Deep Learning in Practice Anis Koubaa

The Vehicle Type Classification Project

Evaluation of the classifier

We first need to load the required libraries

```
%load_ext autoreload
%autoreload 2
# import the necessary packages
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.layers import AveragePooling2D, GlobalAveragePooling2D, Batch
#from tensorflow.keras.applications import Xception
#from tensorflow.keras.applications import resnet50
from tensorflow.keras.applications import MobileNetV2
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
from tensorflow.keras.optimizers import Adam
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 sklearn.metrics import confusion matrix
from imutils import paths
import matplotlib.pyplot as plt
import numpy as np
import argparse
import cv2
import os
import sys
import tensorflow as tf
import h5py
import numpy as np
import sys
```

```
ascii_dev_labels = np.array(hf["devLabels"]).astype("S65")
 devY=np.array(hf["devY"]).astype("int")
 testX= np.array(hf["testX"]).astype("f8")
 ascii_test_labels = np.array(hf["testLabels"]).astype("S65")
 testY=np.array(hf["testY"]).astype("int")
 trainLabels = np.array([n.decode('unicode_escape') for n in ascii_train_labels])
 devLabels = np.array([n.decode('unicode_escape') for n in ascii_dev_labels])
 testLabels = np.array([n.decode('unicode_escape') for n in ascii_test_labels])
 print("trainX.shape: ",trainX.shape)
 print("trainY.shape: ",trainY.shape)
 print("trainLabels.shape: ",trainLabels.shape)
 print("devX.shape: ",devX.shape)
 print("devY.shape: ",devY.shape)
 print("devLabels.shape: ",devLabels.shape)
 print("testX.shape: ",testX.shape)
 print("testY.shape: ",testY.shape)
 print("testLabels.shape: ",testLabels.shape)
 return trainX, trainY, trainLabels, devX,devY,devLabels,testX,testY,testLabels
def load dev_test_dataset_from_hdf5_file(hdf_file_path):
 hf = h5py.File(hdf_file_path, "r")
 devX= np.array(hf["devX"]).astype("f8")
 ascii_dev_labels = np.array(hf["devLabels"]).astype("S65")
 devY=np.array(hf["devY"]).astype("int")
 testX= np.array(hf["testX"]).astype("f8")
 ascii_test_labels = np.array(hf["testLabels"]).astype("S65")
 testY=np.array(hf["testY"]).astype("int")
 devLabels = np.array([n.decode('unicode_escape') for n in ascii_dev_labels])
 testLabels = np.array([n.decode('unicode_escape') for n in ascii_test_labels])
 print("devX.shape: ",devX.shape)
 print("devY.shape: ",devY.shape)
 print("devLabels.shape: ",devLabels.shape)
 print("testX.shape: ",testX.shape)
 print("testY.shape: ",testY.shape)
 print("testLabels.shape: ",testLabels.shape)
 return devX,devY,devLabels,testX,testY,testLabels
import datetime
t0 = datetime.datetime.now()
devX,devY,devLabels,testX,testY,testLabels=load_dev_test_dataset_from_hdf5_file(HDF
t1 = datetime.datetime.now()
print('time to load data: ', (t1-t0))
```

```
TYPE='type'
model_type='mobilenetv2'
#set your user name
user='anis'
iteration='1'
#NOTE: Make sure to set the correct project path with respect to your Google Drive
PROJECT_PATH='/content/drive/My Drive/udemy-deep-learning-in-practice/03-transfer-l
print('PROJECT PATH: ',PROJECT PATH)
HDF5_DATASET_PATH=PROJECT_PATH+'datasets/vehicle-type-dataset-SIZE224-train-dev-tes
print('HDF5 DATASET PATH: ', HDF5 DATASET PATH)
ACCURACY LOSS OUPUT FILE=PROJECT PATH+'log/'+model type+'/'+model type+'-by-'+TYPE+
print('ACCURACY_LOSS_OUPUT_FILE: ', ACCURACY_LOSS_OUPUT_FILE)
TRAINED MODEL=PROJECT_PATH+'trained-models/'+model_type+'/'+'vehicle-classification
print('TRAINED_MODEL: ',TRAINED_MODEL)
CHECKPOINT_PATH = PROJECT_PATH+'checkpoints/'+model_type+'/'+'by-'+TYPE+'-'+model_t
print('CHECKPOINT_PATH: ',CHECKPOINT_PATH)
LOGFILE PATH=PROJECT PATH+'log/'+model type+'/'+model type+'-by-'+TYPE+'-training-l
print('LOGFILE PATH: ',LOGFILE PATH)
PROJECT_PATH: /content/drive/My Drive/udemy-deep-learning-in-practice/03-tran
    HDF5 DATASET PATH: /content/drive/My Drive/udemy-deep-learning-in-practice/03
    ACCURACY_LOSS_OUPUT_FILE: /content/drive/My Drive/udemy-deep-learning-in-prac
    TRAINED MODEL: /content/drive/My Drive/udemy-deep-learning-in-practice/03-tra
    CHECKPOINT PATH: /content/drive/My Drive/udemy-deep-learning-in-practice/03-t
    LOGFILE PATH: /content/drive/My Drive/udemy-deep-learning-in-practice/03-tran
from google.colab import drive
drive.mount('/content/drive')
Drive already mounted at /content/drive; to attempt to forcibly remount, call
sys.path.append(PROJECT PATH)
import anis_koubaa_udemy_computer_vision_lib
from anis_koubaa_udemy_computer_vision_lib import *
```

Load the Dataset

```
def load_dataset_from_hdf5_file(hdf_file_path):
    hf = h5py.File(hdf_file_path, "r")
    trainX= np.array(hf["trainX"]).astype("f8")
    ascii_train_labels = np.array(hf["trainLabels"]).astype("S65")
    trainY=np.array(hf["trainY"]).astype("int")

    devX= np.arrav(hff["devX"]).astype("f8")
```

devX.shape: (75, 224, 224, 3) devY.shape: (75, 7) devLabels.shape: (75,) testX.shape: (76, 224, 224, 3) testY.shape: (76, 7) testLabels.shape: (76,) time to load data: 0:00:03.916500

IMAGE_SIZE=testX.shape[1] print(IMAGE_SIZE)

[→ 224

Dataset Visualization

#anis_koubaa_udemy_computer_vision_lib.plot_sample_from_dataset(trainX, trainLabels

anis_koubaa_udemy_computer_vision_lib.plot_sample_from_dataset(devX, devLabels,rows

С>











anis_koubaa_udemy_computer_vision_lib.plot_sample_from_dataset(testX, testLabels,ro

C→











```
#print('Loading the best model...')
model = tf.keras.models.load_model(CHECKPOINT_PATH)

#tf.compat.v1.keras.experimental.export_saved_model(model, PROJECT_PATH+'tensorflow)
```

Make Predictions

```
print('Loading the best model...', TRAINED MODEL)
t0 = datetime.datetime.now()
test_model = tf.keras.models.load_model(TRAINED_MODEL)
t1 = datetime.datetime.now()
print('time to load the model: ', (t1-t0))
□→ Loading the best model... /content/drive/My Drive/udemy-deep-learning-in-pract
    time to load the model: 0:00:01.978342
# make predictions on the testing set
print("[INFO] evaluating network on the dev dataset...")
test model.evaluate(devX, devY,verbose=0)
[→ [INFO] evaluating network on the dev dataset...
    [1.1841479539871216, 1.0]
print("[INFO] evaluating network on the test dataset...")
test_model.evaluate(testX, testY,verbose=0)
[INFO] evaluating network on the test dataset...
    [1.2156026363372803, 0.9736841917037964]
class_dict,number_of_classes=get_cars_classes_dict(testY,testLabels)
class dict
[ \ \{0: 'car-bus-alltypes',
     1: 'car-sedan-alltypes',
     2: 'car-suv-alltypes',
     3: 'motocycle-bicycle-kids',
     4: 'motocycle-bicycle-racing',
     5: 'motocycle-motorbike-chopper',
     6: 'motocycle-motorbike-sport'}
class_dict[6]
   'motocycle-motorbike-sport'
```

```
len(class_dict)

[→ 7

len(class_dict)

[→ 7
```

Testing on one image

```
index=30
test_image = testX[index]
print(testY[index])
print(class_dict[np.argmax(testY[index])])
anis_koubaa_udemy_computer_vision_lib.display_image(testX, testLabels, index)

[> [0 0 0 0 0 0 1]
    motocycle-motorbike-sport
    Label = motocycle-motorbike-sport
    image shape: (224, 224, 3)
```



import datetime

```
plt.imshow(test_image)
reshaped_image=anis_koubaa_udemy_computer_vision_lib.reshape_image_for_neural_netwo

t0 = datetime.datetime.now()
prediction_array = test_model.predict(reshaped_image)
t1 = datetime.datetime.now()
print('-----')
print ("prediction time: ", t1-t0)
print ("true label: ", testY[index])
print ("predicted label: ", np.argmax(prediction_array[0]))
print ("Car Brand: ",class_dict[np.argmax(testY[index])])
```

```
print ("Car Brand Predicted: ",class_dict[np.argmax(prediction_array[0])])
print ("Confidence: ",prediction_array[0][np.argmax(prediction_array[0])])
print('-----')
```

flatten the image image.shape (150528, 1) reshape the image to be similar to the input feature vector image.shape (1, 224, 224, 3)

prediction time: 0:00:00.667347
true label: [0 0 0 0 0 0 1]

predicted label: 6

Car Brand: motocycle-motorbike-sport

Car Brand Predicted: motocycle-motorbike-sport

Confidence: 0.99122167



5.4651937e-03 9.9122167e-01]]

