



Project Report

Project Name: PCA Face Recognition

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Introduction:

Face recognition is a method of identifying or verifying the identity of an individual using their face. There are various algorithms that can do face recognition but their accuracy might vary. Here I am going to describe how we do face recognition using deep learning in this project. It is a technology in computer vision. In Face recognition / detection we locate and visualize the human faces in any digital image. A facial recognition system uses biometrics to map facial features from a photograph or video. It compares the information with a database of known faces to find a match.

Methodology:

PCA is a statistical approach used for reducing the number of variables in face recognition. In PCA, every image in the training set is represented as a linear combination of weighted eigenvectors called eigenfaces. These eigenvectors are obtained from covariance matrix of a training image set

- Standardize the data.
- Compute the covariance matrix of the features from the dataset.
- Perform eigen decomposition on the covariance matrix.
- Order the eigenvectors in decreasing order based on the magnitude of their corresponding eigen values.
- Determine k, the number of top principal components to select.
- Construct the projection matrix from the chosen number of top principal components.
- Compute the new k-dimensional feature space.

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Import Libraries

First, we need to import matplotlib libraries as well as numpy library.

```
[1] from matplotlib import pyplot as plt
    from matplotlib.image import imread
    import numpy as np
    import os
```

Data loading

```
[2] TRAIN_IMG_FOLDER = (r"/content/sample_data/Training images")
    TEST_IMG_FOLDER = (r"/content/sample_data/TEST")

    train_set_files = os.listdir(TRAIN_IMG_FOLDER)
    test_set_files = os.listdir(TEST_IMG_FOLDER)

    width = 128
    height = 128
```

Check: All data from 'train' is included in 'test'?

```
[3] train_id_file = set([f.split('_')[0] for f in train_set_files])
    test_id_file = set([f.split('_')[0] for f in test_set_files])
```

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```
[3] test_id_file = set([f.split('_')[0] for f in train_set_files])
[3] print(train_id_file <= test_id_file)
```

True

```
[4] print('Train Images:')

train_image_names = os.listdir(TRAIN_IMG_FOLDER)
training_tensor = np.ndarray(shape=(len(train_image_names), height*width), dtype=np.float64)

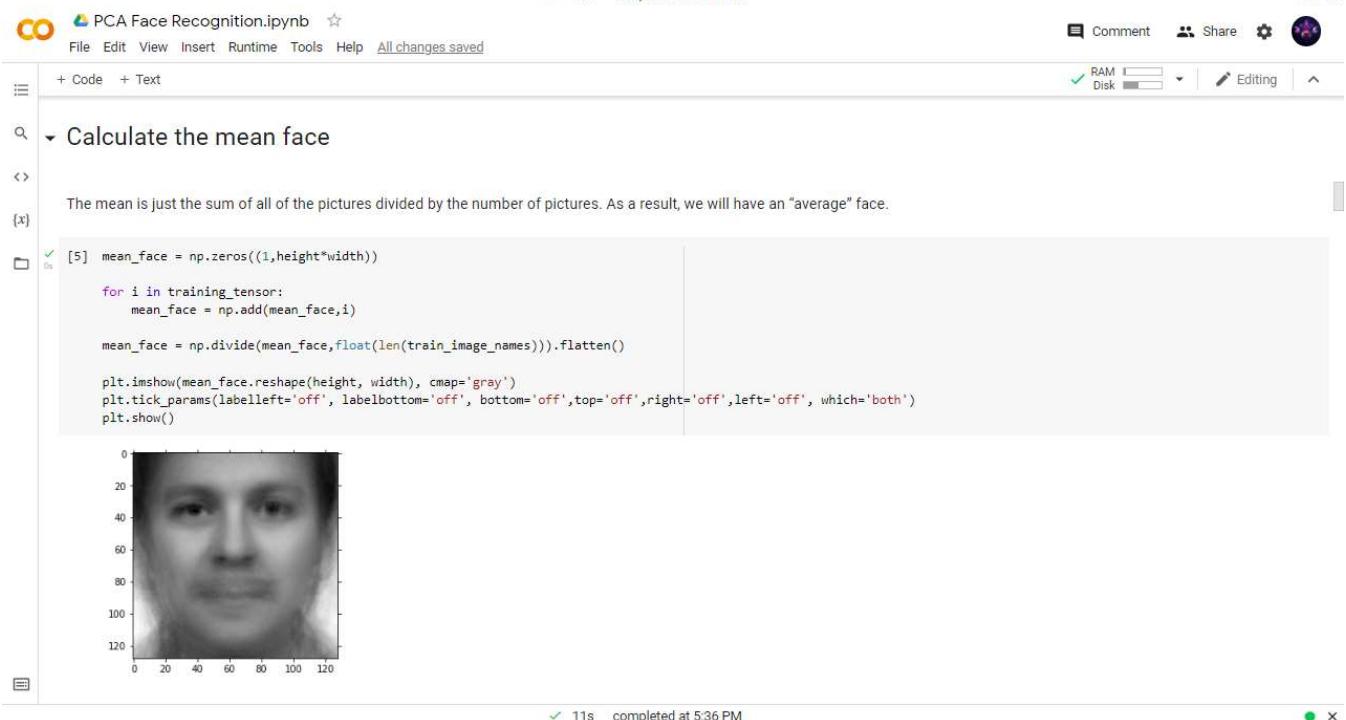
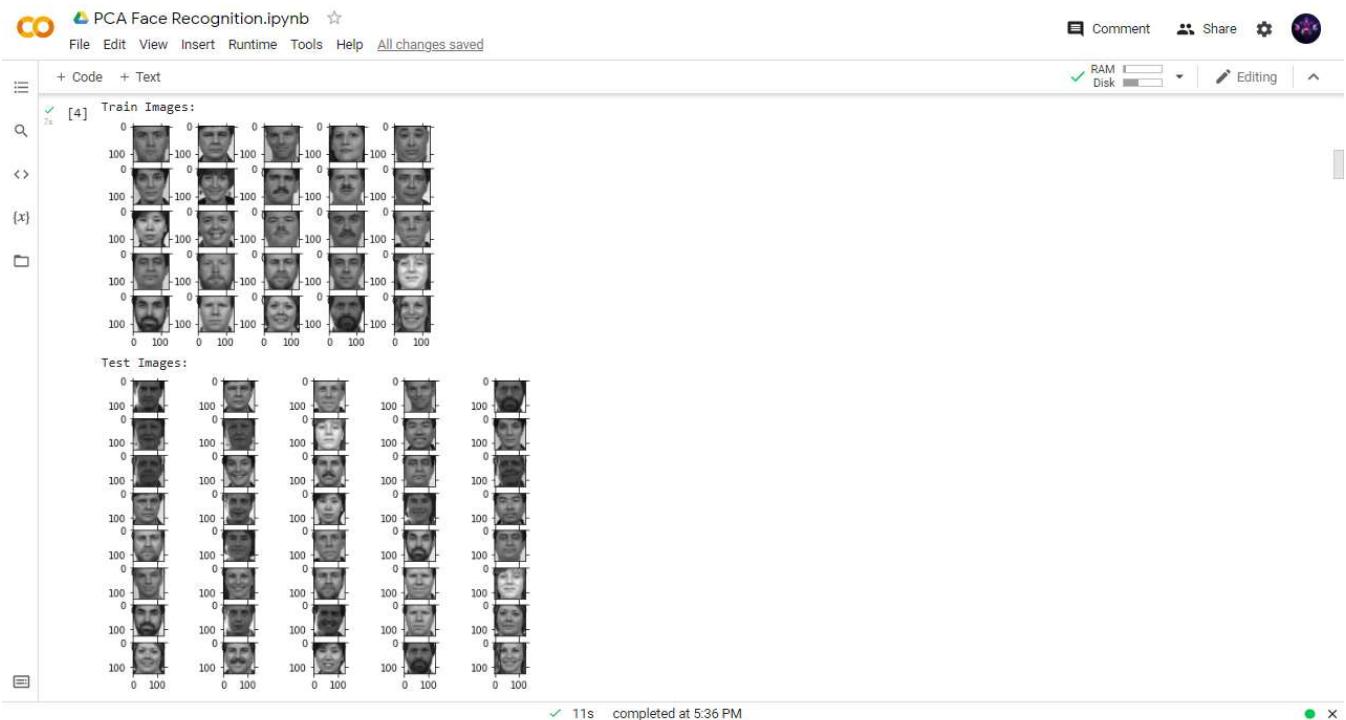
for i in range(len(train_image_names)):
    img = plt.imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[i]))
    training_tensor[i,:] = np.array(img, dtype='float64').flatten()
    plt.subplot(5,5,i+1)
    plt.imshow(img, cmap='gray')
    plt.tick_params(labelleft='off', bottom='off', top='off', right='off', left='off', which='both')
    plt.show()

