



Credit Hours System



Cairo University

Faculty of Engineering

Image Processing and Computer Vision

Project Report

Team #7

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Used Algorithms:

- Face Detection: Viola Jones algorithm: [Paper](#)
- Face Recognition: Eigenfaces algorithm: [Paper](#)

Experiment Results and Analysis

Face Detection:

- 98% accuracy on the Olivetti dataset
- Trained on 837 upfront-face images each of size 24x24 pixels

Points of Strength:

Very high accuracy and only one run over the dataset is required.

Points of weaknesses:

Training takes a lot of time. Also, it needs upfront-close face images dataset which is hard to find on the internet.

Dataset:

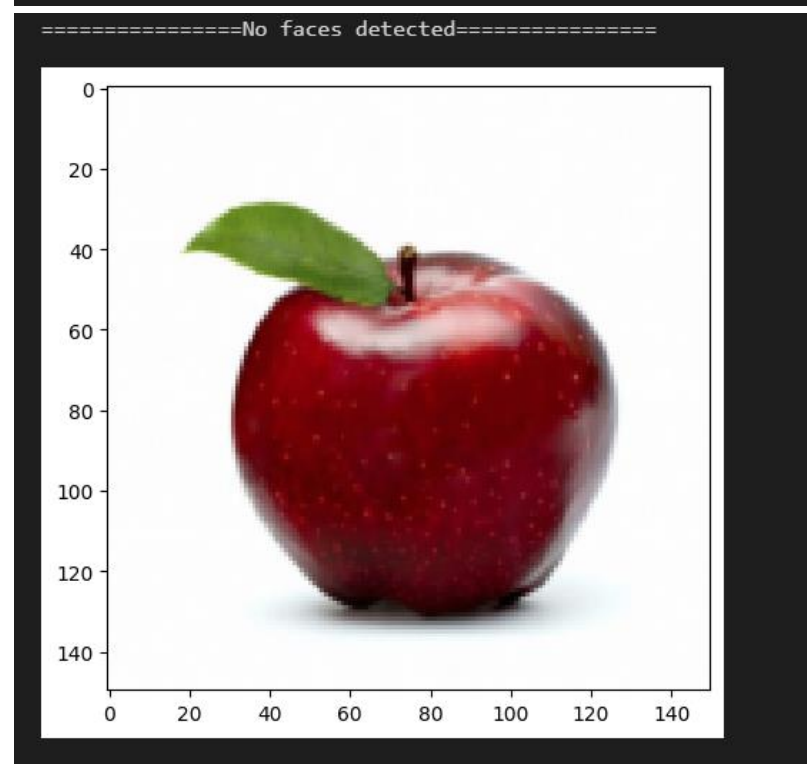
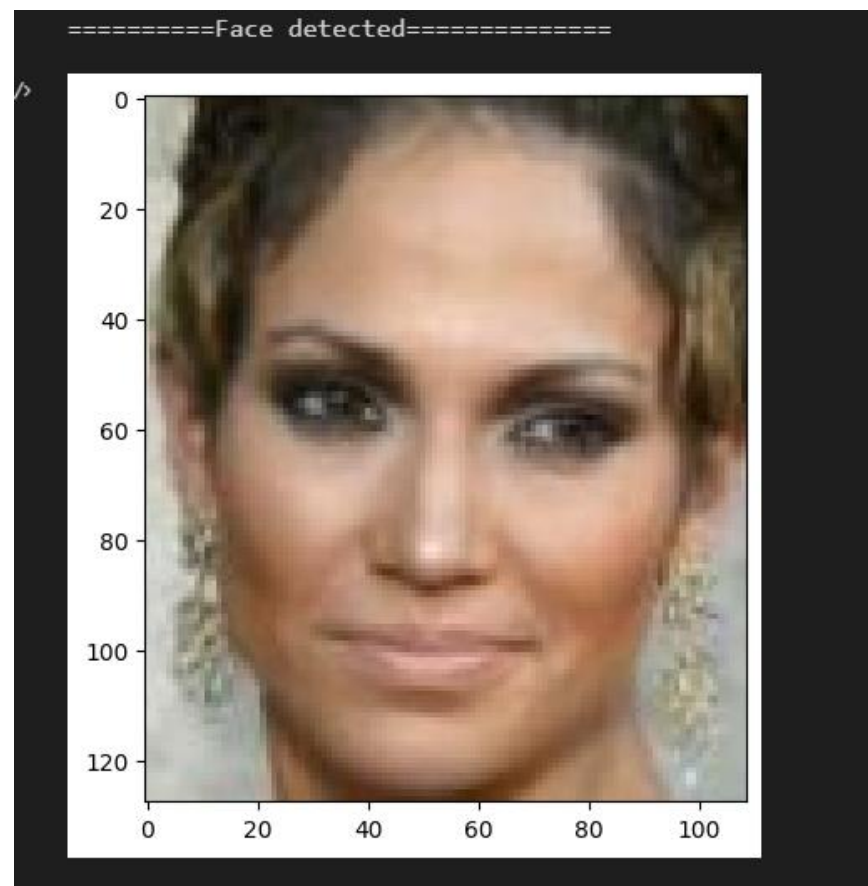
We used the Olivetti faces twice and manually added 37 faces to the training dataset for a total of 837 faces. The 37 faces were manually chosen from LFW dataset and manually cropped to match the required input.

Accuracy:

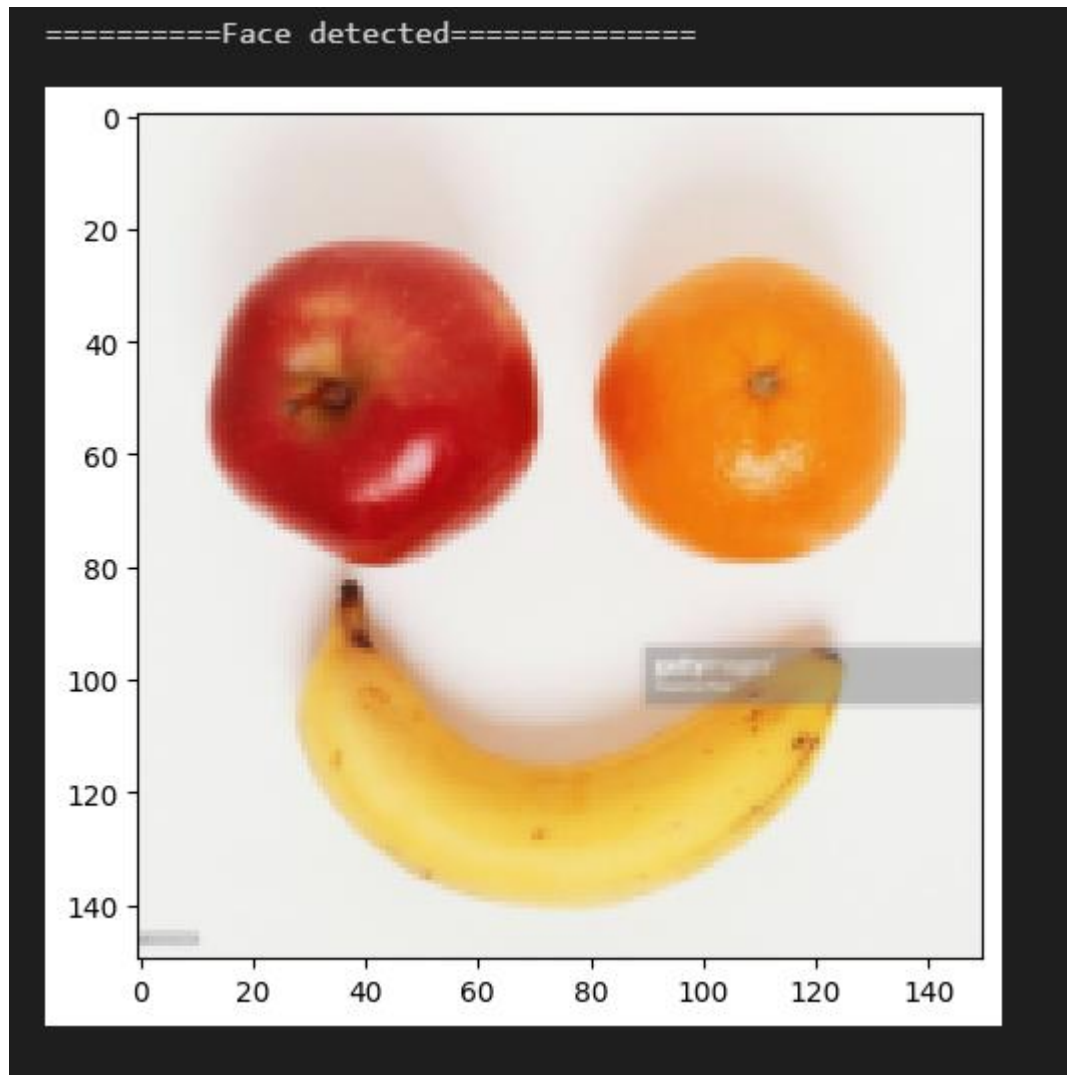
```
Test Dataset Shape: (160, 24, 24)
Predicted:  0
Target:  1
Predicted:  0
Target:  1
Predicted:  0
Target:  1
Accuracy:  98.125 %
Faces detected: 97 / 100 , Accuracy: 97.0%
Non Faces detected: 60 / 60 , Accuracy: 100.0%
```

$$\text{Accuracy} = \frac{\text{Number of correct predictions}}{\text{Total Number of test images}} * 100$$

