Botzer_AI879_HW_Q1_Week9

March 2, 2024

```
[]: # Author: Brandon Botzer
# Date: 2/21/2024
# Class: Penn State - AI 879
```

Problem:

Build your own face recognition classification model, using the following steps:

- 1) face_pos.mat file was build using images of two classes, one containing faces (positive) and another not containing them (negative), see the zip file data.zipDownload data.zip
- 2) Build a cascade detection model and test it on two new images: nature_image.jpg and face_test.jpg
- 3) VOID: Build a ACF detection model and test it on the two new images: (same as above two) Unfortunately, there is no library implement ACF detector except MATLAB source codeLinks to an external site., skip this task if you use python.

Compare the results of two models and discuss.

```
[]: # Imports for functions

from PIL import Image

import skimage as ski
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import patches
import cv2 as cv

# Showing the results of scipy computations
import pandas as pd

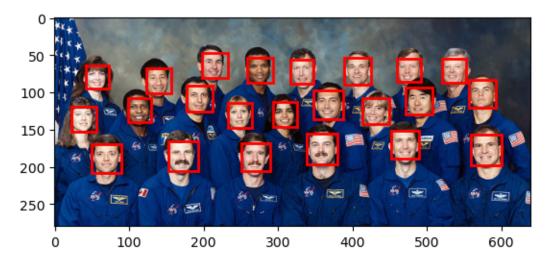
cv.__version__, ski.__version__
```

```
[]: ('4.8.1', '0.22.0')
```

```
[]: # I will at first attempt to do this with a pre-trained file from skimage pretrained_file = ski.data.lbp_frontal_face_cascade_filename()
```

```
[]: # Create the Cascade detector object
     cascade_detector = ski.feature.Cascade(pretrained_file)
[]: # Load in the image with and without faces
     pos_im = ski.io.imread('face_test.jpg')
     pos_im_cv = cv.imread('face_test.jpg')
     neg im = ski.io.imread('nature image.jpg')
     neg_im_cv = cv.imread('nature_image.jpg')
[]: # I had to tinker with the min_size and the max_size of the search window for a_{\sqcup}
     ⇒bit to find the faces
     # Use the cascade detector to locate faces
     face_detected = cascade_detector.detect_multi_scale(img=pos_im,
                                                          scale_factor=1.2,
                                                          step_ratio=1,
                                                          min_size=(20,20),
                                                          max size=(200, 200))
     111
     face_detected is in the form of:
     list of dicts
     Dict have form
     \{'r': int,
     'c': int,
     'width': int,
     'height': int},
     where 'r' represents row position of top left corner of detected window,
     'c' - col position,
     'width' - width of detected window,
     'height' - height of detected window.
     I I I
[]: '\nface_detected is in the form of:\n\nlist of dicts\nDict have form\n{'r':
     int,\n'c': int,\n'width': int,\n'height': int},\nwhere 'r' represents row
     position of top left corner of detected window, \n'c' - col position, \n'width' -
    width of detected window,\n'height' - height of detected window.\n'
[]: # Outputing the image with face mappings as done in:
     # https://scikit-image.org/docs/stable/auto_examples/applications/
      ⇒plot_face_detection.
      ⇔html#sphx-qlr-auto-examples-applications-plot-face-detection-py
     # Show the positive image
     plt.imshow(pos_im)
```

```
# Get the current axis
img_desc = plt.gca()
# Set the color mapping to gray
plt.set_cmap('gray')
# For each of the faces detected, make a red rectangle path
    The patch that is pulled each time is in the form of a dict from above.
    This is how the patch values 'c', 'r', etc. are used below
for patch in face_detected:
    # Add the patch to the image
    img_desc.add_patch(
        patches.Rectangle(
            (patch['c'], patch['r']),
            patch['width'],
            patch['height'],
            fill=False,
            color='r',
            linewidth=2
        )
    )
# Show me everything
plt.show()
```



