```
In []:
from zipfile import ZipFile
ZipFile("/content/Code&Data.zip").extractall("/content/") #/drive/MyDrive

In []:
from scipy.io import loadmat
D = loadmat("/content/Code&Data/PIE.mat")

In []:
Data = D['Data']
Labels = D['Label']
O1.a
```

# ₩ I.a

```
In [ ]:
```

```
import numpy as np
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy score
from sklearn.neighbors import KNeighborsClassifier
import time
train Acc = []
test Acc = []
running time = []
# number of samples per training set = [5,10,15]
for i in range (5,16,5):
   n sub = 21
   trainData = []
   testData = []
   trainLabels = []
   testLabels = []
   for j in range (1,68):
     trainData.append(Data[(j-1)*n sub:(j-1)*n sub+i])
     testData.append(Data[(j-1)*n sub+i:j*n sub])
     trainLabels.append(Labels[(j-1)*n sub:(j-1)*n sub+i])
     testLabels.append(Labels[(j-1)*n_sub+i:j*n_sub])
    # trainData, trainLabels of size number of samples * number of subjects
   trainData = np.array(trainData).reshape(len(trainData*i),900)
   testData = np.array(testData).reshape(len(testData*(n sub-i)),900)
   trainLabels = np.array(trainLabels).reshape(len(trainLabels)*trainLabels[0].shape[0]
, 1)
   testLabels = np.array(testLabels).reshape(len(testLabels)*testLabels[0].shape[0],1)
   start = time.time()
    # standardizing the dataset
   scaler = StandardScaler()
   data scaled train = scaler.fit transform(trainData)
   data scaled test = scaler.transform(testData)
   # calculating covariance matrix and eigenvectors and eigen values
   cov matrix = np.cov(data scaled train.T)
   values, vectors = np.linalg.eigh(cov matrix)
    # sorting based on eigen values
   eig pairs = [(np.abs(values[k]), vectors[:,k]) for k in range(len(values))]
   eig pairs = sorted(eig pairs, key=lambda k: k[0], reverse=True)
```

```
# transforming from deimension 900 to 100 using top eigen vectors
    pca_Transformed_train = data_scaled_train.dot(np.transpose(np.array([a[1] for a in e
ig_pairs[0:100]])))
    pca_Transformed_test = data_scaled_test.dot(np.transpose(np.array([a[1] for a in eig
_pairs[0:100]])))

# using knn classifier
knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(pca_Transformed_train, trainLabels)

# recording train and test accuracies
train_Acc.append(accuracy_score(trainLabels, knn.predict(pca_Transformed_train)))
test_Acc.append(accuracy_score(testLabels, knn.predict(pca_Transformed_test)))
end = time.time()

# recoding running time
running_time.append(end-start)
```

```
import pandas as pd
#summarizing train mse, test mse in a table
table = pd.DataFrame(list(zip(train_Acc,test_Acc,running_time)),columns = ["training mse
using PCA", "Testing mse using PCA", "running time"])
table
```

#### Out[]:

	training mse using PCA	Testing mse using PCA	running time
0	0.877612	0.434701	0.318196
1	0.943284	0.500678	0.392914
2	0.981095	0.975124	0.434237

