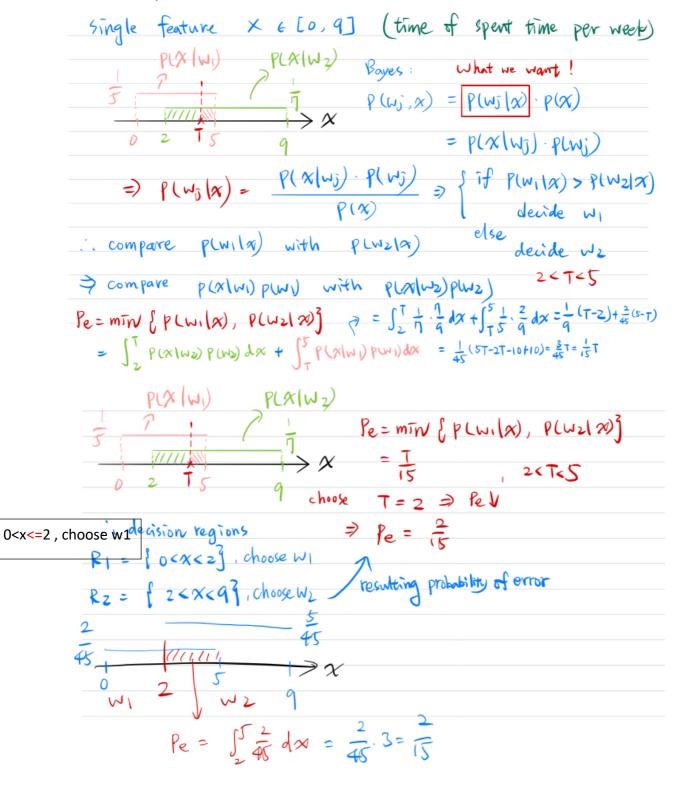
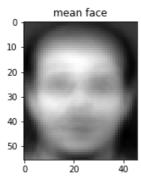
Problem 1: Bayes Decision Rule [1]



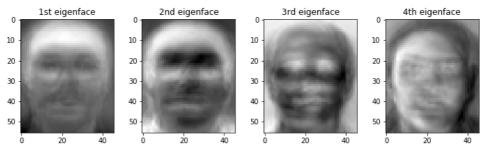
Problem 2: Principal Component Analysis [2]

1. Perform PCA on the training set. Plot the mean face and the first four eigenfaces. use

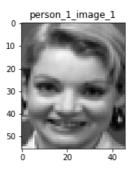
from sklearn.preprocessing import StandardScaler to compute the mean face



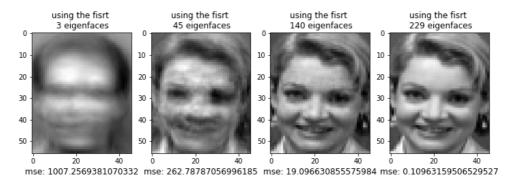
use from sklearn.decomposition import PCA to form eigenfaces



2. and 3. Original image:



Reconstructed images:



Using the first n eigenfaces	3	45	140	229
MSE	1007.25	262.78	19.09	0.10

Table 1: Using the first n eigenfaces to reconstruct the original images and the corresponding mean square error (MSE)

As we using more eigenfaces to reconstruct the original image, less mean square error is obtained.

4. apply the k-nearest neighbors algorithm to classify the testing set images I use the sklearn package to do 3-fold cross-vaildation for me.

By setting the cv parameter to 3 in

cross_val_score(knn, train_knn, train_class, cv=3, scoring='accuracy')
from sklearn.cross_validation import cross_val_score

k n	3	45	140
1	0.7041	0.9291	0.9291
3	0.6166	0.8583	0.8583
5	0.5208	0.7916	0.7541

Table 2: Accuracy of different k and n pairs.

(k: the k of k-nearest neighbors, n: using the first n eigenfaces)

Although the highest accuracy obtained by k-nn is the same when (n,k) = (45,1) and (140,1), I choose (n,k) = (45,1) for reducing to a lower dimension.