Correct Example:



Incorrect Example:



Face Recognition:

- 97% accuracy on the Olivetti Dataset
- 400 images of 40 people
- Each image is 64×64 pixels
- 320 images are used for training
- 80 images are used for testing

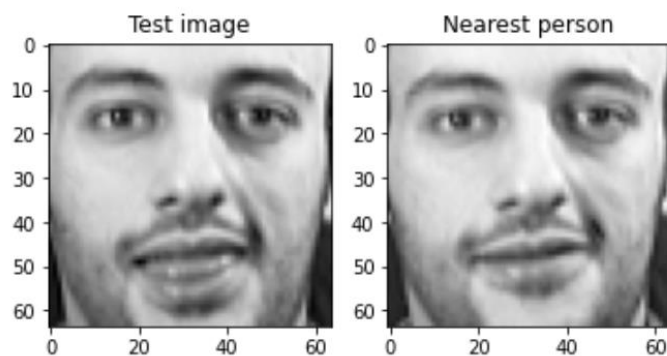
Example of two correct recognitions

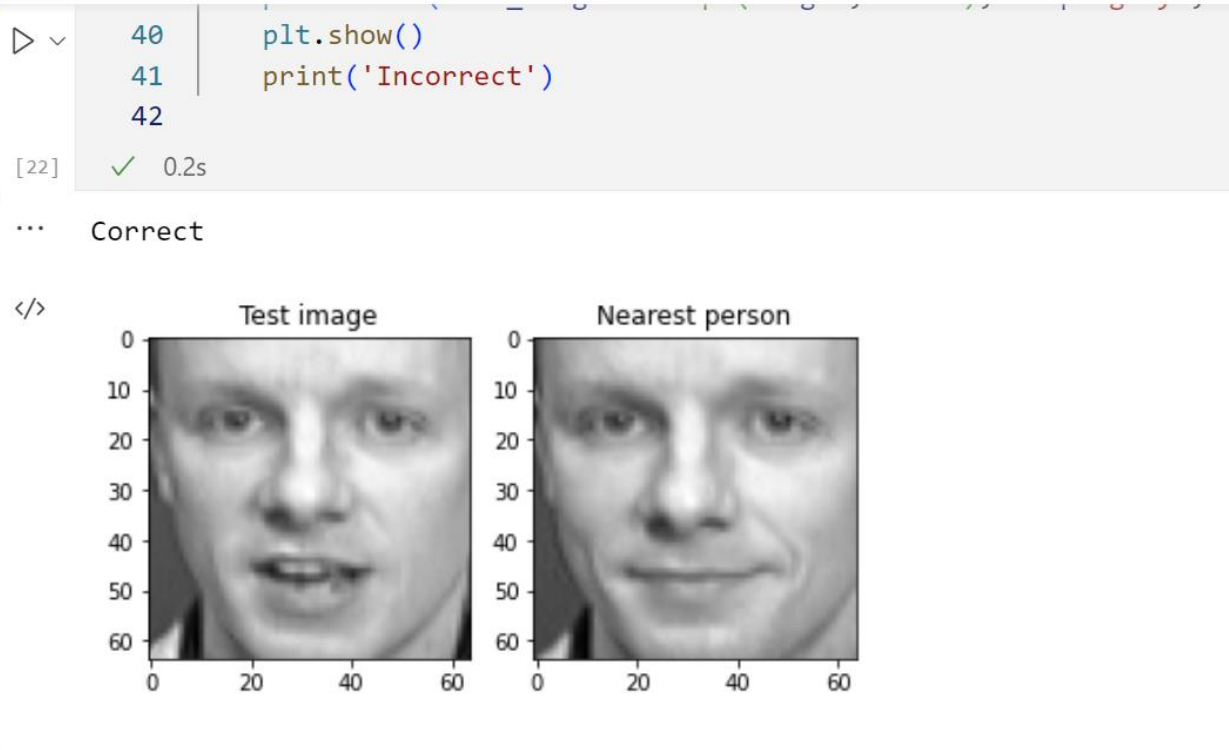
```
40 plt.imshow(test_image.reshape(height, width), cmap='gray', label = 'Te  
41 plt.show()  
42 plt.imshow(train_images[nearest_person,:,:], cmap='gray', label='Neare  
43 plt.show()  
44
```

