

# **CSC320: Project 3**

Face Recognition and Classification with Eigenfaces

**Ramaneek Gill and Ryan D'Souza**

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## Part 1

The dataset of faces used were of popular celebrities (mainly 'Aaron Eckhart', 'Adam Sandler', 'Adrien Brody', 'Andrea Anders', 'Ashley Benson', 'Christina Applegate', 'Dianna Agron', and 'Gillian Anderson'). The images are of size 32 by 32 and have been grayscale by using 0.299 intensity from the red channel, 0.587 from the green channel and 0.114 from the blue channel, just like in project 2. As shown in Figure 1 the cropped out faces are fairly accurate, most images can be aligned with each other but some images only show the side of the face. Thanks to a fairly large dataset we shouldn't be concerned about these 'outliers'.

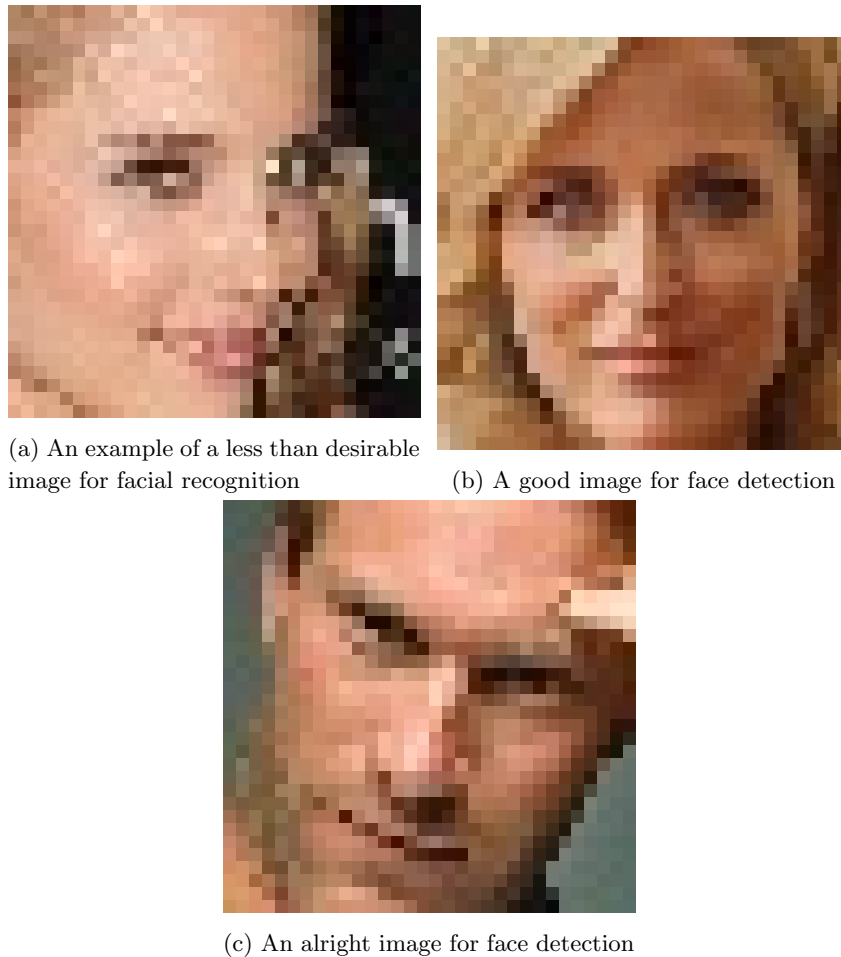


Figure 1: A subset of the dataset of images used

## Part 2

The dataset was separated into 100 training images, 10 validation images, and 10 test images. This was done by just filling an array until it was full, the code is provided in the `getData()` function in the source code (for your convenience it is located in the last section of the report). Figure 2 shows 25 eigenfaces (of the 800 possible ones, 100 for each actor for a total of 8 actors) and the mean face of the training set.



