University at Buffalo School of Engineering and Applied Sciences

CSE 666 - Biometrics Image Analysis Spring 2023

Assignment #1

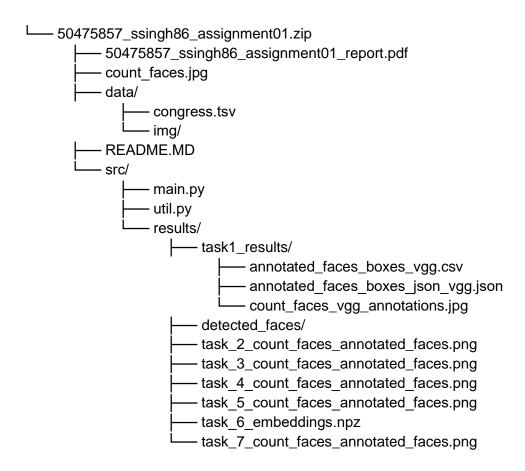
Prepared by

Sakshamdeep Singh - 50475857

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Submission Folder Structure



External Libraries Used:

	OpenCv	Deepface	Face-reconition
Task 1: Annotation			
Task 2: Face Detection	✓		
Task 3: Expression Analysis	✓	✓	
Task 4: Gender Analysis	✓	✓	
Task 5: Face Pose Estimation	✓		✓
Task 6: Feature Extraction	✓		✓
Task 7: Face Recognition	✓		✓

Task 1: Annotation

- For the manual annotation, the VGG Image Annotator was used to annotate images and save the corresponding bounding boxes.
- The VGG Image Annotator can be found at: (https://www.robots.ox.ac.uk/~vgg/software/via/)
- Number of faces annotated = 135.



Figure 1: Faces annotated using VGG Image Annotator

Task 2: Face Detection

- For face detection, OpenCV library's Haar cascades, which are pre-trained machine learning models used for detecting faces, were used.
- Once the faces are detected, we will proceed with the next steps.
- Number of faces detected = 116.



Figure 2: Faces detected using OpenCV Haar cascade.

Evaluation:

- Number of faces detected = 116
- Number of faces annotated = 135
- Accuracy of detecting faces = 116/135 * 100 = 85.92%
- Out of total 135 manually annotated faces, only 116 were identified. Some of the faces were missed due to the following reasons.
 - o Face is tilted on one side so face is not aligned vertically.





Person is looking sideways.







o Person is looking down.



o Occlusion





Task 3: Expression Analysis

- For this task, the deepface Python library was used, which internally uses VGG-Face as the default deep learning model.
- Out of 116 faces detected, each face was passed to the deepface API to obtain the emotion.
- The following <u>six</u> emotions were detected:

o Sad: 47







Neutral: 33





o Fear: 6



o Angry: 24





o Disgust: 1



o Happy: 5







Figure 3: Emotions detected using deepface.

Evaluation:

- Emotions: ['neutral', 'sad', 'angry', 'happy', 'fear', 'disgust']
- Counts: [33, 47, 24, 5, 6, 1]
- Deepface library does a pretty good job at evaluating the sentiments.
- Most of the sentiments, as shown above, were identified correctly.
- One error it made is <u>overclassifying 'sad' emotion</u>, where most of the people in the room seem to have a neutral expression.

Task 4: Gender Analysis

- For gender analysis, the deepface Python library was used, which internally uses
 VGG-Face as the default deep learning model.
- The model identified 107 faces as male and 9 faces as female.



Figure 4: Gender detection using deepface library.

Evaluation:

- Emotions: ['Man', 'Woman']
- Counts: [107, 9]
- Most of the genders, as shown above were identified correctly. However, a lot of women were misclassified as men.
- Total Number of Men in the picture = 88
- Total Number of Women in the picture = 28
- Accuracy of the model = 97/116 * 100 = 83.62%
- Confusion Matrix:

	Man (Ground Truth)	Woman (Ground Truth)
Man (Predicted)	88	19
Woman (Predicted)	19	9

Task 5: Face pose estimation

- For face pose estimation, the yaw angle will be calculated to detect whether the person is looking forward or sideways.
- The face recognition Python library was used to obtain facial landmarks.
- From the facial landmarks, we can obtain the nose bridge and chin to calculate the yaw angle.



Figure 5: Yaw angle detection using face_recognition library.

Evaluation:

- Model didn't seem to perform well using this approach.
- Probably the approach using nose bridge and chin to calculate the yaw angle is not robust enough.
- Adding more landmark features will help with the accuracy but in some of the faces, due to the small image size, fine landmarks are not detected properly.

Task 6: Feature Extraction

- For feature extraction, the face_recognition library was used to obtain the embeddings of each extracted face from the original image.
- Each extracted face was converted into an embedding of a 128-sized NumPy array.
- These embeddings/encodings were used in the next task for face recognition.

Task 7: Face Recognition

- For the given dataset, a dictionary of embeddings was created and stored in memory.
- For each Region of Interest (ROI), an embedding was created and matched with the list of known embedding stored in the above step.
- Euclidean distance was used as a similarity criterion.
- Matches are then stored in detected_faces directory with file name as the identified person's name.

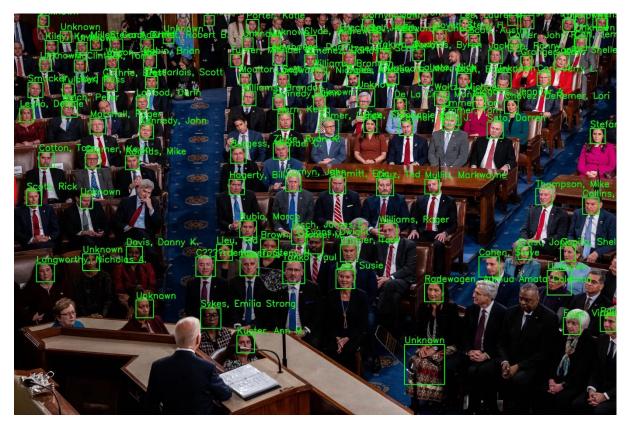


Figure 6: Faces recognized using Euclidean distance similarity function in face_recognition.

Evaluation:

- Number of Known faces: 95
- Number of Unknown faces: 21
- On crosschecking the 95 known faces manually,
 - Number of correctly identified faces = 67
 - Number of incorrectly identified faces = 28
 - Accuracy = 67/95 * 100 = 70.52%
 - o Accuracy is pretty good considering the amount of data we had.

References

- https://www.robots.ox.ac.uk/~vgg/software/via/
- https://github.com/serengil/deepface
- https://face-recognition.readthedocs.io/en/latest/
- http://dlib.net/python/index.html#dlib.get frontal face detector
- https://pypi.org/project/deepface/
- https://pypi.org/project/face-recognition/
- https://docs.opencv.org/4.x/d6/d00/tutorial_py_root.html
- https://docs.opencv.org/3.4/db/d28/tutorial cascade classifier.html
- https://en.wikipedia.org/wiki/Euclidean distance
- https://github.com/ageitgey/face recognition/tree/master/examples
- https://docs.python.org/3/