#### In [ ]:

```
i = 5
mean vectors = []
sw = np.zeros((900,900))
train Acc lda = []
test_Acc_lda = []
running_time_lda = []
for i in range (5, 16, 5):
   n sub = 21
   trainData = []
   testData = []
   trainLabels = []
   testLabels = []
    for j in range (1,68):
     trainData.append(Data[(j-1)*n sub:(j-1)*n sub+i])
     mean vectors.append(np.mean(trainData[-1],axis = 0))
      # calculating with class scatter matrix sw
      sw+= (np.matrix(trainData[-1]-mean vectors[-1]).T).dot(trainData[-1]-mean vectors[
-1])
      testData.append(Data[(j-1)*n sub+i:j*n sub])
     trainLabels.append(Labels[(j-1)*n sub:(j-1)*n sub+i])
     testLabels.append(Labels[(j-1)*n_sub+i:j*n_sub])
    # trainData, trainLabels of size number of samples * number of subjects
    trainData = np.array(trainData).reshape(len(trainData*i),900)
    testData = np.array(testData).reshape(len(testData*(n sub-i)),900)
    trainLabels = np.array(trainLabels).reshape(len(trainLabels)*trainLabels[0].shape[0]
, 1)
    testLabels = np.array(testLabels).reshape(len(testLabels)*testLabels[0].shape[0],1)
```

```
start = time.time()
   #calculating between class scatter matrix sb
   sb = np.zeros((900,900))
   total mean = np.mean(trainData,axis = 0)
   for k in mean_vectors:
       sb+= (i)*((np.matrix(k-total mean).T).dot(np.matrix(k-total mean)))
    # calculating eigenvectors and eigen values
   values, vectors = np.linalg.eigh(np.linalg.inv(sw).dot(sb))
    # sorting based on eigen values
   eig pairs = [(np.abs(values[k]), vectors[:,k]) for k in range(len(values))]
   eig pairs = sorted(eig pairs, key=lambda k: k[0], reverse=True)
    # transforming from deimension 900 to 100 using top eigen vectors
   lda Transformed train = trainData.dot(np.transpose(np.array([a[1] for a in eig pairs
[0:100]])))
   lda Transformed test = testData.dot(np.transpose(np.array([a[1] for a in eig pairs[0
:100]])))
    # using knn classifier
   knn = KNeighborsClassifier(n neighbors=5)
   knn.fit(lda Transformed train, trainLabels)
    # recording train and test accuracies
   train Acc lda.append(accuracy score(trainLabels,knn.predict(lda Transformed train)))
   test Acc lda.append(accuracy score(testLabels,knn.predict(lda Transformed test)))
   end = time.time()
    # recoding running time
   running time lda.append(end-start)
```

```
#summarizing train mse, test mse in a table
table_lda = pd.DataFrame(list(zip(train_Acc_lda,test_Acc_lda,running_time_lda)),columns =
["training mse using LDA","Testing mse using LDA","running time"])
table_lda
```

#### Out[]:

	training mse using LDA	Testing mse using LDA	running time
0	0.853731	0.387127	2.024376
1	0.914925	0.459973	1.738356
2	0.981095	0.972637	2.418785

## In [ ]:

```
#summarizing lda test mse,pca test mse in a table
table_lda_pca = pd.DataFrame(list(zip(test_Acc_lda,test_Acc,running_time_lda,running_time
)),columns = ["Testing mse using LDA",

"Test mse using PCA",

"running time using lda",

"running time using PCA"])
table_lda_pca
```

## Out[]:

	Testing mse using LDA	Test mse using PCA	running time using Ida	running time using PCA
0	0.387127	0.434701	2.024376	0.318196
1	0.459973	0.500678	1.738356	0.392914
2	0.972637	0.975124	2.418785	0.434237

# **Q1.b**

```
In [ ]:
```

```
from sklearn.decomposition import PCA
from sklearn.discriminant analysis import LinearDiscriminantAnalysis
train Acc = []
test Acc = []
running_time = []
# number of samples per training set = [5,10,15]
for i in range (5,16,5):
   n sub = 21
   trainData = []
   testData = []
   trainLabels = []
   testLabels = []
   for j in range (1,68):
     trainData.append(Data[(j-1)*n_sub:(j-1)*n_sub+i])
     testData.append(Data[(j-1)*n_sub+i:j*n_sub])
     trainLabels.append(Labels[(j-1)*n sub:(j-1)*n sub+i])
     testLabels.append(Labels[(j-1)*n sub+i:j*n sub])
    # trainData, trainLabels of size number of samples * number of subjects
   trainData = np.array(trainData).reshape(len(trainData*i),900)
   testData = np.array(testData).reshape(len(testData*(n sub-i)),900)
   trainLabels = np.array(trainLabels).reshape(len(trainLabels)*trainLabels[0].shape[0]
,1)
   testLabels = np.array(testLabels).reshape(len(testLabels)*testLabels[0].shape[0],1)
   start = time.time()
   pca = PCA(n components=100)
   pca Transformed train = pca.fit transform(trainData)
   pca Transformed test = pca.transform(testData)
   knn = KNeighborsClassifier(n neighbors=5)
   knn.fit(pca Transformed train,trainLabels)
    # recording train and test accuracies
   train_Acc.append(accuracy_score(trainLabels,knn.predict(pca_Transformed train)))
   test Acc.append(accuracy score(testLabels,knn.predict(pca Transformed test)))
   end = time.time()
    # recoding running time
   running time.append(end-start)
```

