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# Computer Vision Coursework Submission (INM460)
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# ************ /code/training cv.ipynb **************
# -*- coding: utf-8 -*-
"""training_cv.ipynb
Automatically generated by Colaboratory.
Original file is located at
   https://colab.research.google.com/drive/1tYurtYZECtimLx8lyIGYm580mZeKpq70
!pip install imblearn
!pip install -q -U keras-tuner
from google.colab import drive
drive.mount('/content/drive')
"""# Importing Libraries"""
import matplotlib.pyplot as plt
import numpy as np
import sys
from PIL import Image
sys.modules['Image'] = Image
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import load model
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Flatten, Input, BatchNormalization, Dropout
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Activation
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.applications.resnet50 import preprocess input,
decode predictions
from tensorflow.keras.models import Model
import tensorflow as tf
import keras tuner as kt
from keras.utils.data utils import Sequence
from imblearn.over sampling import RandomOverSampler
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from imblearn.tensorflow import balanced_batch_generator
from sklearn.model_selection import train_test_split
from skimage.feature import hog
import skimage
from sklearn import svm
from sklearn.metrics import accuracy score
from matplotlib import patches
from collections import Counter
import pathlib
import os
import glob
import pandas as pd
import cv2 as cv
import numpy as np
# import face_recognition
from google.colab.patches import cv2_imshow
from PIL import Image
"""# Loading, Splitting and Preprocessing the Dataset"""
train path="/content/drive/MyDrive/Computer-vision/cv-coursework/CW Dataset/train"
# extract and return the first line of a text file as an integer
def txtFileExtract(path):
    f=open(path,'r')
    content=f.readlines()
    return content[0]
# This function loads a dataset from a given path and reads the targets from .txt
files in the subdirectories of the given path
# and stores them in a list, then finds all the .jpeg images in the subdirectories
and returns their paths along with
# the targets list.
def load dataset(path):
  targets = []
  labels=os.listdir(path)
  image path=glob.glob(path+'/*/*.jpeg')
  txt_path=glob.glob(path+'/*/*.txt')
  targets = [txtFileExtract(p) for p in txt path]
  return _image_path, targets
# storing the image path and labels in a variable
im path, labels = load dataset(train path)
# changing the datatype of labels to float32
labels = np.array(labels).astype('float32')
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# check the length of each classes
len(np.where(labels == 1.0)[0])
# saving the labels as y train.npy for reusability
np.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/CW Dataset/y train.npy
',labels)
# loads images from file paths and preprocesses
def load_image(img_path):
  _images = []
  for image path in img path:
    image = cv.imread(image path)
    image = cv.resize(image, (64,64))
    image = cv.normalize(image, None, 0, 1.0, cv.NORM_MINMAX, dtype=cv.CV_32F)
    images.append(image)
  return _images
#this function takes argument all image path and retun hog converted images as an
array
def load_image_to_hog(img_path):
  images = []
  winStride = (8,8)
  padding = (8,8)
  locations = ((10,20),)
  HOG = cv.HOGDescriptor()
  for image path in img path:
    image = cv.imread(image path)
    image = cv.resize(image, (64,64))
    fd, image = hog(image, orientations=8, pixels_per_cell=(16, 16),
                    cells per block=(1, 1), visualize=True, channel axis = -1)
    image = skimage.color.gray2rgb(image)
    _images.append(image)
  return images
# applying the fuction to convert path to original images and hog images as list of
arrays
or images = load image(im path)
hog_images = load_image_to_hog(im_path)
# converting the list to numpy array
or_images = np.array(or_images)
hog images = np.array(hog images)
# saving the numpy arrays
np.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/CW Dataset/hog images
train.npy',hog images)
np.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/CW Dataset/or images t
rain.npy',or images)
```

