Group B Team 2

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# Detecting faces in images using NN and OpenCV - code

#### Downloading tools to measure time and memory usage

!pip install ipython-autotime %load\_ext autotime !pip install wandb

### Importing the necessary packages

import tensorflow as tf
import numpy as np
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from keras.preprocessing import image
from google.colab import drive
import pickle
from IPython.display import Image
from google.colab import files
import matplotlib.pyplot as plt
from sklearn.metrics import confusion\_matrix
import itertools
import cv2
from google.colab.patches import cv2\_imshow
import wandb

# **Connecting to the Google Drive**

drive.mount('/content/drive')

# Loading image data

trainingDatasetPath = "/content/drive/MyDrive/data/training" validationDatasetPath = "/content/drive/MyDrive/data/validation" testingDatasetPath = "/content/drive/MyDrive/data/testing"

trainingDatasetImageGenerator = ImageDataGenerator(rescale=1./255, rotation\_range=20, width\_shift\_range=0.1,

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height_shift_range=0.1,
  shear_range=0.2,
  zoom range=0.3,
  horizontal_flip=True,
  fill mode='nearest')
validationDatasetImageGenerator = ImageDataGenerator(rescale=1./255,
  rotation range=20,
  width shift range=0.1,
  height shift range=0.1,
  shear_range=0.2,
  zoom_range=0.3,
  horizontal flip=True,
  fill_mode='nearest')
testingDatasetImageGenerator = ImageDataGenerator(rescale=1./255)
trainingDataset = trainingDatasetImageGenerator.flow from directory(trainingDatasetPath,
target_size=(128,128), batch_size=64, class_mode='binary')
validationDataset =
validationDatasetImageGenerator.flow from directory(validationDatasetPath,
target size=(128,128), batch size=64, class mode='binary')
testingDataset = testingDatasetImageGenerator.flow from directory(testingDatasetPath,
target_size=(128,128), batch_size=64, class_mode='binary')
Show sample images
plt.imshow(trainingDataset[0][0][0])
plt.imshow(validationDataset[0][0][0])
plt.imshow(testingDataset[0][0][0])
Preparing the structure of the CNN model
model = tf.keras.models.Sequential([
       tf.keras.layers.Conv2D(64, (3,3), padding='same', activation=tf.nn.relu,
input shape=(128, 128, 3)),
  tf.keras.layers.MaxPooling2D((2,2), padding='same'),
  tf.keras.layers.Conv2D(64, (3,3), padding='same', activation=tf.nn.relu),
  tf.keras.layers.MaxPooling2D((2,2), padding='same'),
  tf.keras.layers.Conv2D(128, (3,3), padding='same', activation=tf.nn.relu),
  tf.keras.layers.MaxPooling2D((2,2), padding='same'),
  tf.keras.layers.Conv2D(128, (3,3), padding='same', activation=tf.nn.relu),
  tf.keras.layers.MaxPooling2D((2,2), padding='same'),
  tf.keras.layers.Flatten(),
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tf.keras.layers.Dense(512, activation=tf.nn.relu),

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tf.keras.layers.Dense(1, activation=tf.nn.sigmoid)
])
model.compile(loss='binary_crossentropy',
  optimizer=tf.keras.optimizers.RMSprop(lr=0.001),
  metrics=['binary_accuracy'])
Show the structure of the CNN model
model.summary()
tf.keras.utils.plot_model(model, show_shapes=True)
Learn model, save model and plot the results
wandb.init()
history = model.fit(
  trainingDataset,
  epochs=10,
  validation_data=validationDataset,
  verbose=1
 )
export_model_dir = '/content/drive/MyDrive/model'
model.save(export_model_dir)
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'val'], loc='upper left')
plt.show()
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'val'], loc='upper left')
plt.show()
Load saved model from Google Drive
export_model_dir = '/content/drive/MyDrive/model'
model = tf.keras.models.load model(export model dir)
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#### **Evaluate model**

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wandb.init()
resultsEvaluate = model.evaluate(testingDataset, verbose=1)
print('accuracy:')
print(resultsEvaluate[1])
print('loss:')
print(resultsEvaluate[0])
Show confusion matrix
prediction = np.argmax(model.predict_generator(testingDataset, verbose=1), axis=1)
confusionMatrix = confusion_matrix(testingDataset.classes, prediction)
classes = []
for key in testingDataset.class_indices:
 classes.append(key)
plt.imshow(confusionMatrix, interpolation='nearest', cmap=plt.cm.Blues)
plt.title("Confusion matrix")
numberOfClasses = np.arange(len(classes))
plt.xticks(numberOfClasses, classes, rotation=45)
plt.yticks(numberOfClasses, classes)
confusionMatrix = confusionMatrix.astype('float') / confusionMatrix.sum(axis=1)[:,
np.newaxis]
confusionMatrix = np.around(confusionMatrix, decimals=10)
confusionMatrix[np.isnan(confusionMatrix)] = 0.0
thresh = confusionMatrix.max() / 2.
for i, j in itertools.product(range(confusionMatrix.shape[0]),
range(confusionMatrix.shape[1])):
 plt.text(j, i, confusionMatrix[i, j],
       horizontalalignment="center",
       color="white" if confusionMatrix[i, j] > thresh else "black")
plt.tight_layout()
plt.ylabel('True label')
plt.xlabel('Predicted label')
Use model to predict a given image from a desktop
predictFile = files.upload()
fileName = list(predictFile.keys())[0]
img = image.load_img(fileName, target_size=(128, 128))
img = image.img_to_array(img)
img = np.expand dims(img, axis = 0)
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img = img / 255
imgOpenCV = cv2.imread(fileName)
imgOpenCV = cv2.resize(imgOpenCV, (256,256))

result = model.predict(img, batch_size=1)
print(result)
if (result[0][0] < 0.5):
    imgOpenCV = cv2.rectangle(imgOpenCV, (0,0),(62, 32), (0, 0, 0), -1)
    imgOpenCV = cv2.putText(imgOpenCV, "face", (2,22), cv2.FONT_HERSHEY_PLAIN, 1.5, (255, 255, 255), 2)
else:
    imgOpenCV = cv2.rectangle(imgOpenCV, (0,0),(102, 32), (0, 0, 0), -1)
    imgOpenCV = cv2.putText(imgOpenCV, "no face", (2,22), cv2.FONT_HERSHEY_PLAIN, 1.5, (255, 255, 255), 2)</pre>
cv2_imshow(imgOpenCV)
```