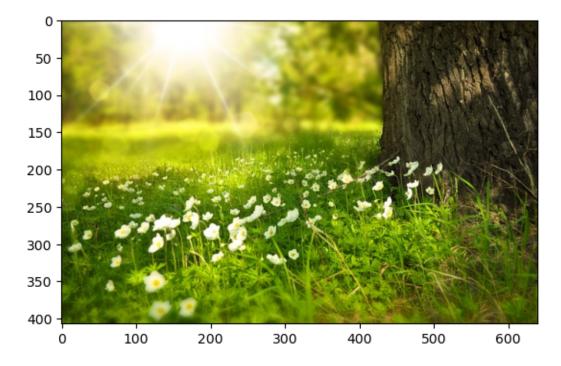
```
[]: # I had to tinker with the min_size and the max_size of the search window for authorized to find the faces

# Use the cascade detector to locate faces
```

```
face_detected_nature = cascade_detector.detect_multi_scale(img=neg_im,
                                                      scale_factor=1.2,
                                                      step_ratio=1,
                                                      min_size=(20,20),
                                                      max_size=(200, 200))
111
face_detected is in the form of:
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Dict have form
\{'r': int,
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where 'r' represents row position of top left corner of detected window,
'c' - col position,
'width' - width of detected window,
'height' - height of detected window.
111
```

[]: '\nface_detected is in the form of:\n\nlist of dicts\nDict have form\n{'r':
 int,\n'c': int,\n'width': int,\n'height': int},\nwhere 'r' represents row
 position of top left corner of detected window,\n'c' - col position,\n'width' width of detected window,\n'height' - height of detected window.\n'

```
[]: # Outputing the image with face mappings as done in:
     # https://scikit-image.org/docs/stable/auto_examples/applications/
      ⇔plot face detection.
      \Rightarrowhtml#sphx-glr-auto-examples-applications-plot-face-detection-py
     # Show the positive image
     plt.imshow(neg_im)
     # Get the current axis
     img_desc = plt.gca()
     # Set the color mapping to gray
     plt.set_cmap('gray')
     # For each of the faces detected, make a red rectangle path
     # The patch that is pulled each time is in the form of a dict from above.
     # This is how the patch values 'c', 'r', etc. are used below
     for patch in face_detected_nature:
         # Add the patch to the image
         img_desc.add_patch(
             patches.Rectangle(
```



From the pre-trained model from skimage, no faces were found in the negative nature image.

This time I have created an XML training file using Matlab using the Local Binary Patterns FeatureType setting.

This had to be done in Matlab prior to this code. The XML file claims to be compatable with OpenCV 4.5.2. Hopefully it will also work with OpenCV 4.8.1 as well as skimage 0.22.0. We shall see...

```
[]: # I had to switch the FeatureType to Local Binary Pattern.

# Using HAAR results in a kernal crash.

# HOG is not found and causes exceptions to be thrown by the openCV classifier

and invalid literals for integers for skimage.

trained_file2 = 'FaceModel_LBP.xml'
```

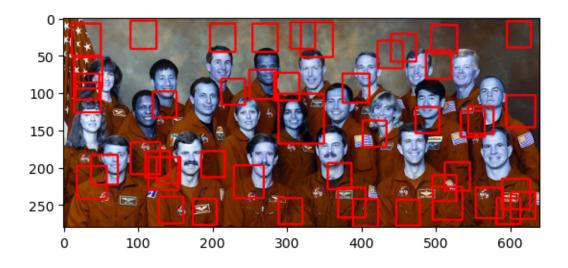
```
# Create the Cascade detector object
cascade_detector2_cv = cv.CascadeClassifier(trained_file2)
cascade_detector2_ski = ski.feature.Cascade(trained_file2)
```

I run a cascade detector out of both OpenCV as well as skimage to see if one performs better with the XML training file.

0.1 OpenCV

```
[]: # I had to tinker with the min_size and the max size of the search window for a_{\sqcup}
      ⇔bit to find the faces
     #OPENCV
     face_detected2_cv = cascade_detector2_cv.detectMultiScale(image=pos_im_cv,
                                                               scaleFactor=1.2,
                                                         minSize=(20,20),
                                                         maxSize=(200, 200))
[]: # Outputing the image with face mappings as done in:
     # https://scikit-image.org/docs/stable/auto_examples/applications/
      ⇔plot_face_detection.
      →html#sphx-qlr-auto-examples-applications-plot-face-detection-py
     # For each of the faces detected, make a red rectangle path
     # The patch that is pulled each time is in the form of a dict from above.
       This is how the patch values 'c', 'r', etc. are used below
     for (x, y, width, height) in face_detected2_cv:
         # Add the patch to the image
         cv.rectangle(pos_im_cv, (x,y), (x+width, y+height), (255,0,0),2)
     # Show me everything
     plt.imshow(pos_im_cv)
```

[]: <matplotlib.image.AxesImage at 0x272d46cf5f0>

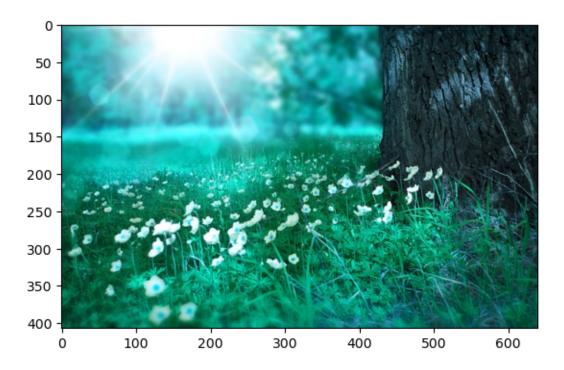


```
#OPENCV
     face_detected2_cv_nature = cascade_detector2_cv.
      →detectMultiScale(image=neg_im_cv,
                                                                scaleFactor=1.2,
                                                          minSize=(20,20),
                                                          maxSize=(200, 200))
[]: # Outputing the image with face mappings as done in:
     # https://scikit-image.org/docs/stable/auto_examples/applications/
      ⇔plot face detection.
      \Rightarrowhtml#sphx-glr-auto-examples-applications-plot-face-detection-py
     # For each of the faces detected, make a red rectangle path
         The patch that is pulled each time is in the form of a dict from above.
        This is how the patch values 'c', 'r', etc. are used below
     for (x, y, width, height) in face_detected2_cv_nature:
         # Add the patch to the image
         cv.rectangle(pos_im_cv, (x,y), (x+width, y+height), (255,0,0),2)
     # Show me everything
     plt.imshow(neg_im_cv)
```

[]: # I had to tinker with the min_size and the max size of the search window for a_{\sqcup}

[]: <matplotlib.image.AxesImage at 0x272d47304d0>

⇔bit to find the faces



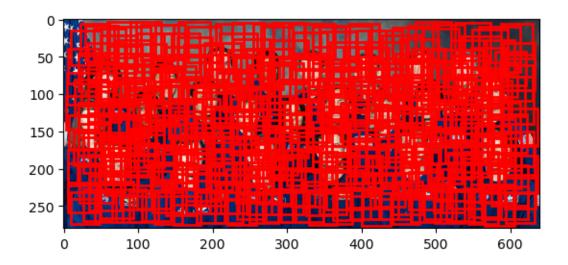
No faces were found in the nature image.

0.2 Skimage

```
[ ]: # # #SKIMAGE
     face_detected2_ski = cascade_detector2_ski.detect_multi_scale(img=pos_im_cv,
                                                          scale_factor=1.2,
                                                          step_ratio=1,
                                                          min_size=(20,20),
                                                          max_size=(200, 200))
     111
     face_detected is in the form of:
     list of dicts
     Dict have form
     {'r': int,
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     'width': int,
     'height': int},
     where 'r' represents row position of top left corner of detected window,
     'c' - col position,
     'width' - width of detected window,
     'height' - height of detected window.'''
```

[]: '\nface_detected is in the form of:\n\nlist of dicts\nDict have form\n{'r':
 int,\n'c': int,\n'width': int,\n'height': int},\nwhere 'r' represents row
 position of top left corner of detected window,\n'c' - col position,\n'width' width of detected window,\n'height' - height of detected window.'

```
[]: # Outputing the image with face mappings as done in:
     # https://scikit-image.org/docs/stable/auto_examples/applications/
      ⇒plot_face_detection.
     →html#sphx-qlr-auto-examples-applications-plot-face-detection-py
     # Show the positive image
     plt.imshow(pos_im)
     # Get the current axis
     img_desc = plt.gca()
     # Set the color mapping to gray
     plt.set_cmap('gray')
     # For each of the faces detected, make a red rectangle path
        The patch that is pulled each time is in the form of a dict from above.
         This is how the patch values 'c', 'r', etc. are used below
     for patch in face_detected2_ski:
         # Add the patch to the image
         img_desc.add_patch(
             patches.Rectangle(
                 (patch['c'], patch['r']),
                 patch['width'],
                 patch['height'],
                 fill=False,
                 color='r',
                 linewidth=2
             )
         )
     # Show me everything
     plt.show()
```



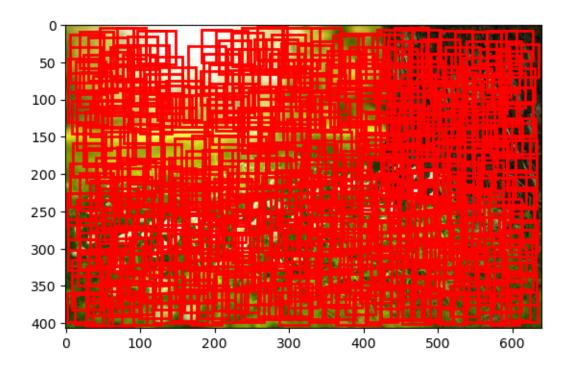
```
[ ]: # # #SKIMAGE
     face_detected2_ski_nature = cascade_detector2_ski.

detect_multi_scale(img=neg_im_cv,
                                                           scale_factor=1.2,
                                                           step_ratio=1,
                                                           min_size=(20,20),
                                                           max_size=(200, 200))
     , , ,
     face_detected is in the form of:
     list of dicts
     Dict have form
     \{'r': int,
     'c': int,
     'width': int,
     'height': int},
     where 'r' represents row position of top left corner of detected window,
     'c' - col position,
     'width' - width of detected window,
     'height' - height of detected window.'''
```

[]: '\nface_detected is in the form of:\n\nlist of dicts\nDict have form\n{'r':
 int,\n'c': int,\n'width': int,\n'height': int},\nwhere 'r' represents row
 position of top left corner of detected window,\n'c' - col position,\n'width' width of detected window,\n'height' - height of detected window.'

```
[]: # Outputing the image with face mappings as done in:
```

```
# https://scikit-image.org/docs/stable/auto_examples/applications/
 \hookrightarrow plot_face_detection.
 →html#sphx-qlr-auto-examples-applications-plot-face-detection-py
# Show the positive image
plt.imshow(neg_im)
# Get the current axis
img_desc = plt.gca()
# Set the color mapping to gray
plt.set_cmap('gray')
# For each of the faces detected, make a red rectangle path
  The patch that is pulled each time is in the form of a dict from above.
# This is how the patch values 'c', 'r', etc. are used below
for patch in face_detected2_ski_nature:
    # Add the patch to the image
    img_desc.add_patch(
        patches.Rectangle(
            (patch['c'], patch['r']),
            patch['width'],
            patch['height'],
            fill=False,
            color='r',
            linewidth=2
        )
    )
# Show me everything
plt.show()
```



The cascade detector for skimage is clearly not working well given the multitude of 'faces' it found in the positive image, and the stupendous amount of false positives it found in the negative image. This probably has to do with the XML file from Matlab being designed with OpenCV in mind.

Looking at the comparisions between OpenCV and skimage, the OpenCV cascade did a much better job with the training file provided from Matlab. The methods to train a XML file nativly using OpenCV have broken in the most recent releases. Skimage bases its training off of OpenCV items which would have hopefully allowed for the XML file to be used by skimage. Clerarly this was not the case. Matlab's output XML file is designed to be used with OpenCV 4.5.2 but seems to also hold with OpenCV version 4.8.1. This probably is why the OpenCV cascade provided better results.