CS559-B HW2

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Problem 1:

(1) Yes, the positive and negative classes are linearly separable, because there can be a line that separates the points (1,1) from the points (0,0) (1,0) (0,1).

(2) Supposed $\eta = 1$, Xb = 1. And if $f(x) \le 0$, f(x) = 0, else f(x) = 1.

| xb | x1 | x2 | y |
|----|-----------|-----------|---|
| 1 | 0 | 1 | 0 |
| 1 | 1 | 1 | 1 |
| 1 | 1 | 0 | 0 |
| 1 | 0 | 0 | 0 |

According to x1 + x2 - 1/2 = 0, w = (wb, w1, w2) = (-1/2, 1, 1).

STEP 1:

①
$$x = (1, 0, 1), w = (-1/2, 1, 1), y = 0$$

$$f(x) = -1/2 + 0 + 1 = 1/2 > 0$$
, $f(x) = 1$

$$\triangle w = \eta * (y - f(x)) * x = (-1, 0, -1)$$

$$(2)$$
 x = (1, 1, 1), w = (-1/2, 1, 1), y = 1

$$f(x) = -1/2 + 1 + 1 = 3/2 > 0, f(x) = 1$$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

(3)
$$x = (1, 1, 0), w = (-1/2, 1, 1), y = 0$$

$$f(x) = -1/2 + 1 + 0 = 1/2 > 0, f(x) = 1$$

$$\triangle w = \eta * (y - f(x)) * x = (-1, -1, 0)$$

$$\textcircled{4}$$
 $x = (1, 0, 0), w = (-1/2, 1, 1), y = 0$

$$f(x) = -1/2 + 0 + 0 = -1/2 < 0, f(x) = 0$$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

$$SUM(\triangle w) = (-2,-1,-1), w = (-5/2, 0, 0)$$

STEP 2:

①
$$\mathbf{x} = (1, 0, 1), \mathbf{w} = (-5/2, 0, 0), \mathbf{y} = 0$$

 $\mathbf{f}(\mathbf{x}) = -5/2 + 0 + 0 = -5/2 < 0, \mathbf{f}(\mathbf{x}) = 0$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

2
$$x = (1, 1, 1), w = (-5/2, 0, 0), y = 1$$

$$f(x) = -5/2 + 0 + 0 = -5/2 < 0, f(x) = 0$$

$$\triangle w = \eta * (y - f(x)) * x = (1, 1, 1)$$

3
$$x = (1, 1, 0), w = (-5/2, 0, 0), y = 0$$

$$f(x) = -5/2 + 0 + 0 = -5/2 < 0, f(x) = 0$$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

$$(4)$$
 $x = (1, 0, 0), w = (-5/2, 0, 0), y = 0$

$$f(x) = -5/2 + 0 + 0 = -5/2 < 0, f(x) = 0$$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

$$SUM(\triangle w) = (1, 1, 1), w = (-3/2, 1, 1)$$

STEP 3:

①
$$x = (1, 0, 1), w = (-3/2, 1, 1), y = 0$$

$$f(x) = -3/2 + 0 + 1 = -1/2 < 0, f(x) = 0$$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

(2)
$$x = (1, 1, 1), w = (-3/2, 1, 1), y = 1$$

$$f(x) = -3/2 + 1 + 1 = 1/2 > 0, f(x) = 1$$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

$$3$$
 $x = (1, 1, 0), w = (-3/2, 1, 1), y = 0$

$$f(x) = -3/2 + 1 + 0 = -1/2 < 0, f(x) = 0$$

$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

$$(4)$$
 $x = (1, 0, 0), w = (-3/2, 1, 1), y = 0$

$$f(x) = -3/2 + 0 + 0 = -3/2 < 0, f(x) = 0$$

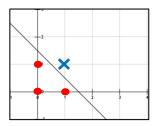
$$\triangle w = \eta * (y - f(x)) * x = (0, 0, 0)$$

$$SUM(\triangle w) = (0, 0, 0)$$

STOP

$$\therefore$$
 w = (-3/2, 1, 1)

: Boundary:
$$-3/2+X1+X2=0$$



Problem 2:

(1) PCA Code:

```
def doPCA(dataMat, K):
    N, D = dataMat.shape
    mean = np.mean(dataMat, axis=0)
    minus_mean = dataMat - mean
    covMultiplyX = np.dot(minus_mean,np.transpose(minus_mean)) * (1 / N)
    eigenVal, eigenVectMultiplyX = np.linalg.eig(covMultiplyX)
    eigenVect = np.dot(np.transpose(minus_mean), eigenVectMultiplyX)
    #print(eigenVal.shape, eigenVectMultiplyX.shape, eigenVect.shape)
    eigenValRank = np.argsort(eigenVal)
    eigenValRank = eigenValRank[:-(K+1):-1]
    # print("val",eigenVal.size,"vec",eigenVect.shape)
    eigenVect_norm = eigenVect / np.linalg.norm(eigenVect, axis=0)
    eigenVectByK = eigenVect[:, eigenValRank]
    eigenVectByK_norm = eigenVect_norm[:, eigenValRank]
```

Use OpenCV to read, process, and visualize images. Numpy is used to calculate the average face, obtain the eigenvalues and eigenvectors and unitization.

