

hw2-Q1

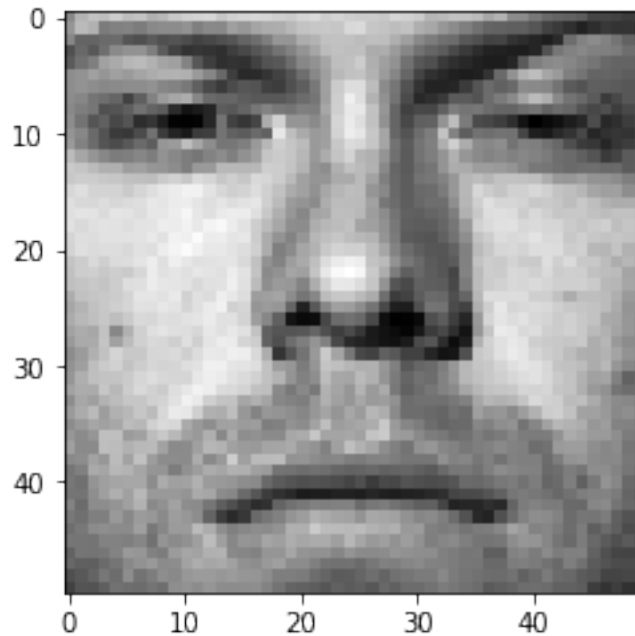
October 10, 2019

```
[12]: import numpy as np
      from scipy import misc
      from matplotlib import pylab as plt
      import matplotlib.cm as cm
      import imageio
      %matplotlib inline
```

```
[13]: #(b)
      #loading data from the training data set
      #https://stackoverflow.com/questions/15345790/
      ↪scipy-misc-module-has-no-attribute-imread
      train_labels, train_data = [], []
      for line in open('./faces/train.txt'):
          im = imageio.imread(line.strip().split()[0])
          train_data.append(im.reshape(2500,))
          train_labels.append(line.strip().split()[1])
      train_data, train_labels = np.array(train_data, dtype=float), np.
      ↪array(train_labels, dtype=int)

      print(train_data.shape, train_labels.shape)
      plt.imshow(train_data[10, :].reshape(50,50), cmap = cm.Greys_r)
      plt.show()
      print(train_data)
```

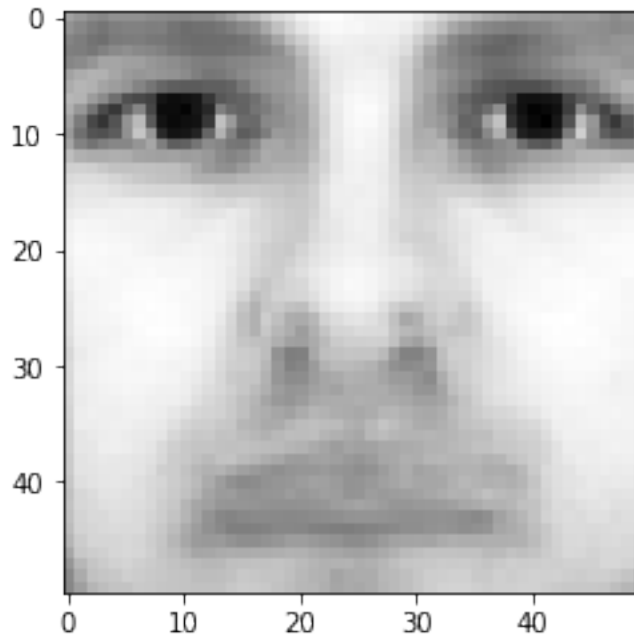
(540, 2500) (540,)



```
[[ 5.  7.  7. ... 13. 20. 31.]
 [65. 74. 74. ... 111. 106. 96.]
 [101. 116. 130. ... 61. 52. 41.]
 ...
 [107. 115. 108. ... 26. 23. 24.]
 [255. 255. 255. ... 22. 22. 22.]
 [174. 178. 202. ... 28. 26. 28.]]
```

```
[14]: # (c)
#Average Face
#Compute the average face from the whole training set
#by summing up every row in X then dividing by the number of faces
average_face = np.sum(train_data, axis = 0) /train_labels.shape[0]
print(average_face)
#Display the average face as a grayscale image.
plt.imshow(average_face.reshape(50,50), cmap = cm.Greys_r)
plt.show()
```

```
[59.25185185 56.10185185 52.42222222 ... 67.22222222 64.61851852
 59.27592593]
```