print('Test Images:')
test_image_names = os.listdir(TEST_IMG_FOLDER)[#i for i in dataset_dir if i not in train_image_names]
testing_tensor = np.ndarray(shape=(len(test_image_names), height*width), dtype=np.float64)

for i in range(len(test_image_names)):
    img = plt.imread(os.path.join(TEST_IMG_FOLDER, test_image_names[i]))
    testing_tensor[i,:] = np.array(img, dtype='float64').flatten()
    plt.subplot(5,5,i+1)
    plt.imshow(img, cmap='gray')
    plt.subplots_adjust(right=1.2, top=1.2)
    plt.tick_params(labelleft='off', bottom='off', top='off', right='off', left='off', which='both')
    plt.show()
```

Train Images:

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Calculation of difference between training vector and mean vector

To normalize the training set, we just simply need to subtract for each picture in the training set the mean that was calculated in the previous step.

The reason why this is necessary is because we want to create a system that is able to represent any face. Therefore, we calculated the elements that all faces have in common (the mean). If we extract this average from the pictures, the features that distinguish each picture from the rest of the set are visible.

```
[6] normalised_training_tensor = np.ndarray(shape=(len(train_image_names), height*width))
for i in range(len(train_image_names)):
    normalised_training_tensor[i] = np.subtract(training_tensor[i],mean_face)
```

Display normalised faces

```
[7] for i in range(len(train_image_names)):
    img = normalised_training_tensor[i].reshape(height,width)
    plt.subplot(5,5,i+1)
    plt.imshow(img, cmap='gray')
    plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
plt.show()
```

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Convenience Matrix

The covariance represents how two variables change together. After the previous step, we have a set of images that have different features, so now we want to see how these features for each individual picture change in relation to the rest of the pictures.

```
[8] cov_matrix=np.cov(normalised_training_tensor)
cov_matrix = np.divide(cov_matrix,25.0)
print('Covariance Matrix Shape:', cov_matrix.shape)
#print('Covariance matrix of X: \n%s' %cov_matrix)
```

Covariance Matrix Shape: (25, 25)

Eigenvector of covariance

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From the covariance we can extract the eigenvectors. Fortunately, there is a function that helps us in this step. There is plenty of information in the internet about eigenvectors but the general idea is that eigenvectors are the vectors of the covariance that describe the direction of the data.

```
[x] [9] #eigenvalues and eigenvectors
eigenvalues, eigenvectors = np.linalg.eig(cov_matrix)
print('eigenvalues.shape: {} eigenvectors.shape: {}'.format(eigenvalues.shape, eigenvectors.shape))

eigenvalues.shape: (25,) eigenvectors.shape: (25, 25)

[10] eig_pairs = [(eigenvalues[index], eigenvectors[:,index]) for index in range(len(eigenvalues))]

# Sort the eigen pairs in descending order:
eig_pairs.sort(reverse=True)
eigenvalues_sort = [eig_pairs[index][0] for index in range(len(eigenvalues))]
eigenvectors_sort = [eig_pairs[index][1] for index in range(len(eigenvalues))]

sorted_ind = sorted(range(eigenvalues.shape[0]), key=lambda k: eigenvalues[k], reverse=True)

eigenvalues_sort = eigenvalues[sorted_ind]
eigenvectors_sort = eigenvectors[sorted_ind]
train_set_files_sort = np.array(train_set_files)[sorted_ind]
```

- Find cumulative variance of each principle component

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```
[11] var_comp_sum = np.cumsum(eigenvalues_sort)/sum(eigenvalues_sort)

# Show cumulative proportion of variance with respect to components
Cumulative proportion of variance explained vector: 0.28481989 0.41576165 0.49714907 0.55777059 0.60785024 0.65427472 0.69489721 0.73168527 0.7641529 0.7916646 0.81546473 0.8366985 0.85752859 0.87642433 0.89418977 0.91069532 0.92659352 0.94017913 0.95309028 0.96500642 0.97612862 0.98491526 0.99275274 1.
```

x-axis for number of principal components kept

```
[11] num_comp = range(1,len(eigenvalues_sort)+1)
plt.title('Cum. Prop. Variance Explain and Components Kept')
plt.xlabel('Principal Components')
plt.ylabel('Cum. Prop. Variance Explained')

plt.scatter(num_comp, var_comp_sum)
plt.show()
```

