

CV Assignment 3

Bayes Implementation of EigenFaces for Face Recognition

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```
In [7]: import os
import numpy as np
from PIL import Image
import matplotlib.pyplot as plt
import cv2
import matplotlib.image as mpimg
import glob
%matplotlib inline
```

Extract and Vectorise Images from Yale Database

```
In [8]: #collect images as classes based on the person ie 15 classes
def yale_db():
    img1 = []
    all_images=[]
    store_location = glob.glob("D:/AnacondaProjects/yalefaces/yalefaces_asgt2/subject*")
    for loc in store_location:
        image = mpimg.imread(loc)
        image = cv2.resize(image,(50,50))
        img1.append(image)
    img = np.array(img1)
    img_train_class1 = []
    img_test1 = []
    for i in range(15): #15 ppl
        img_train1 = []
        for j in range(i,i+10): #10 training images for 1 person
            img_train1.append(img[10*i+j])
        img_train=np.array(img_train1)
        img_train_class1.append(img_train)
        img_test1.append(img[10*i+10+i])
    img_train_class=np.array(img_train_class1)
    img_test=np.array(img_test1)
    return img_train_class,img_test

def vectorize_class_allimg(img):
    if len(img) == 0:
        return np.array([])
    return np.reshape(img,(img.shape[0]*img.shape[1],img.shape[2]*img.shape[3]))
```

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def vectorize_class(img):
    if len(img) == 0:
        return np.array([])
    return np.reshape(img, (img.shape[0], img.shape[1], img.shape[2]*img.shape[3]))

def vectorize_img(img):
    if len(img) == 0:
        return np.array([])
    return np.reshape(img, (img.shape[0], img.shape[1]*img.shape[2]))

img_train_class_orig, img_test_orig = yale_db()
img_train = vectorize_class_allimg(img_train_class_orig)
img_train_class = vectorize_class(img_train_class_orig)
img_test = vectorize_img(img_test_orig)

```

PCA and EigenFaces Computation

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In [9]: dif = []
        for clas in img_train_class:
            for i in range(len(clas)):
                for j in range(i, len(clas)):
                    dif.append(clas[i] - clas[j])
        dif = np.array(dif)
        mean_dif = 0

        mean_i = np.mean(dif, axis=0)
        std_i = np.std(dif, axis=0)
        c = []
        for i in dif:
            c.append((i - mean_i) / std_i)
        c = np.array(c)
        ncovar = np.cov(np.transpose(c))
        evalu, evec = np.linalg.eig(ncovar)

        indices = np.argsort(-evalu)
        evalu = evalu[indices]
        evec = evec[:, indices]

```

Prediction

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In [10]: N = 20

        def coef(img):
            return (np.power(evalu[:N], -1/2) * (evec.T[:N] * img).T).T

```

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eigface_set = []
for clas in img_train_class:
    eigface = 0
    for i in range(len(clas)):
        eigface += coef(clas[i]) / len(clas)
    eigface_set.append(eigface)
eigface_set = np.array(eigface_set)

def predict(img):
    allps = []
    for i in range(eigface_set.shape[0]):
        pr = np.exp(- 0.5 * np.linalg.norm(coef(img)- eigface_set[i,:,:]))
        allps.append(pr)
    pred = np.argmax(allps) + 1
    return pred

```

```

In [6]: for k in range(2,img_test.shape[0],2):
        plt.subplot(1,2,1)
        plt.imshow(np.reshape(img_test[k],(50,50)),cmap='gray')
        plt.title('Expected Class {}'.format(k+1))
        plt.axis('off')
        plt.subplot(1,2,2)
        c=predict(img_test[k])
        plt.imshow(np.reshape(img_train_class[c-1][0],(50,50)),cmap='gray')
        plt.title('Predicted Class {}'.format(c))
        plt.axis('off')
        plt.show()

```

Expected Class 3



Predicted Class 3



Expected Class 5



Predicted Class 5



Expected Class 7



Predicted Class 7



Expected Class 9



Predicted Class 9



Expected Class 11



Predicted Class 11



Expected Class 13



Predicted Class 13



Expected Class 15



Predicted Class 3

