Main ## 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.

Part 1.1 - Face Detection without Rotation on Image

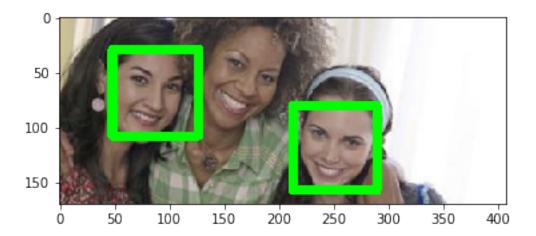
We started off the project with face detection on an image.

```
In [78]:
```

```
# %load /Users/zailchen/Documents/RWorkplace/ADS Project/Spring2018-Project5-grp 9/
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('/Users/zailchen/Documents/RWorkplace/ADS ]
    # loading test image
    img = cv2.imread('/Users/zailchen/Documents/RWorkplace/ADS Project/Spring2018-Pi
    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)
    plt.show()
```

```
In [79]:
```

```
counting_face()
```



Part 1.2 - Face Detection with Rotation

dof rotato point (poc ima angle)

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 [80]:
# %load /Users/zailchen/Documents/RWorkplace/ADS_Project/Spring2018-Project5-grp_9/
#!/usr/bin/env python3
0.00
Created on Sat Apr 21 20:24:20 2018
@author: zailchen
def face dectect image(directory = '../data/test image/cascade/', scaleFactor = 1.3|
    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-
    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 LINE
        return result
```

```
der rocace_porne(pos, rmg, angre).
    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('/Users/zailchen/Desktop/Project5/haarcascate
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])
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)
    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(]
        cv2.putText(img, "Number of faces detected: " + str(faces.shape[0]), (0)
    RGB img = cv2.cvtColor(img, cv2.COLOR BGR2RGB)
    plt.imshow(RGB_img)
    plt.show()
    cv2.imwrite('../output/processed {}'.format(TEST IMAGES NAMES[i+1]),img)
    i +=1
```

In [83]:

face_dectect_image('/Users/zailchen/Documents/RWorkplace/ADS_Project/Spring2018-Proj



In []:

200

400

600

800

1000

1200

1400

1600

1000

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 [87]:

```
# %load /Users/zailchen/Documents/RWorkplace/ADS Project/Spring2018-Project5-grp 9/1
#!/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 LINE
        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('/Users/zailchen/Documents/RWorkplace/ADS ]
    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)
```

Part 2 - Object Detection API using Tensorflow

In []:

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.

Part 2.1 - Object Detection API with Tensorflow on Image

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

```
In [89]:

# %load /Users/zailchen/Documents/RWorkplace/ADS_Project/Spring2018-Project5-grp_9/.

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 wa.
```

```
# # Imports
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(cwd)
sys.path.append("../lib")
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 la
# ## 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 he
# By default we use an "SSD with Mobilenet" model here. See the [detection mode.
# 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
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
label map = label map util.load labelmap(PATH TO LABELS)
categories = label map util.convert label map to categories(label map, max num (
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
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])
# 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 co
        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_detect:
        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[
```

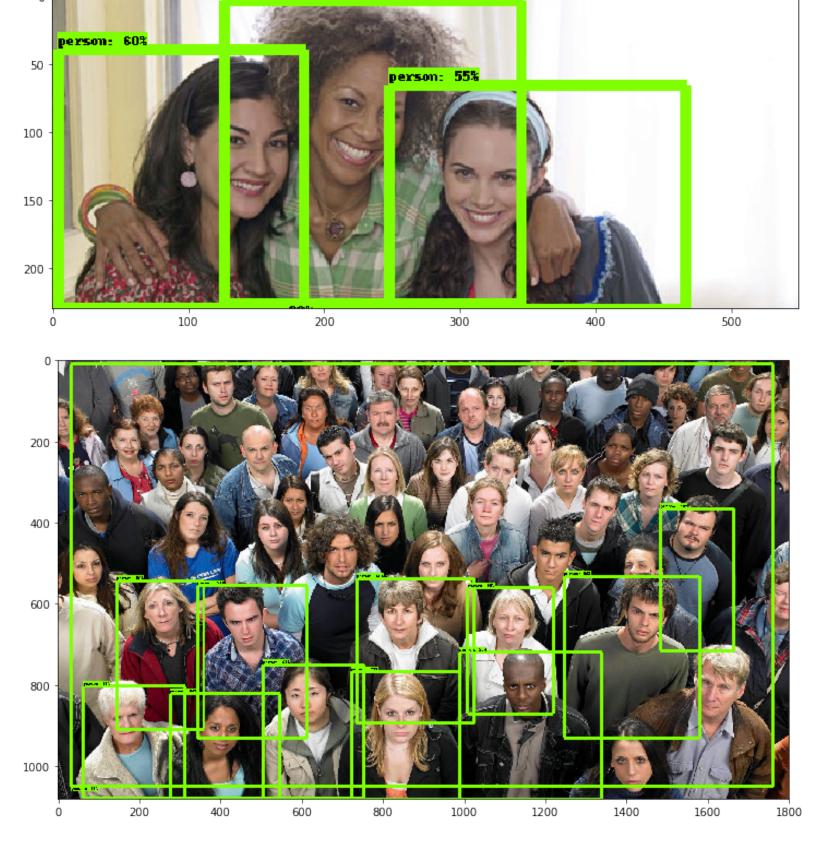
```
output dict['detection boxes'] = output dict['detection boxes'][0]
      output dict['detection scores'] = output dict['detection scores'][0]
      if 'detection masks' in output dict:
        output dict['detection masks'] = output dict['detection masks'][0]
  return output dict
for image_path in TEST_IMAGE_PATHS:
  image = Image.open(image path)
  # the array based representation of the image will be used later in order to |
  # 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, No
  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)
```

'detection_classes'][0].astype(np.uint8)

In [90]:

objectDetection(directory = '/Users/zailchen/Documents/RWorkplace/ADS_Project/Spring





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 [93]:
```

```
# %load /Users/zailchen/Documents/RWorkplace/ADS_Project/Spring2018-Project5-grp_9/.

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(cwd)
sys.path.append("../lib")
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 la
# ## 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 he
# By default we use an "SSD with Mobilenet" model here. See the [detection mode.
# 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
DAME TO CEDE - MODEL NAME + '/frogon informed graph ph'
```

```
# 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
    label map = label map util.load labelmap(PATH TO LABELS)
    categories = label map util.convert label map to categories(label map, max num (
    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],
                  food digt-(image tengent image no 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) & 0xFF == ord('q'):
    cv2.destroyAllWindows()
    cap.release()
    break
```

In [94]:

```
objectDetectionCap()
```