'motocycle-motorbike-sport'

```
# for each image in the testing set we need to find the index of the
# label with corresponding largest predicted probability
predicted label = np.argmax(testY[index])
print("true label: ", np.argmax(prediction array[0]))
print("predicted label: ", predicted_label)
T→ true label: 6
    predicted label: 6
reshaped test image = anis koubaa udemy computer vision lib.reshape image for neura
p = test model.predict(reshaped test image)
print(p)
class_dict[np.argmax(p[0])]
flatten the image
    image.shape (150528, 1)
    reshape the image to be similar to the input feature vector
    image.shape (1, 224, 224, 3)
    [[2.7564276e-04 4.7135452e-04 9.2403946e-04 8.2937750e-04 8.1273646e-04
```

Testing with the whole dataset

```
prediction array all test dataset = test model.predict(testX, verbose=1)
predicted labels all test dataset = np.argmax(prediction array all test dataset, ax
# show a nicely formatted classification report
print(classification report(testY.argmax(axis=1), predicted labels_all test dataset
 □→ 1/3 [=======>.....] - ETA: 0sWARNING:tensorflow:Callbacks met
    3/3 [======= ] - 0s 19ms/step
                                 precision
                                              recall f1-score
                                                                 support
               car-bus-alltypes
                                      1.00
                                                1.00
                                                          1.00
                                                                      13
                                      0.92
                                                1.00
                                                          0.96
             car-sedan-alltypes
                                                                      12
               car-suv-alltypes
                                      1.00
                                                0.89
                                                          0.94
                                                                      9
         motocycle-bicycle-kids
                                      0.93
                                                1.00
                                                          0.96
                                                                      13
       motocycle-bicycle-racing
                                                0.90
                                                          0.95
                                      1.00
                                                                      10
    motocycle-motorbike-chopper
                                      1.00
                                                1.00
                                                          1.00
                                                                      9
      motocycle-motorbike-sport
                                      1.00
                                                1.00
                                                          1.00
                                                                      10
                                                          0.97
                                                                      76
                       accuracy
                                                                      76
                                                          0.97
                      macro avg
                                      0.98
                                                0.97
                   weighted avg
                                      0.98
                                                0.97
                                                          0.97
                                                                      76
# compute the confusion matrix and and use it to derive the raw
# accuracy, sensitivity, and specificity
model confusion matrix = confusion matrix(testY.argmax(axis=1), predicted labels al
# true_positive = model_confusion_matrix[0, 0]
# true negative= model confusion matrix[1, 1]
# false positive=model confusion matrix[0, 1]
# false negative=model confusion matrix[1, 0]
# total = sum(sum(model confusion matrix))
# acc = (true positive + true negative) / total
# sensitivity = true positive / (true positive + false negative)
# precision = true_positive / (true_positive + false positive)
# recall = sensitivity
# specificity = true negative / (false positive + true negative)
```

precision = true positive / (true positive + false positive)

print("model confusion matrix: \n", model confusion matrix)

#print("sensitivity, recall: {:.4f}".format(sensitivity))

#print("specificity: {:.4f}".format(specificity))

#print("acc: {:.4f}".format(acc))

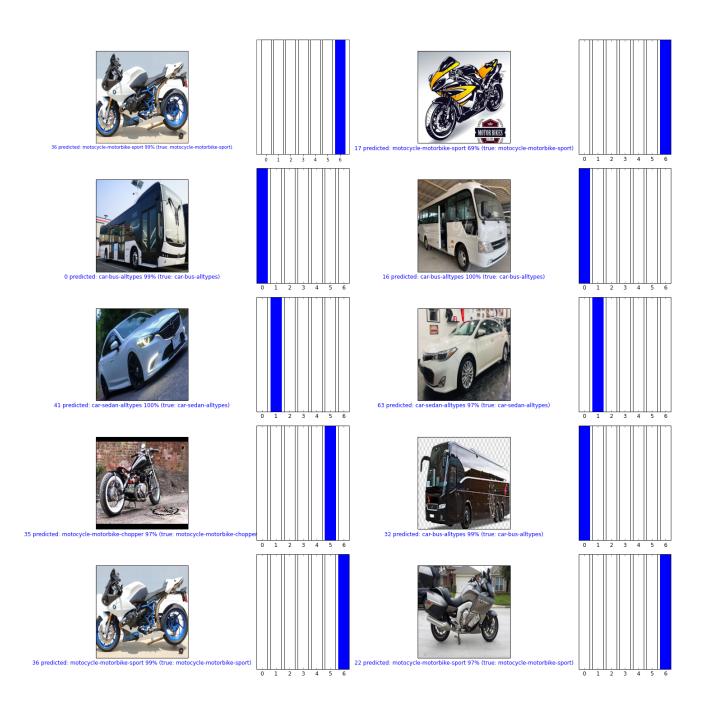
show the confusion matrix, accuracy, sensitivity, and specificity

#anis_koubaa_udemy_computer_vision_lib.plot_loss_accuracy(H,EPOCHS=25, output_file=

Double-click (or enter) to edit

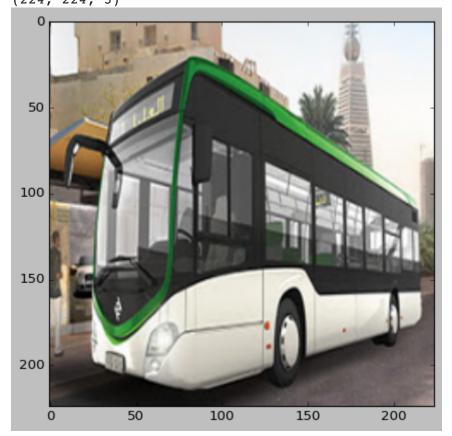
```
#classes_names=['covid', 'normal']
#print(testX.shape)
#print(test_binary_labels.shape)
#anis_koubaa_udemy_computer_vision_lib.plot_car_sample_predictions(testX, predictio
anis_koubaa_udemy_computer_vision_lib.plot_car_sample_predictions_v2(testX, testY,
```

 \Box



```
None
(224, 224)

RGB
(224, 224, 3)
(224, 224, 3)
```



```
prediction_my_image = test_model.predict(reshaped_test_image)
print(prediction_my_image)
class dict[np.argmax(prediction my image)]
flatten the image
    image.shape (150528, 1)
    reshape the image to be similar to the input feature vector
    image.shape (1, 224, 224, 3)
    [[9.9798238e-01 2.2007358e-04 3.4697400e-04 2.9347328e-04 5.6409533e-04
      3.5445308e-04 2.3848382e-04]]
    'car-bus-alltypes'
def top_five(prediction_my_image, class_dict):
   sorted_array=np.argsort(prediction_my_image)[0][-5:]
   sorted array=np.flip(sorted array)
   sorted array
   for s in sorted array:
   #print(class dict[s])
   #print('id:',s,', brand:'+class_dict[s]+', confidence:',prediction_my_image[0,s]
       print('id: %3d, brand:%-10s, confidence: %6.3f'%(s,class dict[s],prediction
```

top five(prediction my image, class dict)

reshaped_test_image = anis_koubaa_udemy_computer_vision_lib.reshape_image_for_neura

```
4, brand:motocycle-bicycle-racing, confidence: 0.001
          5, brand:motocycle-motorbike-chopper, confidence: 0.000
    id:
          2, brand:car-suv-alltypes, confidence: 0.000
    id:
    id:
          3, brand:motocycle-bicycle-kids, confidence: 0.000
t0=datetime.datetime.now()
ms test=anis koubaa udemy computer vision lib.plot misclassifications(testX, testY,
t1=datetime.datetime.now()
print('time to find misclassifications: ',(t1-t0))
¬ number of misclassications: 2
    number of images: 75
    {40: ['car-suv-alltypes', 'car-sedan-alltypes'], 46: ['motocycle-bicycle-racin
    40 predicted: car-sedan-alltypes 100% (true: car-suv-alltypes)
    time to find misclassifications: 0:00:00.317932
prediction array all dev dataset = test model.predict(devX, verbose=1)
[→ 1/3 [======>.....] - ETA: 0sWARNING:tensorflow:Callbacks met
    3/3 [=======] - 0s 19ms/step
t0=datetime.datetime.now()
ms dev=anis koubaa udemy computer vision lib.plot misclassifications(devX, devY, de
t1=datetime.datetime.now()
print('time to find misclassifications: ',(t1-t0))
□ number of misclassications: 0
    number of images: 75
    rate of misclassications: 0.0 %
    {}
    <Figure size 3200x3200 with 0 Axes>
    time to find misclassifications: 0:00:00.018239
```

0, brand:car-bus-alltypes, confidence: 0.998

□ id:

id: