MobileNet implementation on Cellphone

```
import tensorflow as tf
from tensorflow.keras import Sequential, Model
from tensorflow.keras.layers import Dense, Dropout, Flatten, BatchNormalization
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from keras.applications import MobileNet
from keras.applications.mobilenet import preprocess input
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
# Drive mounting
from google.colab import drive
drive.mount('/content/drive')
    Mounted at /content/drive
img width=224
img height=224
batch_size=10
# Initializing training dataset
train data dir="/content/drive/My Drive/cell phone/training"
datagen = ImageDataGenerator(rescale=1./255)
train_generator = datagen.flow_from_directory(directory=train_data_dir,
                                              target size = (img width, img height),
                                               class mode = 'binary',
                                              batch_size=batch_size)
```

Found 245 images belonging to 2 classes.

Initializing validation dataset
validation_dir="/content/drive/My Drive/cell_phone/val"
val_generator=datagen.flow_from_directory(validation_dir,target_size=(img_width,img_height),classes=['cat','dog'],batch_size

Found 0 images belonging to 2 classes.

mobile=tf.keras.applications.mobilenet.MobileNet()

mobile.summary()

conv_pw_1 (Conv2D)	(None,	112, 112, 64)	2048
conv_pw_1_bn (BatchNormaliza	(None,	112, 112, 64)	256
conv_pw_1_relu (ReLU)	(None,	112, 112, 64)	0
conv_pad_2 (ZeroPadding2D)	(None,	113, 113, 64)	0
conv_dw_2 (DepthwiseConv2D)	(None,	56, 56, 64)	576
conv_dw_2_bn (BatchNormaliza	(None,	56, 56, 64)	256
conv_dw_2_relu (ReLU)	(None,	56, 56, 64)	0
conv_pw_2 (Conv2D)	(None,	56, 56, 128)	8192
conv_pw_2_bn (BatchNormaliza	(None,	56, 56, 128)	512
conv_pw_2_relu (ReLU)	(None,	56, 56, 128)	0
conv_dw_3 (DepthwiseConv2D)	(None,	56, 56, 128)	1152
conv_dw_3_bn (BatchNormaliza	(None,	56, 56, 128)	512
conv_dw_3_relu (ReLU)	(None,	56, 56, 128)	0

conv_pw_3 (Conv2D)	(None,	56,	56,	128)	16384
conv_pw_3_bn (BatchNormaliza	(None,	56,	56,	128)	512
conv_pw_3_relu (ReLU)	(None,	56,	56,	128)	0
conv_pad_4 (ZeroPadding2D)	(None,	57,	57,	128)	0
conv_dw_4 (DepthwiseConv2D)	(None,	28,	28,	128)	1152
conv_dw_4_bn (BatchNormaliza	(None,	28,	28,	128)	512
conv_dw_4_relu (ReLU)	(None,	28,	28,	128)	0
conv_pw_4 (Conv2D)	(None,	28,	28,	256)	32768
conv_pw_4_bn (BatchNormaliza	(None,	28,	28,	256)	1024
conv_pw_4_relu (ReLU)	(None,	28,	28,	256)	0
conv_dw_5 (DepthwiseConv2D)	(None,	28,	28,	256)	2304
conv_dw_5_bn (BatchNormaliza	(None,	28,	28,	256)	1024
conv_dw_5_relu (ReLU)	(None,	28,	28,	256)	0
conv_pw_5 (Conv2D)	(None,	28,	28,	256)	65536
conv_pw_5_bn (BatchNormaliza	(None,	28,	28,	256)	1024
conv_pw_5_relu (ReLU)	(None,	28,	28,	256)	0
cony nad 6 (ZanoDadding)D)	/ None	20	20	2561	α

[#] This says that not to use dense values mentioned in VGG (Not to use VGG layers)
for layers in mobile.layers:
 layers.trainable=False

