donttouchyourface

Donttouchyourface is a simple machine learning application that verbally warns the user if they touch their face.

The things you need to build your own

- CUDA (Optional)
 - Used during the training process to save time
- YOLOv3
 - · Used for real-time object detection
- · Your data set to train and test the model
 - At least 60 pictures (30 pictures for face and 30 pictures for hand)
- OpenCV
 - Used for capturing images from webcam and image processing
- PyGame
 - Used for playing sound files
- Camera

A quick summary of what I did

- Step 1- Created a small dataset which contains images of my face and hands
- · Step 2- Trained a YOLOv3 model using the dataset
- Step 3- After 3 hours of training, the model learned to detect and localize my face and hands in an image
- Step 4- Used the AWS Polly speech synthesis tool to create a few funny audio files to warn my self
- Step 5- Finally, I created a Python script that captures images from my webcam, uses the pre-trained model for inference and warns me if my hand gets too close to my face

Alt Text

Link to the Project Video: https://www.linkedin.com/embed/feed/update/urn:li:ugcPost:6647733944861159424?compact=1

My ancient test system

- GPU: Nvidia GTX 1050Ti 4GB
- CPU: Intel i5-3470
- RAM: 12GB
- OS: Ubuntu 18.04

Detailed steps

Step 1- Create a small dataset which contains images of your face and hands

- 1.A: Setup dataset directory and class name files
 - Create a new directory called dataset
 - mkdir dataset
 - cd dataset
 - Create a new .txt file called classes.txt
 - Open classes.txt with your favorite text editor such as nano and gedit etc
 - Type the following class names in a new line:
 - face
 - hand
 - Save the file. We will use this file dataset labeling process.

- 1.B: Install cheese to take pictures
- sudo apt install cheese
- 1.C: Take pictures of your hands and face
- I took 30 hand pictures and 30 face pictures. Remember the more the better!
- 25 for training and 5 for testing for each class.
- You will have all your pictures in one folder, so make sure the dataset is shuffled.
- Example image dataset:
- 1.D: Label your data set using the labeling tool
- There are several different tools that you can use to complete this step. I used https://github.com/tzutalin/labellmg
- Follow their instruction to install labelimg then run:
- python3 labelImg.py /home/melih/Desktop/dataset /home/melih/Desktop/dataset/classes.txt
- Click on the Change Save Dir button and choose your dataset directory. Because we want our images and their labels to be in the same directory.
- The default save format is Pascal/VOC, so change it to YOLO.
- Press on w to draw a rectangle in the image to label.
- Select the class type
- Save the image and go to the next image.
- Repeat this process for all the images in the dataset.

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· Example image dataset with labels:

• 1.E: Selecting Training and Testing Data

- This the final step in dataset preparation
- · YOLO looks for three different files in the dataset directory.
 - classes.names
 - This is a copy of the classes.txt that we have created earlier. It just has a different name, so you can copypaste classes.txt and change its name to classes.names
 - train.txt
 - This text file stores the training image locations. Create a text file and save training image locations.
 - test.txt
 - This text file stores the training image locations. Create a text file and save testing image locations.
 - Example train.txt and test.txt files:

Step 2- Train a YOLOv3 model using your dataset

. 2.A: Install Darknet:

- Follow the instructions at the link or below: https://pjreddie.com/darknet/install/
 - Download Darknet
 - git clone https://github.com/pjreddie/darknet.git
 - cd darknet
 - Before start building anything, you need to edit the Makefile for OpenCV and if choose to use GPU (You need to
 install CUDA first. See the next step). Make sure you have the following lines in the Makefile. ```

- GPU=1
- CUDNN=0
- OPENCV=1
- OPENMP=0
- DEBUG=0
- NUMPY=1 ```
- After editing Makefile type:
 - make
- How to install CUDA -> https://developer.nvidia.com/cuda-downloads
 - You can test the installation and see your CUDA version by typing:
 - nvcc --version

· 2.B: Share information about our custom dataset with YOLO

We need to tell YOLO how many numbers of classes and class names, where to find our custom datasets and where to
save models. To do that, you need to create a configuration file. You can use custom.data file as an example, but you
have to change the paths unless your user name is melih and your data set is located in Desktop:). You need to have the
following lines in this file.

• • • •

- o classes= 2
- train = /home/melih/Desktop/dataset/train.txt
- valid = /home/melih/Desktop/dataset/test.txt
- names = /home/melih/Desktop/dataset/classes.names
- o backup = /home/melih/Desktop/backup ````

• 2.C: Download the pre-trained model (darknet53.conv.74)

- darknet53.conv.74 is trained on Imagenet dataset. We want to use it because it is better to begin with some pre-trained weights than complete random weights.
 - wget https://pjreddie.com/media/files/darknet53.conv.74

• 2.D: Train a model using your data set

- You can follow the YOLO training instruction at https://pjreddie.com/darknet/yolo/. Also, I explain what you need to do to train your network in the next step as well.
- It took about 3 hours to start getting good results.
 - I overclocked my GPU to save time using the Green with Envy tool (Optional)
 - https://gitlab.com/leinardi/gwe
 - I am not sure how much time overclocking actually saves me. I probably spent more time on figuring out how to overclock my GPU:/
- Training YOLO in details:
 - ./darknet detector train cfg/custom.data cfg/yolov3-voc.cfg darknet53.conv.74
 - cfg/custom.data: The file tells where to find your dataset, class names and trained models
 - cfg/yolov3-voc.cfg: YOLO network architecture configuration.
 - I had to reduce the batch size and the number of subdivisions because my GPU has only 4 GB of memory.
 - batch=16
 - subdivisions=16
 - darknet53.conv.74: pre-trained weights

Step 3- Test the trained model

- After 3 hours of training, the model should learn to detect and localize your face and hands in an image.
- 3.A: Find the model in /backup

- 3.B: You will need a webcam for testing.
- · 3.C: Test the model
 - You can use the YOLO demo to quickly test the accuracy of your model.
 - ./darknet detector demo cfg/custom.data cfg/yolov3-voc_inf.cfg yolov3-voc.backup
 - yolov3-voc.backup: Your model
 - cfg/yolov3-voc_inf.cfg: We need to comment on the training config lines and uncomment testing config lines. In
 other words, you need to make sure, the batch size and the number of subdivision values are set to 1.
 - batch=1
 - subdivisions=1

Step 4- Create audio files to play

- Use the AWS Polly speech synthesis tool to create a few funny audio files to warn my self. I uploaded the audio files that I created if you want to skip this part.
 - 4.A: You have to create an AWS account.
 - 4.B: https://aws.amazon.com/polly/
 - 4.C: Alternatively, You can record your voice or use any other audio file as well (Optional)

Step 5- Create a Python script

- Finally, we need to create a Python script that captures images from the web webcam uses the pre-trained model for inference and warns us if our hands get too close to my face
- 5A: Install Anaconda (Optional)
 - Create a virtual environment (Python 3.6) with the following packages installed
 - OpenCV
 - Numpy
 - PyGame (Used it to play audio files
 - python3 -m pip install -U pygame --use
- 5B: You can either use the Python wrapper that I created for YOLOv3 and write your application or use the corona.py as an example.
 - https://github.com/mlherd/darknet/tree/python36_wrapper
 - This probably is the most tricky part of this tutorial.
 - If you decide to use my version of darkent.py, first you should check the changes I did in the darknet repo.
 - In the Makefile, you need to use the right path to the numpy installation directory.
 - After these changes, you need to rebuild YOLO. ```
 - cd darknet
 - make clean
 - make ```
 - To detect touches, I used the rectangle overlap algorithm to check if a hand overlaps with a face.

Final Result:

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Good luck and stay sterilized!