[39] ✓ 0.2s

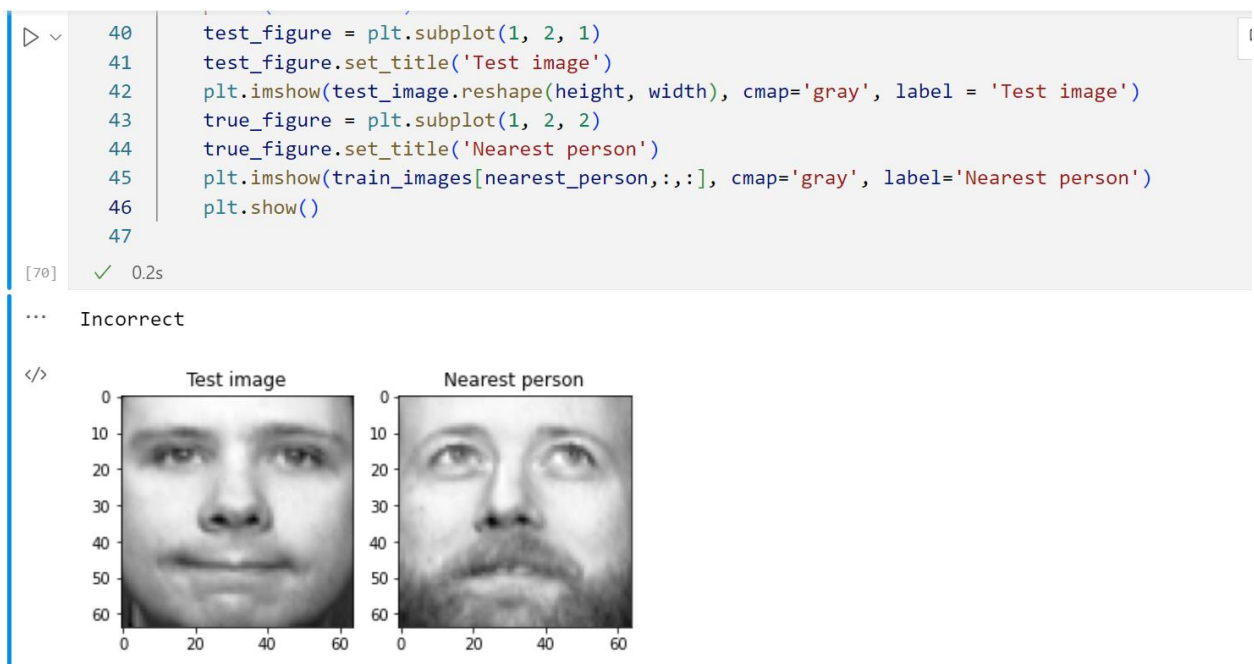
... Correct

</>





Example of an incorrect recognition



Accuracy:

```
33     threshold = float('inf')
34     if (min_distance < threshold):
35         if (y_train[nearest_person] == y_test[index]):
36             if (min_distance > max_correct_distances):
37                 max_correct_distances = min_distance
38             if (min_distance < min_correct_distances):
39                 min_correct_distances = min_distance
40             sum_correct += min_distance
41             count_correct += 1
42         else:
43             if (min_distance < min_incorrect_distances):
44                 min_incorrect_distances = min_distance
45             if (min_distance > max_incorrect_distances):
46                 max_incorrect_distances = min_distance
47             sum_incorrect += min_distance
48
49     print(f'Accuracy: {count_correct/trials*100}%')
```

[93] ✓ 2.3s

... Accuracy: 96.8%

Accuracy is calculated as $\frac{\text{number of correct predictions}}{\text{Total number of predictions}} \times 1000 \text{ trials}$

- Points of strength: The algorithm does not need much time to train because it sees the training data only once (1 epoch)
- Points of weakness: The images must be close ups of faces. The algorithm did not perform well for Datasets where faces were not the main element in the image (15% accuracy for the LFW Dataset).

This is expected of the Eigenfaces algorithm, because it does not try to extract certain features from the images that would make the difference between one face and the other. Rather, it makes the assumption that the faces distribution over the whole image space (which is all possible combinations of pixels \equiv all possible *width* \times *height* images) is not random. Based on this assumption, the algorithm aims to calculate the Eigenvectors (called Eigenfaces) that best describe the distribution of face images over the images space.

Work Division

Team Member	Task
Ahmed Mohamed Ismail	Face Detection
Moaz Mohamed Elsherbini	
Mostafa Ashraf Ahmed	Face Recognition
Nader Youhanna Adib	