计算机视觉第一次作业

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一、实验主题

基于Google Teachable Machine 实现简单的电子设备识别

二、实验目的

通过学习一定的相关知识,学习如何搭建计算机视觉需要用到的软件环境,以及使用Google Teachable Machine实现一些简单的物体识别,并将模型导出,在本地使用。

三、实验内容

1、配置环境

要实现本次实验以及后续可能需要完成的其他实验,配置好相应的软件环境至关重要。

根据Google Teachable Machine上提供的示例代码的注释了解到,要在本地运行已训练的模型,需要TensorFlow库的支持。

TensorFlow是一个强大的机器学习框架,可以支持深度学习的各种算法,是目前最受欢迎的机器学习开源专案,很多大型电商使用的客服系统就是基于TensorFlow开发。

1)建立虚拟环境

为了防止和其他python版本和对应的库混用,用Anaconda建立一个新的虚拟python环境,如图:

Anaconda Prompt (conda)

```
(base) C:\Users\72331>conda create --name tensorf1ow python=3.9_
```

再激活该环境:

```
(base) C:\Users\72331>conda activate tensorflow
(tensorflow) C:\Users\72331>
```

2)安装TensorFlow

在此处的python环境我们使用了python3.9,需要搭配TensorFlow2.5以正常使用。使用pip指令安装,如图:

```
(tensorflow) C:\Users\72331>pip install tensorflow==2.5_
```

3)安装OpenCV

因为之后的测试样例常常用到OpenCV,一并将其配置到虚拟环境中,如图:

```
(tensorflow) C:\Users\72331>pip install opencv-python_
```

4)测试安装完毕

使用VSCode运行一段测试代码(代码如下)来测试是否已经安装完成TensorFlow和OpenCV:

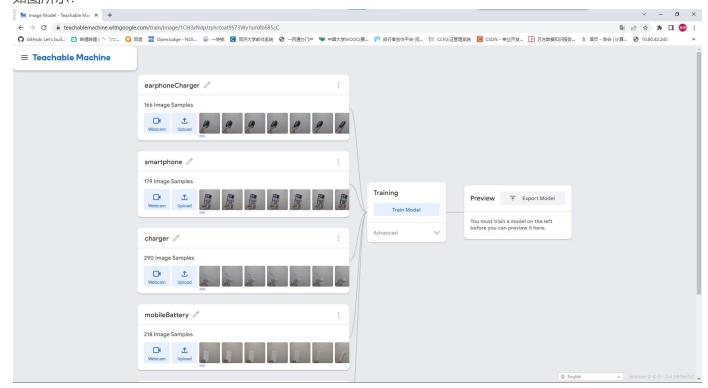
```
import tensorflow as tf
import cv2
print(tf)
print(cv2)
```

```
PS D:\Grade II\Computer Vision\作业 源代码 & 'C:\Users\72331\.conda\envs\tensorflow\python.exe' 'c:\Users\72331\.vscode\extensions\ms-python.python-20 23.4.1\pythonFiles\lib\python\debugpy\adapter/...\.debugpy\launcher' '51998' '--' 'D:\Grade II\Computer Vision\作业 源代码\test.py' 2023-03-22 18:28:07.990736: W tensorflow\stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'cudart64_110.dll'; dlerror: cudart64_110.dll not found 2023-03-22 18:28:08.001909: I tensorflow\stream_executor/cuda/cudart_stub.cc:29] Ignore above cudart dlerror if you do not have a GPU set up on your machine. kmodule 'tensorflow' from 'C:\Users\72331\\.conda\\envs\\tensorflow\\lib\\site-packages\\tensorflow\\__init__.py'> kmodule 'cv2' from 'C:\Users\\72331\\.conda\\envs\\tensorflow\\lib\\site-packages\\cv2\\__init__.py'> PS D:\Grade II\Computer Vision\frac{\tensorflow}\\lib\\site-packages\\cv2\\__init__.py'>
```