Cumulative proportion of variance explained vector:

Principal Components	Cum. Prop. Variance Explained
1	0.28481989
2	0.41576165
3	0.49714907
4	0.55777059
5	0.60785024
6	0.65427472
7	0.69489721
8	0.73168527
9	0.7641529
10	0.7916646
11	0.81546473
12	0.8366985
13	0.85752859
14	0.87642433
15	0.89418977
16	0.91069532
17	0.92659352
18	0.94017913
19	0.95309028
20	0.96500642
21	0.97612862
22	0.98491526
23	0.99275274
24	1.0

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Choose the necessary of principal components:

```
[12] reduced_data = np.array(eigvectors_sort[:25]).transpose()
reduced_data.shape
(x) (25, 25)

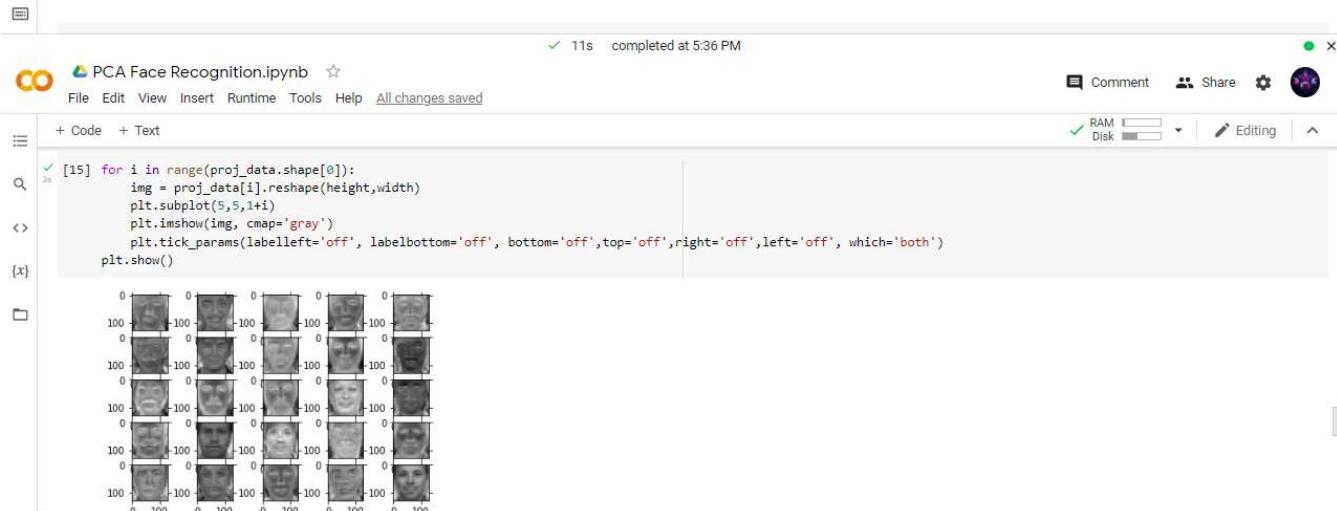
[13] print(training_tensor.transpose().shape, reduced_data.shape)
(16384, 25) (25, 25)
```

Calculate eigenfaces

Each eigenvector is multiplied by the whole normalized training set matrix and as a result, we will have the same amount of eigenfaces as images in our training set.

```
[14] proj_data = np.dot(training_tensor.transpose(),reduced_data)
proj_data = proj_data.transpose()
proj_data.shape
(25, 16384)
```

Plot eigen faces



Finding weights for each traning image

Each normalized face in the training set multiplies each eigenface. Consequently, there will be N set of weights with M elements (N = amount of pictures in the training set, M = number of eigenfaces).

```
[16] w = np.array([np.dot(proj_data,i) for i in normalised_training_tensor])
print(w.shape)
```

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Reconizing all test images

```
[x] [17] def recogniser(test_image_names, train_image_names, proj_data, w, t0=2e8, prn=False):
    count = 0
    num_images = 0
    correct_pred = 0

    result = []
    wts = []

    #False match rate (FMR)
    FMR_count = 0

    #False non-match rate (FNMR)
    FNMR_count = 0

    test_image_names2 = sorted(test_image_names)

    for img in test_image_names2:
        img = imread(os.path.join(TEST_IMG_FOLDER, test_image_names[i]))
        unknown_face = plt.imread(os.path.join(TEST_IMG_FOLDER, img))
        num_images += 1

        unknown_face_vector = np.array(unknown_face, dtype='float64').flatten()
        normalised_uface_vector = np.subtract(unknown_face_vector, mean_face)

        w_unknown = np.dot(proj_data, normalised_uface_vector)
        norms = np.linalg.norm(w - w_unknown, axis=1)
        index = np.argmin(norms)

        wts.append([count, norms[index]])

        if prn: print('Input:' + '.'.join(img.split('.')[0:-1]), end='\t')
        count += 1

        match = img.split('_')[0] == train_image_names[index].split('_')[0]
        if norms[index] < t0: # It's a face
            if match:
                if prn: print('Matched:' + train_image_names[index], end = '\t')
                correct_pred += 1
                result.append(1)
            else:
                if prn: print('F/Matched:' + train_image_names[index], end = '\t')
                result.append(0)
                FMR_count += 1
        else:
            if match:
                if prn: print('Unknown face!' + train_image_names[index], end = '\t')
                FNMR_count += 1
            else:
                pass
                correct_pred += 1

        if prn: print(norms[index], end=' ')
    if prn: print()
```

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```
[x] [17] w_unknown = np.dot(proj_data, normalised_uface_vector)
    diff = w - w_unknown
    norms = np.linalg.norm(diff, axis=1)
    index = np.argmin(norms)

    wts.append([count, norms[index]])

    if prn: print('Input:' + '.'.join(img.split('.')[0:-1]), end='\t')
    count += 1

    match = img.split('_')[0] == train_image_names[index].split('_')[0]
    if norms[index] < t0: # It's a face
        if match:
            if prn: print('Matched:' + train_image_names[index], end = '\t')
            correct_pred += 1
            result.append(1)
        else:
            if prn: print('F/Matched:' + train_image_names[index], end = '\t')
            result.append(0)
            FMR_count += 1
    else:
        if match:
            if prn: print('Unknown face!' + train_image_names[index], end = '\t')
            FNMR_count += 1
        else:
            pass
            correct_pred += 1

    if prn: print(norms[index], end=' ')
if prn: print()
```

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[17] FMR = FMR_count/num_images
FNMR = FNMR_count/num_images