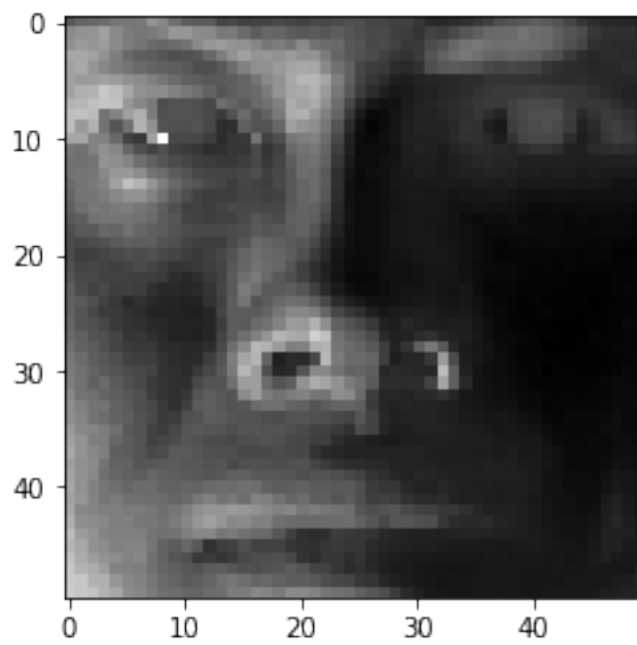
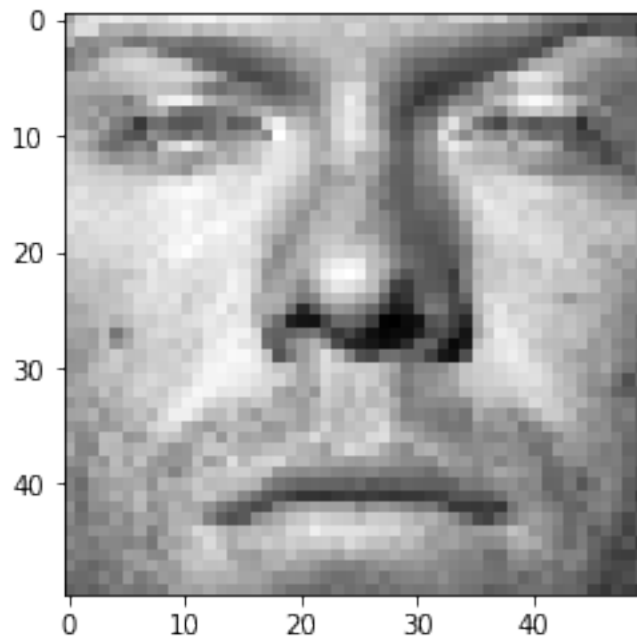


```
[15]: #(d)
#Mean Subtraction.
#Subtract average face  $\mu$  from every row in X.
sub_avg_face = train_data - average_face

#Pick a face image after mean subtraction from the new X and display that image
→ in grayscale.
plt.imshow(sub_avg_face[10, :].reshape(50,50), cmap = cm.Greys_r)
plt.show()

#Do the same thing for the test set Xtest using the pre-computed average face  $\mu$ 
→ in (c).
#loading data from the testing data set
test_labels, test_data = [], []
for line in open('./faces/test.txt'):
    im = imageio.imread(line.strip().split()[0])
    test_data.append(im.reshape(2500,))
    test_labels.append(line.strip().split()[1])
test_data, test_labels = np.array(test_data, dtype=float), np.
→ array(test_labels, dtype=int)

#Subtract average face  $\mu$  from every row in Xtest.
sub_avg_face_test = test_data - average_face
plt.imshow(sub_avg_face_test[10, :].reshape(50,50), cmap = cm.Greys_r)
plt.show()
```



