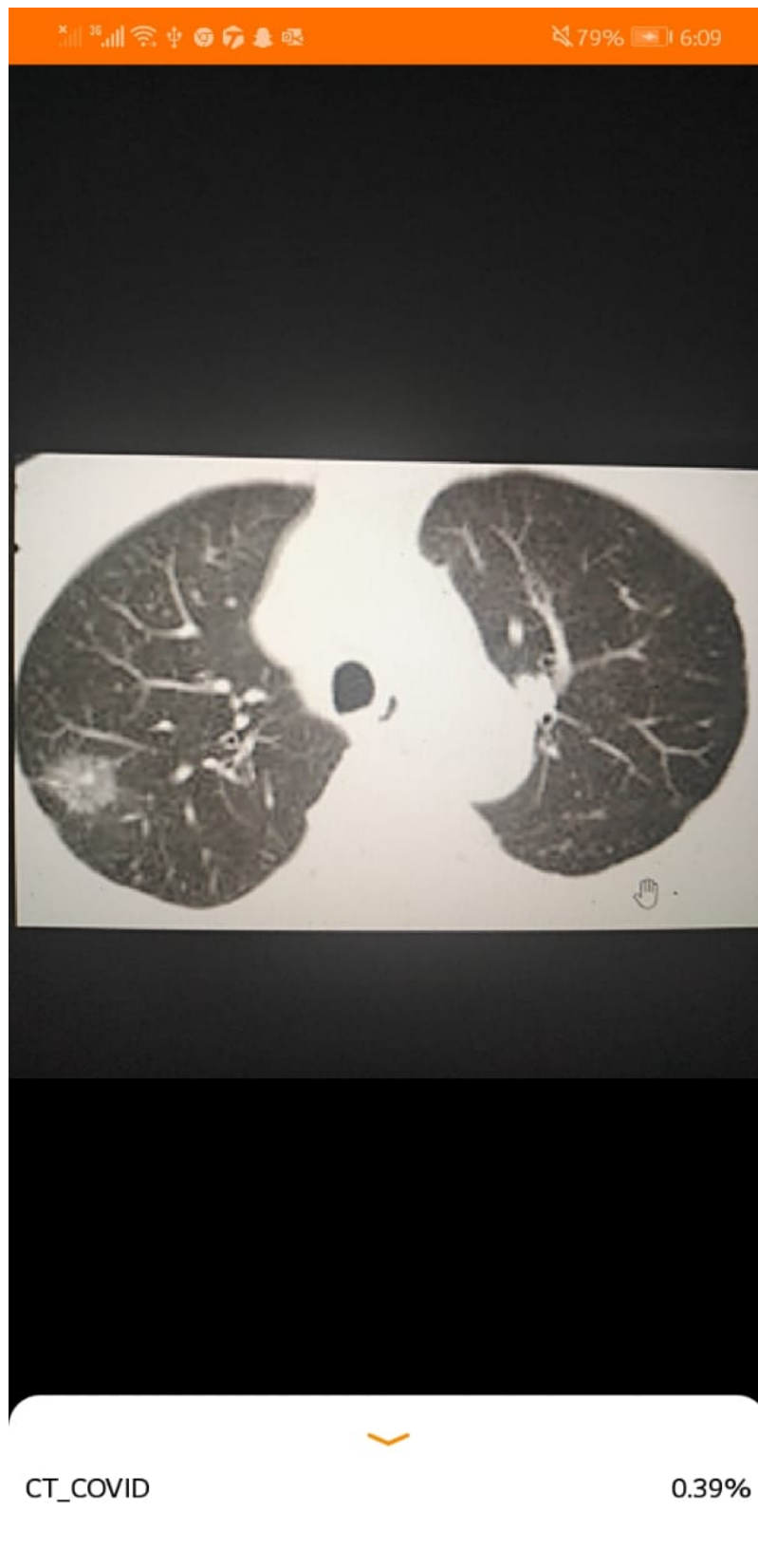
Open Android studio and open Image-Classificaion-master, build the application.



The following picture shows the directory structure. CT\_COVID and CT\_NonCOVID contain the original images. Fold\*\_\* shows temporary images. Logfile shows the logfiles.







Make Android Application - Flutter

The following picture shows the android studio, and we have to change the assets.