```

        print('Correct predictions: {} / {} = {} \t'.format(correct_pred, num_images, correct_pred/num_images), end=' ')
        print('FMR: {} \t'.format(FMR), end=' ')
        print('FNMR: {} \t'.format(FNMR))

    return wts, result, correct_pred, num_images, FMR, FNMR

```

wts, result, correct_pred, num_images, FMR, FNMR = recogniser(test_image_names, train_image_names, proj_data, w, t0=2e8, prn=True)

Input	Matched	F/Matched	Value
Input:00770_960530_fa.jpg	Matched:00770_960530_fa.jpg	0.0	
Input:00770_960530_fa_a.jpg	F/Matched:00744_941201_fa.jpg	40443667.467302494	
Input:00771_941205_fa.jpg	Matched:00771_941205_fa.jpg	0.0	
Input:00771_941205_fb.jpg	Matched:00771_941205_fa.jpg	12411902.37011687	
Input:00772_941201_fa.jpg	Matched:00772_941201_fa.jpg	0.0	
Input:00772_941201_fb.jpg	F/Matched:00763_941201_fa.jpg	21167349.767040696	
Input:00773_941201_fa.jpg	Matched:00773_941201_fa.jpg	0.0	
Input:00773_941201_fb.jpg	Matched:00773_941201_fa.jpg	21385298.893085312	
Input:00775_941205_fa.jpg	Matched:00775_941205_fa.jpg	0.0	
Input:00775_941205_fb.jpg	Matched:00775_941205_fa.jpg	9590985.449027378	
Input:00779_941205_fa.jpg	Matched:00779_941205_fa.jpg	0.0	
Input:00779_941205_fb.jpg	Matched:00779_941205_fa.jpg	38922068.407121904	
Input:00781_941205_fa.jpg	Matched:00781_941205_fa.jpg	0.0	
Input:00781_941205_fb.jpg	F/Matched:00804_941205_fa.jpg	14277462.667984799	
Input:00787_941205_fa.jpg	Matched:00787_941205_fa.jpg	0.0	
Input:00787_941205_fb.jpg	F/Matched:00744_941201_fa.jpg	32827038.49216738	
Input:00794_941205_fa.jpg	Matched:00794_941205_fa.jpg	0.0	
Input:00794_941205_fb.jpg	F/Matched:00775_941205_fa.jpg	13696831.306772245	
Input:00797_941205_fa.jpg	Matched:00797_941205_fa.jpg	0.0	
Input:00797_941205_fb.jpg	Matched:00797_941205_fa.jpg	22738402.814766508	
Totals: 00804_941205_fa.jpg	Matched:00804_941205_fa.jpg	0.0	

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[17] Input:00772_941201_fb.jpg
Input:00773_941201_fa.jpg
Input:00773_941201_fb.jpg
Input:00775_941205_fa.jpg
Input:00775_941205_fb.jpg
Input:00779_941205_fa.jpg
Input:00779_941205_fb.jpg
Input:00781_941205_fa.jpg
Input:00781_941205_fb.jpg
Input:00787_941205_fa.jpg
Input:00787_941205_fb.jpg
Input:00794_941205_fa.jpg
Input:00794_941205_fb.jpg
Input:00797_941205_fa.jpg
Input:00797_941205_fb.jpg
Input:00804_941205_fa.jpg
Input:00804_941205_fb.jpg
Input:00806_941205_fa.jpg
Input:00806_941205_fb.jpg
Input:00807_941205_fa.jpg
Input:00807_941205_fb.jpg
Input:00809_941205_fa.jpg
Input:00809_941205_fb.jpg
Input:00816_941205_fa.jpg
Input:00816_941205_fb.jpg
Input:00876_960530_fa.jpg
Input:00876_960530_fb.jpg
Input:00879_960530_fa.jpg
Input:00879_960530_fb.jpg
Input:00894_960530_fa.jpg
Input:00894_960530_fb.jpg
Input:00900_960530_fa.jpg
Input:00900_960530_fb.jpg
Input:00903_960530_fa.jpg
Input:00903_960530_fb.jpg

Input	Matched	F/Matched	Value
Input:00772_941201_fb.jpg	Matched:00763_941201_fa.jpg	21167349.767040696	
Input:00773_941201_fa.jpg	Matched:00773_941201_fa.jpg	0.0	
Input:00773_941201_fb.jpg	Matched:00773_941201_fa.jpg	21385298.893085312	
Input:00775_941205_fa.jpg	Matched:00775_941205_fa.jpg	0.0	
Input:00775_941205_fb.jpg	Matched:00775_941205_fa.jpg	9590985.449027378	
Input:00779_941205_fa.jpg	Matched:00779_941205_fa.jpg	0.0	
Input:00779_941205_fb.jpg	Matched:00779_941205_fa.jpg	38922068.407121904	
Input:00781_941205_fa.jpg	Matched:00781_941205_fa.jpg	0.0	
Input:00781_941205_fb.jpg	F/Matched:00804_941205_fa.jpg	14277462.667984799	
Input:00787_941205_fa.jpg	Matched:00787_941205_fa.jpg	0.0	
Input:00787_941205_fb.jpg	F/Matched:00744_941201_fa.jpg	32827038.49216738	
Input:00794_941205_fa.jpg	Matched:00794_941205_fa.jpg	0.0	
Input:00794_941205_fb.jpg	F/Matched:00775_941205_fa.jpg	13696831.306772245	
Input:00797_941205_fa.jpg	Matched:00797_941205_fa.jpg	0.0	
Input:00797_941205_fb.jpg	Matched:00797_941205_fa.jpg	22738402.814766508	
Totals: 00804_941205_fa.jpg	Matched:00804_941205_fa.jpg	0.0	

Correct predictions: 21/40 = 0.525 FMR: 0.475 FNMR: 0.0

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```
[18] def rg(r):
    if r: return 'g'
    else: return 'r'

c1 = [rg(r) for r in result]

x=[x[0] for x in wts]
y=[y[1] for y in wts]
plt.scatter(x,y, color=c1, label = 'Distance measure (true ang false pred.)')

x2=[x[0] for x in wts]
y2=[2.7e7 for y in wts]

plt.plot(x2,y2, label = 'Empirical error threshold')
plt.legend()
plt.grid()

plt.show()
```

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```
[19] CPR_list, t0_list, FMR_list, FNMR_list = [], [], [], []
for t0 in np.linspace(start=0, stop=1e8, num=20):
    print('{:e}'.format(t0), end=' ')
    wts, result, correct_pred, num_images, FMR, FNMR = recogniser(test_image_names, train_image_names, proj_data, w, t0)