(a) 25 eigenfaces from the 800 in our dataset



(b) The mean face of the eigenspace of 800 eigenfaces

Figure 2: Data generated from part 2

## Part 3

The results from running `p3.py`:

Training, validation, and test sets created!

Projection matrix and mean eigenface computed!

This match was incorrect, validation image 0 incorrectly matched actor 0 with 4

This match was incorrect, validation image 1 incorrectly matched actor 0 with 2

This match was incorrect, validation image 3 incorrectly matched actor 0 with 2

This match was incorrect, validation image 4 incorrectly matched actor 0 with 2

This match was incorrect, validation image 6 incorrectly matched actor 0 with 1

Performance on the validation set using 2 eigenfaces was: 22.5

Best setting for face recognition is to use 2 eigenfaces

Performance on the validation set using 5 eigenfaces was: 38.75

Best setting for face recognition is to use 5 eigenfaces

Performance on the validation set using 10 eigenfaces was: 47.5

Best setting for face recognition is to use 10 eigenfaces

Performance on the validation set using 20 eigenfaces was: 61.25

Best setting for face recognition is to use 20 eigenfaces

Performance on the validation set using 50 eigenfaces was: 66.25

Best setting for face recognition is to use 50 eigenfaces

Performance on the validation set using 80 eigenfaces was: 65.0

Best setting for face recognition is to use 50 eigenfaces

Performance on the validation set using 100 eigenfaces was: 66.25

Best setting for face recognition is to use 50 eigenfaces

Performance on the validation set using 150 eigenfaces was: 66.25

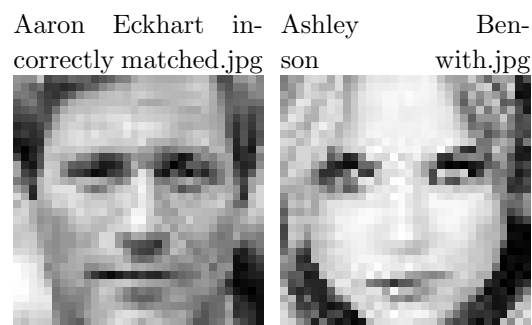
Best setting for face recognition is to use 50 eigenfaces

Performance on the validation set using 200 eigenfaces was: 67.5

Best setting for face recognition is to use 200 eigenfaces

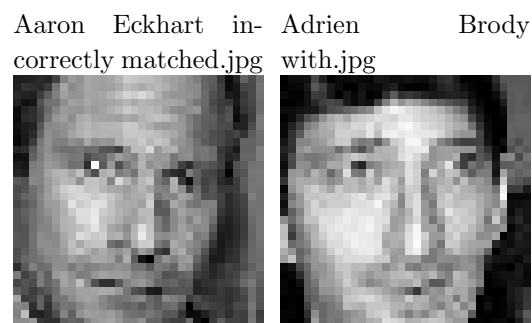
Performance on the test set using 200 eigenfaces was: 50.0

Figure 3a,4a,5a,6a,7a show the 0th, 1st, 3rd, 4th, and 6th validation images in the validation set were incorrectly recognized as the images from the training set as shown in Figure 3b,4b,5b,6b,7b respectively when using only 2 eigenfaces to compute facial recognitions.



(a) This image was in- correctly matched with Figure 3b (b) This recognition was computed using only 2 eigenfaces

Figure 3: Data generated from part 3



(a) This image was in- correctly matched with Figure 4b (b) This recognition was computed using only 2 eigenfaces

Figure 4: Data generated from part 3

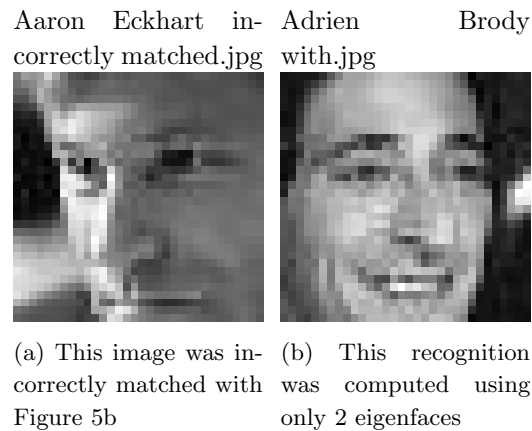


Figure 5: Data generated from part 3

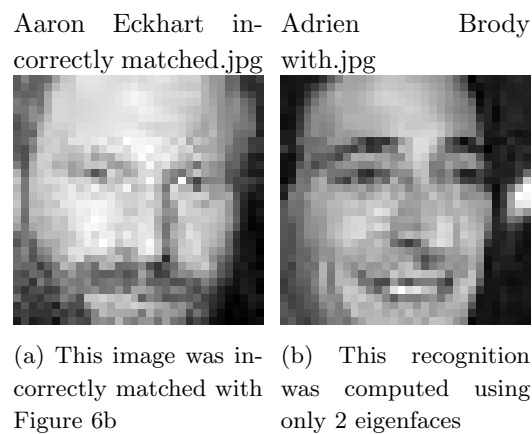


Figure 6: Data generated from part 3

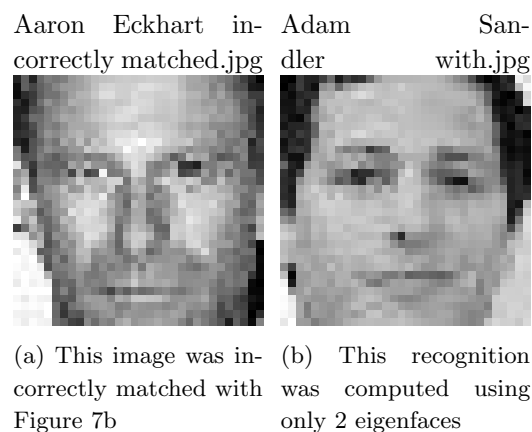


Figure 7: Data generated from part 3

## Part 4

The output of the program:

Performance on the validation set using 2 eigenfaces was: 75.0

Performance on the validation set using 5 eigenfaces was: 73.75

Performance on the validation set using 10 eigenfaces was: 85.0

Performance on the validation set using 20 eigenfaces was: 88.75

Performance on the validation set using 50 eigenfaces was: 92.5

Performance on the validation set using 80 eigenfaces was: 91.25

Performance on the validation set using 100 eigenfaces was: 92.5

Performance on the test set using 150 eigenfaces was: 62.5

**p3.py:**

```
from pylab import *
5 import numpy as np
import random
import matplotlib.cbook as cbook
import random
import time
10 from scipy.misc import imread
from scipy.misc import imsave
from scipy.misc import imresize
import matplotlib.image as mpimg
import matplotlib.pyplot as plt
15 import os

matplotlib
gray()

20 os.chdir('C:/Users/Ramaneek/SkyDrive/Documents/Github/CSC320-Winter-2014/project 3/')

#global variables
act = ['Aaron Eckhart', 'Adam Sandler', 'Adrien Brody', 'Andrea Anders', 'Ashley
      Benson', 'Christina Applegate', 'Dianna Agron', 'Gillian Anderson']
25 training_set = np.zeros((len(act)*100, 32*32)) - 1 #need to use zeros()!!
validation_set = np.zeros((len(act)*10, 32*32)) - 1 #32*32 since a 32x32 matrix is
      flattened
test_set = np.zeros((len(act)*10, 32*32)) - 1

#fills in the global variable data
30 def getData():
    count_tr = 0 #training
    count_va = 0 #validation
    count_te = 0 #testing
    k = 135 #every actor has at least 135 pics
35
    for a in act:
        name = a.split()[1].lower()
        for i in range(k):
            if i == k:
40                 print "You might want to get more images or lower the training set amount"
                exit()

            if os.path.isfile("cropped/"+name+str(i)+".jpg"):
                #print "JPG"
45                 img = imread("cropped/"+name+str(i)+".jpg")
            elif os.path.isfile("cropped/"+name+str(i)+".png"):
                #print "PNG"
                img = imread("cropped/"+name+str(i)+".png")
            else: #couldn't open this image
```

```

50         #print "trying next image"
        continue

        #need to convert img to gray scale
        gray_img = 0.299*img[:, :, 0] + 0.587*img[:, :, 1] + 0.114*img[:, :, 2]
55         #if training_set[(act.index(a)+1)*100,-1] == -1: #get 100 training images
            for this actor
        if count_tr < (act.index(a)+1) * 100:
            training_set[count_tr][:] = gray_img.flatten()
            count_tr += 1
        elif count_va < (act.index(a)+1) * 10: #get 10 validation images for this
            actor
60         validation_set[count_va][:] = gray_img.flatten()
            count_va += 1
        elif count_te < (act.index(a)+1) * 10: #get 10 test images for this actor
            test_set[count_te][:] = gray_img.flatten()
            count_te += 1
65         else: #got all the pictures needed for this actor, move on to next actor
            break
        print "Training, validation, and test sets created!"

def pca(X):
70     """ Principal Component Analysis
        input: X, matrix with training data stored as flattened arrays in rows
        return: projection matrix (with important dimensions first), variance and mean.
        From: Jan Erik Solem, Programming Computer Vision with Python
        #http://programmingcomputervision.com/
75     """

    # get dimensions
    num_data,dim = X.shape

    # center data
80     mean_X = X.mean(axis=0)
    X = X - mean_X

    if dim>num_data:
85         # PCA - compact trick used
        M = dot(X,X.T) # covariance matrix
        e,EV = linalg.eigh(M) # eigenvalues and eigenvectors
        tmp = dot(X.T,EV).T # this is the compact trick
        V = tmp[::-1] # reverse since last eigenvectors are the ones we want
90         S = sqrt(e)[:,::-1] # reverse since eigenvalues are in increasing order
        for i in range(V.shape[1]):
            V[:,i] /= S
    else:
        # PCA - SVD used
95         U,S,V = linalg.svd(X)
        V = V[:num_data] # only makes sense to return the first num_data

    # return the projection matrix, the variance and the mean
    print "Projection matrix and mean eigenface computed!"
100    return V,S,mean_X

```



```

def ssd(x, y):
    return sum((x.flatten().astype(float)-y.flatten().astype(float))**2)

105 def display_save_25_comps(V, im_shape):
    '''Display 25 components in V'''
    figure()
    for i in range(25):
        plt.subplot(5, 5, i+1)
110     plt.axis('off')
        gray()
        imshow(V[i,:].reshape(im_shape))
    savefig('report/display_save_25_comps.jpg')
    show()

115 #####

getData() #get the data from the cropped images and fill in some of the global
         variables

120 projection_M, variance, mean_img = pca(training_set) #we actually don't even need the
         variance in this project
display_save_25_comps(projection_M, (32,32))
average_face = np.reshape(mean_img, (32,32)) #keep this to show later on in the report
imsave("report/mean_face.jpg", average_face)

125 validation_settings = [2, 5, 10, 20, 50, 80, 100, 150, 200] # the top k eigenfaces

#Validation set
max_performance = 0 #for determining the most correct setting for how many eigenfaces
                    to use
best_setting = 0
130 num_incorrect = 0 #for displaying incorrect matches for reporting purposes

for setting in validation_settings:
    num_correct = 0
    for i in range(validation_set.shape[0]):

135         #project the validation image on to the eigenface space
        val_proj_img = np.dot(projection_M[:setting], validation_set[i] - mean_img)
        validation_name_index = int(i/10)

140         #compute the closest projected training face to identify validation_set[i]
        min = Infinity
        for j in range(training_set.shape[0]):
            train_proj_img = np.dot(projection_M[:setting], training_set[j] - mean_img)
            ssd_value = ssd(val_proj_img, train_proj_img)

145             if ssd_value < min:
                #print "found new min:", int(j/100)
                min = ssd_value
                #the index of the actor's name
                training_name_index = int(j/100)

150             raw_training_name_index = j

```

```

    #print ssd_value
    #print validation_name, training_name
    if validation_name_index == training_name_index:
155         num_correct += 1
        #print "\t found match"
    elif num_incorrect < 5:
        num_incorrect += 1
        print "This match was incorrect, validation image", i, "incorrectly matched
            actor", validation_name_index, "with ", training_name_index
160         imsave("report/"+str(num_incorrect)+"_"+act[validation_name_index]+"_incorrectly_matched.jpg",
            np.reshape(validation_set[i], (32,32)))
        imsave("report/"+str(num_incorrect)+"_"+act[training_name_index]+"_with.jpg",
            np.reshape(training_set[raw_training_name_index], (32,32)))

performance = num_correct * 100.0 / validation_set.shape[0]
165 print "Performance on the validation set using", setting, "eigenfaces was: ",
    performance
    if max_performance < performance:
        max_performance = performance
        best_setting = setting
    print "Best setting for face recognition is to use", best_setting, "eigenfaces"
170
#test set
num_correct = 0
for i in range(test_set.shape[0]):
    test_proj_img = np.dot(projection_M[:best_setting], test_set[i] - mean_img)
175     test_name_index = int(i/10)

    #compute the closest projected training face to identify validation_set[i]
    min = Infinity
    for j in range(training_set.shape[0]):
180         train_proj_img = np.dot(projection_M[:best_setting], training_set[j] - mean_img)
        ssd_value = ssd(test_proj_img, train_proj_img)
        if ssd_value < min:
            min = ssd_value
            #the index of the actor's name
185             training_name_index = int(j/100)

    if test_name_index == training_name_index:
        num_correct += 1

190 performance = num_correct * 100.0 / test_set.shape[0]
print "Performance on the test set using", best_setting, "eigenfaces was: ",
    performance

# Part 3 is done

195
## Part 4 begin
print "\n\n\nNOW TESTING GENDER RECOGNITION\n\n\n"

```

```

#Validation set
200 max_performance = 0 #for determining the most correct setting for how many eigenfaces
    to use
best_setting = 0
num_incorrect = 0 #for displaying incorrect matches for reporting purposes

for setting in validation_settings:
205     num_correct = 0
    for i in range(validation_set.shape[0]):
        #project the validation image on to the eigenface space
        val_proj_img = np.dot(projection_M[:setting], validation_set[i] - mean_img)
        validation_name_index = int(i/10)

210         #compute the closest projected training face to identify validation_set[i]
        min = Infinity
        for j in range(training_set.shape[0]):
            train_proj_img = np.dot(projection_M[:setting], training_set[j] - mean_img)
215             ssd_value = ssd(val_proj_img, train_proj_img)
            if ssd_value < min:
                min = ssd_value
                training_name_index = int(j/100)

220         if validation_name_index >= 3 and training_name_index >= 3: #both female
            num_correct += 1
            #print "found match"
        elif validation_name_index < 3 and training_name_index < 3: #both male
            num_correct += 1
225         elif num_incorrect < 5:
            num_incorrect += 1
            #print "This match was incorrect, validation image", i, "incorrectly matched
                actor", validation_name_index, "with ", training_name_index
            #imsave(str(num_incorrect)+act[validation_name_index]+"gender_incorrectly_matched_with"+act

230
        performance = num_correct * 100.0 / validation_set.shape[0]
        print "Performance on the validation set using", setting, "eigenfaces was: ",
            performance
        if max_performance < performance:
            max_performance = performance
235         best_setting = setting

#test set
num_correct = 0
for i in range(test_set.shape[0]):
240     test_proj_img = np.dot(projection_M[:best_setting], test_set[i] - mean_img)
    validation_name_index = int(i/10)
    #compute the closest projected training face to identify validation_set[i]
    min = Infinity
    for j in range(training_set.shape[0]):
245         train_proj_img = np.dot(projection_M[:best_setting], training_set[j] - mean_img)
        ssd_value = ssd(test_proj_img, train_proj_img)
        if ssd_value < min:
            min = ssd_value

```

```
        training_name = int(j/100)

250     if validation_name_index >= 3 and training_name_index >= 3: #both female
        num_correct += 1
        #print "found match"
    elif validation_name_index < 3 and training_name_index < 3: #both male
255         num_correct += 1

performance = num_correct * 100.0 / test_set.shape[0]
print "Performance on the test set using", best_setting, "eigenfaces was: ",
    performance

260 # End part 4
```

## get data.py

```
#for downloading the image data
from pylab import *
import numpy as np
import matplotlib.pyplot as plt
5 import matplotlib.cbook as cbook
import random
import time
from scipy.misc import imread
from scipy.misc import imresize
10 import matplotlib.image as mpimg
import os
from scipy.ndimage import filters
import urllib

15

os.chdir('C:/Users/Ramaneek/SkyDrive/Documents/Github/CSC320-Winter-2014/project 3/')

act = ['Aaron Eckhart', 'Adam Sandler', 'Adrien Brody', 'Andrea Anders', 'Ashley
    Benson', 'Christina Applegate', 'Dianna Agron', 'Gillian Anderson']

20

def timeout(func, args=(), kwargs={}, timeout_duration=0.5, default=None):
25     '''From:
    http://code.activestate.com/recipes/473878-timeout-function-using-threading/'''
    import threading
    class InterruptableThread(threading.Thread):
        def __init__(self):
30             threading.Thread.__init__(self)
            self.result = None

        def run(self):
            try:
```

```
35         self.result = func(*args, **kwargs)
        except:
            self.result = default

        it = InterruptableThread()
40        it.start()
        it.join(timeout_duration)
        if it.isAlive():
            return False
        else:
45            return it.result

testfile = urllib.URLopener()

50 #Note: you need to create the uncropped folder first in order
    #for this to work

for a in act:
    name = a.split()[1].lower()
55    i = 0
    for line in open("faces_subset.txt"):
        if a in line:
            # line.split()[-2] gives the x1,y1,x2,y2 coordinates for specific image in
                this line
            coordinates = line.split()[-2].split(',') #an array of x1,y1,x2,y2 coordinates
60
            #filename = actorname + number of pics of this actor + filename
            filename = name+str(i)+'.'+line.split()[4].split('.')[1]
            #A version without timeout (uncomment in case you need to
                #unsupress exceptions, which timeout() does)
65            #testfile.retrieve(line.split()[4], "uncropped/"+filename)
            #timeout is used to stop downloading images which take too long to download
            timeout(testfile.retrieve, (line.split()[4], "uncropped/"+filename), {}, 30)
            print "saved original file:" + filename

70        if os.path.isfile("uncropped/"+filename): #if the image has been saved
            #need to open this image and crop it
            #no point converting it to grayscale here since imsave saves it as a 3D
                array

            #check to see if image can be read
75            try:
                image_cropped = imread("uncropped/"+filename)
                image_cropped = image_cropped[int(coordinates[1]):int(coordinates[3]) ,
                    int(coordinates[0]):int(coordinates[2])]
                #now lets resize image to 32x32
                image_cropped = imresize(image_cropped, (32,32))
80                imsave("cropped/"+filename, image_cropped)
                print "saved cropped file: " + filename
            except Exception:
                print "cant open file: " + filename
                continue
```

```
85         else:
            print "CONTINUING!!!!!!!!!!!!!!!!!!!!!!"
            continue

90     i += 1
```