5. Choose (n,k) = (45,1)

	precision	recall	f1-score	support
1	1.00	1.00	1.00	4
2	1.00	1.00	1.00	4
3	1.00	1.00	1.00	4
4	1.00	1.00	1.00	4
5	1.00	0.75	0.86	4
6	1.00	1.00	1.00	4
7	1.00	1.00	1.00	4
8	1.00	1.00	1.00	4
9	1.00	1.00	1.00	4
10	1.00	0.75	0.86	4
11	1.00	0.75	0.86	4
12	1.00	1.00	1.00	4
13	1.00	1.00	1.00	4
14	1.00	1.00	1.00	4
15	0.67	1.00	0.80	4
16	0.80	1.00	0.89	4
17	1.00	1.00	1.00	4
18	1.00	1.00	1.00	4
19	1.00	0.75	0.86	4
20	1.00	0.75	0.86	4
21	1.00	1.00	1.00	4
22	1.00	1.00	1.00	4
23	1.00	1.00	1.00	4
24	1.00	1.00	1.00	4
25	1.00	1.00	1.00	4
26	1.00	1.00	1.00	4
27	1.00	1.00	1.00	4
28	1.00	0.75	0.86	4
29	1.00	1.00	1.00	4
30	0.80	1.00	0.89	4
31	1.00	1.00	1.00	4
32	1.00	1.00	1.00	4
33	1.00	1.00	1.00	4
34	1.00	1.00	1.00	4
35	1.00	0.75	0.86	4
36	1.00	1.00	1.00	4
37	0.80	1.00	0.89	4
38	0.80	1.00	0.89	4
39	1.00	1.00	1.00	4
40	0.80	1.00	0.89	4
total	0.97	0.96	0.96	160

Figure 1: The recognition rate of the testing set

The average recognition rate achieved 97% showed in Figure 1.

1.

1.	
class	Three random patches
Banana	0 2 2 4 4 6 6 6 8 8 100 12 12 12 14 1 14 1 15 15 0 5 10 15
Fountain	0 - 2 - 2 - 4 - 4 - 6 - 8 - 10 - 10 - 12 - 12 - 12 - 12 - 14 - 14 - 15 - 0 5 10 15 0 5 10 15
Reef	0 - 2 - 2 - 4 - 4 - 4 - 4 - 4 - 4 - 4 - 4
Tractor	2 - 4 - 4 - 6 - 8 - 10 - 12 - 12 - 14 - 14 - 14 - 14 - 14 - 14

Two of the sets of patches can be classified to the true class by seeing three patches of each set (Banana and Reef). I can't classify the other two sets of patches (Fountain and Tractor).

2.

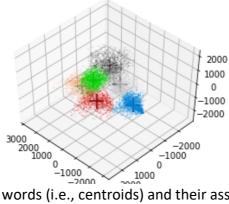
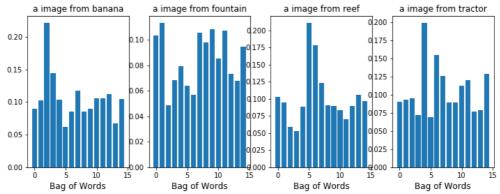


Figure 2: Bag of words (i.e., centroids) and their associated features (i.e., patches) in this PCA subspace.

3. Choose one image from each category and visualize its BoW using histogram plot.



4. Adopt the k-nearest neighbors classifier (k-NN) to perform classification on X_test using the above BoW features.

	Classification Accuracy
Banana	0.67
Fountain	0.35
Reef	0.56
Tractor	0.54
average	0.53

Table 3: The classification accuracy for each category and an average accuracy

Reference:

- [1] http://vllab.ee.ntu.edu.tw/uploads/1/1/1/6/111696467/dlcv w1.pdf
- [2] https://scikit-learn.org/stable/modules/generated/sklearn.decomposition.PCA.html
- [3] http://scikit-image.org/docs/dev/api/skimage.util.html