main(1)

April 26, 2018

1 Main

1.1 Part 1 - Face Detection

In this project, we aim to construct a face detection model. We used a method haar to extract features. After that, by applying extracted features to cascade method, we were able to dectect people's faces and also count the number of faces through pictures as well as webcam. Pre-trained cascade was used from OpenCV.

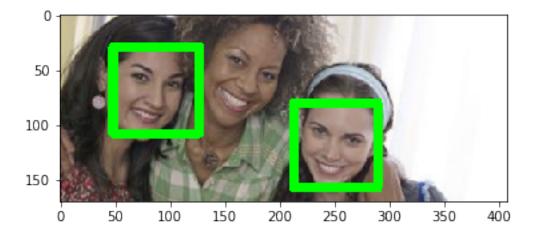
1.1.1 Part 1.1 - Face Detection without Rotation on Image

We started off the project with face detection on an image.

plt.show()

```
In [233]: os.chdir('../doc')
          # %load ../lib/count_face.py
          def counting_face():
                  import numpy as np
                  import cv2
                  from matplotlib import pyplot as plt
                  # loading OpenCV cascade for haar method with frontal face
                  face_cascade = cv2.CascadeClassifier('../lib/haarcascade_frontalface_default.x
                  # loading test image
                  img = cv2.imread('../data/test_image/104.jpg')
                  gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
                  faces = face_cascade.detectMultiScale(gray, 1.2, 5)
                  # implementing model
                  for (x,y,w,h) in faces:
                      cv2.rectangle(img,(x,y),(x+w,y+h),(0,255,0),8)
                  # showing image
                  RGB_img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
                  plt.imshow(RGB_img)
```

In [234]: counting_face()



1.1.2 Part 1.2 - Face Detection with Rotation

After implementation of Part 1, we realized the OpenCV front face cascade could not detect rotated face. Hence, we made some adjustments and declared additional functions to allow our model to analyze images from different rotation angle.

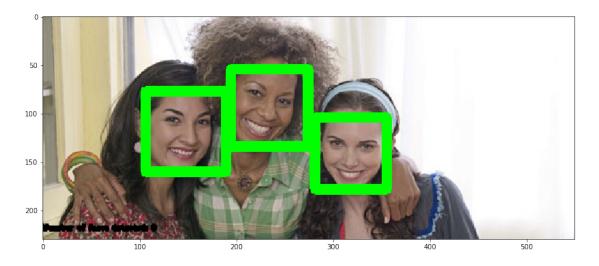
```
In [241]: # %load ../lib/counting_faces_image.py
          #!/usr/bin/env python3
          Created on Sat Apr 21 20:24:20 2018
          Qauthor: zailchen
          ,,,,,,
          def face_dectect_image(directory = '../data/test_image/cascade/', scaleFactor = 1.3, m
              import numpy as np
              import cv2
              import tensorflow
              import os
              from matplotlib import pyplot as plt
                # I followed Harrison Kingsley's work for this
                # Much of the source code is found https://pythonprogramming.net/haar-cascade-fa
              def rotate_image(img, angle):
                  if angle == 0: return img
                  # print("checked for shape".format(image.shape))
```

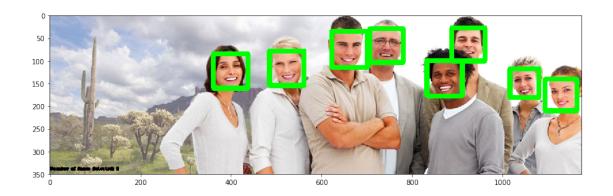
```
height, width = img.shape[:2]
    rot_mat = cv2.getRotationMatrix2D((width/2, height/2), angle, 0.9)
    result = cv2.warpAffine(img, rot_mat, (width, height), flags=cv2.INTER_LINEAR)
    return result
def rotate_point(pos, img, angle):
    if angle == 0: return pos
    x = pos[:,0] - img.shape[1]*0.4
    y = pos[:,1] - img.shape[0]*0.4
    newx = x*cos(radians(angle)) + y*sin(radians(angle)) + img.shape[1]*0.4
    newy = -x*sin(radians(angle)) + y*cos(radians(angle)) + img.shape[0]*0.4
    return np.array((newx, newy, pos[:,2], pos[:,3]), int).T
face_cascade = cv2.CascadeClassifier('../lib/haarcascade_frontalface_default.xml')
PATH_TO_TEST_IMAGES_DIR = directory
TEST_IMAGES_NAMES = os.listdir(directory)
TEST_IMAGE_PATHS = [os.path.join(PATH_TO_TEST_IMAGES_DIR, TEST_IMAGES_NAMES[i]) for
n = len(TEST_IMAGE_PATHS)
i = 0
for image in TEST_IMAGE_PATHS:
    img = cv2.imread(image)
    length = int(max(img.shape[0:2]))
    height = int(min(img.shape[0:2]))
    gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    for angle in [0, -30, 30]:
        rimg = rotate_image(gray, angle)
        faces = face_cascade.detectMultiScale(rimg, scaleFactor, minNeighbors)
        if len(faces):
                faces = rotate_point(faces, img, -angle)
                break
    if len(faces) == 0:
        print("No faces found")
    else:
        for (x,y,w,h) in faces:
            cv2.rectangle(img,(x,y),(x+w,y+h),(0,255,0),10)
        cv2.rectangle(img, ((0,img.shape[0] - 50*int(height/1080))), (620 * int(length))
        cv2.putText(img, "Number of faces detected: " + str(faces.shape[0]), (0,im
```

```
RGB_img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
plt.figure(figsize=(14,10))
plt.imshow(RGB_img)
plt.show()

cv2.imwrite('../output/processed_{{}}'.format(TEST_IMAGES_NAMES[i+1]),img)
i +=1
```

In [242]: face_dectect_image('../data/test_image/cascade')







1.1.3 Part 1.3 - Real Time Face Detection with WebCam

After we improved our model, we wanted to further develop our model. Therefore, in this part, we implemented real time face detection using WebCam.

```
In [225]: # %load ../lib/counting_faces_webcam.py
    #!/usr/bin/env python3
    """
    Created on Sun Apr 22 16:14:00 2018

    @author: zailchen
    """

    def face_dectect_webcam(scaleFactor = 1.3, minNeighbors = 5):

        import numpy as np
        import cv2
        from math import sin, cos, radians

        def rotate_image(img, angle):
            if angle == 0: return img
            # print("checked for shape".format(image.shape))
            height, width = img.shape[:2]
            rot_mat = cv2.getRotationMatrix2D((width/2, height/2), angle, 0.9)
```

```
result = cv2.warpAffine(img, rot_mat, (width, height), flags=cv2.INTER_LINEAR)
    return result
def rotate_point(pos, img, angle):
    if angle == 0: return pos
    x = pos[:,0] - img.shape[1]*0.4
   y = pos[:,1] - img.shape[0]*0.4
   newx = x*cos(radians(angle)) + y*sin(radians(angle)) + img.shape[1]*0.4
   newy = -x*sin(radians(angle)) + y*cos(radians(angle)) + img.shape[0]*0.4
    return np.array((newx, newy, pos[:,2], pos[:,3]), int).T
face_cascade = cv2.CascadeClassifier('../lib/haarcascade_frontalface_default.xml')
cap = cv2.VideoCapture(0)
while 1:
    ret, img = cap.read()
    gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    for angle in [0, -45, 45]:
        rimg = rotate_image(gray, angle)
        faces = face_cascade.detectMultiScale(rimg, scaleFactor, minNeighbors)
        if len(faces):
                faces = rotate_point(faces, img, -angle)
                break
    for (x,y,w,h) in faces:
        cv2.rectangle(img,(x,y),(x+w,y+h),(255,0,0),2)
    cv2.rectangle(img, ((0,img.shape[0] -25)),(270, img.shape[0]), (255,255,255),
    if type(faces) == tuple:
        cv2.putText(img, "Number of faces detected: 1" , (0,img.shape[0] -10), cv2
    else:
        cv2.putText(img, "Number of faces detected: " + str(faces.shape[0]), (0,im
    cv2.imshow('img',img)
    if cv2.waitKey(25) & OxFF == ord('q'):
          cv2.destroyAllWindows()
          cap.release()
          break
```

```
Press Q to close the camera

'''
face_dectect_webcam()
```

1.2 Part 2 - Object Detection API using Tensorflow

After we constructed our model in Part 1, we realized there exist some limitations in cascade model. Cascade model tends to have lower accuracy in side faces or partially showed faces. Also, cascade cannot detect highly rotated faces. To overcome such limitations, a popular and powerful approach is the use of tensorflow. In this section, we implement object detection with a pretrained model, Tensorflow Object Detection API. This model requires intallation of tensorflow. Further instruction of the installation can be referred to https://github.com/tensorflow/models/tree/master/research/object_detection. This model can detect and categorize object, including person, bottle, cellphone, etc. However, cascada model would result better if only faces are showed on an image while this API model would result better if more parts of human body are showed.

1.2.1 Part 2.1 - Object Detection API with Tensorflow on Image

Similar to Part 1.1, we started off with object detection using image.

import tarfile

import tensorflow as tf

```
In [227]: # %load ../lib/tensorflowFn.py

def objectDetection(directory = '../data/test_image/tensorflow'):
    import cv2
    # coding: utf-8
    # # Object Detection Demo
    # Welcome to the object detection inference walkthrough! This notebook will walk
    # # Imports

import numpy as np
import os
import six.moves.urllib as urllib
import sys
```

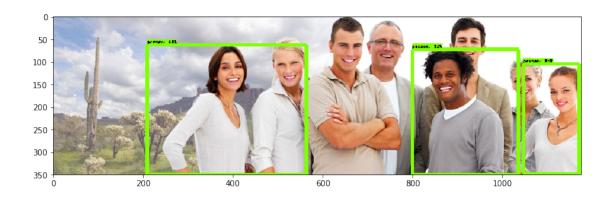
```
import zipfile
from collections import defaultdict
from io import StringIO
from matplotlib import pyplot as plt
from PIL import Image
# This is needed since the notebook is stored in the object_detection folder.
#cwd = os.getcwd()
os.chdir('../lib/object_detection')
#sys.path.append("..")
from object_detection.utils import ops as utils_ops
if tf.__version__ < '1.4.0':
  raise ImportError('Please upgrade your tensorflow installation to v1.4.* or late
# ## Env setup
# This is needed to display the images.
get_ipython().magic('matplotlib inline')
# ## Object detection imports
# Here are the imports from the object detection module.
from utils import label_map_util
from utils import visualization_utils as vis_util
# # Model preparation
# ## Variables
# Any model exported using the `export_inference_graph.py` tool can be loaded here
# By default we use an "SSD with Mobilenet" model here. See the [detection model 2
# What model to download.
MODEL_NAME = 'ssd_mobilenet_v1_coco_2017_11_17'
MODEL_FILE = MODEL_NAME + '.tar.gz'
DOWNLOAD_BASE = 'http://download.tensorflow.org/models/object_detection/'
# Path to frozen detection graph. This is the actual model that is used for the ob-
```

```
PATH_TO_CKPT = MODEL_NAME + '/frozen_inference_graph.pb'
# List of the strings that is used to add correct label for each box.
PATH_TO_LABELS = os.path.join('data', 'mscoco_label_map.pbtxt')
NUM_CLASSES = 90
# ## Download Model
if False:
    opener = urllib.request.URLopener()
    opener.retrieve(DOWNLOAD_BASE + MODEL_FILE, MODEL_FILE)
    tar_file = tarfile.open(MODEL_FILE)
    for file in tar_file.getmembers():
      file_name = os.path.basename(file.name)
      if 'frozen_inference_graph.pb' in file_name:
        tar_file.extract(file, os.getcwd())
# ## Load a (frozen) Tensorflow model into memory.
detection_graph = tf.Graph()
with detection_graph.as_default():
  od_graph_def = tf.GraphDef()
  with tf.gfile.GFile(PATH_TO_CKPT, 'rb') as fid:
    serialized_graph = fid.read()
    od_graph_def.ParseFromString(serialized_graph)
    tf.import_graph_def(od_graph_def, name='')
# ## Loading label map
# Label maps map indices to category names, so that when our convolution network p
label_map = label_map_util.load_labelmap(PATH_TO_LABELS)
categories = label_map_util.convert_label_map_to_categories(label_map, max_num_cla
category_index = label_map_util.create_category_index(categories)
# ## Helper code
def load_image_into_numpy_array(image):
  (im_width, im_height) = image.size
  return np.array(image.getdata()).reshape(
```

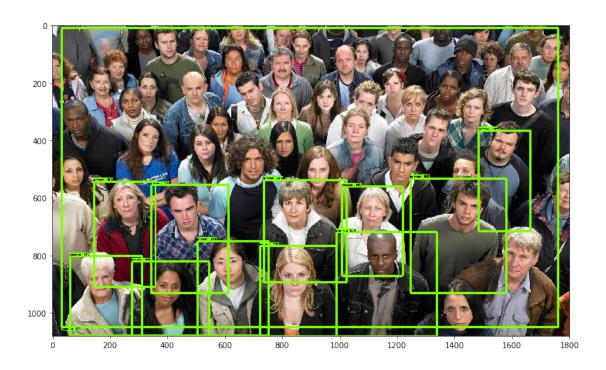
(im_height, im_width, 3)).astype(np.uint8)

```
# # Detection
# For the sake of simplicity we will use only 2 images:
# image1.jpg
# image2.jpg
# If you want to test the code with your images, just add path to the images to th
PATH_TO_TEST_IMAGES_DIR = directory
TEST_IMAGES_NAMES = os.listdir(directory)
TEST_IMAGE_PATHS = [os.path.join(PATH_TO_TEST_IMAGES_DIR, TEST_IMAGES_NAMES[i]) for
# Size, in inches, of the output images.
IMAGE\_SIZE = (12, 8)
def run_inference_for_single_image(image, graph):
  with graph.as_default():
    with tf.Session() as sess:
      # Get handles to input and output tensors
      ops = tf.get_default_graph().get_operations()
      all_tensor_names = {output.name for op in ops for output in op.outputs}
      tensor_dict = {}
      for key in [
          'num_detections', 'detection_boxes', 'detection_scores',
          'detection_classes', 'detection_masks'
      ]:
        tensor_name = key + ':0'
        if tensor_name in all_tensor_names:
          tensor_dict[key] = tf.get_default_graph().get_tensor_by_name(
              tensor_name)
      if 'detection_masks' in tensor_dict:
        # The following processing is only for single image
        detection_boxes = tf.squeeze(tensor_dict['detection_boxes'], [0])
        detection_masks = tf.squeeze(tensor_dict['detection_masks'], [0])
        # Reframe is required to translate mask from box coordinates to image coor
        real_num_detection = tf.cast(tensor_dict['num_detections'][0], tf.int32)
        detection_boxes = tf.slice(detection_boxes, [0, 0], [real_num_detection, -
        detection_masks = tf.slice(detection_masks, [0, 0, 0], [real_num_detection
        detection_masks_reframed = utils_ops.reframe_box_masks_to_image_masks(
            detection_masks, detection_boxes, image.shape[0], image.shape[1])
        detection_masks_reframed = tf.cast(
            tf.greater(detection_masks_reframed, 0.5), tf.uint8)
        # Follow the convention by adding back the batch dimension
        tensor_dict['detection_masks'] = tf.expand_dims(
            detection_masks_reframed, 0)
      image_tensor = tf.get_default_graph().get_tensor_by_name('image_tensor:0')
```

```
# Run inference
                    output_dict = sess.run(tensor_dict,
                                           feed_dict={image_tensor: np.expand_dims(image, 0)})
                    # all outputs are float32 numpy arrays, so convert types as appropriate
                    output_dict['num_detections'] = int(output_dict['num_detections'][0])
                    output_dict['detection_classes'] = output_dict[
                        'detection_classes'][0].astype(np.uint8)
                    output_dict['detection_boxes'] = output_dict['detection_boxes'][0]
                    output_dict['detection_scores'] = output_dict['detection_scores'][0]
                    if 'detection_masks' in output_dict:
                      output_dict['detection_masks'] = output_dict['detection_masks'][0]
                return output_dict
              for image_path in TEST_IMAGE_PATHS:
                image = Image.open(image_path)
                # the array based representation of the image will be used later in order to pre
                # result image with boxes and labels on it.
                image_np = load_image_into_numpy_array(image)
                # Expand dimensions since the model expects images to have shape: [1, None, None
                image_np_expanded = np.expand_dims(image_np, axis=0)
                # Actual detection.
                output_dict = run_inference_for_single_image(image_np, detection_graph)
                # Visualization of the results of a detection.
                vis_util.visualize_boxes_and_labels_on_image_array(
                    image_np,
                    output_dict['detection_boxes'],
                    output_dict['detection_classes'],
                    output_dict['detection_scores'],
                    category_index,
                    instance_masks=output_dict.get('detection_masks'),
                    use_normalized_coordinates=True,
                    line_thickness=8)
                plt.figure(figsize=IMAGE_SIZE)
                plt.imshow(image_np)
              os.chdir('../../doc')
In [243]: #os.chdir('../../doc')
          #os.getcwd()
          objectDetection('../../data/test_image/tensorflow')
```







1.2.2 Part 2.2 - Real Time Object Detection API with Tensorflow using WebCam

Part 2.1 resulted in highly accurate result detecting most of the objects into their corresponding categories. We further improved the model by implementing the model using WebCam as an input. The result was again very accurate.

In [230]: # %load ../lib/tensorflowFnVideo.py

```
def objectDetectionCap():
    import cv2
    import numpy as np
    import os
    import six.moves.urllib as urllib
    import sys
    import tarfile
    import tensorflow as tf
    import zipfile

from collections import defaultdict
    from io import StringIO
    from matplotlib import pyplot as plt
    from PIL import Image
```

This is needed since the notebook is stored in the object_detection folder.

```
#cwd = os.getcwd()
os.chdir('../lib/object_detection')
#sys.path.append("..")
from object_detection.utils import ops as utils_ops
if tf.__version__ < '1.4.0':
  raise ImportError('Please upgrade your tensorflow installation to v1.4.* or late
# ## Env setup
# This is needed to display the images.
get_ipython().magic('matplotlib inline')
# ## Object detection imports
# Here are the imports from the object detection module.
from utils import label_map_util
from utils import visualization_utils as vis_util
# # Model preparation
# ## Variables
# Any model exported using the `export_inference_graph.py` tool can be loaded here
# By default we use an "SSD with Mobilenet" model here. See the [detection model 2
# What model to download.
MODEL_NAME = 'ssd_mobilenet_v1_coco_2017_11_17'
MODEL_FILE = MODEL_NAME + '.tar.gz'
DOWNLOAD_BASE = 'http://download.tensorflow.org/models/object_detection/'
# Path to frozen detection graph. This is the actual model that is used for the ob-
PATH_TO_CKPT = MODEL_NAME + '/frozen_inference_graph.pb'
# List of the strings that is used to add correct label for each box.
PATH_TO_LABELS = os.path.join('data', 'mscoco_label_map.pbtxt')
NUM_CLASSES = 90
```

```
# ## Download Model
#
    opener = urllib.request.URLopener()
#
    opener.retrieve(DOWNLOAD_BASE + MODEL_FILE, MODEL_FILE)
    tar_file = tarfile.open(MODEL_FILE)
#
    for file in tar_file.getmembers():
      file_name = os.path.basename(file.name)
      if 'frozen_inference_graph.pb' in file_name:
        tar_file.extract(file, os.getcwd())
   # ## Load a (frozen) Tensorflow model into memory.
   detection_graph = tf.Graph()
   with detection_graph.as_default():
     od_graph_def = tf.GraphDef()
     with tf.gfile.GFile(PATH_TO_CKPT, 'rb') as fid:
       serialized_graph = fid.read()
       od_graph_def.ParseFromString(serialized_graph)
       tf.import_graph_def(od_graph_def, name='')
   # ## Loading label map
    # Label maps map indices to category names, so that when our convolution network p
   label_map = label_map_util.load_labelmap(PATH_TO_LABELS)
   categories = label_map_util.convert_label_map_to_categories(label_map, max_num_cla
   category_index = label_map_util.create_category_index(categories)
   cap = cv2.VideoCapture(0)
   with detection_graph.as_default():
     with tf.Session(graph=detection_graph) as sess:
      ret = True
      while (ret):
         ret,image_np = cap.read()
         image_np_expanded = np.expand_dims(image_np, axis=0)
         image_tensor = detection_graph.get_tensor_by_name('image_tensor:0')
         boxes = detection_graph.get_tensor_by_name('detection_boxes:0')
         scores = detection_graph.get_tensor_by_name('detection_scores:0')
         classes = detection_graph.get_tensor_by_name('detection_classes:0')
         num_detections = detection_graph.get_tensor_by_name('num_detections:0')
          (boxes, scores, classes, num_detections) = sess.run(
                  [boxes, scores, classes, num_detections],
                  feed_dict={image_tensor: image_np_expanded})
         vis_util.visualize_boxes_and_labels_on_image_array(
```

```
image_np,
                        np.squeeze(boxes),
                        np.squeeze(classes).astype(np.int32),
                        np.squeeze(scores),
                        category_index,
                        use_normalized_coordinates=True,
                        line_thickness=8)
                    cv2.imshow('image', cv2.resize(image_np,(1280,800)))
                    if cv2.waitKey(25) & OxFF == ord('q'):
                        cv2.destroyAllWindows()
                        cap.release()
                        break
              os.chdir('../../doc')
In [231]: '''
          Press Q to close the camera
          111
          objectDetectionCap()
In [246]:
Out[246]: '/Users/zailchen/Documents/RWorkplace/ADS_Project/Spring2018-Project5-grp_9/doc'
```