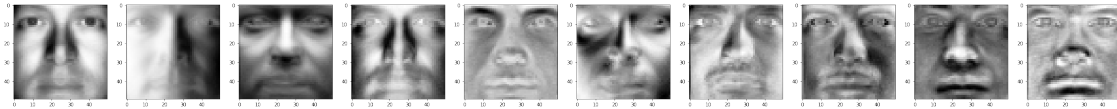
```
[17]: #(e)
      #Eigenface.
      #Perform Singular Value Decomposition (SVD) on training set X ( $X = U\Sigma V^T$ ) to get  $U$ 
      ↪ matrix  $V^T$ 
```

```

u, s, vt = np.linalg.svd(sub_avg_face)
s = np.diag(s)
#create 10 subplots and set appropriate size
fig, ax = plt.subplots(nrows=1, ncols=10)
fig.set_size_inches(40,40)

#We refer to  $v_i$ , the  $i$ -th row of  $V^T$ , as  $i$ -th eigenface.
#Display the first 10 eigenfaces as 10 images in grayscale
for i in range(10):
    ax[i].imshow(vt[i, :].reshape(50,50), cmap = cm.Greys_r)
plt.show()

```



```

[7]: #(f)
#Low-rank Approximation.
#we can approximate  $X$  by  $\hat{X} = U[:, :r] \Sigma[:, :r] V^T[:, :r]$ .
#The matrix  $\hat{X}$  is called rank- $r$  approximation of  $X$ .

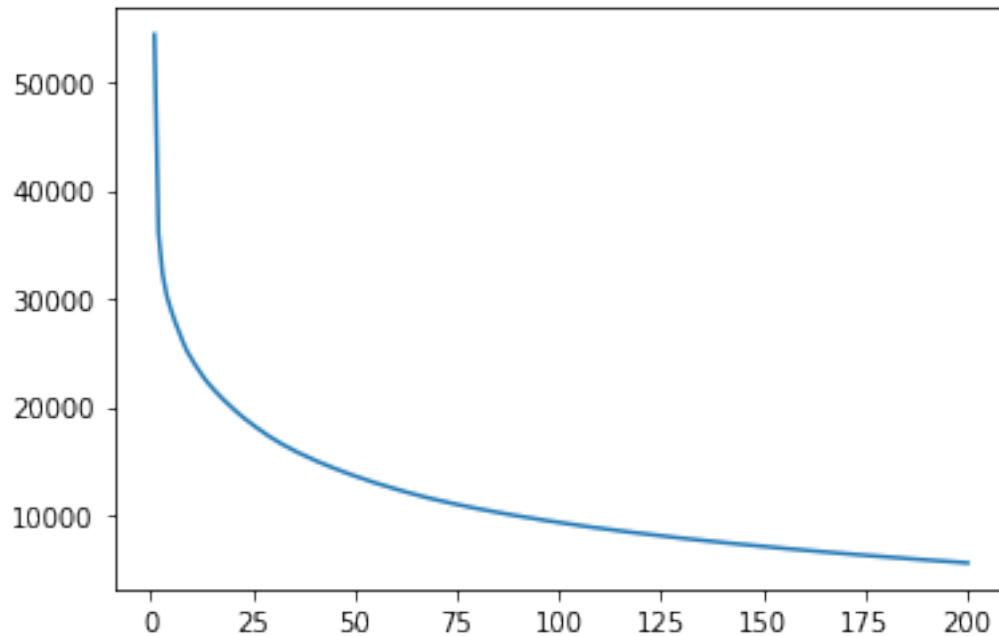
err = []
for r in range(1,201):
    temp = u[:, :r].dot(s[:, :r])
    Xr = temp.dot(vt[:, :r])
    err_fnorm = np.linalg.norm(sub_avg_face - Xr, ord='fro')
    err.append(err_fnorm)
#Plot the rank- $r$  approximation error  $\|X - \hat{X}\|_F$  as a function of  $r$  when  $r = 1, \dots, 200$ .
plt.plot(range(1,201),err)

```

```

[7]: [<matplotlib.lines.Line2D at 0x111425e10>]