```
# loading the hog and original images which can be said as x_train
hog images =
np.load('/content/drive/MyDrive/Computer-vision/CW Folder PG/CW Dataset/hog images
train.npv')
or images =
np.load('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/CW_Dataset/or_images_t
rain.npy')
https://www.linkedin.com/pulse/some-tricks-handling-imbalanced-dataset-image-m-farh
an-tandia/
datagen = ImageDataGenerator(
    rotation range=10, # rotate the image by a maximum of 10 degrees
    width_shift_range=0.1, # shift the image horizontally by a maximum of 10%
    height_shift_range=0.1, # shift the image vertically by a maximum of 10%
    shear_range=0.2, # shear the image by a maximum of 20%
    zoom_range=0.2, # zoom into the image by a maximum of 20%
    horizontal_flip=True, # flip the image horizontally
    vertical flip=False # do not flip the image vertically
)
# this method creates and balnced dataset with image aggumentation technique
applied for a sample of image the library used for balancing are keras and imblearn
#https://www.linkedin.com/pulse/some-tricks-handling-imbalanced-dataset-image-m-far
han-tandia/
class BalancedDataGenerator(Sequence):
  def __init__(self, x, y, datagen, batch_size=64):
      self.datagen = datagen
      self.batch size = min(batch size, x.shape[0])
      datagen.fit(x)
      self.gen, self.steps_per_epoch =
balanced_batch_generator(x.reshape(x.shape[0], -1), y, sampler=RandomOverSampler(),
batch size=self.batch size, keep sparse=True)
      self._shape = (self.steps_per_epoch * batch_size, *x.shape[1:])
  def len (self):
      return self.steps per epoch
  def __getitem__(self, idx):
      x_batch, y_batch = self.gen.__next__()
      x batch = x batch.reshape(-1, *self. shape[1:])
      return self.datagen.flow(x_batch, y_batch, batch_size=self.batch_size).next()
#splitting traing and validation set which are 80% and 20% respectively
x train, x val, y train, y val = train test split(or images, labels, test size =
0.2, random_state=42, stratify=labels, shuffle=True)
x train hog, x val hog, y train hog, y val hog = train test split(hog images,
labels, test_size = 0.2, random_state=42, stratify=labels, shuffle=True)
```

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# balancing the splitted datset and applying batches for train and validation set
balanced_gen = BalancedDataGenerator(x_train, y_train, datagen, batch_size=64)
balanced gen val = BalancedDataGenerator(x val, y val, datagen, batch size=64)
# balancing the splitted datset and applying batches for train and validation set
for HOG images
balanced_gen_hog = BalancedDataGenerator(x_train_hog, y_train_hog, datagen,
batch size=64)
balanced_gen_val_hog = BalancedDataGenerator(x_val_hog, y_val_hog, datagen,
batch size=64)
# extracting x and y values from the generator
x, y = [], []
for i in range(len(balanced_gen)):
    batch_x, batch_y = balanced_gen[i]
    x.append(batch_x)
    y.append(batch_y)
x = np.concatenate(x)
y = np.concatenate(y)
# ploting the frequency for Balanced dataset
plt.hist(y)
plt.xlabel('integer labels')
plt.ylabel('frequency')
plt.show()
counter = Counter(labels)
counts = list(counter.values())
categories = list(counter.keys())
# ploting the frequency for imbalanced train dataset
plt.bar(categories, counts)
# add labels
plt.xlabel('Category')
plt.ylabel('Frequency')
plt.title('Train set data frequency')
# show plot
plt.show()
counter = Counter(y val)
counts = list(counter.values())
categories = list(counter.keys())
# ploting the frequency for imbalanced test dataset
plt.bar(categories, counts)
# add labels
```

```
plt.xlabel('Category')
plt.ylabel('Frequency')
plt.title('Test set data frequency')
# show plot
plt.show()
# it converts the xtrain and y_train arrays into tensor dataset
train_dataset = tf.data.Dataset.from_tensor_slices((x_train, y_train))
train dataset = train dataset.shuffle(buffer size=1000,
reshuffle each iteration=True)
val dataset = tf.data.Dataset.from_tensor_slices((x_val, y_val))
val dataset = val dataset.shuffle(buffer size=1000, reshuffle each iteration=True)
# it converts the xtrain and y_train arrays into tensor hog dataset
train_dataset_hog = tf.data.Dataset.from_tensor_slices((x_train_hog, y_train_hog))
train_dataset_hog = train_dataset_hog.shuffle(buffer_size=1000,
reshuffle_each_iteration=True)
val dataset hog = tf.data.Dataset.from_tensor_slices((x_val_hog, y_val_hog))
val dataset hog = val dataset hog.shuffle(buffer size=1000,
reshuffle each iteration=True)
# apply batches for all datasets
train dataset = train dataset.batch(64)
val dataset = val dataset.batch(64)
train_dataset_hog = train_dataset_hog.batch(64)
val_dataset_hog = val_dataset_hog.batch(64)
"""# MLP with Hog Feature Descriptor
.. .. ..
# https://www.tensorflow.org/tutorials/keras/keras tuner
# this function creates MLP architecture with hyperparameter optimization
def model mlp ar(hp):
  # initializing the dropout value from range 0 to 1
  hp_rate = hp.Float('rate', min_value=0, max_value=1.0, step=0.25)