    CPR_list.append(correct_pred/num_images)
    t0_list.append(t0)
    FMR_list.append(FMR)
    FNMR_list.append(FNMR)
```

t0	CPR	FMR	FNMR
0.000000e+00	Correct predictions: 19/40 = 0.475	FMR: 0.0	FNMR: 0.525
5.263158e+06	Correct predictions: 32/40 = 0.8	FMR: 0.0	FNMR: 0.2
1.052632e+07	Correct predictions: 33/40 = 0.825	FMR: 0.0	FNMR: 0.175
1.578947e+07	Correct predictions: 30/40 = 0.75	FMR: 0.1	FNMR: 0.15
2.105263e+07	Correct predictions: 30/40 = 0.75	FMR: 0.125	FNMR: 0.125
2.631579e+07	Correct predictions: 32/40 = 0.8	FMR: 0.175	FNMR: 0.025
3.157895e+07	Correct predictions: 32/40 = 0.8	FMR: 0.175	FNMR: 0.025
3.684211e+07	Correct predictions: 31/40 = 0.775	FMR: 0.2	FNMR: 0.025
4.210526e+07	Correct predictions: 29/40 = 0.725	FMR: 0.275	FNMR: 0.0
4.736842e+07	Correct predictions: 26/40 = 0.65	FMR: 0.35	FNMR: 0.0
5.263158e+07	Correct predictions: 25/40 = 0.625	FMR: 0.375	FNMR: 0.0
5.789474e+07	Correct predictions: 22/40 = 0.55	FMR: 0.45	FNMR: 0.0
6.315789e+07	Correct predictions: 22/40 = 0.55	FMR: 0.45	FNMR: 0.0
6.842105e+07	Correct predictions: 22/40 = 0.55	FMR: 0.45	FNMR: 0.0
7.368421e+07	Correct predictions: 22/40 = 0.55	FMR: 0.45	FNMR: 0.0
7.894737e+07	Correct predictions: 22/40 = 0.55	FMR: 0.45	FNMR: 0.0
8.421053e+07	Correct predictions: 22/40 = 0.55	FMR: 0.45	FNMR: 0.0
8.947368e+07	Correct predictions: 22/40 = 0.55	FMR: 0.45	FNMR: 0.0
9.473684e+07	Correct predictions: 21/40 = 0.525	FMR: 0.475	FNMR: 0.0
1.000000e+08	Correct predictions: 21/40 = 0.525	FMR: 0.475	FNMR: 0.0

```
[20] x1=t0_list
y1=FMR_list
```

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```
[20]: x1=t0_list
y1=FMR_list

x2=t0_list
y2=FNMR_list

{x}
x3=t0_list
y3=CPR_list

plt.plot(x1,y1, ls='--', color='r', label='FMR')
plt.plot(x2,y2, ls='-.', color='b', label='FNMR')
plt.plot(x3,y3, color='g', label='Correct prediction using threshold')

plt.grid()
plt.legend()

<matplotlib.legend.Legend at 0x7fbec23ae7d0>
```

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Visualization of prediction result on all test images

```
count      = 0
num_images = 0
correct_pred = 0
def Visualization(img, train_image_names, proj_data, w, t0):
    global count, highest_min, num_images, correct_pred
    unknown_face      = plt.imread(os.path.join(TEST_IMG_FOLDER, img))
    num_images += 1
    unknown_face_vector = np.array(unknown_face, dtype='float64').flatten()
    normalised_uface_vector = np.subtract(unknown_face_vector, mean_face)

    plt.subplot(40,2,1+count)
    plt.imshow(unknown_face, cmap='gray')
    plt.title('Input:'+'.'.join(img.split('.')[2:]))
    plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
    count+=1

    w_unknown = np.dot(proj_data, normalised_uface_vector)
    diff = w - w_unknown
    norms = np.linalg.norm(diff, axis=1)
    index = np.argmin(norms)

    if norms[index] < t0: # It's a face

        match = img.split('_')[0] == train_image_names[index].split('_')[0]
        if img.split('.')[0] == train_image_names[index].split('.')[0]:
            correct_pred += 1

    plt.subplot(40,2,1+count)
    if norms[index] < t0: # It's a face
```

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```

+ Code + Text
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if norms[index] < t0: # It's a face

    match = img.split('_')[0] == train_image_names[index].split('_')[0]
    if img.split('.')[0] == train_image_names[index].split('.')[0]:
        if match:
            plt.title('Matched:'+'.'.join(train_image_names[index].split('.')[2]), color='g')
            plt.title('Matched:', color='g')
            plt.imshow(imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[index])), cmap='gray')
            #img = plt.imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[i]))
            correct_pred += 1
        else:
            plt.title('Matched:'+'.'.join(train_image_names[index].split('.')[2]), color='r')
            plt.title('False matched:', color='r')
            plt.imshow(imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[index])), cmap='gray')
    else:
        if img.split('.')[0] not in [i.split('.')[0] for i in train_image_names] and img.split('.')[0] != 'apple':
            if img.split('_')[0] not in [i.split('_')[0] for i in train_image_names]:
                plt.title('Unknown face', color='g')
                correct_pred += 1
            else:
                plt.title('Unknown face', color='r')

    plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
    plt.subplots_adjust(right=1.2, top=2.5)

count+=1

fig = plt.figure(figsize=(5, 30))

test_image_names2 = sorted(test_image_names)
for i in range(len(test_image_names2)):
    Visualization(test_image_names2[i], train_image_names, proj_data,w, t0=2.7e7)

```

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```

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count+=1

fig = plt.figure(figsize=(5, 30))

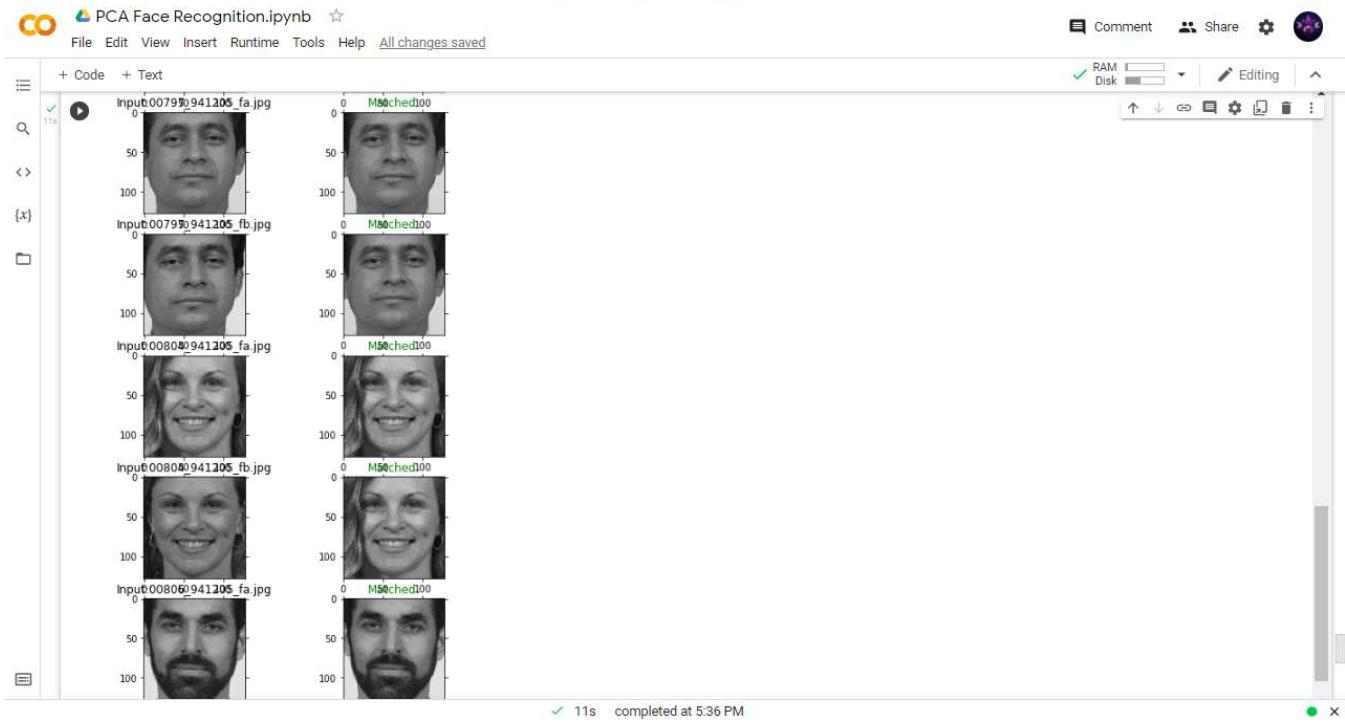
test_image_names2 = sorted(test_image_names)
for i in range(len(test_image_names2)):
    Visualization(test_image_names2[i], train_image_names, proj_data,w, t0=2.7e7)

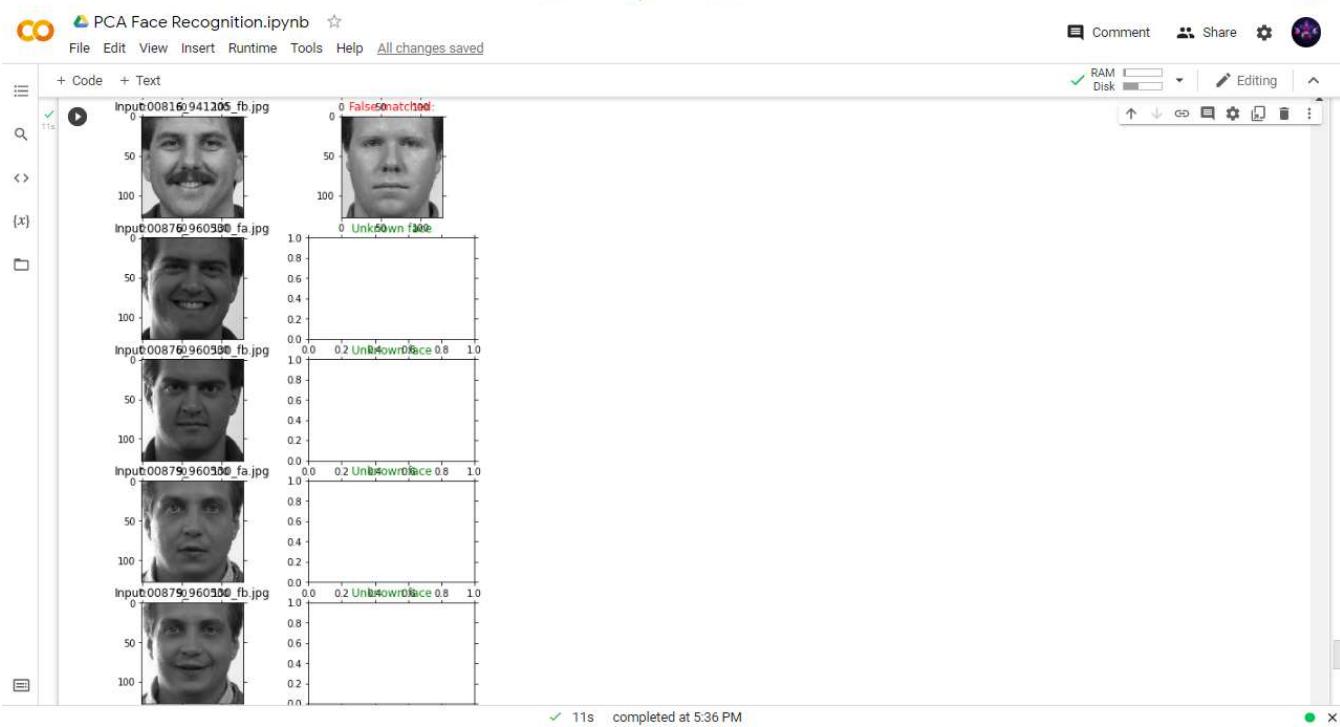
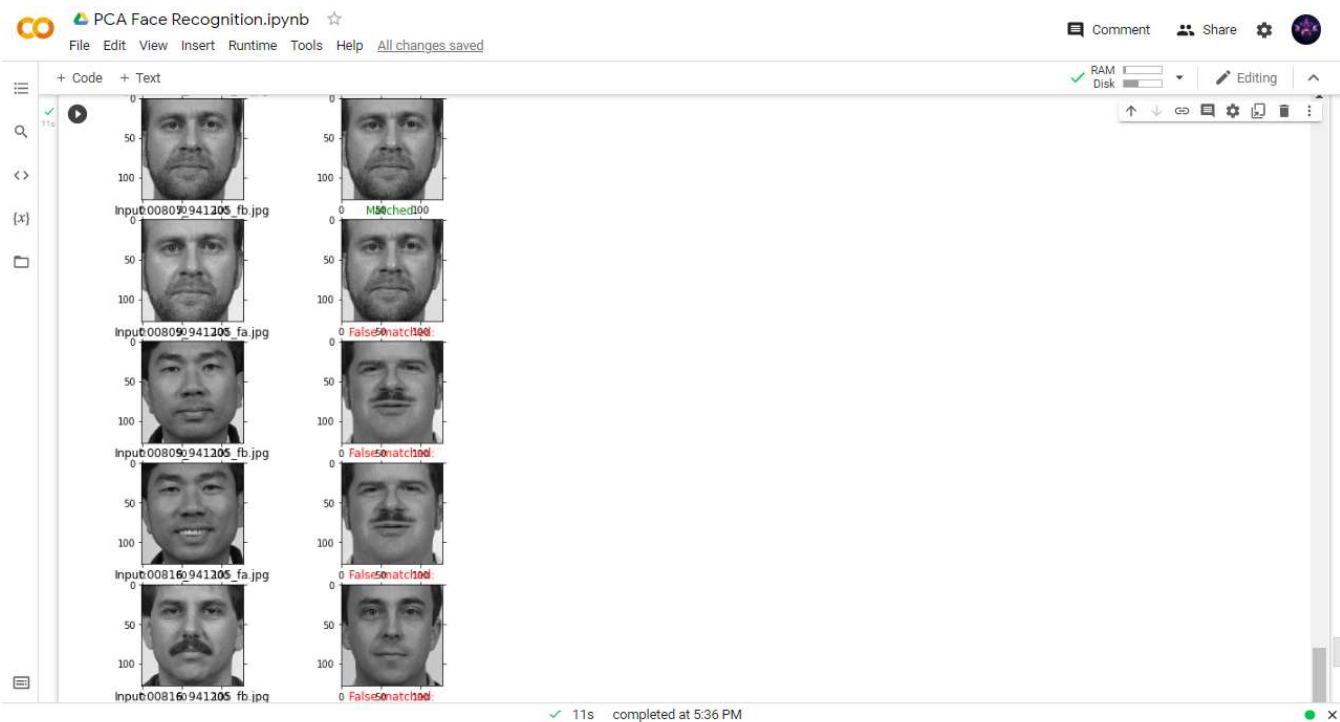
plt.show()

