A PROJECT ON

"FACE MASK DETECTION USING MobileNetV3Small"

Submitted to:

KIIT Deemed to be University, Bhubaneswar

In Partial Fulfilment of the Requirement for the

Award of BACHELOR'S DEGREE IN INFORMATION TECHNOLOGY

INPUT : CODE

LANGUAGE: PYTHON

VERSION: Python 3.9.12

IDE USED: JUPYTER

SUBMITTED BY:

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PROBLEM STATEMENT:

The outspread of the corona virus types has caused pandemic across the globe. The virus is known for severe acute respiratory issues. The ailment it causes is called coronavirus disease 2019 or (COVID-19) affecting our day-to-day life disrupting world trade and movements. Wearing a mask is one of the effective methods to prevent us from being affected by the vicious COVID virus. Due to carelessness and unawareness of the local public towards this issue a facemask detection system can be used by the authorities to keep a check on them and create awareness about the same. This project titled Face Mask Detection has been developed using a machine learning technique known as transfer learning and by using image classification method of MobileNetV3small .Steps followed while building the model are data collection, data preprocessing, data splitting, model-rectification, model testing and model implementation. Model is capable of predicting if individuals are wearing or not wearing masks at an accuracy of 99.81 percent. We also explored using other image classification models like VGG16 with an accuracy of 96 percent, MobileNetV2 with an accuracy of 97.21 percent and ResNet101 with an accuracy of 97.79 percent to detect masks. We have implemented the Transfer learning method that uses already acquired knowledge by avoiding overfitting of the models.

INPUT CODE:

```
import numpy as np
import os
import matplotlib.pyplot as plt
from imutils import paths
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.models import Sequential
             tensorflow.keras.preprocessing.image
from
                                                           import
ImageDataGenerator
from tensorflow.keras.preprocessing.image import img to array
from tensorflow.keras.preprocessing.image import load img
from tensorflow.keras.utils import to categorical
from sklearn.preprocessing import LabelBinarizer
from sklearn.model selection import train test split
from sklearn.metrics import classification report
from tensorflow.keras.applications import MobileNetV3Small
From tensorflow.keras.applications.mobilenet v3 import
preprocess input
```

```
from tensorflow.keras.layers import AveragePooling2D
from tensorflow.keras.layers import Dropout
from tensorflow.keras.layers import Flatten
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Input
from tensorflow.keras.models import Model
dataset=r'C:\Users\KIIT\Videos\MobileNetV3\dataset'
image site=list(paths.list images(dataset))
image site
images=[]
category=[]
for a in image site:
    label=a.split(os.path.sep)[-2]
    category.append(label)
    image=load img(a,target size=(224,224))
    image=img to array(image)
    image=preprocess input(image)
    images.append(image)
images=np.array(images,dtype='float32')
category=np.array(category)
Images.shape
category.shape
lb=LabelBinarizer()
category=lb.fit transform(category)
category=to categorical(category)
Category
train X, test X, train Y, test Y=train test split(images, category, t
est size=0.25,stratify=category,random state=10)
train X.shape
test X.shape
train Y.shape
test Y.shape
aug=ImageDataGenerator(rotation range=20,
zoom range=0.15,
width shift range=0.2,
shear range=0.15,
horizontal flip=True,
vertical flip=True)
```

```
baseModel=MobileNetV3Small(
    input tensor=Input(shape=(224,224,3)),
    alpha=1.0,
    minimalistic=False,
    include top=False,
    weights="imagenet",
    classes=1000,
   pooling=None,
    include preprocessing=True,
)
def init (mdl):
        super(MobileNetV3Small, mdl). init ()
        mdl.conv1 = tf.keras.layers.Conv2D(filters=16,
                                            kernel size=(3, 3),
                                            strides=2,
                                            padding="same")
       mdl.bn1 = tf.keras.layers.BatchNormalization()
              mdl.bneck1 = BottleNeck(in size=16, exp size=16,
out size=16, s=2, is se existing=True, NL="RE", k=3)
              mdl.bneck2 = BottleNeck(in size=16, exp size=72,
out size=24, s=2, is se existing=False, NL="RE", k=3)
             mdl.bneck3 = BottleNeck(in size=24, exp size=88,
out size=24, s=1, is se existing=False, NL="RE", k=3)
             mdl.bneck4 = BottleNeck(in size=24, exp size=96,
out size=40, s=2, is se existing=True, NL="HS", k=5)
             mdl.bneck5 = BottleNeck(in size=40, exp size=240,
out size=40, s=1, is se existing=True, NL="HS", k=5)
             mdl.bneck6 = BottleNeck(in size=40, exp size=240,
out size=40, s=1, is se existing=True, NL="HS", k=5)
             mdl.bneck7 = BottleNeck(in size=40, exp size=120,
out size=48, s=1, is se existing=True, NL="HS", k=5)
             mdl.bneck8 = BottleNeck(in size=48, exp size=144,
out size=48, s=1, is se existing=True, NL="HS", k=5)
             mdl.bneck9 = BottleNeck(in size=48, exp size=288,
out size=96, s=2, is se existing=True, NL="HS", k=5)
            mdl.bneck10 = BottleNeck(in size=96, exp size=576,
out size=96, s=1, is se existing=True, NL="HS", k=5)
            mdl.bneck11 = BottleNeck(in size=96, exp size=576,
out size=96, s=1, is se existing=True, NL="HS", k=5)
       mdl.conv2 = tf.keras.layers.Conv2D(filters=576,
                                            kernel size=(1, 1),
                                            strides=1,
                                            padding="same")
       mdl.bn2 = tf.keras.layers.BatchNormalization()
```

```
mdl.avgpool
tf.keras.layers.AveragePooling2D(pool size=(7, 7),
strides=1)
        mdl.conv3 = tf.keras.layers.Conv2D(filters=1280,
                                             kernel size=(1, 1),
                                             strides=1,
                                             padding="same")
        mdl.conv4 = tf.keras.layers.Conv2D(filters=NUM CLASSES,
                                             kernel size=(1, 1),
                                             strides=1,
                                             padding="same",
activation=tf.keras.activations.softmax)
def call(mdl, inputs, training=None, mask=None):
        x = mdl.conv1(inputs)
        x = mdl.bn1(x, training=training)
        x = h swish(x)
        x = mdl.bneck1(x, training=training)
        x = mdl.bneck2(x, training=training)
        x = mdl.bneck3(x, training=training)
        x = mdl.bneck4(x, training=training)
        x = mdl.bneck5(x, training=training)
        x = mdl.bneck6(x, training=training)
        x = mdl.bneck7(x, training=training)
        x = mdl.bneck8(x, training=training)
        x = mdl.bneck9(x, training=training)
        x = mdl.bneck10(x, training=training)
        x = mdl.bneck11(x, training=training)
        x = mdl.conv2(x)
        x = mdl.bn2(x, training=training)
        x = h swish(x)
        x = mdl.avgpool(x)
        x = mdl.conv3(x)
        x = h swish(x)
        x = mdl.conv4(x)
        return x
```

baseModel.summary()

```
headModel=baseModel.output
headModel=AveragePooling2D(pool size=(7,7))(headModel)
headModel=Flatten(name='Flatten') (headModel)
headModel=Dense(1280,activation='swish') (headModel)
headModel=Dropout(0.5) (headModel)
headModel=Dense(2, activation='softmax')(headModel)
model=Model(inputs=baseModel.input,outputs=headModel)
model.summary()
learning rate=0.001
Epochs=20
BS=12
opt=Adam(lr=learning rate,decay=learning rate/Epochs)
model.compile(loss='binary crossentropy',optimizer=opt,metrics=[
'accuracy'])
history=model.fit(
    aug.flow(train X, train Y, batch size=BS),
    steps per epoch=len(train X)//BS,
    validation data=(test X, test Y),
    validation steps=len(test X)//BS,
    epochs=Epochs
)
model.save(r'C:\Users\KIIT\Videos\MobileNetV3\mobilenetv3Small.m
odel')
predict=model.predict(test X,batch size=BS)
predict=np.argmax(predict,axis=1)
print("CLASSIFICATION REPORT :\n")
print(classification report(test Y.argmax(axis=1),predict,target
names=lb.classes ))
history.history.keys()
import matplotlib.pyplot as plt
plt.plot(history.history['accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train'], loc='upper left')
plt.show()
plt.plot(history.history['loss'])
plt.title('model loss')
```

```
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train'], loc='upper left')
plt.show()
train datagen = ImageDataGenerator(zoom range=0.5)
train generator = train datagen.flow from directory(dataset,
target size=(224, 224), batch size=BS, class mode='categorical',
shuffle=True)
valid datagen = ImageDataGenerator()
valid generator=valid datagen.flow from directory(dataset, target
size=(224,224),batch size=BS,class mode='categorical',
shuffle=True)
test datagen = ImageDataGenerator()
test generator = test datagen.flow from directory(dataset,
target size=(224, 224), batch size=BS, class mode='categorical',
shuffle=False)
import itertools
       plot confusion matrix(cm,
                                    classes,
                                                  normalize=True,
title='Confusion matrix', cmap=plt.cm.Blues):
    plt.figure(figsize=(10,10))
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    plt.title(title)
    plt.colorbar()
    tick marks = np.arange(len(classes))
    plt.xticks(tick marks, classes, rotation=45)
    plt.yticks(tick marks, classes)
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        cm = np.around(cm, decimals=2)
```

```
cm[np.isnan(cm)] = 0.0
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')
    thresh = cm.max() / 2.
                           itertools.product(range(cm.shape[0]),
fori,
                   in
range(cm.shape[1])):
        plt.text(j, i, cm[i, j],
                 horizontalalignment="center",
                       color="white" if cm[i, j] > thresh else
"black")
   plt.tight layout()
   plt.ylabel('True label')
    plt.xlabel('Predicted label')
target names = []
for key in train generator.class indices:
    target names.append(key)
from sklearn.metrics import confusion matrix
Y pred = model.predict generator(test generator)
y pred = np.argmax(Y pred, axis=1)
print('Confusion Matrix')
cm = confusion matrix(test generator.classes, y pred)
plt.figure(figsize=(20, 20))
plot_confusion_matrix(cm, target_names, title='Confusion
Matrix')
plt.title('Confusion matrix')
plt.show()
```

MODEL TESTING:

```
tensorflow.keras.applications.mobilenet v2
                                                              import
from
preprocess input
from tensorflow.keras.preprocessing.image import img to array
from tensorflow.keras.models import load model
import numpy as np
import cv2
import os
from imutils.video import VideoStream
import imutils
def detect and predict mask(frame, faceNet, maskNet):
    (h, w) = frame.shape[:2]
blob=cv2.dnn.blobFromImage(frame, 1.0, (300, 300), (104.0, 177.0, 123.
0))
    faceNet.setInput(blob)
    detections=faceNet.forward()
    faces=[]
    locs=[]
    preds=[]
    for i in range(0,detections.shape[2]):
        confidence=detections[0,0,i,2]
        if confidence>0.5:
        #we need the X,Y coordinates
            box=detections[0,0,i,3:7]*np.array([w,h,w,h])
             (startX, startY, endX, endY) = box.astype('int')
             (startX, startY) = (max(0, startX), max(0, startY))
             (endX, endY) = (min(w-1, endX), min(h-1, endY))
            face=frame[startY:endY, startX:endX]
            face=cv2.cvtColor(face,cv2.COLOR BGR2RGB)
            face=cv2.resize(face,(224,224))
            face=img to array(face)
            face=preprocess input(face)
            faces.append(face)
            locs.append((startX,startY,endX,endY))
        if len(faces)>0:
```

```
faces=np.array(faces,dtype='float32')
             preds=maskNet.predict(faces,batch size=12)
         return (locs,preds)
 maskNet=load model(r'C:\Users\KIIT\Videos\Face Mask Detection\mo
 bilenet v3.model')
 prototxtPath=os.path.sep.join([r'C:\Users\KIIT\Videos\Face Mask
 Detection\face detector','deploy.prototxt'])
 weightsPath=os.path.sep.join([r'C:\Users\KIIT\Videos\Face Mask D
 etection\face detector','res10 300x300 ssd iter 140000.caffemode
 faceNet=cv2.dnn.readNet(prototxtPath,weightsPath)
 vs=VideoStream(src=0).start()
 while True:
     frame=vs.read()
     frame=imutils.resize(frame, width=400)
     (locs,preds) = detect and predict mask(frame,faceNet,maskNet)
     for (box,pred) in zip(locs,preds):
          (startX, startY, endX, endY) = box
          (mask, withoutMask) = pred
         label='Mask' if mask>withoutMask else 'No Mask'
         color=(0,255,0) if label=='Mask' else (0,0,255)
 cv2.putText(frame, label, (startX, startY-10), cv2.FONT HERSHEY SIMP
 LEX, 0.45, color, 2)
         rY=(endY - startY)//2
              cv2.circle(frame,((endX - startX)//2+startX,(endY -
 startY) //2+startY) ,rY,color,3)
     cv2.imshow("Face Mask Detector",frame)
     key=cv2.waitKey(1) & 0xFF
     if key==ord('o'):
         break
 cv2.destroyAllWindows()
 vs.stop()
 #
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```