Haar Cascade, Face Detection

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0.1 Face detection using OpenCV

One older (from around 2001), but still popular scheme for face detection is a Haar cascade classifier; these classifiers in the OpenCV library and use feature-based classification cascades that learn to isolate and detect faces in an image. You can read the original paper proposing this approach here.

Let's see how face detection works on an exampe in this notebook.



To use a face detector, we'll first convert the image from color to grayscale. For face detection this is perfectly fine to do as there is plenty non-color specific structure in the human face for our detector to learn on.



Next we load in the fully trained architecture of the face detector, found in the file detector_architectures/ haarcascade_frontalface_default.xml,and use it on our image to find faces!

A note on parameters

How many faces are detected is determined by the function, detectMultiScale which aims to detect faces of varying sizes. The inputs to this function are: (image, scaleFactor, minNeighbors); you will often detect more faces with a smaller scaleFactor, and lower value for minNeighbors, but raising these values often produces better matches. Modify these values depending on your input image.

The output of the classifier is an array of detections; coordinates that define the dimensions of a bounding box around each face. Note that this always outputs a bounding box that is square in dimension.

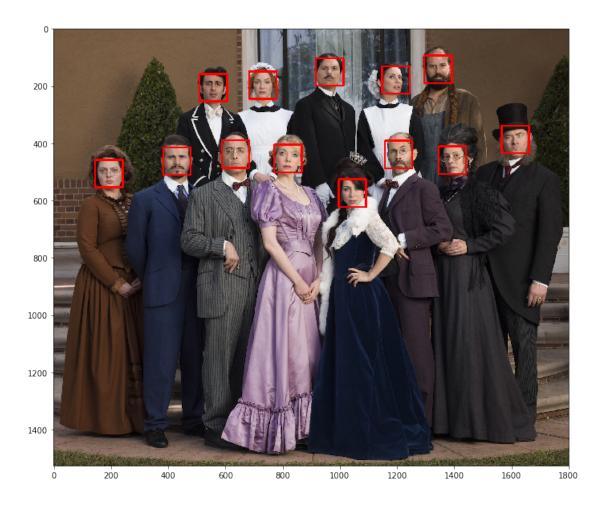
```
In [52]: # print out the detections found
        print ('We found ' + str(len(faces)) + ' faces in this image')
        print ("Their coordinates and lengths/widths are as follows")
        print ('======"')
        print (faces)
We found 13 faces in this image
Their coordinates and lengths/widths are as follows
_____
ΓΓ1295
        94
            96
                 961
[ 917 103
            96
                 96]
[1148 131
            96
                 96]
[ 683 149
            96
                 967
Γ 510 158
            96
                 967
[1565 339
            96
                 96]
[ 588 390
            96
                 96]
[1157 391
                 96]
            96
[ 771 405
            96
                 96]
「 383 414
            96
                 961
[1345 411
            96
                 96]
[ 146 458
            96
                 96]
[ 996 526
                 96]]
            96
```

Let's plot the corresponding detection boxes on our original image to see how well we've done.

```
In [53]: img_with_detections = np.copy(image)  # make a copy of the original image to plot rect

# loop over our detections and draw their corresponding boxes on top of our original im
for (x,y,w,h) in faces:
    # draw next detection as a red rectangle on top of the original image.
    # Note: the fourth element (255,0,0) determines the color of the rectangle,
    # and the final argument (here set to 5) determines the width of the drawn rectangle
    cv2.rectangle(img_with_detections,(x,y),(x+w,y+h),(255,0,0),5)

# display the result
plt.figure(figsize=(20,10))
plt.imshow(img_with_detections)
Out [53]: <matplotlib.image.AxesImage at Ox7f0ec1050ac8>
```



In []: