

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:

Face recognition ml model is a statistical approach used for reducing the number of variables in face recognition. In Face recognition model, every image in the training set is represented as a linear combination of weighted eigenvectors called Eigen faces. 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.

Screenshot Of Model:

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 = ("content/sample_data/Training images")
TEST_IMG_FOLDER = ("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 train_set_files])
```

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```
[3] test_id_file = set([f.split('_')[0] for f in train_set_files])
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(5,5,1+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()
```

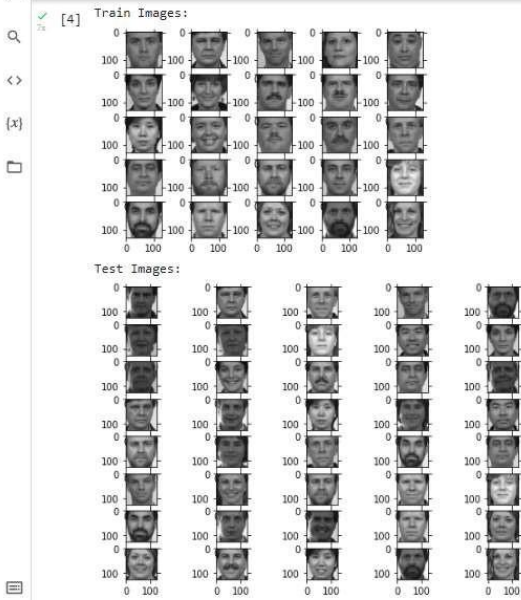
Train Images:



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Calculate the mean face

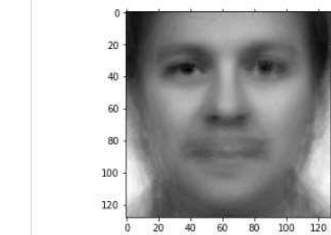
The mean is just the sum of all of the pictures divided by the number of pictures. As a result, we will have an "average" face.

```
[5] 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()
```



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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.

```
[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))]
eigvectors_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]
eigvectors_sort = eigenvectors[sorted_ind]
train_set_files_sort = np.array(train_set_files)[sorted_ind]

Find cumulative variance of each principle component
```

```
[11] 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)
```

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

[11] # 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()
```

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 0.83669985
0.85752859 0.87642433 0.89418977 0.91069532 0.92659352 0.94017913
0.95309028 0.96500642 0.97612862 0.98491526 0.99275274 1.
1.]
```

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Choose the necessary of principal components:

```

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

(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

PCA Face Recognition.ipynb

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```

[15] 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()

```

Finding weights for each training 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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(25, 25)

Reconizing all test images

```
[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)
```

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```
[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('.')[0:2]), end='\n')
    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='\n')
            correct_pred += 1
            result.append(1)
        else:
            if prn: print('F/Matched: '+train_image_names[index], end='\n')
            result.append(0)
            FMR_count += 1
    else:
        if match:
            if prn: print('Unknown face!'+train_image_names[index], end='\n')
            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
FMR = FMR_count/num_images

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

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

wts, result, correct_pred, num_images, FMR, FMR = recogniser(test_image_names, train_image_names, prof_data, w, t0=2e5, prn=True)

Input:00770_960530_fa.jpg Matched:00770_960530_fa.jpg 0.0
Input:00770_960530_fa.jpg F/Matched:00744_941201_fa.jpg 40443667.467302494
Input:00771_941205_fa.jpg Matched:00771_941205_fa.jpg 0.0
Input:00771_941205_fa.jpg Matched:00771_941205_fa.jpg 12411902.37011687
Input:00772_941201_fa.jpg Matched:00772_941201_fa.jpg 0.0
Input:00772_941201_fa.jpg F/Matched:00763_941201_fa.jpg 21167349.767040696
Input:00773_941201_fa.jpg Matched:00773_941201_fa.jpg 0.0
Input:00773_941201_fa.jpg Matched:00773_941201_fa.jpg 21385298.893085312
Input:00775_941205_fa.jpg Matched:00775_941205_fa.jpg 0.0
Input:00775_941205_fa.jpg Matched:00775_941205_fa.jpg 9590985.449027378
Input:00779_941205_fa.jpg Matched:00779_941205_fa.jpg 0.0
Input:00779_941205_fa.jpg Matched:00779_941205_fa.jpg 38922068.407121904
Input:00781_941205_fa.jpg Matched:00781_941205_fa.jpg 0.0
Input:00781_941205_fa.jpg F/Matched:00804_941205_fa.jpg 14277462.667984759
Input:00787_941205_fa.jpg Matched:00787_941205_fa.jpg 0.0
Input:00787_941205_fa.jpg F/Matched:00744_941201_fa.jpg 32827038.49216738
Input:00794_941205_fa.jpg Matched:00794_941205_fa.jpg 0.0
Input:00794_941205_fa.jpg F/Matched:00775_941205_fa.jpg 13696831.306772245
Input:00797_941205_fa.jpg Matched:00797_941205_fa.jpg 0.0
Input:00797_941205_fa.jpg Matched:00797_941205_fa.jpg 22738402.814766508
Input:00804_941205_fa.jpg Matched:00804_941205_fa.jpg 0.0
```

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```
[17] Input:00770_960530_fa.jpg Matched:00770_960530_fa.jpg 0.0
Input:00770_960530_fa.jpg F/Matched:00763_941201_fa.jpg 21167349.767040696
Input:00771_941201_fa.jpg Matched:00771_941201_fa.jpg 0.0
Input:00771_941201_fa.jpg Matched:00771_941201_fa.jpg 12411902.37011687
Input:00772_941201_fa.jpg Matched:00772_941201_fa.jpg 0.0
Input:00772_941201_fa.jpg F/Matched:00763_941201_fa.jpg 21167349.767040696
Input:00773_941201_fa.jpg Matched:00773_941201_fa.jpg 0.0
Input:00773_941201_fa.jpg Matched:00773_941201_fa.jpg 21385298.893085312
Input:00775_941205_fa.jpg Matched:00775_941205_fa.jpg 0.0
Input:00775_941205_fa.jpg Matched:00775_941205_fa.jpg 9590985.449027378
Input:00779_941205_fa.jpg Matched:00779_941205_fa.jpg 0.0
Input:00779_941205_fa.jpg Matched:00779_941205_fa.jpg 38922068.407121904
Input:00781_941205_fa.jpg Matched:00781_941205_fa.jpg 0.0
Input:00781_941205_fa.jpg F/Matched:00804_941205_fa.jpg 14277462.667984759
Input:00787_941205_fa.jpg Matched:00787_941205_fa.jpg 0.0
Input:00787_941205_fa.jpg F/Matched:00744_941201_fa.jpg 32827038.49216738
Input:00794_941205_fa.jpg Matched:00794_941205_fa.jpg 0.0
Input:00794_941205_fa.jpg F/Matched:00775_941205_fa.jpg 13696831.306772245
Input:00797_941205_fa.jpg Matched:00797_941205_fa.jpg 0.0
Input:00797_941205_fa.jpg Matched:00797_941205_fa.jpg 22738402.814766508
Input:00804_941205_fa.jpg Matched:00804_941205_fa.jpg 0.0
Input:00804_941205_fa.jpg Matched:00804_941205_fa.jpg 23695586.065497693
Input:00806_941205_fa.jpg Matched:00806_941205_fa.jpg 0.0
Input:00806_941205_fa.jpg Matched:00806_941205_fa.jpg 22783423.946304042
Input:00807_941205_fa.jpg Matched:00807_941205_fa.jpg 0.0
Input:00807_941205_fa.jpg Matched:00807_941205_fa.jpg 19176199.70993461
Input:00809_941205_fa.jpg F/Matched:00761_941201_fa.jpg 13086008.769753503
Input:00809_941205_fa.jpg F/Matched:00761_941201_fa.jpg 22463983.150025778
Input:00816_941205_fa.jpg F/Matched:00766_941201_fa.jpg 16977907.094736833
Input:00816_941205_fa.jpg F/Matched:00794_941205_fa.jpg 10808932.965668682
Input:00876_960530_fa.jpg F/Matched:00744_941201_fa.jpg 55433907.515527114
Input:00876_960530_fa.jpg F/Matched:00744_941201_fa.jpg 44354768.59399727
Input:00879_960530_fa.jpg F/Matched:00797_941205_fa.jpg 44897093.13547301
Input:00879_960530_fa.jpg F/Matched:00797_941205_fa.jpg 42228054.24150416
Input:00894_960530_fa.jpg F/Matched:00781_941205_fa.jpg 40255117.39147155
Input:00894_960530_fa.jpg F/Matched:00781_941205_fa.jpg 37625178.90044492
Input:00900_960530_fa.jpg F/Matched:00745_941201_fa.jpg 55566607.45522714
Input:00900_960530_fa.jpg F/Matched:00745_941201_fa.jpg 50361147.60444721
Input:00903_960530_fa.jpg F/Matched:00745_941201_fa.jpg 91323857.85102612
Input:00903_960530_fa.jpg F/Matched:00745_941201_fa.jpg 54705102.14316325
Correct predictions: 21/40 = 0.525 FMR: 0.475 FMR: 0.0
```

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

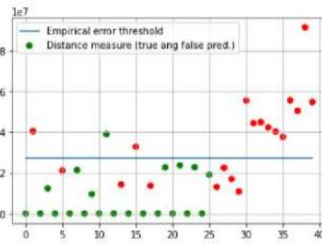
    cl = [rg(r) for r in result]

    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()
```



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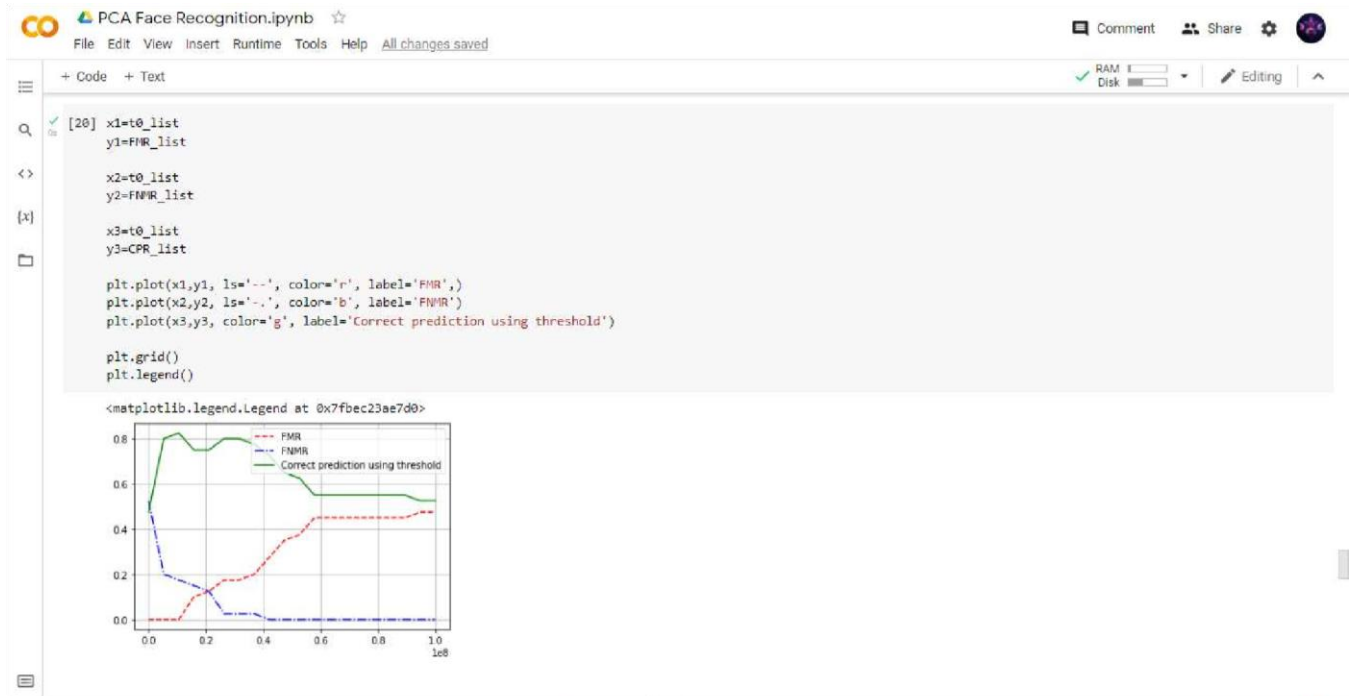
```
[19] CPR_list, t0_list, FMR_list, FNMR_list = [], [], [], []
for t0 in np.linspace(start=0, stop=1e8, num=20):
    print('{:e}'.format(t0), end=' ')
    wts, result, correct_pred, num_images, FMR, FNMR = recogniser(test_image_names, train_image_names,proj_data,w, t0)

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

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=FNMR_list
```

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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:{}'.format(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)

    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:
```

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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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+ Code + Text

RAM Disk Editing

```

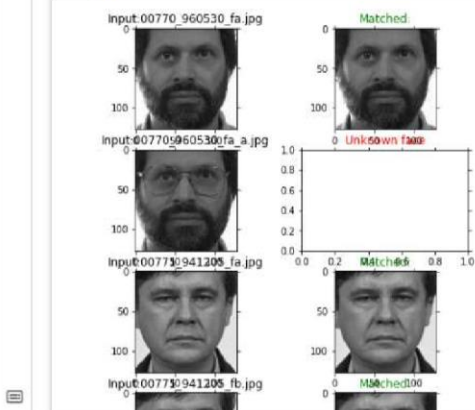
count+=1

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

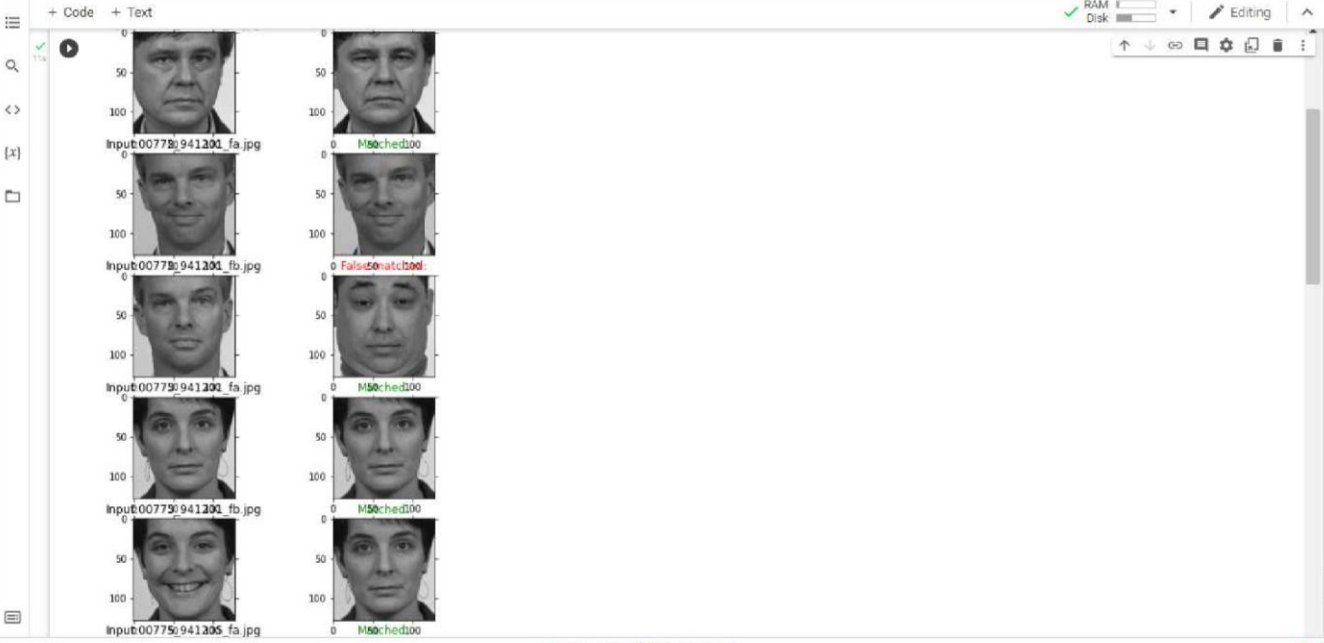
test_image_names2 = sorted(test_image_names)
for i in range(len(test_image_names2)):
    Visualization(test_image_names2[i], train_image_names, proj_data, w, t0=2.7e7)

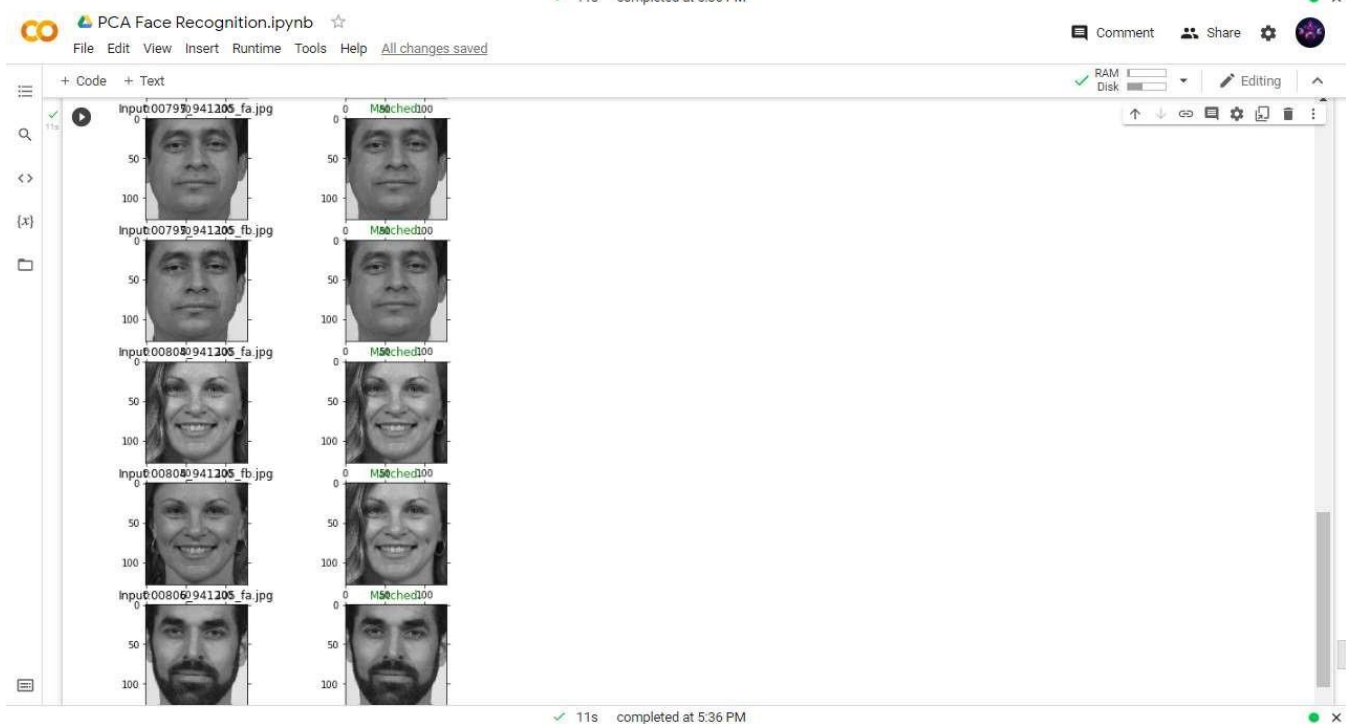
plt.show()

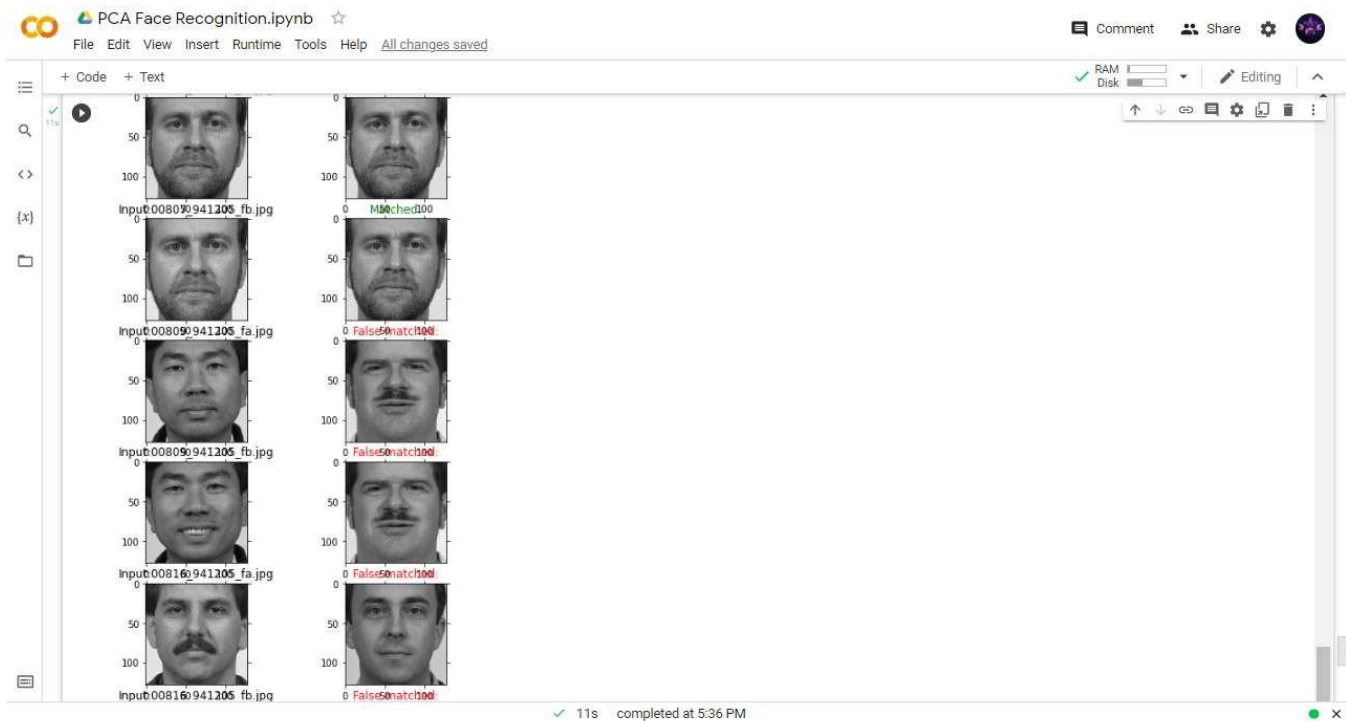
```

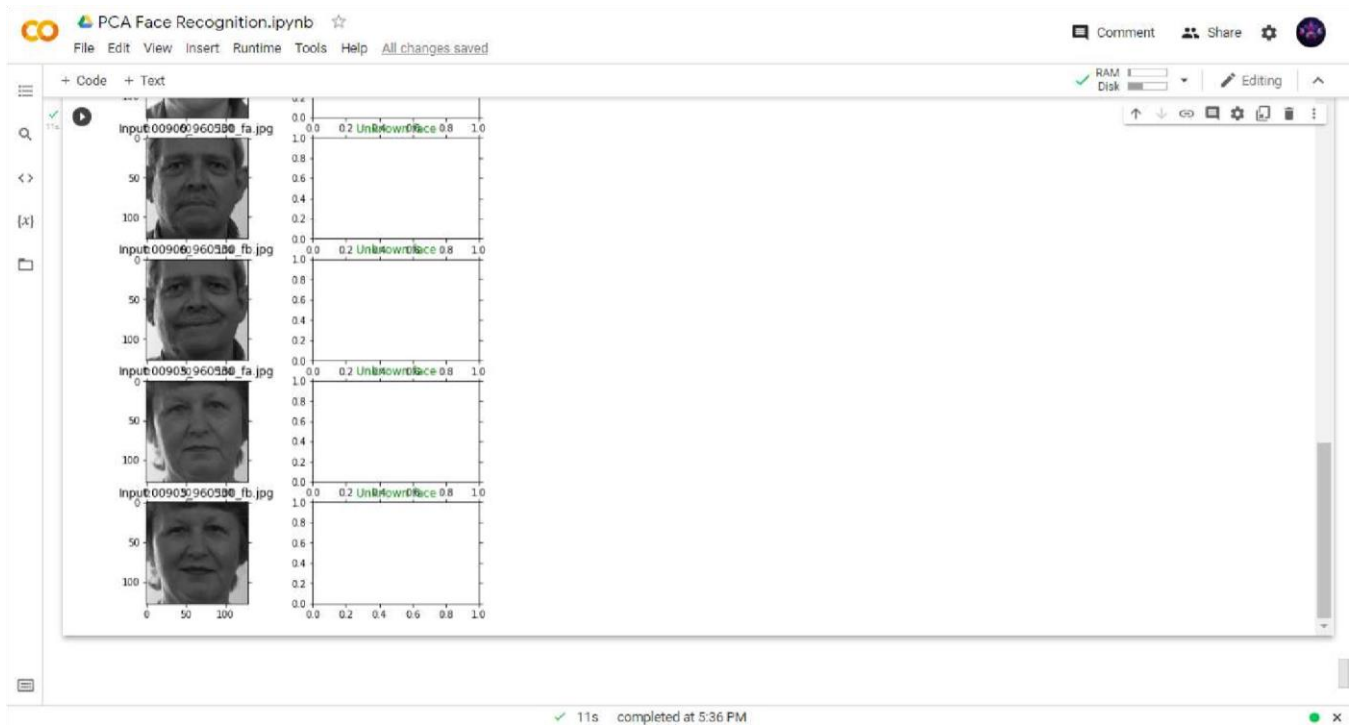


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Code:

```
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( eigenvectors))]

```

```

# 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]
eigvectors_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 Explained 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(eigvectors_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('.')[0:2]), end='\t')
    count+=1
    match = img.split('.')[0] == train_image_names[index].split('.')[0]
    if norms[index] < t0: # It's a face
        if match:
            correct_pred += 1
        if prn: print('Matched:' + train_image_names[index], end = '\t')
        result.append(1)
    else:
        if prn: print('F/Matched:' + train_image_names[index], end = '\t')
        result.append(0)
        FMR_count += 1
        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: {}/{} = {}'.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 and 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('.')[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 Face recognition ml model is that with the help of this project we learn Artificial intelligence and coding of face recognition in python.