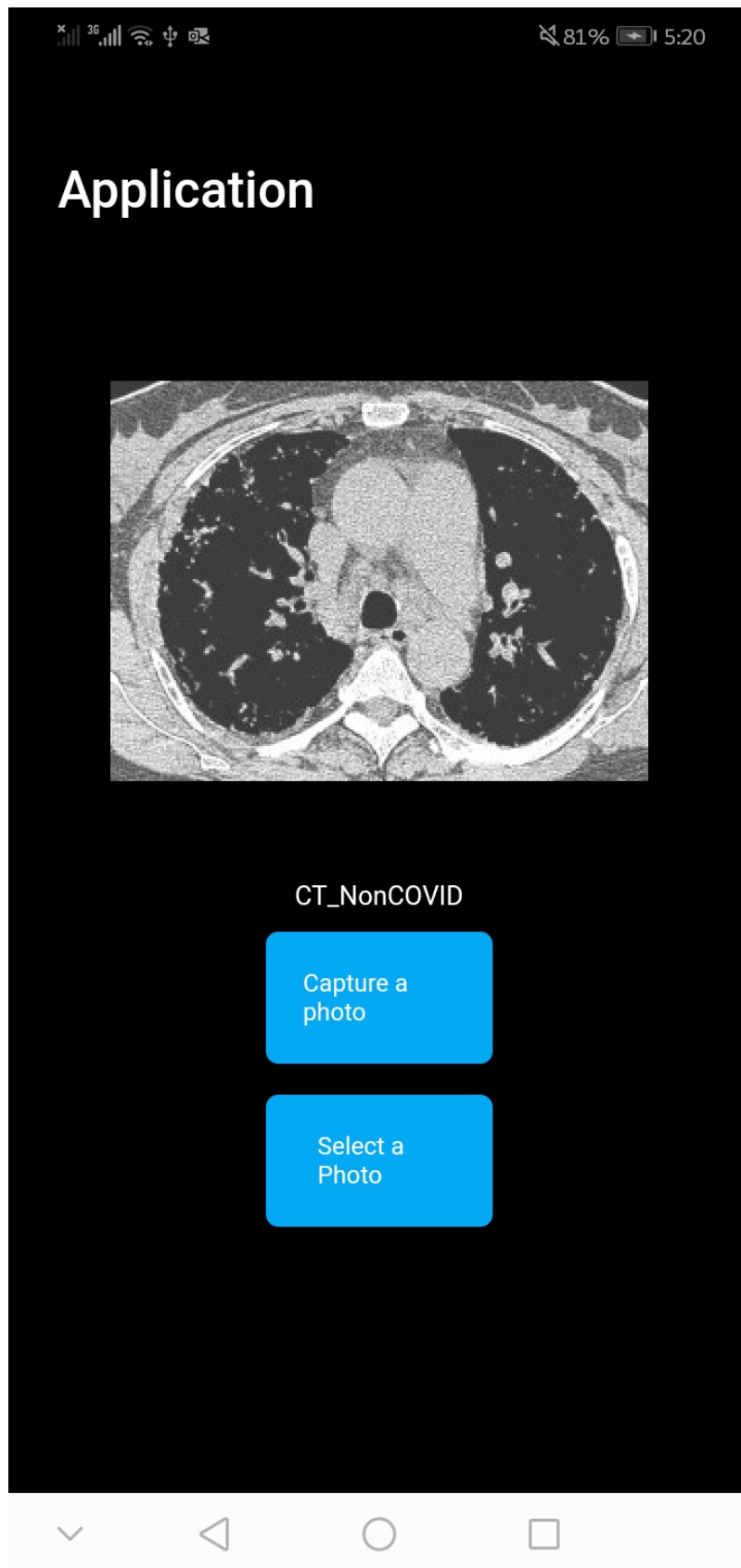
Replace

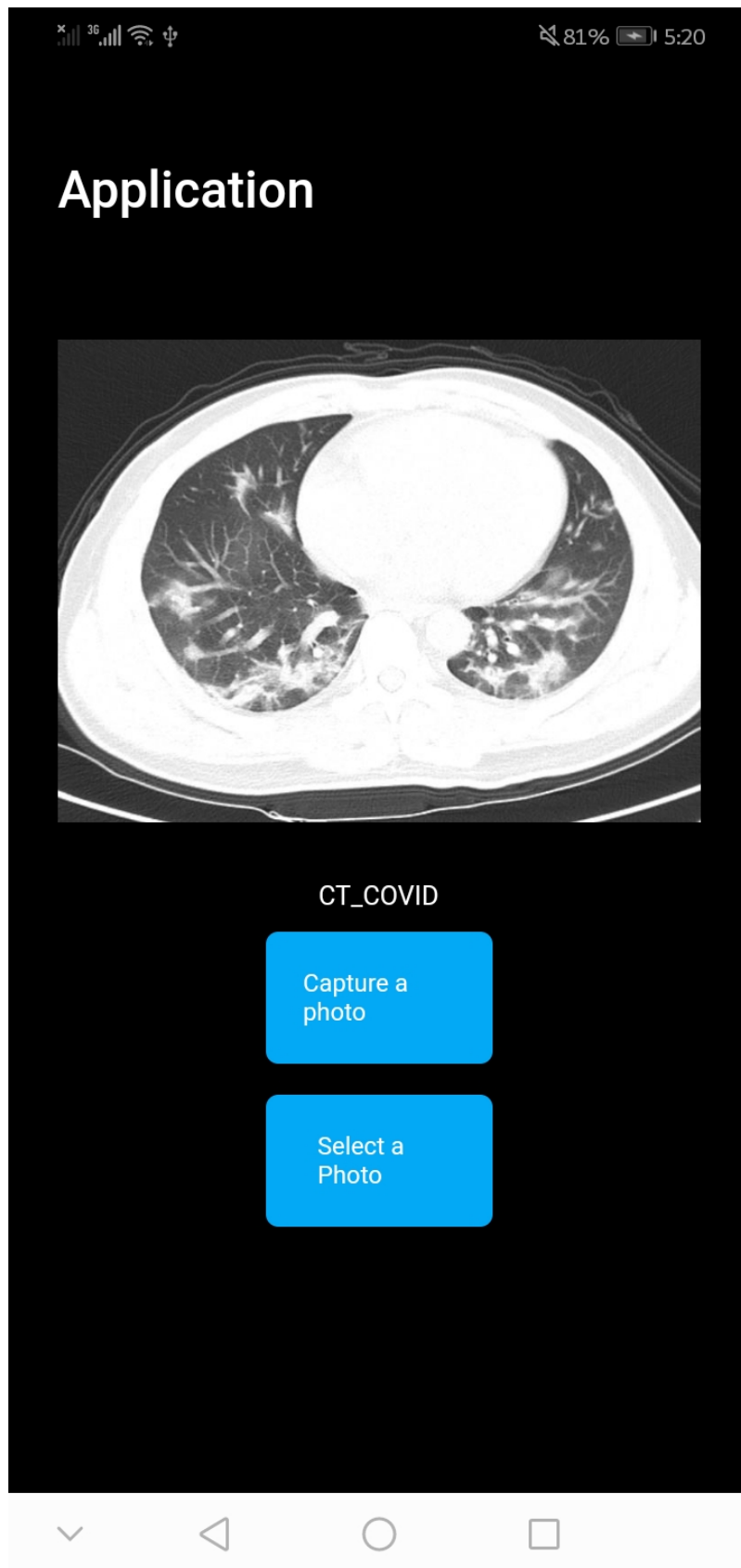
G:/Application/Tree Application/Cat-Dog-Classifier-main/assets

With

Finalmodel/qmodel and rename model with model\_unquant

Open Android studio and open Cat-Dog-Classifier-main, build the application.





## 1.4 Tree classification

We included only cross-validation + image augmentation with variation in picture sizes in a simple train machine learning model.