```



```
[8]: #(g)
#Eigenface Feature.
#Write a function to generate r-dimensional feature matrix F and Ftest for
#→ training images X and test images Xtest, respectively
#(to get F, multiply X to the transpose of first r rows of VT
#F should have same number of rows as X and r columns

def train_featurematrix_generator(r):
    #multiply X to the transpose of first r rows of VT
    F = np.dot(sub_avg_face, vt[:r,:].T)
    return F

def test_featurematrix_generator(r):
    #multiply X to the transpose of first r rows of VT
    Ftest = np.dot(sub_avg_face_test, vt[:r,:].T)
    return Ftest

F = train_featurematrix_generator(10)
Ftest = test_featurematrix_generator(10)
print(F.shape)
print(Ftest.shape)
```

```
(540, 10)
```

```
(100, 10)
```

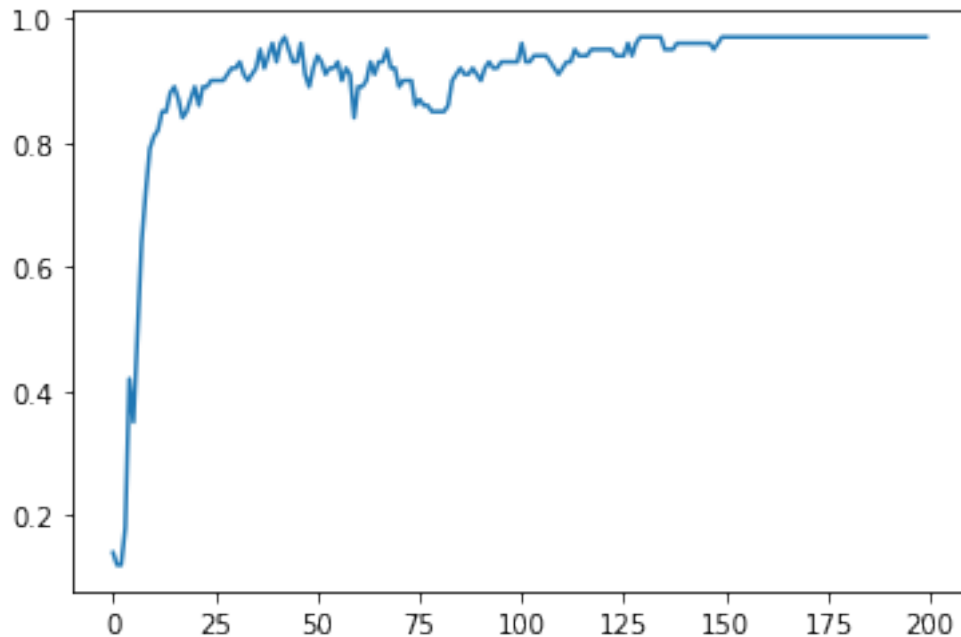
```
[9]: from sklearn.linear_model import LogisticRegression
```

```
[10]: # import warnings filter
import warnings
# ignore all future warnings
warnings.simplefilter('ignore')
```

```
[11]: #(h)
#Face Recognition.
#Extract training and test features for  $r = 10$ .
#Train a Logistic Regression model using  $F$  and test on  $F_{test}$ .
#Report the classification accuracy on the test set.
#Plot the classification accuracy on the test set as a function of  $r$  when  $r = \underline{\hspace{1cm}}$ 
 $\rightarrow 1, 2, \dots, 200$ .
accuracy = []
logreg = LogisticRegression(multi_class='ovr')

for r in range(1, 201):
    F = train_featurematrix_generator(r)
    Ftest = test_featurematrix_generator(r)
    model = logreg.fit(F, train_labels)
    pred = model.predict(Ftest)
    score = model.score(Ftest, test_labels)
    accuracy.append(score)

plt.plot(accuracy)
plt.show()
```



```
[13]: F = train_featurematrix_generator(10)
      Ftest = test_featurematrix_generator(10)
      model = LogisticRegression().fit(F, train_labels)
      pred = model.predict(Ftest)
      score = model.score(Ftest, test_labels)
      print(score)
```

0.79

```
[ ]:
```