Training the model
model=Sequential()
model.add(mobile)

```
model.add(Flatten())
model.add(Dense(128,activation='relu',))
model.add(Dropout(0.5))
model.add(BatchNormalization())
model.add(Dense(1,activation="sigmoid"))
model.compile(optimizer="adam",loss="binary crossentropy",metrics=['accuracy'])
history=model.fit generator(generator=train generator, steps per epoch=len(train generator), epochs = 30,
                              validation data=val generator, validation steps=len(val generator)
                              , verbose = 2)
     WARNING:tensorflow:From <ipython-input-12-624fd0d036d0>:3: Model.fit_generator (from tensorflow.python.keras.engine
     Instructions for updating:
     Please use Model.fit, which supports generators.
     Epoch 1/30
     25/25 - 111s - loss: 0.4701 - accuracy: 0.7796
     Epoch 2/30
     25/25 - 9s - loss: 0.2606 - accuracy: 0.8980
     Epoch 3/30
     25/25 - 9s - loss: 0.1737 - accuracy: 0.9551
     Epoch 4/30
     25/25 - 9s - loss: 0.2366 - accuracy: 0.9184
     Epoch 5/30
     25/25 - 9s - loss: 0.1381 - accuracy: 0.9510
     Epoch 6/30
     25/25 - 9s - loss: 0.1014 - accuracy: 0.9755
     Epoch 7/30
     25/25 - 9s - loss: 0.0886 - accuracy: 0.9714
     Epoch 8/30
     25/25 - 9s - loss: 0.0999 - accuracy: 0.9755
     Epoch 9/30
     25/25 - 9s - loss: 0.0627 - accuracy: 0.9918
     Epoch 10/30
     25/25 - 9s - loss: 0.0859 - accuracy: 0.9714
     Epoch 11/30
     25/25 - 9s - loss: 0.0678 - accuracy: 0.9755
     Epoch 12/30
```

```
25/25 - 9s - loss: 0.0766 - accuracy: 0.9714
Epoch 13/30
25/25 - 9s - loss: 0.0487 - accuracy: 0.9837
Epoch 14/30
25/25 - 9s - loss: 0.0731 - accuracy: 0.9755
Epoch 15/30
25/25 - 9s - loss: 0.0868 - accuracy: 0.9673
Epoch 16/30
25/25 - 9s - loss: 0.0707 - accuracy: 0.9755
Epoch 17/30
25/25 - 9s - loss: 0.0729 - accuracy: 0.9755
Epoch 18/30
25/25 - 10s - loss: 0.0551 - accuracy: 0.9837
Epoch 19/30
25/25 - 9s - loss: 0.0420 - accuracy: 0.9837
Epoch 20/30
25/25 - 9s - loss: 0.0591 - accuracy: 0.9796
Epoch 21/30
25/25 - 9s - loss: 0.0697 - accuracy: 0.9837
Epoch 22/30
25/25 - 9s - loss: 0.0604 - accuracy: 0.9837
Epoch 23/30
25/25 - 9s - loss: 0.0572 - accuracy: 0.9714
Epoch 24/30
25/25 - 9s - loss: 0.0836 - accuracy: 0.9796
Epoch 25/30
25/25 - 9s - loss: 0.0544 - accuracy: 0.9878
Epoch 26/30
25/25 - 9s - loss: 0.0274 - accuracy: 0.9959
Epoch 27/30
25/25 - 9s - loss: 0.0782 - accuracy: 0.9796
Epoch 28/30
```

from tensorflow.keras.preprocessing import image

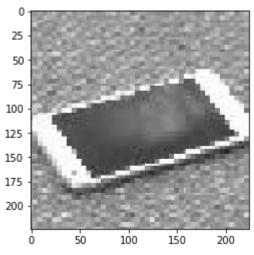
```
img=image.load_img("/content/img130.png",target_size=(img_width,img_height))
plt.imshow(img)
img=image.img_to_array(img)
```

```
img=img/255.0
img = np.expand_dims(img, axis=0)
img_class = np.argmax(model.predict(img),axis=1)

if(model.predict(img)<=0.5):
    print('cellphone=NO')

else:
    print('cellphone=YES')</pre>
```

cellphone=YES



```
img=image.load_img("/content/img82.png",target_size=(img_width,img_height))
plt.imshow(img)
img=image.img_to_array(img)
img=img/255.0
img = np.expand_dims(img, axis=0)

if(model.predict(img)<=0.5):
    print('cellphone=NO')

else:
    print('cellphone=YES')</pre>
```

cellphone=NO