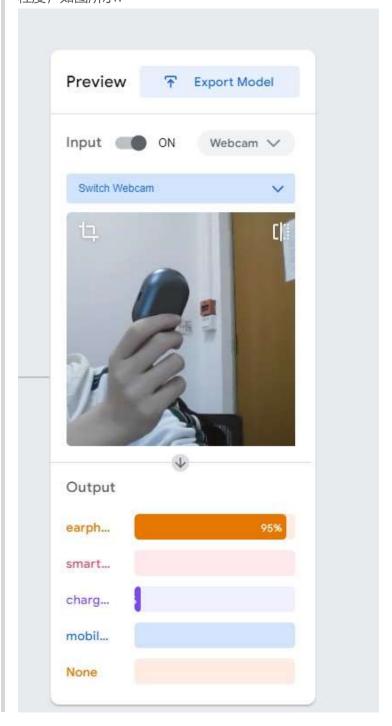
2、开始初步训练

进入Teachable Machine后,点击开始使用,训练图片项目。

进入项目后,将分类分别命名成对应电子设备的名字,然后将webcam打开,分别采集每一种物品的样本图片,如图所示:

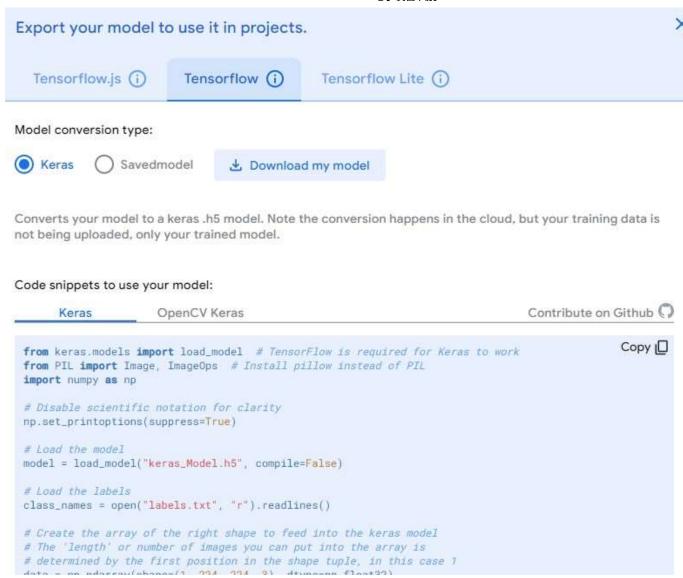


分类建立完成,并添加好样本后,开始训练模型。等待一段时间后,可以在网页上即时预览到自己的模型准确程度,如图所示:



3、导出模型到本地

选择Export Model,选中TensorFlow,Download my model,将模型下载到本地。

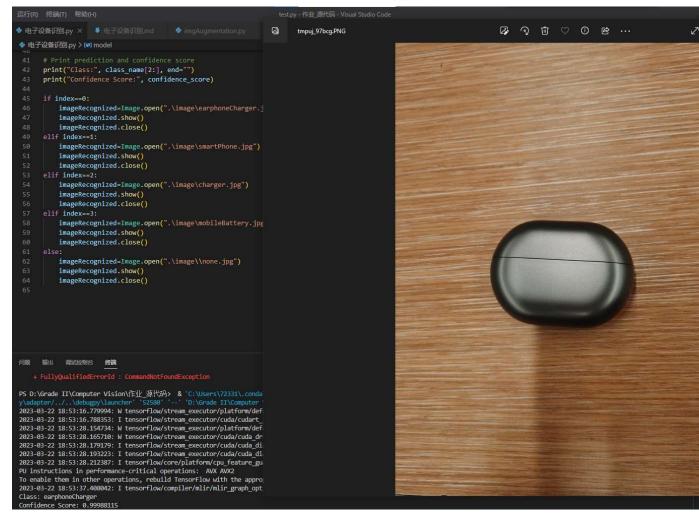


然后把下面的代码复制到一个python源文件里,将这个源文件和刚才下载的模型放在同一目录下运行,即可对某张本地的图片进行识别,如图所示:

```
| FullyQualifiedErrorId: CommandNotFoundException

P5 D:\Grade II\Computer Vision\作业源代码> & 'C:\Users\72331\.conda\envs\tensorFlow\python.exe' 'c:\Users\72331\.vscode\extensions\ms-python.python-2023.4.1\pythonFiles\lib\python\debugp \\partial_{\tensor}\partial_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tensor}\psi_{\tens
```