Eigenfaces (K = 30):



From the top left to the bottom right are the 10 eigen faces corresponding to the largest 10 eigen values.

(2) Reconstruction Code:

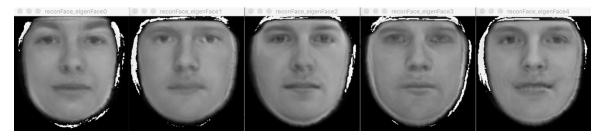
```
def reconstructTestingImages(test_dataMat, eigenVect , meanface):
    N = test_dataMat.shape[0]
    test_minus_mean = test_dataMat - meanface
    reconSystem = np.dot(test_minus_mean, eigenVect)
    reconFace = meanface + np.dot(reconSystem, np.transpose(eigenVect))

Error = []
    row, col = reconFace.shape
    for i in range(row):
        sub = (reconFace[i,:]-test_dataMat[i,:]) ** 2
        error_sum = np.sum(sub)
        Error.append(error_sum)
    reconstructionError = np.sum(Error) / N
    print("reconstructionError: ",reconstructionError)

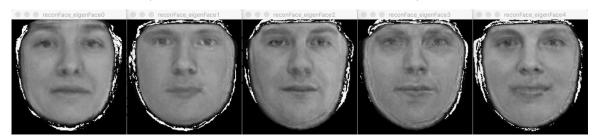
return reconFace, Error, reconstructionError
```

Project the centralized test sets onto the eigen faces to get the reconstruction faces and reconstruction error.

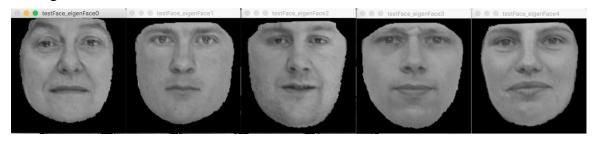
Reconstructed Faces (K=30, Reconstruction error=): 15581741):



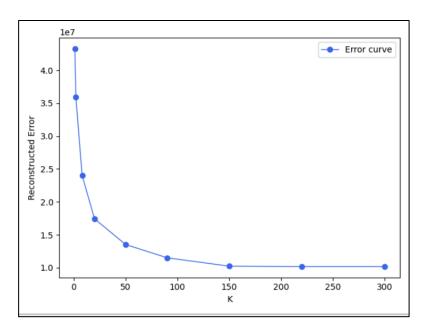
Reconstructed Faces (K=150, Reconstruction error = 10232190):



Testing Faces:

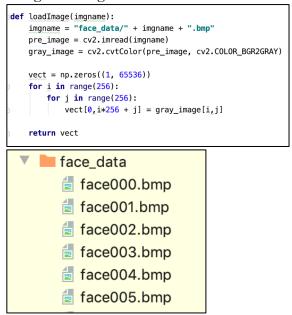


(3) Reconstruction Errors Curve:



The x-axis is K values, and the y-axis is reconstruction error.

Image reading instruction:



The original images needs to be placed in the face data folder in the project directory.