  # initializing the input neurons from range 32 to 512
  hp_units = hp.Int('units', min_value=32, max_value=512, step=32)
  # initializing the keras sequential layers
  model mlp = Sequential()
  #flattening the input tensor
  model mlp.add(Flatten())
  # Adding dense layer with batch normalization and dropouts
  model mlp.add(Dense(units = hp units, activation='relu'))
```

```
model_mlp.add(BatchNormalization())
  model mlp.add(Dropout(rate=hp rate))
  model mlp.add(Dense(units = hp units, activation='sigmoid'))
  model mlp.add(BatchNormalization())
  model mlp.add(Dropout(rate=hp rate))
  model mlp.add(Dense(units = hp units, activation='sigmoid'))
  model_mlp.add(BatchNormalization())
  model_mlp.add(Dropout(rate=hp_rate))
  model mlp.add(Dense(units = hp units, activation='sigmoid'))
  model mlp.add(BatchNormalization())
  model mlp.add(Dropout(rate=hp rate))
  model mlp.add(Dense(units = hp units, activation='sigmoid'))
  model mlp.add(BatchNormalization())
  model_mlp.add(Dropout(rate=hp_rate))
  model mlp.add(Dense(3, activation='sigmoid'))
  # initializing the learning rate
  hp_learning_rate = hp.Choice('learning_rate', values=[1e-2, 1e-3, 1e-4])
  Apply optimization function to the model
model mlp.compile(optimizer=keras.optimizers.Adam(learning rate=hp learning rate),
                loss=keras.losses.SparseCategoricalCrossentropy(from logits=True),
                metrics=['accuracy'])
  return model mlp
# initializing a Hyperband tuner from Keras Tuner for hyperparameter optimization
for imbalanced dataset
tuner_mlp_unbal = kt.Hyperband(model_mlp_ar,
                     objective='val accuracy',
                     max epochs=100,
                     factor=3,
                     directory='my dir n1',
                     project name='mlp unbal')
# initializing a Hyperband tuner from Keras Tuner for hyperparameter optimization
for balanced dataset
tuner mlp bal = kt.Hyperband(model mlp ar,
                     objective='val accuracy',
                     max epochs=100,
                     factor=3,
                     directory='my dirn2',
                     project name='mlp bal2')
# initializing early stopping with patience 10 and monitoring validation loss
early_stopping = EarlyStopping(monitor = 'val_loss', patience = 10)
```

```
tuner_mlp_unbal.search(train_dataset_hog,epochs=100,validation_data=val_dataset_hog
, callbacks=[early_stopping])
# Geta the best hyperparameters
best hps=tuner mlp unbal.get best hyperparameters(num trials=1)[0]
print(f""" units: {best hps.get('units')} , Learning Rate:
{best_hps.get('learning_rate')}, Dropout: {best_hps.get('rate')}
""")
# load the best optimized model and traing with unbalanced dataset
model mlp unbal = tuner mlp unbal.hypermodel.build(best hps)
model mlp unbal hist = model mlp unbal.fit(train dataset, epochs=100,
validation data=val dataset, callbacks=[early stopping] )
val acc per epoch = model mlp unbal hist.history['val accuracy']
best epoch unbal = val acc per epoch.index(max(val acc per epoch)) + 1
print('Best epoch in TrainSet_imbal: %d' % (best_epoch_unbal))
# saving the best model
model mlp unbal.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/Models/mo
del mlp hyp unbal.h5')
#evaluating the model
model mlp unbal.evaluate(train dataset)
# applying Hyperband searchin balanced datset
early stopping = EarlyStopping(monitor = 'val loss', patience = 10)
try:
tuner mlp bal.search(balanced gen hog,epochs=100,validation data=balanced gen val h
og, callbacks=[early stopping])
except:
best hps mlp balanced=tuner mlp bal.get best hyperparameters(num trials=1)[0]
# load the best optimized model and traing with unbalanced dataset
model_mlp_bal = tuner_mlp_bal.hypermodel.build(best_hps_mlp_balanced)
model mlp bal hist = model_mlp_bal.fit(train_dataset, epochs=100,
validation data=val dataset, callbacks=[early stopping] )
val acc per epoch bal = model mlp bal hist.history['val accuracy']
best_epoch_bal = val_acc_per_epoch_bal.index(max(val_acc_per_epoch_bal)) + 1
print('Best epoch: %d' % (best epoch bal,))
# saving the best model
model_mlp_bal.save('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/Models/mode
1 mlp hyp bal.h5')
```

```
"""# CNN Baseline
#initializing the baseline CNN model
model cnn = Sequential()
model cnn.add(layers.Conv2D(64,(3,3),activation='relu', input shape=(64,64,3)))
model cnn.add(layers.MaxPooling2D((2, 2)))
model_cnn.add(Flatten())
model_cnn.add(layers.Dense(64, activation='relu'))
model cnn.add(Dense(3))
# Model summary
model cnn.summary()
# Initializing ealy stooping
early_stopping= tf.keras.callbacks.EarlyStopping(
    monitor='val_loss',
    patience=10,
# Apply optimization function to the model
model_cnn.compile(optimizer='adam',
              loss=tf.keras.losses.SparseCategoricalCrossentropy(from logits=True),
              metrics=['accuracy'])
# traning the model with imbalanced datset
model_cnn_unbal = model_cnn.fit(train_dataset, epochs=100,
callbacks=[early stopping], validation data=val dataset)