### 1.4.1 python imports

```
# This section contains the list of modules and imports require to execute the code.

# We ignored some warnings.
import warnings
warnings.filterwarnings("ignore")
warnings.filterwarnings("ignore", message="libpng warning: iCCP: known incorrect sRGB_
↪profile")
warnings.filterwarnings("ignore", message="'rm' is not recognized as an")

# We run the code on the CPU, so the following 2 lines disable GPU-based execution.
import os
os.environ["CUDA_VISIBLE_DEVICES"] = "-1"

from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
import seaborn as sn
import matplotlib.pyplot as plt
from pylab import rcParams

from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
import seaborn as sn
import matplotlib.pyplot as plt
from pylab import rcParams
from PIL import Image
import tensorflow as tf
import os
seed_value= 0
import os
os.environ['PYTHONHASHSEED']=str(seed_value)
import random
random.seed(seed_value)
import numpy as np
np.random.seed(seed_value)
import numpy as np
import tensorflow as tf
tf.random.set_seed(seed_value)
from tensorflow.keras.utils import to_categorical
from sklearn.model_selection import StratifiedKFold
from PIL import Image
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Conv2D, Dropout, Flatten, MaxPooling2D
from tensorflow.keras import layers
from tensorflow.keras import optimizers
import numpy as np
import pandas as pd
```

(continues on next page)

(continued from previous page)

```

from sklearn.model_selection import train_test_split
from tensorflow.keras.utils import to_categorical
import tensorflow as tf

# CV2 module is required to read PNG images.
import cv2
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import OneHotEncoder
import sys
tf.compat.v1.logging.set_verbosity(tf.compat.v1.logging.ERROR)

```

## 1.4.2 Helper functions

```

# This section contains helper functions.

# Plot the confusion matrix.
def textplotting(model,x_test,y_test):
    y_pred = model.predict(x_test)
    y_pred = np.argmax(y_pred, axis=1)
    y_test = np.argmax(y_test, axis=1)
    y_pred = y_pred.reshape(-1,1)
    y_test = y_test.reshape(-1,1)
    print(confusion_matrix(y_test, y_pred))
    print(accuracy_score(y_test, y_pred))

# Plot the confusion matrix.
def metric(best_model,x_train,y_train,x_test,y_test):

    y_pred = best_model.predict(x_test)
    sn.set(font_scale=2)
    rcParams['figure.figsize'] = 7, 7
    confusion_matrix = pd.crosstab(y_test.argmax(axis=1), y_pred.argmax(axis=1),
    ↪rownames=['Actual'], colnames=['Predicted'])
    sn.heatmap(confusion_matrix, annot=True)

    plt.savefig("Test.png")
    plt.clf()
    confusion_matrix = pd.crosstab(y_train.argmax(axis=1), best_model.predict(x_train).
    ↪argmax(axis=1), rownames=['Actual'], colnames=['Predicted'])
    sn.heatmap(confusion_matrix, annot=True)

    plt.savefig("Train.png")
    plt.clf()

# Plot the parameter reduction.
def simplot(parameterslist):
    plt.clf()
    sn.set(font_scale=1)
    rcParams['figure.figsize'] = 10, 10
    plt.rcParams["axes.edgecolor"] = "black"
    plt.rcParams["axes.linewidth"] = 1
    lists = sorted(parameterslist.items())
    x, y = zip(*lists)
    plt.minorticks_on()
    fig,ax=plt.subplots()

```

(continues on next page)

(continued from previous page)

```

ax.plot(x, y, marker="o")
ax.grid(True)
ax.patch.set_edgecolor('black')
ax.patch.set_linewidth('1')
ax.set_xscale('log')
ax.set_xlabel("Number of Parameters")
ax.set_ylabel("Accuracy")
plt.savefig("Parameters.png")
plt.show()

# Save parameter reduction file in a particular format.
def saveLogfile(NeuronsInLayer1,NeuronsInLayer2,IntermediateAccuracy,
↪IntermediateParameters):
    data = np.column_stack([NeuronsInLayer1,NeuronsInLayer2,IntermediateAccuracy,
↪IntermediateParameters])
    datafile_path = "./Logfile.txt"
    np.savetxt(datafile_path , data, fmt=['%f','%d','%f','%d'])

# This segment is used to place images in specific directories.
def makedirectory(name,x_train,y_train,x_test, y_test,x_val, y_val):
    base_dir = './fold'+str(name)
    if not os.path.isdir(base_dir):
        os.mkdir(base_dir)

    train_dir = os.path.join(base_dir, 'train')
    if not os.path.isdir(train_dir):
        os.mkdir(train_dir)

    validation_dir = os.path.join(base_dir, 'validation')
    if not os.path.isdir(validation_dir):
        os.mkdir(validation_dir)

    test_dir = os.path.join(base_dir, 'test')
    if not os.path.isdir(test_dir):
        os.mkdir(test_dir)

# Make a folder for train,test, and validation set for each category.
for loop in folders:
    temptrain = os.path.join(train_dir, loop)
    if not os.path.isdir(temptrain):
        os.mkdir(temptrain)
    tempvalid = os.path.join(validation_dir, loop)
    if not os.path.isdir(tempvalid):
        os.mkdir(tempvalid)
    temptest = os.path.join(test_dir, loop)

    if not os.path.isdir(temptest):
        os.mkdir(temptest)

    for loop2 in range(0,len(y_train)):
        if y_train[loop2]==loop:

```

(continues on next page)

(continued from previous page)

```
        im = Image.fromarray(x_train[loop2])
        im.save(temptrain+os.sep+str(loop2)+".jpeg")
    for loop2 in range(0,len(y_test)):
        if y_test[loop2]==loop:
            im = Image.fromarray(x_test[loop2])
            im.save(tempptest+os.sep+str(loop2)+".jpeg")
    for loop2 in range(0,len(y_val)):
        if y_val[loop2]==loop:
            im = Image.fromarray(x_val[loop2])
            im.save(tempvalid+os.sep+str(loop2)+".jpeg")

# Read images
def load_images_from_folder(folder,shape):
    images = []
    for filename in os.listdir(folder):
        img = cv2.imread(os.path.join(folder, filename))
        if img is not None:
            img = cv2.resize(img, (shape, shape))
            images.append(img)

    return images
```

### 1.4.3 Step 1 - Train the model.

```
# The folders in which images are placed.
folders = [
    'Ash',
    'Beech',
    'Hornbeam',
    'Mountainoak',
    'Sycamoremaple',
]

label_encoder = LabelEncoder()
enc = OneHotEncoder(handle_unknown='ignore')

# Any number of image generators can be specified and used, but we used 4 different
↳ image augmentation architectures.
gen=[1,2,3,4]

# Images can be reshaped to various sizes to reduce the training time and neural
↳ network size.
shapes = [50,100,200,300]

# For each shape
for shape in shapes:
    print("Shape:", shape)
    X = load_images_from_folder(folders[0],shape)
    Y = [folders[0]]*len(X)
    cat = 1
    for loop in range(1,len(folders)):
        print("Processing : ",folders[loop])
        tempX = load_images_from_folder(folders[loop],shape)
        tempY = [folders[loop]]*len(tempX)
```

(continues on next page)



(continued from previous page)

```

Y = Y + tempY
X = X + tempX

X = np.array(X)
Y= np.array(Y)
skf = StratifiedKFold(n_splits=5)
skf.get_n_splits(X, Y)
fold_no = 1
acc_per_fold = {}
loss_per_fold = {}
input_shape = (shape , shape , 3)
batch_size=1
# For cross validation
for train_index, test_index in skf.split(X, Y):
    x_train, x_test = X[train_index], X[test_index]
    y_train, y_test = Y[train_index], Y[test_index]
    x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size=0.1,
    random_state=1)
    makedirs(str(fold_no)+"_"+str(shape),x_train,y_train,x_test, y_test,x_val, y_
    val)
    train_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'train')
    validation_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'validation')
    test_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'test')
    y_train = to_categorical(label_encoder.fit_transform(y_train))
    y_test = to_categorical(label_encoder.fit_transform(y_test))
    y_val = to_categorical(label_encoder.fit_transform(y_val))

    model = Sequential()
    model = Sequential()
    model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
    model.add(MaxPooling2D(pool_size = (2, 2)))
    #model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
    #model.add(MaxPooling2D(pool_size = (2, 2)))
    # Use the best number of neurons for the first layer in this layer obtained from
    the previous step.
    #model.add(Conv2D(32, kernel_size=(3,3), input_shape=input_shape))
    #model.add(Conv2D(64, kernel_size=(3,3)))
    # Use the best number of neurons for the second layer in this layer obtained from
    the previous step.
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Flatten())
    model.add(Dense(128, activation=tf.nn.relu))
    model.add(Dropout(0.2))
    model.add(Dense(50, activation=tf.nn.relu))
    model.add(Dropout(0.2))
    model.add(Dense(20, activation=tf.nn.relu))
    model.add(Dropout(0.2))
    model.add(Dense(len(folders), activation=tf.nn.softmax))

    # For each generator
    for selectgen in gen:

```

(continues on next page)

(continued from previous page)

```

if selectgen==1:
    train_datagen = ImageDataGenerator(
        rescale=1./255)
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==2:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
elif selectgen==3:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        featurewise_center=True,
        featurewise_std_normalization=True,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)
else:
    train_datagen = ImageDataGenerator(
        rescale=1./255,
        rotation_range=40,
        width_shift_range=0.2,
        height_shift_range=0.2,
        shear_range=0.2,
        featurewise_center=True,
        featurewise_std_normalization=True,
        zoom_range=0.2,
        brightness_range=[0.2,1.0],
        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest')
    test_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

validation_generator = test_datagen.flow_from_directory(
    validation_dir,
    target_size=(shape , shape ),
    batch_size=batch_size,
    class_mode='categorical')

model.compile(optimizer='adam',
              loss='categorical_crossentropy',
              metrics=['accuracy'])

```

(continues on next page)

(continued from previous page)

```

history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.n // batch_size,
    epochs=30,
    validation_data=validation_generator,
    validation_steps=validation_generator.n // batch_size,
    verbose=0)

    #print("DONE")
test_generator = test_datagen.flow_from_directory(
    test_dir,
    target_size=(shape , shape),
    batch_size=batch_size ,
    class_mode='categorical')
test_loss, test_acc = model.evaluate_generator(test_generator, steps=50)
print('test acc:', test_acc)
if selectgen not in acc_per_fold:
    acc_per_fold[selectgen] = [test_acc * 100]
    loss_per_fold[selectgen] = [test_loss]
else:
    acc_per_fold[selectgen].append(test_acc * 100)
    loss_per_fold[selectgen].append(test_loss)
#os.system("rm -rf fold"+str(fold_no)+"_" +str(shape))
fold_no = fold_no + 1

# For each generator, print the results for each fold.
for g in gen:
    print("Generator",g)
    acc_per_foldt = acc_per_fold[g]
    loss_per_foldt = loss_per_fold[g]
    print('-----')
    print('Score per fold')
    for i in range(0, len(acc_per_foldt)):
        print('-----')
        print(f'> Fold {i+1} - Loss: {loss_per_foldt[i]} - Accuracy: {acc_per_
        foldt[i]}%')
        print('-----')
    print('Average scores for all folds:')
    print(f'> Accuracy: {np.mean(acc_per_foldt)} (+- {np.std(acc_per_foldt)})')
    print(f'> Loss: {np.mean(loss_per_foldt)}')
    print('-----')

```

#### 1.4.4 Step 2 - Convnet Model to TFlite And AddDetadata

```

gen = [3]
shapes = [224]
for shape in shapes:
    X = load_images_from_folder(folders[0],shape)
    Y = [folders[0]]*len(X)
    cat = 1
    for loop in range(1,len(folders)):
        print("Processing : ",folders[loop])
        tempX = load_images_from_folder(folders[loop],shape)
        tempY = [folders[loop]]*len(tempX)

```

(continues on next page)

(continued from previous page)

```

Y = Y + tempY
X = X + tempX

X = np.array(X)
Y= np.array(Y)
skf = StratifiedKFold(n_splits=5)
input_shape = (shape , shape , 3)
for selectgen in gen:
    batch_size=1
    acc_per_fold = []
    loss_per_fold = []
    fold_no = 0
    for train_index, test_index in skf.split(X, Y):
        x_train, x_test = X[train_index], X[test_index]
        y_train, y_test = Y[train_index], Y[test_index]
        x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_
→size=0.1, random_state=1)
        makedirs(str(fold_no)+"_"+str(shape),x_train,y_train,x_test, y_test,
→x_val, y_val)
        train_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'train')
        validation_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape),
→'validation')
        test_dir = os.path.join("fold"+str(fold_no)+"_"+str(shape), 'test')
        y_train = to_categorical(label_encoder.fit_transform(y_train))
        y_test = to_categorical(label_encoder.fit_transform(y_test))
        y_val = to_categorical(label_encoder.fit_transform(y_val))
        #print("Shape ",str(shape))
        model = Sequential()
        model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        model.add(Conv2D(32, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
        model.add(MaxPooling2D(pool_size = (2, 2)))
        #model.add(Conv2D(64, (3, 3), activation="relu", input_shape=input_shape))
        #model.add(MaxPooling2D(pool_size = (2, 2)))
        # Use the best number of neurons for the first layer in this layer_
→obtained from the previous step.
        #model.add(Conv2D(32, kernel_size=(3,3), input_shape=input_shape))
        #model.add(Conv2D(64, kernel_size=(3,3)))
        # Use the best number of neurons for the second layer in this layer_
→obtained from the previous step.
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Flatten())
        model.add(Dense(128, activation=tf.nn.relu))
        model.add(Dropout(0.2))
        model.add(Dense(50, activation=tf.nn.relu))
        model.add(Dropout(0.2))
        model.add(Dense(20, activation=tf.nn.relu))
        model.add(Dropout(0.2))
        model.add(Dense(len(folders), activation=tf.nn.softmax))
        param = model.count_params()

    if selectgen==1:

```

(continues on next page)

(continued from previous page)

```

        train_datagen = ImageDataGenerator(
            rescale=1./255)
        test_datagen = ImageDataGenerator(rescale=1./255)
    elif selectgen==2:
        train_datagen = ImageDataGenerator(
            rescale=1./255,
            rotation_range=40,
            brightness_range=[0.2,1.0],
            horizontal_flip=True,
            vertical_flip=True,
            fill_mode='nearest')
        test_datagen = ImageDataGenerator(rescale=1./255)
    elif selectgen==3:
        train_datagen = ImageDataGenerator(
            rescale=1./255,
            rotation_range=40,
            featurewise_center=True,
            featurewise_std_normalization=True,
            brightness_range=[0.2,1.0],
            horizontal_flip=True,
            vertical_flip=True,
            fill_mode='nearest')
        test_datagen = ImageDataGenerator(rescale=1./255)
    else:
        train_datagen = ImageDataGenerator(
            rescale=1./255,
            rotation_range=40,
            width_shift_range=0.2,
            height_shift_range=0.2,
            shear_range=0.2,
            featurewise_center=True,
            featurewise_std_normalization=True,
            zoom_range=0.2,
            brightness_range=[0.2,1.0],
            horizontal_flip=True,
            vertical_flip=True,
            fill_mode='nearest')
        test_datagen = ImageDataGenerator(rescale=1./255)

    train_generator = train_datagen.flow_from_directory(
        train_dir,
        target_size=(shape , shape ),
        batch_size=batch_size,
        class_mode='categorical')

    validation_generator = test_datagen.flow_from_directory(
        validation_dir,
        target_size=(shape , shape ),
        batch_size=batch_size,
        class_mode='categorical')

    model.compile(optimizer='adam',
                  loss='categorical_crossentropy',
                  metrics=['accuracy'])

    history = model.fit_generator(

```

(continues on next page)

(continued from previous page)

```

        train_generator,
        steps_per_epoch=train_generator.n // batch_size,
        epochs=30,
        validation_data=validation_generator,
        validation_steps=validation_generator.n // batch_size,
        verbose=0)

    #print("DONE")
    test_generator = test_datagen.flow_from_directory(
        test_dir,
        target_size=(shape , shape),
        batch_size=batch_size ,
        class_mode='categorical')

    test_loss, test_acc = model.evaluate_generator(test_generator, steps=test_
    ↪generator.n // batch_size)

    acc_per_fold.append(test_acc * 100)
    loss_per_fold.append(test_loss)
    fold_no = fold_no + 1

    ### In this step we will simply save the model
    model_save_path = "."
    tf.saved_model.save(model, model_save_path)
    converter = tf.lite.TFLiteConverter.from_saved_model(model_save_path)
    tflite_model = converter.convert()

    # Save the simple model.
    open("model.tflite", "wb").write(tflite_model)

    # Generate Quantized Model
    import tensorflow_model_optimization as tfmot
    quantize_model = tfmot.quantization.keras.quantize_model

    # q_aware stands for for quantization aware.
    q_aware_model = quantize_model(model)
    # `quantize_model` requires a recompile.
    q_aware_model.compile(optimizer='adam',
        loss='categorical_crossentropy',
        metrics=['accuracy'])
    history = q_aware_model.fit_generator(
        train_generator,
        steps_per_epoch=train_generator.n // batch_size,
        epochs=30,
        validation_data=validation_generator,
        validation_steps=validation_generator.n // batch_size,
        verbose=1)
    q_aware_model.summary()

    _, baseline_model_accuracy = model.evaluate(x_test/255, y_test, verbose=0)
    ↪_, q_aware_model_accuracy = q_aware_model.evaluate(x_test/255, y_test,
    ↪verbose=0)

    print('Baseline test accuracy:', baseline_model_accuracy)
    print('Quant test accuracy:', q_aware_model_accuracy)

    textplotting(model,x_train/255,y_train)

```

(continues on next page)

(continued from previous page)

```

textplotting(model,x_test/255,y_test)
textplotting(q_aware_model,x_train/255,y_train)
textplotting(q_aware_model,x_test/255,y_test)

#q_aware_model.save("qalsome.hd5")
### In this step we will simply save the model
model_save_path = "./"
tf.saved_model.save(q_aware_model, model_save_path)
converter = tf.lite.TFLiteConverter.from_saved_model(model_save_path)
tflite_model = converter.convert()
# Save the quantized model.
open("qmodel.tflite", "wb").write(tflite_model)

# We can save any model for the 5 fold validation, so we exit the code_
↪after the training of the first fold.

```

### 1.4.5 Step 3 - Add Metadata to Mode

```

import os
# Change values in metadata_writer_for_image_classifier.py
#_MODEL_INFO = {
#    "model.tflite":
#        ModelSpecificInfo(
#            name="model",
#            version="v1",
#            image_width=100,
#            image_height=100,
#            image_min=0,
#            image_max=1,
#            mean=[0],
#            std=[255],
#            num_classes=5,
#            author="Alpha")
#}

# Change labels in labels.txt file

os.mkdir("Finalmodel")
os.system("python ./metadata_writer_for_image_classifier.py --model_file=model.tflite_
↪--label_file=labels.txt --export_directory=./Finalmodel")

```

### 1.4.6 Step 4 - Make Android Application

Make Android Application - TensorFlow Lite

The following picture shows the android studio, and we have to change the assets.

