from keras.applications import imagenet\_utils from keras.applications.vgg16 import preprocess\_input, decode\_predictions import random import glob import matplotlib.pyplot as plt import warnings #from IPython.display import Image #from PIL import Image warnings.simplefilter(action='ignore', category=FutureWarning) %matplotlib inline In [17]: #Organize data into train , valid and test os.chdir('C:\\Users\\User\\Desktop\\Projects\\Signlanguage') if os.path.isdir('train/0') is False: os.mkdir('train') os.mkdir('valid') os.mkdir('test') **for** i **in** range (0,10): shutil.move(f'{i}','train') os.mkdir(f'valid/{i}') os.mkdir(f'test/{i}') valid\_samples=random.sample(os.listdir(f'train/{i}'),30) for j in valid\_samples: shutil.move(f'train/{i}/{j}', f'valid/{i}') test\_samples=random.sample(os.listdir(f'train/{i}'),5) for k in test\_samples: shutil.move(f'train/{i}/{k}', f'test/{i}') os.chdir('../..') train\_path='C:\\Users\\User\\Desktop\\Projects\\Signlanguage\\train' In [18]: valid\_path='C:\\Users\\User\\Desktop\\Projects\\Signlanguage\\valid' test\_path='C:\\User\\Desktop\\Projects\\Signlanguage\\test' train\_batches = ImageDataGenerator(preprocessing\_function=tf.keras.applications.mobilenet.preprocess\_input) \ .flow\_from\_directory(directory=train\_path, target\_size=(224,224), batch\_size=10) valid\_batches = ImageDataGenerator(preprocessing\_function=tf.keras.applications.mobilenet.preprocess\_input) \ .flow\_from\_directory(directory=valid\_path, target\_size=(224,224), batch\_size=10) test\_batches = ImageDataGenerator(preprocessing\_function=tf.keras.applications.mobilenet.preprocess\_input) \ .flow\_from\_directory(directory=test\_path, target\_size=(224,224), batch\_size=10, shuffle=False) Found 1712 images belonging to 10 classes. Found 300 images belonging to 10 classes. Found 50 images belonging to 10 classes. assert train\_batches.n==1712 In [20]: assert valid\_batches.n==300 assert test\_batches.n==50 assert train\_batches.num\_classes==valid\_batches.num\_classes==test\_batches.num\_classes==10 mobile=tf.keras.applications.mobilenet.MobileNet() In [22]: mobile.summary() Model: "mobilenet\_1.00\_224" Layer (type) Output Shape Param # input\_2 (InputLayer) [(None, 224, 224, 3)] conv1 (Conv2D) (None, 112, 112, 32) 864 conv1\_bn (BatchNormalization (None, 112, 112, 32) 128 conv1\_relu (ReLU) (None, 112, 112, 32) conv\_dw\_1 (DepthwiseConv2D) (None, 112, 112, 32) 288 conv\_dw\_1\_bn (BatchNormaliza (None, 112, 112, 32) 128 conv\_dw\_1\_relu (ReLU) (None, 112, 112, 32) 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) 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) 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) conv\_dw\_3 (DepthwiseConv2D) (None, 56, 56, 128) 1152 conv\_dw\_3\_bn (BatchNormaliza (None, 56, 56, 128) conv\_dw\_3\_relu (ReLU) (None, 56, 56, 128) 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) (None, 57, 57, 128) conv\_pad\_4 (ZeroPadding2D) 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) 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) 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) 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) conv\_pad\_6 (ZeroPadding2D) (None, 29, 29, 256) conv\_dw\_6 (DepthwiseConv2D) (None, 14, 14, 256) 2304 conv\_dw\_6\_bn (BatchNormaliza (None, 14, 14, 256) 1024 conv\_dw\_6\_relu (ReLU) (None, 14, 14, 256) conv\_pw\_6 (Conv2D) (None, 14, 14, 512) 131072 conv\_pw\_6\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_pw\_6\_relu (ReLU) (None, 14, 14, 512) (None, 14, 14, 512) conv\_dw\_7 (DepthwiseConv2D) 4608 conv\_dw\_7\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_dw\_7\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_7 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_7\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_pw\_7\_relu (ReLU) (None, 14, 14, 512) conv\_dw\_8 (DepthwiseConv2D) (None, 14, 14, 512) 4608 conv\_dw\_8\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_dw\_8\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_8 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_8\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_pw\_8\_relu (ReLU) (None, 14, 14, 512) conv\_dw\_9 (DepthwiseConv2D) (None, 14, 14, 512) 4608 conv\_dw\_9\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_dw\_9\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_9 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_9\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_pw\_9\_relu (ReLU) (None, 14, 14, 512) conv\_dw\_10 (DepthwiseConv2D) (None, 14, 14, 512) 4608 conv\_dw\_10\_bn (BatchNormaliz (None, 14, 14, 512) 2048 conv\_dw\_10\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_10 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_10\_bn (BatchNormaliz (None, 14, 14, 512) 2048 conv\_pw\_10\_relu (ReLU) (None, 14, 14, 512) conv\_dw\_11 (DepthwiseConv2D) (None, 14, 14, 512) 4608 conv\_dw\_11\_bn (BatchNormaliz (None, 14, 14, 512) 2048 conv\_dw\_11\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_11 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_11\_bn (BatchNormaliz (None, 14, 14, 512) 2048 conv\_pw\_11\_relu (ReLU) (None, 14, 14, 512) conv\_pad\_12 (ZeroPadding2D) (None, 15, 15, 512) conv\_dw\_12 (DepthwiseConv2D) (None, 7, 7, 512) 4608 conv\_dw\_12\_bn (BatchNormaliz (None, 7, 7, 512) 2048 conv\_dw\_12\_relu (ReLU) (None, 7, 7, 512) conv\_pw\_12 (Conv2D) (None, 7, 7, 1024) 524288 conv\_pw\_12\_bn (BatchNormaliz (None, 7, 7, 1024) 4096 conv\_pw\_12\_relu (ReLU) (None, 7, 7, 1024) conv\_dw\_13 (DepthwiseConv2D) (None, 7, 7, 1024) 9216 conv\_dw\_13\_bn (BatchNormaliz (None, 7, 7, 1024) 4096 conv\_dw\_13\_relu (ReLU) (None, 7, 7, 1024) conv\_pw\_13 (Conv2D) (None, 7, 7, 1024) 1048576 conv\_pw\_13\_bn (BatchNormaliz (None, 7, 7, 1024) 4096 conv\_pw\_13\_relu (ReLU) (None, 7, 7, 1024) global\_average\_pooling2d\_1 ( (None, 1024) 0 reshape\_1 (Reshape) (None, 1, 1, 1024) dropout (Dropout) (None, 1, 1, 1024) conv\_preds (Conv2D) (None, 1, 1, 1000) 1025000 reshape\_2 (Reshape) (None, 1000) 0 predictions (Activation) (None, 1000) 0 Total params: 4,253,864 Trainable params: 4,231,976 Non-trainable params: 21,888 # Modifing Model In [23]: x=mobile.layers[-6].output output=Dense(units=10, activation='softmax')(x) model=keras.Model(inputs=mobile.input, outputs=output) In [26]: for layer in model.layers[:-23]: In [29]: layer.trainable=False model.summary() In [30]: Model: "model" Layer (type) Output Shape Param # input\_2 (InputLayer) [(None, 224, 224, 3)] conv1 (Conv2D) (None, 112, 112, 32) 864 conv1\_bn (BatchNormalization (None, 112, 112, 32) 128 conv1\_relu (ReLU) (None, 112, 112, 32) 0 conv\_dw\_1 (DepthwiseConv2D) (None, 112, 112, 32) 288 conv\_dw\_1\_bn (BatchNormaliza (None, 112, 112, 32) 128 conv\_dw\_1\_relu (ReLU) (None, 112, 112, 32) (None, 112, 112, 64) conv\_pw\_1 (Conv2D) 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) conv\_dw\_2 (DepthwiseConv2D) (None, 56, 56, 64) 576 conv\_dw\_2\_bn (BatchNormaliza (None, 56, 56, 64) conv\_dw\_2\_relu (ReLU) (None, 56, 56, 64) conv\_pw\_2 (Conv2D) 8192 (None, 56, 56, 128) conv\_pw\_2\_bn (BatchNormaliza (None, 56, 56, 128) 512 conv\_pw\_2\_relu (ReLU) (None, 56, 56, 128) 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) 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) conv\_pad\_4 (ZeroPadding2D) (None, 57, 57, 128) (None, 28, 28, 128) conv\_dw\_4 (DepthwiseConv2D) 1152 conv\_dw\_4\_bn (BatchNormaliza (None, 28, 28, 128) 512 (None, 28, 28, 128) conv\_dw\_4\_relu (ReLU) (None, 28, 28, 256) 32768 conv\_pw\_4 (Conv2D) conv\_pw\_4\_bn (BatchNormaliza (None, 28, 28, 256) 1024 conv\_pw\_4\_relu (ReLU) (None, 28, 28, 256) (None, 28, 28, 256) conv\_dw\_5 (DepthwiseConv2D) 2304 conv\_dw\_5\_bn (BatchNormaliza (None, 28, 28, 256) 1024 (None, 28, 28, 256) conv\_dw\_5\_relu (ReLU) 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) conv\_pad\_6 (ZeroPadding2D) (None, 29, 29, 256) conv\_dw\_6 (DepthwiseConv2D) (None, 14, 14, 256) 2304 conv\_dw\_6\_bn (BatchNormaliza (None, 14, 14, 256) 1024 conv\_dw\_6\_relu (ReLU) (None, 14, 14, 256) (None, 14, 14, 512) conv\_pw\_6 (Conv2D) 131072 conv\_pw\_6\_bn (BatchNormaliza (None, 14, 14, 512) 2048 (None, 14, 14, 512) conv\_pw\_6\_relu (ReLU) conv\_dw\_7 (DepthwiseConv2D) 4608 (None, 14, 14, 