#saving the model
model cnn.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/Models/model cn
n.h5')
# training model with balanced datset
model_cnn_bal = model_cnn.fit(balanced_gen, epochs=100,
callbacks=[early stopping], validation data=balanced gen val)
#saving the model
model cnn.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/Models/model cn
n_bal.h5')
"""# RESNET 50"""
# initalizing input size for resnet 50
input res = keras.Input(shape=(64,64,3))
# loading resnet 50 from keras
model res = keras.applications.ResNet50(weights=None,input tensor=input res)
```

```
model_res.summary()
# Compile the model
model res.compile(
    optimizer=tf.keras.optimizers.Adam(1e-3),
    loss="sparse categorical crossentropy",
    metrics=["accuracy"],
)
# initalizing early stooping with patience as 10 and monitor on val_accuracy
early stopping= tf.keras.callbacks.EarlyStopping(
    monitor='val accuracy',
    patience=10,
)
# Training the model on imbalanced dataset
model_res_imbal.fit(train_dataset, epochs=100, callbacks=[early_stopping],
validation data=val dataset)
#saving the model trained on imbalanced dataset
model res imbal.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/Models/mo
del_res.h5')
# Training the model on balanced dataset
model_res_bal.fit(balanced_gen, epochs=100, callbacks=[early_stopping_monitor],
validation_data=balanced_gen_val,)
#saving the model trained on balanced dataset
model res bal.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/Models/mode
l resv2.1.h5')
# ******** /code/test_cv.ipynb *************
# -*- coding: utf-8 -*-
"""test cv.ipynb
Automatically generated by Colaboratory.
Original file is located at
    https://colab.research.google.com/drive/1ceHnbzdXcqKCUWBmTq2AINnShmLorIfD
!pip install retina-face
from google.colab import drive
drive.mount('/content/drive')
```

```
import matplotlib.pyplot as plt
from matplotlib import patches
from retinaface import RetinaFace
import itertools
import numpy as np
import sys
from PIL import Image
sys.modules['Image'] = Image
from collections import Counter
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import load model
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Flatten, Input
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Activation
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.applications.resnet50 import preprocess input,
decode predictions
from tensorflow.keras.models import Model
from retinaface import RetinaFace
import torch
from sklearn.model selection import train test split
from skimage.feature import hog
import skimage
from sklearn import svm
from sklearn.metrics import accuracy score
from sklearn.metrics import confusion matrix
import sklearn
from sklearn.metrics import multilabel confusion matrix, ConfusionMatrixDisplay,
classification report
import pathlib
import os
import glob
import pandas as pd
import cv2 as cv
import numpy as np
# import face recognition
from google.colab.patches import cv2 imshow
from PIL import Image
```

```
"""# Loading Test Dataset"""
test path="/content/drive/MyDrive/Computer-vision/cv-coursework/CW Dataset/test"
# extract and return the first line of a text file as an integer
def txtFileExtract(path):
    f=open(path,'r')
    content=f.readlines()
    return content[0]
# This function loads a dataset from a given path and reads the targets from .txt
files in the subdirectories of the given path
# and stores them in a list, then finds all the .jpeg images in the subdirectories
and returns their paths along with
# the targets list.
def load_dataset(path):
  targets = []
  labels=os.listdir(path)
  _image_path=glob.glob(path+'/*/*.jpeg')
  txt path=glob.glob(path+'/*/*.txt')
  targets = [txtFileExtract(p) for p in txt path]
  return image path, targets
# loads images from file paths and preprocesses
def load image(img path):
  images = []
  for image_path in img_path:
    image = cv.imread(image path)
    image = cv.resize(image, (64,64))
    image = cv.normalize(image, None, 0, 1.0, cv.NORM MINMAX, dtype=cv.CV 32F)
    images.append(image)
  return _images
#this function takes argument all image path and retun hog converted images as an
array
def load_image_to_hog(img_path):
  images = []
  winStride = (8,8)
  padding = (8,8)
  locations = ((10,20),)
  HOG = cv.HOGDescriptor()
  for image path in img path:
    image = cv.imread(image_path)
    image = cv.resize(image, (64,64))
    fd, image = hog(image, orientations=8, pixels per cell=(16, 16),
                    cells_per_block=(1, 1), visualize=True, channel_axis = -1)
    image = skimage.color.gray2rgb(image)
    images.append(image)
  return _images
```

```
# storing the image path and labels in a variable
im path, labels = load dataset(test path)
# changing the datatype of labels to float32
labels = np.array(labels).astype('float32')
# check the length of each classes
np.where( labels == 2.0)[0]
# saving the labels as y_test.npy for reusability
np.save('/content/drive/MyDrive/Computer-vision/CW Folder PG/CW Dataset/y test.npy'
,labels)
# loading the labels
labels=np.load('/content/drive/MyDrive/Computer-vision/CW Folder PG/CW Dataset/y te
st.npy')
# applying the fuction to convert path to original images and hog images as list of
arrays
or images = load image(im path)
hog images = load image to hog(im path)
# saving the numpy arrays
np.save('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/CW_Dataset/hog_images_
test.npy',hog images)
np.save('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/CW_Dataset/or_images_t
est.npy',_or_images)
# converting the list to numpy array
or images = np.array( or images)
hog images = np.array(hog images)
# loading the hog and original images which can be said as x test
or images =
np.load('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/CW_Dataset/or_images_t
est.npy')
hog images =
np.load('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/CW_Dataset/hog_images_
test.npy')
# it converts the xtrain and y_train arrays into tensor dataset
test dataset = tf.data.Dataset.from tensor slices((or images, labels))
test_dataset = test_dataset.shuffle(buffer_size=1000,
reshuffle each iteration=True)
# it converts the xtrain and y_train arrays into tensor hog dataset
test dataset hog = tf.data.Dataset.from tensor slices((hog images, labels))
test dataset hog = test dataset hog.shuffle(buffer size=1000,
reshuffle each iteration=True)
```

```
# apply batches for all datasets
test_dataset = test_dataset.batch(64)
test dataset hog = test dataset hog.batch(64)
"""# HOG + MLP
### HOG+MLP with imbalanced Dataset
#loading the model
model mlp =
tf.keras.models.load model('/content/drive/MyDrive/Computer-vision/CW Folder PG/Mod
els/model mlp hyp unbal.h5')
#evaluating the model
model_mlp.evaluate(test_dataset_hog)
pred_mlp_unbal = model_mlp.predict(hog_images)
# getting the highest value using argmax and round off
predicted classes = np.argmax(pred mlp unbal, axis=1)
predicted classes
predicted classes = np.array(predicted classes)
print(classification_report(labels,predicted_classes))
cmx = confusion matrix(labels.tolist(),predicted classes.tolist())
#https://datascience.stackexchange.com/a/40068
classes = ['without_mask', 'with_mask', 'with_Improper_mask']
def plot_confusion_matrix(cm, classes,
                          normalize=False,
                          title='Confusion matrix',
                          cmap=plt.cm.Blues):
    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)
    fmt = '.2f' if normalize else 'd'
    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, format(cm[i, j], fmt),
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")
```

```
plt.ylabel('True label')
    plt.xlabel('Predicted label')
    plt.tight layout()
plot confusion matrix(cmx,classes)
"""### HOG+MLP with balanced Dataset"""
model mlp bal =
tf.keras.models.load_model('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/Mod
els/model mlp hyp bal.h5')
model mlp bal.evaluate(test dataset hog)
pred mlp bal = model mlp bal.predict(hog images)
predicted_mlp_bal = np.argmax(pred_mlp_bal, axis=1)
cmx = confusion_matrix(labels.tolist(),predicted_mlp_bal.tolist())
#https://datascience.stackexchange.com/a/40068
classes = ['without mask', 'with mask', 'with Improper mask']
def plot confusion matrix(cm, classes,
                          normalize=False,
                          title='Confusion matrix',
                          cmap=plt.cm.Blues):
    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)
    fmt = '.2f' if normalize else 'd'
    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, format(cm[i, j], fmt),
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")
    plt.ylabel('True label')
    plt.xlabel('Predicted label')
    plt.tight layout()
print(classification_report(labels,predicted_mlp_bal))
plot confusion matrix(cmx,classes)
"""# CNN Baseline
```

```
### CNN baseline with imbalanced dataset
model cnn =
tf.keras.models.load_model('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/Mod
els/model cnn.h5')
model_cnn.summary()
model_cnn.evaluate(test_dataset)
pred cnn unbal = model cnn.predict(or images)
predicted cnn unbal = np.argmax(pred cnn unbal, axis=1)
cm cnn unbal= confusion matrix(labels.tolist(),predicted cnn unbal.tolist())
print(classification_report(labels,predicted_cnn_unbal))
plot_confusion_matrix(cm_cnn_unbal,classes)
"""### CNN baseline with Balanced Dataset"""
model cnn bal =
tf.keras.models.load model('/content/drive/MyDrive/Computer-vision/CW Folder PG/Mod
els/model_cnn_bal.h5')
model cnn bal.evaluate(test dataset)
pred cnn bal = model cnn bal.predict(or images)
predicted cnn bal = np.argmax(pred cnn bal, axis=1)
cm_cnn_bal= confusion_matrix(labels.tolist(),predicted_cnn_bal.tolist())
print(classification report(labels, predicted cnn bal))
plot confusion matrix(cm cnn bal,classes)
"""# RESNET 50
### RESNET 50 with imbalanced dataset
model res =
tf.keras.models.load_model('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/Mod
els/model res.h5')