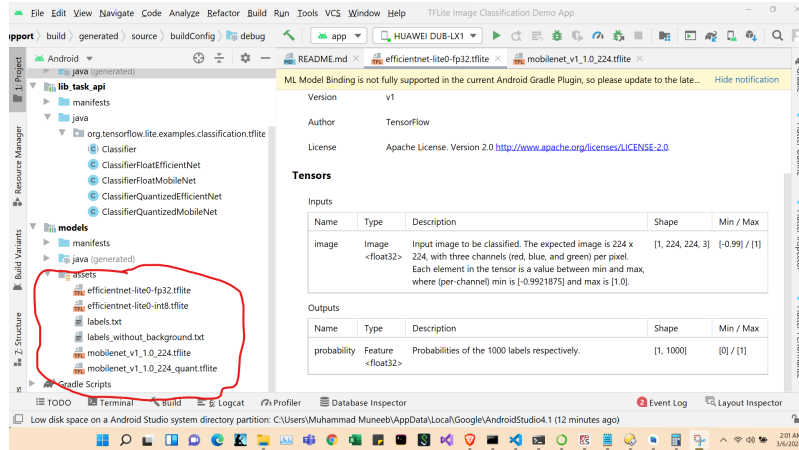
Replace

G:/Application/Tree Application/Tensorflow Application/models/src/main/assets

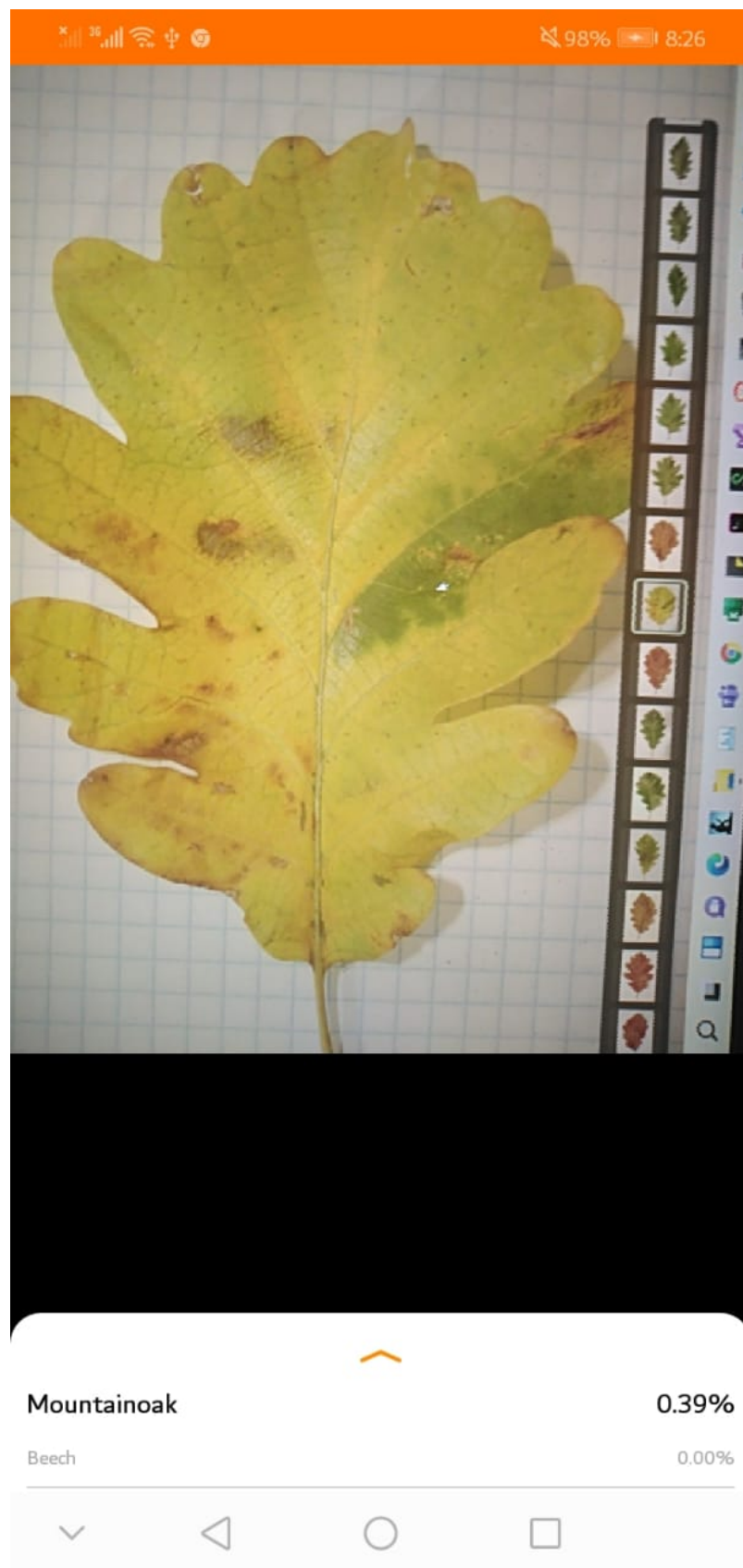
With

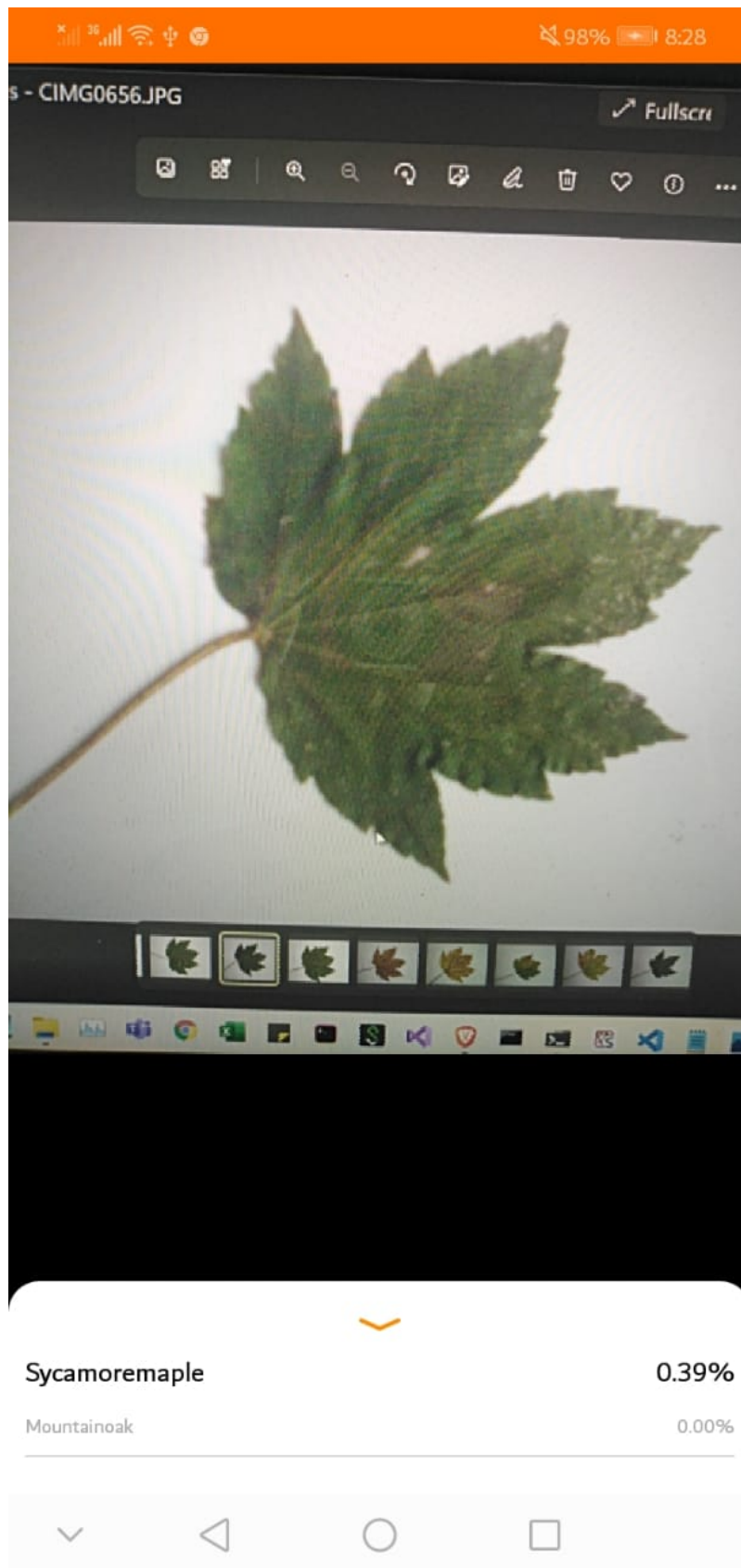
Finalmodel/model and rename model with efficientnet-lite0-int8