```

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Code:

```

from matplotlib import pyplot as plt
from matplotlib.image import imread
import numpy as np
import os

TRAIN_IMG_FOLDER = (r"/content/sample_data/Training images")
TEST_IMG_FOLDER = (r"/content/sample_data/TEST")

train_set_files = os.listdir(TRAIN_IMG_FOLDER)
test_set_files = os.listdir(TEST_IMG_FOLDER)

width = 128
height = 128

train_id_file = set([f.split('_')[0] for f in train_set_files])
test_id_file = set([f.split('_')[0] for f in test_set_files])
print(train_id_file <= test_id_file)

print('Train Images:')

train_image_names = os.listdir(TRAIN_IMG_FOLDER)
training_tensor = np.ndarray(shape=(len(train_image_names), height*width), dtype=np.float64)

for i in range(len(train_image_names)):
    img = plt.imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[i]))
    training_tensor[i] = np.array(img, dtype='float64').flatten()

```

```

plt.subplot(5,5,1+i)
plt.imshow(img, cmap='gray')
plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
plt.show()

print('Test Images:')
test_image_names = os.listdir(TEST_IMG_FOLDER) #[i for i in dataset_dir if i not in train_image_names]
testing_tensor = np.ndarray(shape=(len(test_image_names), height*width), dtype=np.float64)

for i in range(len(test_image_names)):
    img = imread(os.path.join(TEST_IMG_FOLDER, test_image_names[i]))
    testing_tensor[i,:] = np.array(img, dtype='float64').flatten()
    plt.subplot(8,5,1+i)
    plt.imshow(img, cmap='gray')
    plt.subplots_adjust(right=1.2, top=1.2)
    plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
plt.show()

mean_face = np.zeros((1,height*width))

for i in training_tensor:
    mean_face = np.add(mean_face,i)

mean_face = np.divide(mean_face, float(len(train_image_names))).flatten()

plt.imshow(mean_face.reshape(height, width), cmap='gray')
plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
plt.show()

normalised_training_tensor = np.ndarray(shape=(len(train_image_names), height*width))

for i in range(len(train_image_names)):
    normalised_training_tensor[i] = np.subtract(training_tensor[i],mean_face)

for i in range(len(train_image_names)):
    img = normalised_training_tensor[i].reshape(height,width)
    plt.subplot(5,5,1+i)
    plt.imshow(img, cmap='gray')
    plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
plt.show()

cov_matrix=np.cov(normalised_training_tensor)
cov_matrix = np.divide(cov_matrix,25.0)

```

```

print('Covariance Matrix Shape:', cov_matrix.shape)
#print('Covariance matrix of X: \n%s' %cov_matrix)

#eigenvalues and eigenvectors
eigenvalues, eigenvectors, = np.linalg.eig(cov_matrix)
print('eigenvalues.shape: {} eigenvectors.shape: {}'.format(eigenvalues.shape, eigenvectors.shape))

eig_pairs = [(eigenvalues[index], eigenvectors[:,index]) for index in range(len(eigenvalues))]

# Sort the eigen pairs in descending order:
eig_pairs.sort(reverse=True)
eigenvalues_sort = [eig_pairs[index][0] for index in range(len(eigenvalues))]
eigenvectors_sort = [eig_pairs[index][1] for index in range(len(eigenvalues))]

sorted_ind = sorted(range(eigenvalues.shape[0]), key=lambda k: eigenvalues[k], reverse=True)

eigenvalues_sort = eigenvalues[sorted_ind]
eigenvectors_sort = eigenvectors[sorted_ind]
train_set_files_sort = np.array(train_set_files)[sorted_ind]

var_comp_sum = np.cumsum(eigenvalues_sort)/sum(eigenvalues_sort)

# Show cumulative proportion of variance with respect to components
print("Cumulative proportion of variance explained vector: \n%s" %var_comp_sum)

# x-axis for number of principal components kept
num_comp = range(1,len(eigenvalues_sort)+1)
plt.title('Cum. Prop. Variance Explain and Components Kept')
plt.xlabel('Principal Components')
plt.ylabel('Cum. Prop. Variance Explained')

plt.scatter(num_comp, var_comp_sum)
plt.show()

reduced_data = np.array(eigenvectors_sort[:25]).transpose()
reduced_data.shape

print(training_tensor.transpose().shape, reduced_data.shape)

proj_data = np.dot(training_tensor.transpose(), reduced_data)
proj_data = proj_data.transpose()
proj_data.shape

for i in range(proj_data.shape[0]):
    img = proj_data[i].reshape(height,width)
    plt.subplot(5,5,1+i)
    plt.imshow(img, cmap='gray')
    plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')