512) conv\_dw\_7\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_dw\_7\_relu (ReLU) (None, 14, 14, 512) (None, 14, 14, 512) conv\_pw\_7 (Conv2D) 262144 conv\_pw\_7\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_pw\_7\_relu (ReLU) (None, 14, 14, 512) (None, 14, 14, 512) conv\_dw\_8 (DepthwiseConv2D) 4608 conv\_dw\_8\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_dw\_8\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_8 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_8\_bn (BatchNormaliza (None, 14, 14, 512) 2048 (None, 14, 14, 512) conv\_pw\_8\_relu (ReLU) conv\_dw\_9 (DepthwiseConv2D) 4608 (None, 14, 14, 512) conv\_dw\_9\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_dw\_9\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_9 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_9\_bn (BatchNormaliza (None, 14, 14, 512) 2048 conv\_pw\_9\_relu (ReLU) (None, 14, 14, 512) conv\_dw\_10 (DepthwiseConv2D) (None, 14, 14, 512) 4608 conv\_dw\_10\_bn (BatchNormaliz (None, 14, 14, 512) 2048 (None, 14, 14, 512) conv\_dw\_10\_relu (ReLU) conv\_pw\_10 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_10\_bn (BatchNormaliz (None, 14, 14, 512) 2048 conv\_pw\_10\_relu (ReLU) (None, 14, 14, 512) 4608 conv\_dw\_11 (DepthwiseConv2D) (None, 14, 14, 512) conv\_dw\_11\_bn (BatchNormaliz (None, 14, 14, 512) 2048 conv\_dw\_11\_relu (ReLU) (None, 14, 14, 512) conv\_pw\_11 (Conv2D) (None, 14, 14, 512) 262144 conv\_pw\_11\_bn (BatchNormaliz (None, 14, 14, 512) 2048 conv\_pw\_11\_relu (ReLU) (None, 14, 14, 512) conv\_pad\_12 (ZeroPadding2D) (None, 15, 15, 512) conv\_dw\_12 (DepthwiseConv2D) (None, 7, 7, 512) 4608 conv\_dw\_12\_bn (BatchNormaliz (None, 7, 7, 512) 2048 conv\_dw\_12\_relu (ReLU) (None, 7, 7, 512) conv\_pw\_12 (Conv2D) (None, 7, 7, 1024) 524288 conv\_pw\_12\_bn (BatchNormaliz (None, 7, 7, 1024) 4096 conv\_pw\_12\_relu (ReLU) (None, 7, 7, 1024) conv\_dw\_13 (DepthwiseConv2D) (None, 7, 7, 1024) 9216 conv\_dw\_13\_bn (BatchNormaliz (None, 7, 7, 1024) 4096 conv\_dw\_13\_relu (ReLU) (None, 7, 7, 1024) (None, 7, 7, 1024) 1048576 conv\_pw\_13 (Conv2D) conv\_pw\_13\_bn (BatchNormaliz (None, 7, 7, 1024) 4096 (None, 7, 7, 1024) conv\_pw\_13\_relu (ReLU) global\_average\_pooling2d\_1 ( (None, 1024) dense (Dense) 10250 (None, 10) Total params: 3,239,114 Trainable params: 1,873,930 Non-trainable params: 1,365,184 model.compile(optimizer=Adam(learning\_rate=0.0001), loss='categorical\_crossentropy', metrics=['accuracy']) In [31]: model.fit(x=train\_batches, validation\_data=valid\_batches,epochs=10,verbose=2) 172/172 - 245s - loss: 0.6597 - accuracy: 0.8072 - val\_loss: 0.2458 - val\_accuracy: 0.9367 Epoch 2/10 172/172 - 220s - loss: 0.1166 - accuracy: 0.9836 - val\_loss: 0.1046 - val\_accuracy: 0.9733 Epoch 3/10 172/172 - 220s - loss: 0.0711 - accuracy: 0.9912 - val\_loss: 0.0692 - val\_accuracy: 0.9767 Epoch 4/10 172/172 - 220s - loss: 0.0366 - accuracy: 0.9982 - val\_loss: 0.0533 - val\_accuracy: 0.9767 Epoch 5/10 172/172 - 217s - loss: 0.0239 - accuracy: 0.9988 - val\_loss: 0.0525 - val\_accuracy: 0.9833 Epoch 6/10 172/172 - 225s - loss: 0.0174 - accuracy: 0.9988 - val\_loss: 0.0423 - val\_accuracy: 0.9833 Epoch 7/10 172/172 - 221s - loss: 0.0216 - accuracy: 0.9977 - val\_loss: 0.0505 - val\_accuracy: 0.9833 Epoch 8/10 172/172 - 221s - loss: 0.0160 - accuracy: 0.9988 - val\_loss: 0.0528 - val\_accuracy: 0.9800 Epoch 9/10 172/172 - 215s - loss: 0.0109 - accuracy: 0.9988 - val\_loss: 0.0342 - val\_accuracy: 0.9833 Epoch 10/10 172/172 - 214s - loss: 0.0069 - accuracy: 1.0000 - val\_loss: 0.0315 - val\_accuracy: 0.9867 Out[32]: <tensorflow.python.keras.callbacks.History at 0x25502b12d00> # Predict sign language digits test\_labels=test\_batches.classes In [33]: predictions=model.predict(x=test\_batches, verbose=0) In [34]: cm=confusion\_matrix(y\_true=test\_labels, y\_pred=predictions.argmax(axis=1)) In [37]: In [42]: def plot\_confusion\_matrix(cm, classes, normalize=False, title='Confusion matrix', cmap=plt.cm.Blues): This function prints and plots the confusion matrix. Normalization can be applied by setting `normalize=True`. 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:

else:

In [43]:

In [45]:

Out[43]: {'0': 0,

'1': 1,

'4': 4,

'6': 6, '7': 7, '8': 8, '9': 9}

2

3

4

5

6

10

0

0

8 0

0

0

print(cm)

thresh = cm.max() / 2.

plt.tight\_layout()

test\_batches.class\_indices

plt.ylabel('True label')
plt.xlabel('Predicted label')

Confusion matrix, without normalization

confusion Matrix

0 0

0 0 0

0 0

0 0

0 0 5

0 0 0

0 0 0 0

0 0 0 0 0 0

0 0 0 0 0

0 ト 2 3 き 5 6 1 8 9 ゆ
Predicted label

0 0 0 0 0 0 0

0 0 0

0 0

0 0

0 0

plt.text(j, i, cm[i, j],

cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]

for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):

color="white" if cm[i, j] > thresh else "black")

print('Confusion matrix, without normalization')

horizontalalignment="center",

cm\_plot\_labels=['0','1','2','3','4','5','6','7','8','9','10']

plot\_confusion\_matrix(cm=cm, classes=cm\_plot\_labels,title='confusion Matrix')

- 3

- 2

-1

print("Normalized confusion matrix")

# Classifying sign language digits using modified mobile nets model and plotting the results on

confusion matrix

from tensorflow.keras.models import Sequential

from tensorflow.keras.metrics import categorical\_crossentropy

from tensorflow.keras.preprocessing.image import ImageDataGenerator

from tensorflow.keras.optimizers import Adam

from sklearn.metrics import confusion\_matrix

from keras.preprocessing import image

from tensorflow.keras.layers import Activation, Dense, Flatten, BatchNormalization, Conv2D, MaxPool2D

import numpy as np
import tensorflow as tf
from tensorflow import keras

import itertools

import os