model_res.evaluate(test_dataset)
pred res unbal = model res.predict(or images)
predicted_res_unbal = np.argmax(pred_res_unbal, axis=1)
cm res unbal= confusion matrix(labels.tolist(),predicted res unbal.tolist())
```

```
print(classification_report(labels,predicted_res_unbal))
plot_confusion_matrix(cm_res_unbal,classes)
class names = {0: "without mask", 1: "with mask", 2: "improper mask"}
idx = [8,255,166,350,455]
for i in idx:
  fig, axs = plt.subplots(1, 2, figsize=(10, 5))
  # Plot the original image in the first subplot
  class_name = class_names[int(labels[i])]
  axs[0].imshow(or images[i])
  print(class name)
  axs[0].set title("Original:"+str(class name))
  # Plot the predicted label in the second subplot
  class name pred = class names[int(predicted res unbal[i])]
  axs[1].imshow(or images[i])
  axs[1].set_title("predicted:"+ str(class_name_pred))
  # Display the plot
  plt.show()
"""### RESNET 50 with balanced dataset"""
model res balanced =
tf.keras.models.load_model('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/Mod
els/model resv2.1.h5')
model res balanced.evaluate(test dataset)
pred_res_bal = model_res_balanced.predict(or_images)
predicted res bal = np.argmax(pred res bal, axis=1)
cm res bal= confusion matrix(labels.tolist(),predicted res bal.tolist())
print(classification report(labels, predicted res bal))
plot_confusion_matrix(cm_res_bal,classes)
"""# Processing Video"""
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
# Loading the video
cv.VideoCapture('/content/drive/MyDrive/Computer-vision/CW Folder PG/Video/TestVide
o.mp4')
# Getting the frames per second and size of the video
```

```
fps = int(cap.get(cv.CAP_PROP_FPS))
frame_width = int(cap.get(3))
frame_height = int(cap.get(4))
size = (frame_width, frame_height)
# Creating a dictionary of class names for the face mask classification
class names = {0: "without mask", 1: "with mask", 2: "improper mask"}
print('FPS:', fps)
print('size:',size)
# Create a VideoWriter object for the output file
fourcc = cv.VideoWriter_fourcc(*'XVID')
out = cv.VideoWriter('output.mp4', fourcc, fps, size)
# Initializing counters for frame and detected faces
count = 0
number of frames = int(cap.get(cv.CAP PROP FRAME COUNT))
# Starting the video processing loop
while True:
  # Read the current frame
  s, img = cap.read()
  if not s:
    break
  # Detecting faces in the current frame using a RetinaFace model
  ret faces = RetinaFace.detect faces(img)
  # If faces are detected, classify them and draw boxes with labels on the image
  if ret faces is not None:
    count += 1
    try:
      # Get the coordinates of the detected face
      x1,y1,x2,y2 = ret faces['face 1']['facial area'][0],
ret_faces['face_1']['facial_area'][1], ret_faces['face_1']['facial_area'][2],
ret_faces['face_1']['facial_area'][3]
      # Drawing a rectangle around the face on the image
      img = cv.rectangle(img, (int(x1), int(y1)), (int(x2), int(y2)), (128, 0, 0),
2)
      # Croping the detected face
      croped faces = img[y1:y2, x1:x2]
      # Creating a frame count string
      frame count = 'Frame: ' + str(count)
    except:
      continue
    # Resizing the cropped face to 64x64 pixels and normalize its values
    _face = cv.resize(croped_faces,(64,64))
    _face = cv.normalize(_face, None, 0, 1.0, cv.NORM_MINMAX, dtype=cv.CV_32F)
    face = face[tf.newaxis,:,:,:]
    # Classifying the face using a pre-trained model and get its class index
    pred = model_res.predict(_face)
    class idx = tf.argmax(pred, axis=-1)
    class idx = class idx.numpy()
```

```
# Get the class name from the dictionary using the class index
    class_name = class_names[int(class_idx)]
    # Get the class name from the dictionary using the class index
    w,h= cv.getTextSize(class_name,cv.FONT_HERSHEY_SIMPLEX, 0.5, 2)
    img = cv.rectangle(img, (int(x1), int(y1)-25), (int(x1)+w[0], int(y1)), (0, 0, 0)
0), -1)
    img = cv.rectangle(img, (int(x1), int(y1)), (int(x2), int(y2)), (0, 255, 0), 2)
    img = cv.putText(img, class_name, (int(x1), int(y1)-10),
cv.FONT_HERSHEY_SIMPLEX, 0.5, (0, 255, 0), 2)
    img = cv.putText(img, frame_count, (int(frame_width)-150,
int(frame_height)-500), cv.FONT_HERSHEY_SIMPLEX, 0.75, (255, 0, 0), 2)
    pil image = Image.fromarray(img)
    pil image.show()
  out.write(img)
  if cv.waitKey(1) & 0xFF == ord('q'):
cap.release()
out.release()
cv.destrovAllWindows()
print("success")
# ********** /test function.ipynb *************
# -*- coding: utf-8 -*-
"""test functions.ipynb
Automatically generated by Colaboratory.
Original file is located at
    https://colab.research.google.com/drive/1cubsXJXGICTE9-kIPnTFkbnyr5IL3tWj
# Computer Vision Coursework Submission (INM460)
**Student name, ID and cohort:** Sam Clastine Jesumuthu (220038747) - PG
# Notebook Setup
In this section you should include all the code cells required to test your
coursework submission. Specifically:
### Mount Google Drive
from google.colab import drive
drive.mount('/content/drive')
!pip install retina-face
```

"""### Define Local Path

In the next cell you should assign to the variable `GOOGLE_DRIVE_PATH_AFTER_MYDRIVE` the relative path of this folder in your Google Drive.

IMPORTANT: you have to make sure that **all the files required to test your functions are loaded using this variable** (as was the case for all lab tutorials). In other words, do not use in the notebook any absolute paths. This will ensure that the markers can run your functions. Also, **do not use** the magic command `%cd` to change directory.

.....

import os

TODO: Fill in the Google Drive path where you uploaded the CW_folder_PG
Example: GOOGLE_DRIVE_PATH_AFTER_MYDRIVE = 'Colab Notebooks/Computer
Vision/CW folder PG'

GOOGLE_DRIVE_PATH_AFTER_MYDRIVE =
'/content/drive/MyDrive/Computer-vision/CW_Folder_PG'
GOOGLE_DRIVE_PATH = os.path.join('drive', 'My Drive',
GOOGLE_DRIVE_PATH_AFTER_MYDRIVE)
print(os.listdir(GOOGLE_DRIVE_PATH))

"""### Load packages

In the next cell you should load all the packages required to test your functions.

import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import load_model
import cv2 as cv
from retinaface import RetinaFace
from matplotlib import animation, rc
from IPython.display import HTML
import imageio
from skimage.transform import resize

"""### Load models

In the next cell you should load all your trained models for easier testing of your functions. Avoid to load them within `MaskDetection` and `MaskDetectionVideo` to avoid having to reload them each time.

```
hog mlp imbalanced = load model(os.path.join(GOOGLE DRIVE PATH,
'Models', 'model mlp hyp unbal.h5'))
hog mlp balanced = load model(os.path.join(GOOGLE DRIVE PATH,
'Models', 'model mlp hyp bal.h5'))
baseline cnn imbalanced = load model(os.path.join(GOOGLE DRIVE PATH,
'Models', 'model cnn.h5'))
baseline cnn balanced = load model(os.path.join(GOOGLE DRIVE PATH,
'Models','model_cnn_bal.h5'))
Resnet_50_imbalanced = load_model(os.path.join(GOOGLE_DRIVE_PATH,
'Models', 'model res.h5'))
Resnet 50 balanced = load model(os.path.join(GOOGLE DRIVE PATH,
'Models', 'model resv2.1.h5'))
"""# Load Test Dataset"""
x test =
np.load('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/CW_Dataset/or_images_t
est.npy')
x test hog =
np.load('/content/drive/MyDrive/Computer-vision/CW Folder PG/CW Dataset/hog images
test.npy')
y test
=np.load('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/CW_Dataset/y_test.npy
"""# Test MaskDetection
This section should allow a quick test of the `MaskDetection` function. First, add
cells with the code needed to load the necessary subroutines to make
`MaskDetection` work.
def modelPlot(x_test_hog,x_test,y_test,model,modelName):
  if modelName == 'hog-mlp-imbalanced' or 'hog-mlp-balanced':
    prediction = model.predict(x_test_hog)
  else:
    prediction = model.predict(x test)
  prediction = np.argmax(prediction, axis=1)
  class names = {0: "without mask", 1: "with mask", 2: "improper mask"}
  fig, axs = plt.subplots(1, 4, figsize=(20, 10))
  predicted class name = class names[int(prediction[8])]
  class name = class names[int(y test[8])]
  axs[0].imshow(cv.cvtColor(x_test[8], cv.COLOR_BGR2RGB))
  axs[0].set title("Original:"+str(class name)+", predicted:"+
str(predicted class name))
  class name = class names[int(y test[255])]
  predicted class name = class names[int(prediction[255])]
  axs[1].imshow(cv.cvtColor(x test[255], cv.COLOR BGR2RGB))
```

```
axs[1].set_title("Original:"+str(class_name)+", predicted:"+
str(predicted_class_name))
  class name = class_names[int(y_test[166])]
  predicted class name = class names[int(prediction[166])]
  axs[2].imshow(cv.cvtColor(x_test[166], cv.COLOR_BGR2RGB))
  axs[2].set title("Original:"+str(class name)+", predicted:"+
str(predicted class name))
  class_name = class_names[int(y_test[455])]
  predicted class name = class names[int(prediction[455])]
  axs[3].imshow(cv.cvtColor(x_test[455], cv.COLOR_BGR2RGB))
  axs[3].set_title("Original:"+str(class_name)+", predicted:"+
str(predicted class name))
  plt.show()
def MaskDetection(x_test, x_test_hog, y_test, modelName):
  if modelName == 'hog-mlp-imbalanced':
    modelPlot(x test hog,x test,y test,hog mlp imbalanced,modelName)
  if modelName== 'hog-mlp-balanced':
    prediction = hog mlp balanced.predict(x test hog)
    modelPlot(x_test_hog,x_test,y_test,hog_mlp_balanced,modelName)
  if modelName== 'baseline-cnn-imbalanced':
    baseline cnn prediction = baseline cnn imbalanced.predict(x test)
    modelPlot(x test hog,x test,y test,baseline cnn_imbalanced,modelName)
  if modelName== 'baseline-cnn-balanced':
    baseline cnn prediction = baseline cnn balanced.predict(x test)
    modelPlot(x test hog,x test,y test,baseline cnn balanced,modelName)
  if modelName== 'Resnet-50-imbalanced':
    resNet 50 prediction = Resnet 50 imbalanced.predict(x test)
    modelPlot(x test_hog,x_test,y_test,Resnet_50_imbalanced,modelName)
  if modelName== 'Resnet-50-balanced':
    resNet 50 prediction = Resnet 50 balanced.predict(x test)
    modelPlot(x_test_hog,x_test,y_test,Resnet_50_balanced,modelName)
"""Then, make a call to the `MaskDetection` function to see what results it
produces. You must also indicate the syntax needed to test your different
models."""
# where model type can be one of
  # 'hog-mlp-imbalanced'
  # 'hog-mlp-balanced'
  # 'baseline-cnn-balanced'
  # 'baseline-cnn-imbalanced'
  # 'Resnet-50-balanced'
```

```
# 'Resnet-50-imbalanced'
MaskDetection(x_test, x_test_hog,y_test,'Resnet-50-balanced')
"""# Test MaskDetectionVideo
This section should allow a quick test of the `MaskDetectionVideo` function. First,
add cells with the code needed to load the necessary subroutines to make
`MaskDetectionVideo` work.
def MaskDetectionVideo(file):
  # Loading the video
  cap = cv.VideoCapture(file)
  fps = int(cap.get(cv.CAP PROP FPS))
  # Getting the frames per second and size of the video
  frame width = int(cap.get(3))
  frame height = int(cap.get(4))
  size = (frame_width, frame_height)
  # Creating a dictionary of class names for the face mask classification
  class names = {0: "without mask", 1: "with mask", 2: "improper mask"}
  print('FPS:', fps)
  print('size:',size)
  # Create a VideoWriter object for the output file
  fourcc = cv.VideoWriter_fourcc(*'XVID')
  out = cv.VideoWriter('output.mp4', fourcc, fps, size)
  # Initializing counters for frame and detected faces
  count = 0
  number of frames = int(cap.get(cv.CAP PROP FRAME COUNT))
  # Starting the video processing loop
  while True:
    # Read the current frame
    s, img = cap.read()
    if not s:
      break
    # Detecting faces in the current frame using a RetinaFace model
    ret_faces = RetinaFace.detect_faces(img)
    # If faces are detected, classify them and draw boxes with labels on the image
    if ret faces is not None:
      count += 1
      try:
        # Get the coordinates of the detected face
        x1,y1,x2,y2 = ret_faces['face_1']['facial_area'][0],
ret_faces['face_1']['facial_area'][1], ret_faces['face_1']['facial_area'][2],
ret_faces['face_1']['facial_area'][3]
        # Drawing a rectangle around the face on the image
        img = cv.rectangle(img, (int(x1), int(y1)), (int(x2), int(y2)), (128, 0,
0), 2)
        # Croping the detected face
        croped faces = img[y1:y2, x1:x2]
        # Creating a frame count string
```

```
frame_count = 'Frame: ' + str(count)
      except:
        continue
        # Resizing the cropped face to 64x64 pixels and normalize its values
      face = cv.resize(croped faces,(64,64))
      _face = cv.normalize(_face, None, 0, 1.0, cv.NORM_MINMAX, dtype=cv.CV 32F)
      face = face[tf.newaxis,:,:,:]
      # Classifying the face using a pre-trained model and get its class index
      pred = Resnet_50_imbalanced.predict(_face)
      class_idx = tf.argmax(pred, axis=-1)
      class idx = class idx.numpy()
      # Get the class name from the dictionary using the class index
      class name = class names[int(class idx)]
      # Get the class name from the dictionary using the class index
      w,h= cv.getTextSize(class_name,cv.FONT_HERSHEY_SIMPLEX, 0.5, 2)
      img = cv.rectangle(img, (int(x1), int(y1)-25), (int(x1)+w[0], int(y1)), (0,
0, 0), -1)
      img = cv.rectangle(img, (int(x1), int(y1)), (int(x2), int(y2)), (0, 255, 0),
2)
      img = cv.putText(img, class_name, (int(x1), int(y1)-10),
cv.FONT HERSHEY SIMPLEX, 0.5, (0, 255, 0), 2)
      img = cv.putText(img, frame count, (int(frame width)-150,
int(frame height)-500), cv.FONT HERSHEY SIMPLEX, 0.75, (255, 0, 0), 2)
    out.write(img)
    if cv.waitKey(1) & 0xFF == ord('q'):
        break
  # Release the resources
  cap.release()
  out.release()
  cv.destroyAllWindows()
  video = imageio.mimread('/content/output.mp4',memtest=False) #Loading video
  video = [resize(frame, (540,940))[..., :3] for frame in video]
  HTML(display video(video).to html5 video())
  print("success")
#https://github.com/TUIlmenauAMS/Videocoding/blob/main/seminars/vc seminar01 suppor
t.ipynb
def display video(video):
    fig = plt.figure(figsize=(7,7)) #Display size specification
    mov = []
    for i in range(len(video)): #Append videos one by one to mov
        img = plt.imshow(video[i], animated=True)
        plt.axis('off')
        mov.append([img])
    #Animation creation
    anime = animation.ArtistAnimation(fig, mov, interval=50, repeat delay=1000)
```

```
plt.close()
return anime
```

"""Then, make a call to the `MaskDetectionVideo` function to see what results it produces."""

MaskDetectionVideo('/content/drive/MyDrive/Computer-vision/CW_Folder_PG/Video/TestV
ideo.mp4')

#https://github.com/TUIlmenauAMS/Videocoding/blob/main/seminars/vc_seminar01_suppor t.ipynb

video = imageio.mimread('/content/output.mp4',memtest=False) #Loading video video = [resize(frame, (540,940))[..., :3] for frame in video] HTML(display_video(video).to_html5_video()) #Inline video display in HTML5