```

```

plt.show()

w = np.array([np.dot(proj_data,i) for i in normalised_training_tensor])
print(w.shape)

def recogniser(test_image_names, train_image_names,proj_data,w, t0=2e8, prn=False):

    count      = 0
    num_images = 0
    correct_pred = 0

    result = []
    wts = []

    #False match rate (FMR)
    FMR_count = 0

    #False non-match rate (FNMR)
    FNMR_count = 0

    test_image_names2 = sorted(test_image_names)

    for img in test_image_names2:
        #img = imread(os.path.join(TEST_IMG_FOLDER, test_image_names[i]))
        unknown_face = plt.imread(os.path.join(TEST_IMG_FOLDER, img))
        num_images += 1

        unknown_face_vector = np.array(unknown_face, dtype='float64').flatten()
        normalised_uface_vector = np.subtract(unknown_face_vector,mean_face)

        w_unknown = np.dot(proj_data, normalised_uface_vector)
        diff = w - w_unknown
        norms = np.linalg.norm(diff, axis=1)
        index = np.argmin(norms)

        wts.append([count, norms[index]])

        if prn: print('Input:'+'.'.join(img.split('.')[0:-1]), end='\t')
        count+=1

        match = img.split('_')[0] == train_image_names[index].split('_')[0]
        if norms[index] < t0: # It's a face
            if match:
                if prn: print('Matched:' + train_image_names[index], end = '\t')
                correct_pred += 1
                result.append(1)
            else:
                if prn: print('F/Matched:' +train_image_names[index], end = '\t')
                result.append(0)
                FMR_count += 1
        else:
            if match:

```

```

        if prn: print('Unknown face! '+train_image_names[index], end = '\t')
        FNMR_count +=1

    else:
        pass
    correct_pred += 1

if prn: print(norms[index], end=' ')
if prn: print()

FMR = FMR_count/num_images
FNMR = FNMR_count/num_images

print('Correct predictions: {} / {} = {} \t\t'.format(correct_pred, num_images, correct_pred/num_images), end=' ')
print('FMR: {} \t'.format(FMR), end=' ')
print('FNMR: {} \t'.format(FNMR))

return wts, result, correct_pred, num_images, FMR, FNMR

```

wts, result, correct_pred, num_images, FMR, FNMR =recogniser(test_image_names, train_image_names,proj_data,w, t0=2e8, prn=True)

```

def rg(r):
    if r: return 'g'
    else: return 'r'
cl = [rg(r) for r in result]

x=[x[0] for x in wts]
y=[y[1] for y in wts]
plt.scatter(x,y, color=cl, label = 'Distance measure (true ang false pred.)')

x2=[x[0] for x in wts]
y2=[2.7e7 for y in wts]

plt.plot(x2,y2, label = 'Empirical error threshold')
plt.legend()
plt.grid()

plt.show()

```

CPR_list, t0_list, FMR_list, FNMR_list = [], [], [], []
for t0 in np.linspace(start=0, stop=1e8, num=20):
 print('{:e}'.format(t0), end=' ')

```

wts, result, correct_pred, num_images, FMR, FNMR = recogniser(test_image_names,
train_image_names,proj_data,w, t0)

CPR_list.append(correct_pred/num_images)
t0_list.append(t0)
FMR_list.append(FMR)
FNMR_list.append(FNMR)

x1=t0_list
y1=FMR_list

x2=t0_list
y2=FNMR_list

x3=t0_list
y3=CPR_list

plt.plot(x1,y1, ls='--', color='r', label='FMR')
plt.plot(x2,y2, ls='-.', color='b', label='FNMR')
plt.plot(x3,y3, color='g', label='Correct prediction using threshold')

plt.grid()
plt.legend()

count      = 0
num_images = 0
correct_pred = 0

def Visualization(img, train_image_names,proj_data,w, t0):
    global count,highest_min,num_images,correct_pred
    unknown_face      = plt.imread(os.path.join(TEST_IMG_FOLDER, img))
    num_images       += 1
    unknown_face_vector = np.array(unknown_face, dtype='float64').flatten()
    normalised_uface_vector = np.subtract(unknown_face_vector,mean_face)

    plt.subplot(40,2,1+count)
    plt.imshow(unknown_face, cmap='gray')
    plt.title('Input:'+'.'.join(img.split('.')[0:2]))
    plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
    count+=1

    w_unknown = np.dot(proj_data, normalised_uface_vector)
    diff   = w - w_unknown
    norms = np.linalg.norm(diff, axis=1)
    index = np.argmin(norms)

    plt.subplot(40,2,1+count)
    if norms[index] < t0: # It's a face

```

```

match = img.split('_')[0] == train_image_names[index].split('_')[0]
#if img.split('.')[0] == train_image_names[index].split('.')[0]:
if match:
    #plt.title('Matched:'+'.'.join(train_image_names[index].split('.')[2]))
, color='g')
    plt.title('Matched:', color='g')
    plt.imshow(imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[index])), cmap='gray')
#img = plt.imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[i]))
)
    correct_pred += 1
else:
    #plt.title('Matched:'+'.'.join(train_image_names[index].split('.')[2]))
, color='r')
    plt.title('False matched:', color='r')
    plt.imshow(imread(os.path.join(TRAIN_IMG_FOLDER, train_image_names[index])), cmap='gray')
else:
    #if img.split('.')[0] not in [i.split('.')[0] for i in train_image_names] and img.split('.')[0] != 'apple':
    if img.split('_')[0] not in [i.split('_')[0] for i in train_image_names]:
        plt.title('Unknown face', color='g')
        correct_pred += 1
    else:
        plt.title('Unknown face', color='r')

plt.tick_params(labelleft='off', labelbottom='off', bottom='off', top='off', right='off', left='off', which='both')
plt.subplots_adjust(right=1.2, top=2.5)

count+=1

fig = plt.figure(figsize=(5, 30))

test_image_names2 = sorted(test_image_names)
for i in range(len(test_image_names2)):
    Visualization(test_image_names2[i], train_image_names, proj_data, w, t0=2.7e7)

plt.show()

```

Conclusion:

Conclusion for this project is that with the help of this project we learn Artificial intelligence and coding of face recognition in python. The best part is that we've learned a lot by creating this face application.