也就是说,我们已经成功将识别结果导出来了,现在我们就已经可以对它进行一些其他操作,比如,根据识别结果展示对应的图片,如图所示:

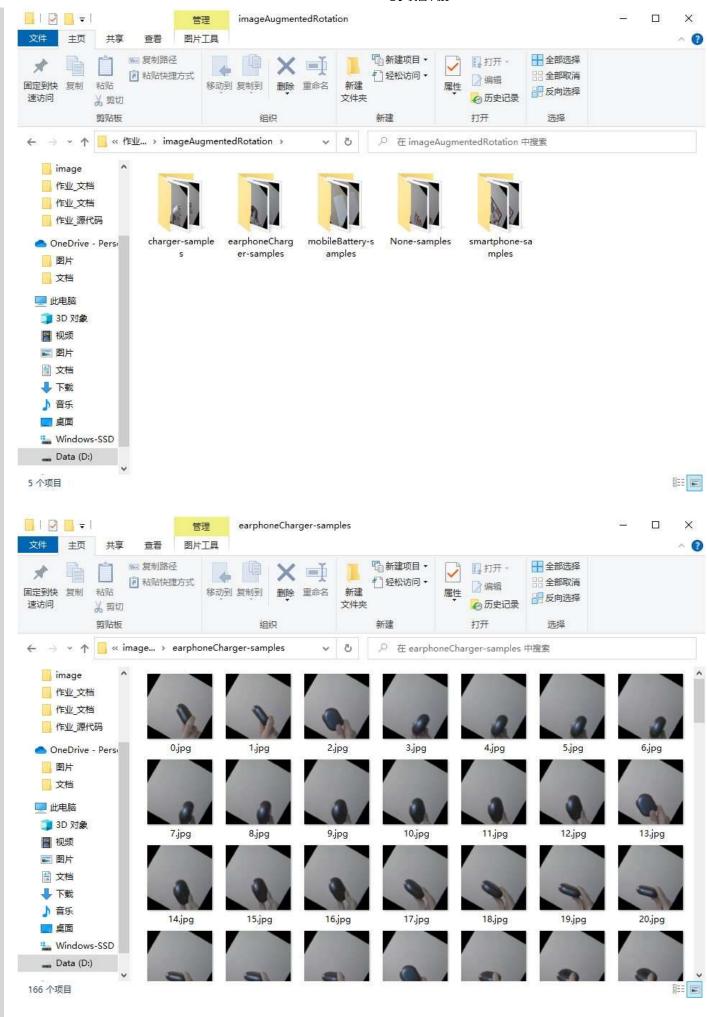


4、尝试数据增强

通过实验,发现有时候会识别错误或者识别不到,考虑通过数据增强的方式,预处理原始数据,然后重新导入 样本到训练模型。

预处理采用了PIL库中的各项函数来对某张图片进行修改(翻转、旋转等),然后批量处理输出到新文件夹中,最后将样本重新导入到网站中重新训练。(代码附在报告结尾)

以下是经过数据增强后的样本文件夹(以左右翻转、耳机充电仓为例):



重新训练模型后重复之前的实验, 发现识别的精度有所提高。

```
# 图像识别
from keras.models import load model # TensorFlow is required for Keras to work
from PIL import Image, ImageOps # Install pillow instead of PIL
import numpy as np
# Disable scientific notation for clarity
np.set printoptions(suppress=True)
# Load the model
model = load_model("keras_Model.h5", compile=False)
# Load the labels
class names = open("labels.txt", "r").readlines()
# Create the array of the right shape to feed into the keras model
# The 'length' or number of images you can put into the array is
# determined by the first position in the shape tuple, in this case 1
data = np.ndarray(shape=(1, 224, 224, 3), dtype=np.float32)
# Replace this with the path to your image
image = Image.open(".\image\IMG 20230321 160416.jpg").convert("RGB")
# resizing the image to be at least 224x224 and then cropping from the center
size = (224, 224)
image = ImageOps.fit(image, size, Image.Resampling.LANCZOS)
# turn the image into a numpy array
image_array = np.asarray(image)
# Normalize the image
normalized_image_array = (image_array.astype(np.float32) / 127.5) - 1
# Load the image into the array
data[0] = normalized_image_array
# Predicts the model
prediction = model.predict(data)
index = np.argmax(prediction)
class_name = class_names[index]
confidence_score = prediction[0][index]
# Print prediction and confidence score
print("Class:", class_name[2:], end="")
print("Confidence Score:", confidence_score)
if index==0:
    imageRecognized=Image.open(".\image\earphoneCharger.jpg")
    imageRecognized.show()
    imageRecognized.close()
elif index==1:
    imageRecognized=Image.open(".\image\smartPhone.jpg")
    imageRecognized.show()
    imageRecognized.close()
elif index==2:
    imageRecognized=Image.open(".\image\charger.jpg")
    imageRecognized.show()
    imageRecognized.close()
elif index==3:
    imageRecognized=Image.open(".\image\mobileBattery.jpg")
    imageRecognized.show()
    imageRecognized.close()
```

```
else:
    imageRecognized=Image.open(".\image\\none.jpg")
    imageRecognized.show()
    imageRecognized.close()
# 数据批量预处理
from PIL import Image
import os
import imageAugFunctions as iaf
if __name__ == '__main__':
    deviceClasses=['charger-samples','earphoneCharger-samples',
             'mobileBattery-samples','None-samples','smartphone-samples']
    for deviceClass in deviceClasses:
        path='D:\Grade II\Computer Vision\作业_源代码\imageSamples\\'+deviceClass
        filelist=os.listdir(path)
        path all=[]
        for f in filelist:
            if f.endswith('.jpg'):
                path_all.append(os.path.join(path,f))
        pil ims=[]
        for p in path_all:
            pil_ims.append(Image.open(p))
        print(len(pil_ims))
        counting=0
        savingPathRotation='.\imageAugmentedRotation\\'+deviceClass+'\\'
        savingPathGray='.\imageAugmentedGray\\'+deviceClass+'\\'
        savingPathFlipLR='.\imageAugmentedFlipLR\\'+deviceClass+'\\'
        savingPathFlipTB='.\imageAugmentedFlipTB\\'+deviceClass+'\\'
        for pilCurrent in pil_ims:
            iaf.Rotation(pilCurrent, counting, savingPathRotation, 60)
            iaf.grayScaleConversion(pilCurrent,counting,savingPathGray)
            iaf.FlipLeftRight(pilCurrent,counting,savingPathFlipLR)
            iaf.FlipTopBottom(pilCurrent,counting,savingPathFlipTB)
            counting=counting+1
# 数据批量预处理所用函数
from PIL import Image
import os
def Rotation(pilCurrent,index,path,angel):
    savingPath=path+(str)(index)+'.jpg'
    pilRotate=pilCurrent.rotate(angel)
    pilRotate.save(savingPath)
def grayScaleConversion(pilCurrent,index,path):
    savingPath=path+(str)(index)+'.jpg'
    pilRotate=pilCurrent.convert('L')
    pilRotate.save(savingPath)
def FlipLeftRight(pilCurrent,index,path):
    savingPath=path+(str)(index)+'.jpg'
    pilRotate=pilCurrent.transpose(Image.FLIP_LEFT_RIGHT)
    pilRotate.save(savingPath)
def FlipTopBottom(pilCurrent,index,path):
    savingPath=path+(str)(index)+'.jpg'
    pilRotate=pilCurrent.transpose(Image.FLIP_TOP_BOTTOM)
    pilRotate.save(savingPath)
```