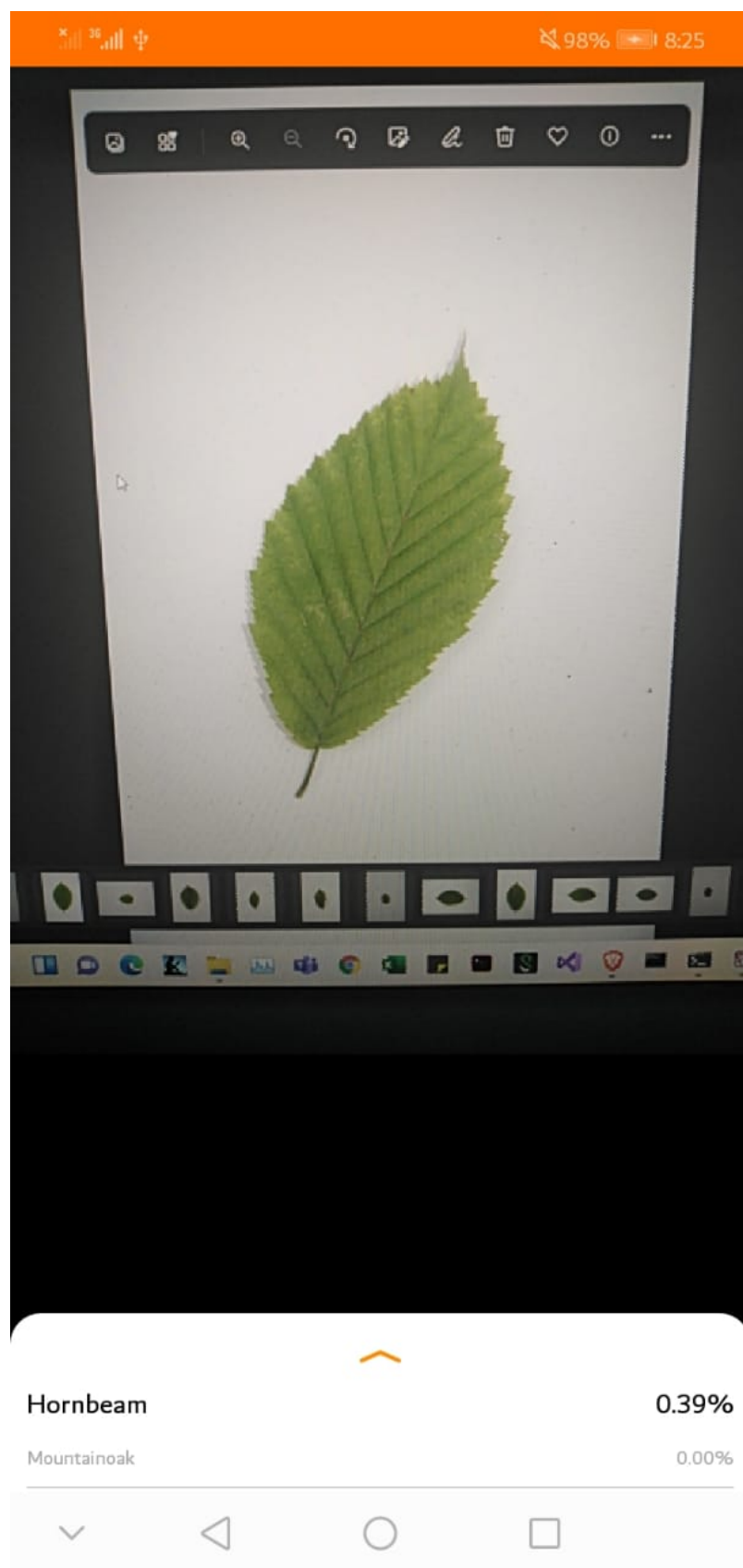
Open Android studio and open Image-Classificaion-master, build the application.

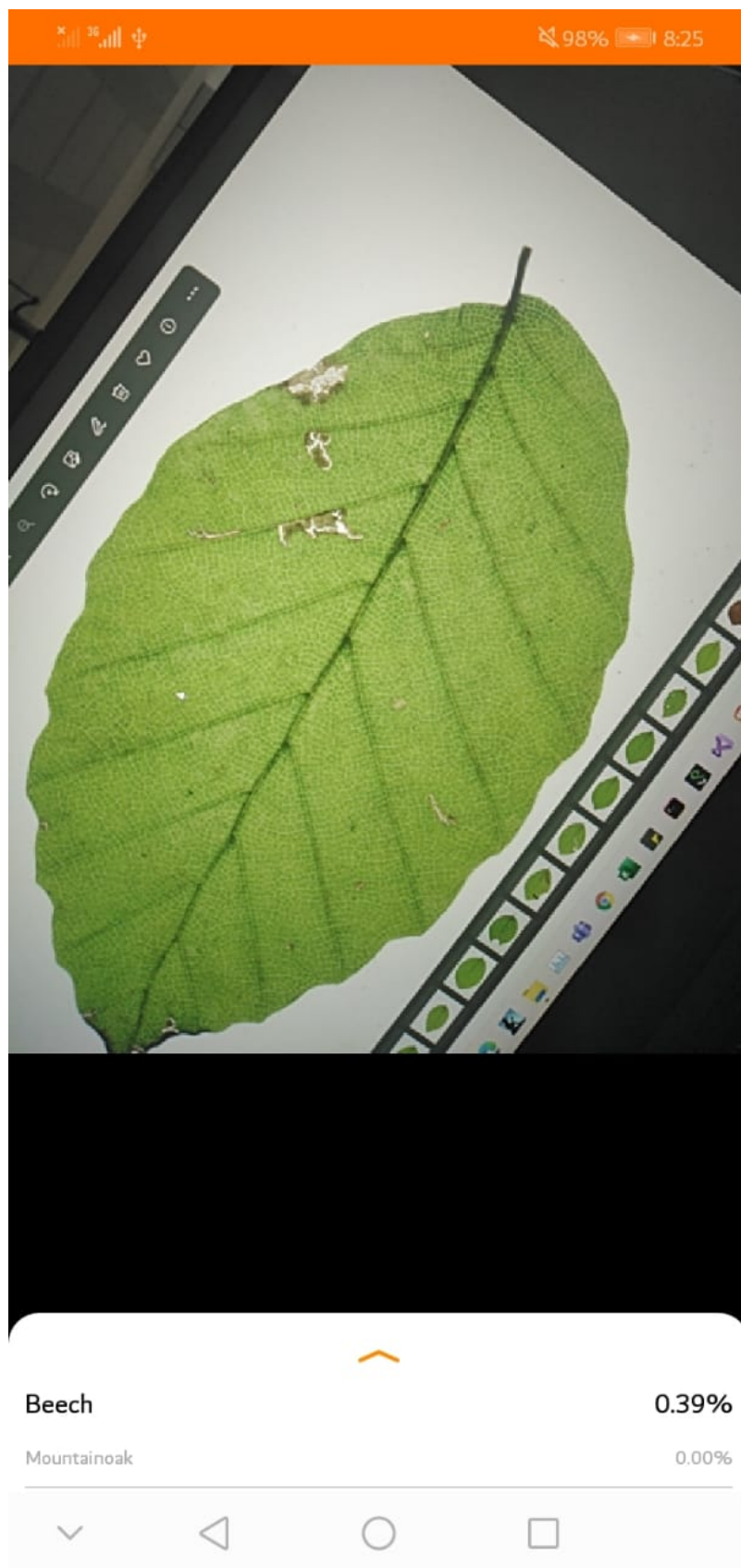


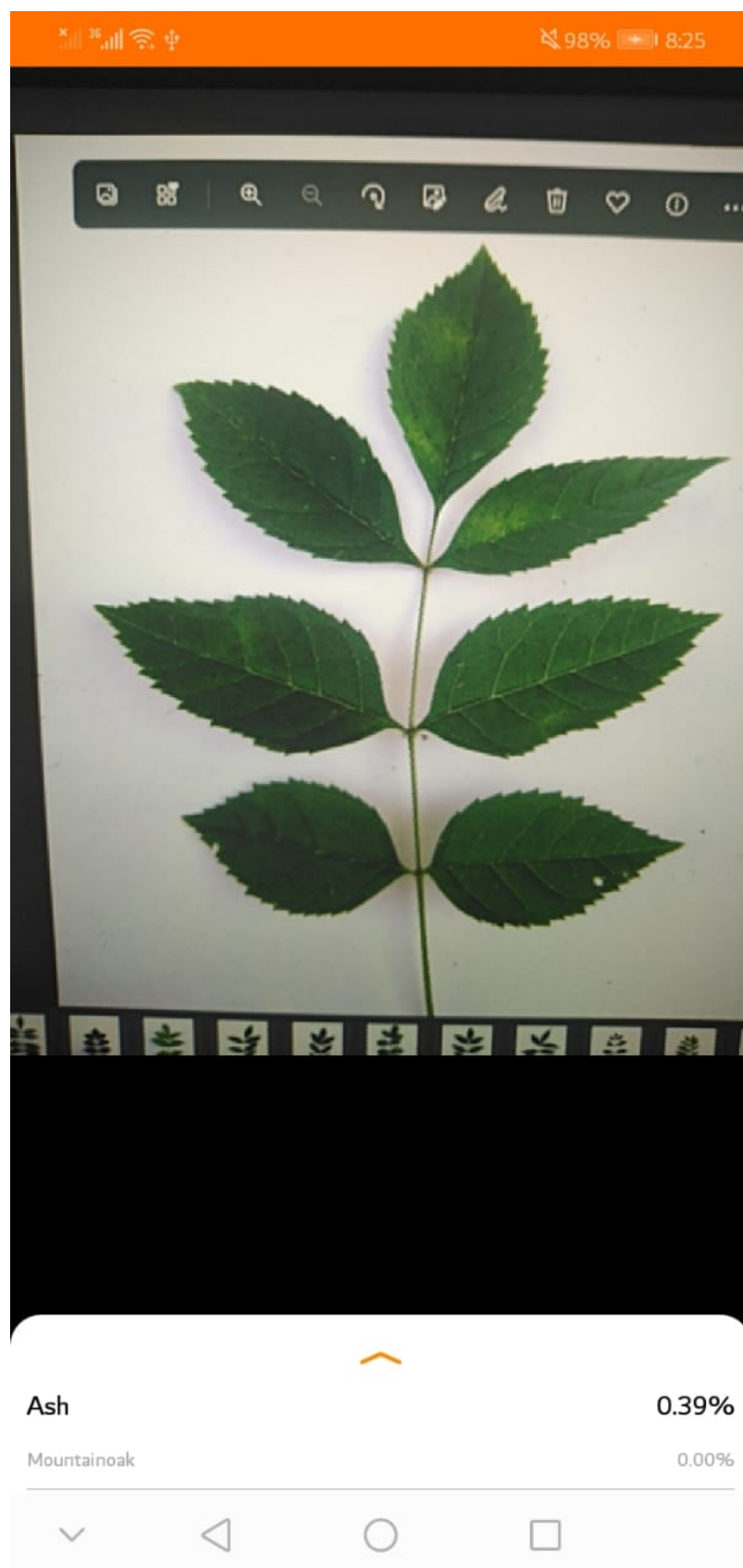












Make Android Application - Flutter

The following picture shows the android studio, and we have to change the assets.

Replace

G:/Application/Tree Application/Cat-Dog-Classifier-main/assets

With

Finalmodel/qmodel and rename model with model\_unquant

Open Android studio and open Cat-Dog-Classifier-main, build the application.