Source code:

```
from PIL import Image
import cv2
import numpy as np
import matplotlib.pyplot as plt
def loadImage(imgname):
   imgname = "face_data/" + imgname + ".bmp"
   pre_image = cv2.imread(imgname)
   gray_image = cv2.cvtColor(pre_image, cv2.COLOR_BGR2GRAY)
   vect = np.zeros((1, 65536))
   for i in range(256):
      for j in range(256):
          vect[0,i*256 + j] = gray_image[i,j]
   return vect
def doPCA(dataMat, K):
   N, D = dataMat.shape
   mean = np.mean(dataMat, axis=0)
   minus_mean = dataMat - mean
   covMultiplyX = np.dot(minus_mean,np.transpose(minus_mean)) * (1 / N)
   eigenVal, eigenVectMultiplyX = np.linalg.eig(covMultiplyX)
   eigenVect = np.dot(np.transpose(minus_mean), eigenVectMultiplyX)
   #print(eigenVal.shape, eigenVectMultiplyX.shape, eigenVect.shape)
   eigenValRank = np.argsort(eigenVal)
   eigenValRank = eigenValRank[:-(K+1):-1]
   # print("val",eigenVal.size,"vec",eigenVect.shape)
   eigenVect_norm = eigenVect / np.linalg.norm(eigenVect, axis=0)
   eigenVectByK = eigenVect[:, eigenValRank]
   eigenVectByK_norm = eigenVect_norm[:, eigenValRank]
   return eigenVectByK, eigenVectByK_norm, mean
def visualizeEigenFace(eigenVect, times):
   row,col = eigenVect.shape
   time = 0
   eigenFaces = []
```

```
imgnames = []
   while time < times:</pre>
      imgname = "eigenFace"+str(time)
      eigenFace = np.zeros((256, 256))
      for i in range (256):
          for j in range(256):
             eigenFace[i,j] = int(eigenVect[i*256+j, time])
      eigenFace = eigenFace.astype('uint8')
      # row, col = eigenFace.shape
      # for i in range(row):
            for j in range(col):
               print(eigenFace[i, j])
      eigenFaces.append(eigenFace)
      imgnames.append(imgname)
      time = time + 1
   return imgnames,eigenFaces
def reconstructTestingImages(test_dataMat, eigenVect , meanface):
   N = test_dataMat.shape[0]
   test_minus_mean = test_dataMat - meanface
   reconSystem = np.dot(test_minus_mean, eigenVect)
   reconFace = meanface + np.dot(reconSystem, np.transpose(eigenVect))
   Error = []
   row, col = reconFace.shape
   for i in range(row):
      sub = (reconFace[i,:]-test_dataMat[i,:]) ** 2
      error_sum = np.sum(sub)
      Error.append(error_sum)
   reconstructionError = np.sum(Error) / N
   #print("reconstructionError: ",reconstructionError)
   return reconFace, Error, reconstructionError
def drawingErrorsCurve(training_dataMat, test_dataMat):
   klist = [1,2,8,20,50,90,150,220,300]
   reconErrorList = []
   for K in range(np.size(klist)):
      k = klist[K]
```

```
eigenface, eigenVect, meanface = doPCA(training_dataMat, k)
      reconFace, errorList, reconError = reconstructTestingImages(test_dataMat,
eigenVect, meanface)
      print(reconError)
      reconErrorList.append(reconError)
   plt.plot(klist, reconErrorList, 'ro-', color='#4169E1', linewidth=1,
label='Error curve')
   plt.legend(loc="upper right")
   plt.xlabel('K')
   plt.ylabel('Reconstructed Error')
   plt.show()
if name == " main ":
   # Data preprocessing
   traning_num = 157
   test_num = 20
   # training data
   training_dataMat = np.zeros((traning_num, 65536))
   for i in range(traning_num):
      if i != 103: # NO.103 picture is missing
          num = str(i).zfill(3)
          filename = "face"+num
          vect = loadImage(filename)
          training_dataMat[i,]=vect
   # 157 X 65536
   print(training_dataMat.shape)
   # Doing PCA and return eigenVects
   eigenface, eigenVect, meanface = doPCA(training_dataMat, 30)
   # Visualize the number of 10 eigen faces
   eigennames, eigenFaces = visualizeEigenFace(eigenface, 10)
   labelname = "eigeenFace_"
   for e in range(10):
      eigennames[e] = labelname + eigennames[e]
      cv2.imshow(eigennames[e], eigenFaces[e])
   # obtaining test data
   test dataMat = np.zeros((test num, 65536))
   for i in range(test_num):
```

```
num = str(i+traning_num).zfill(3)
      filename = "face"+num
      vect = loadImage(filename)
      test_dataMat[i,]=vect
   # 20 X 65536
   print(test_dataMat.shape)
   # Reconstruct testing images
   reconFace, errorList, reconError =
reconstructTestingImages(test_dataMat,eigenVect,meanface)
   print(reconFace.shape)
   # Visualizing test faces
   testFaceNum = 5
   testnames, reconFaces = visualizeEigenFace(np.transpose(test dataMat),
testFaceNum)
   labelname = "testFace_"
   for t in range(testFaceNum):
      testnames[t] = labelname + testnames[t]
      cv2.imshow(testnames[t], reconFaces[t])
   # Visualizing reconstucted faces
   reconFaceNum = 5
   reconnames,reconFaces =
visualizeEigenFace(np.transpose(reconFace),reconFaceNum)
   labelname = "reconFace "
   for r in range(reconFaceNum):
      reconnames[r] = labelname + reconnames[r]
      cv2.imshow(reconnames[r], reconFaces[r])
   # Visualizing reconstucted error
   print("Reconstucted error(K=30):", reconError)
   # Visualizing mean face
   mean_eigenFace = np.zeros((256, 256))
   for i in range(256):
      for j in range(256):
          mean_eigenFace[i,j] = int(meanface[i*256+j])
   mean_eigenFace = mean_eigenFace.astype('uint8')
   # cv2.imshow("meanface", mean_eigenFace)
   # Drawing Errors Curve
   drawingErrorsCurve(training_dataMat,test_dataMat)
```

cv2.waitKey()
cv2.destroyAllWindows()