## In [ ]:

```
table = pd.DataFrame(list(zip(train_Acc,test_Acc,running_time)),columns = ["training mse
using PCA", "Testing mse using PCA", "running time"])
table
```

#### Out[]:

	training mse using PCA	Testing mse using PCA	running time
0	0.907463	0.448694	0.259994
1	0.947761	0.516961	0.357522
2	0.986070	0.980100	0.362947

#### In [ ]:

```
train_Acc_lda = []
test_Acc_lda = []
running_time_lda = []
```

```
# number of samples per training set = [5,10,15]
for i in range (5, 16, 5):
   n sub = 21
   trainData = []
   testData = []
   trainLabels = []
   testLabels = []
   for j in range (1,68):
     trainData.append(Data[(j-1)*n sub:(j-1)*n sub+i])
     testData.append(Data[(j-1)*n sub+i:j*n sub])
     trainLabels.append(Labels[(j-1)*n sub:(j-1)*n sub+i])
     testLabels.append(Labels[(j-1)*n sub+i:j*n sub])
    # trainData, trainLabels of size number of samples * number of subjects
   trainData = np.array(trainData).reshape(len(trainData*i),900)
   testData = np.array(testData).reshape(len(testData*(n sub-i)),900)
   trainLabels = np.array(trainLabels).reshape(len(trainLabels)*trainLabels[0].shape[0]
, 1)
   testLabels = np.array(testLabels).reshape(len(testLabels)*testLabels[0].shape[0],1)
   start = time.time()
   lda = LinearDiscriminantAnalysis(n components = 66)
   lda Transformed train = lda.fit transform(trainData,trainLabels)
   lda Transformed test = lda.transform(testData)
   knn = KNeighborsClassifier(n neighbors=5)
   knn.fit(lda Transformed train, trainLabels)
    # recording train and test accuracies
   train Acc lda.append(accuracy score(trainLabels,knn.predict(lda Transformed train)))
   test Acc lda.append(accuracy score(testLabels,knn.predict(lda Transformed test)))
   end = time.time()
    # recoding running time
   running time lda.append(end-start)
```

```
table_lda = pd.DataFrame(list(zip(train_Acc_lda,test_Acc_lda,running_time_lda)),columns =
["training mse using LDA","Testing mse using LDA","running time"])
table_lda
```

# Out[]:

	training mse using LDA	Testing mse using LDA	running time
0	1.0	0.527985	0.326060
1	1.0	0.754410	0.612968
2	1.0	0.985075	1.014390

## In [ ]:

```
table_lda_pca = pd.DataFrame(list(zip(test_Acc_lda,test_Acc,running_time_lda,running_time
)),columns = ["Testing mse using LDA",

"Test mse using PCA",

"running time using lda",

"running time using PCA"])
table_lda_pca
```

#### Out[]:

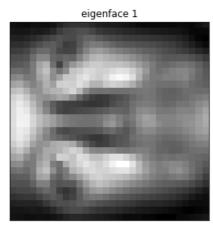
	Testing mse using	Test mse using	running time uslA	running time using
0	<b>LDA</b> 0.527985	<b>PCA</b> 0.448694	lda 0.326060	<b>PCA</b> 0.259994
1	0.754410	0.516961	0.612968	0.357522
2	0.985075	0.980100	1.014390	0.362947

# **Q1.C**

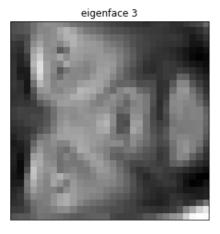
#### In [ ]:

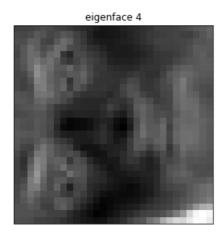
```
import matplotlib.pyplot as plt
mean vec = np.mean(Data, axis=0)
centered data = Data-mean vec
#using np.linalg.svd(singlular value decomposition to generate eigen faces)
U, S, V = np.linalg.svd(centered_data)
eigen_faces = V[:5].reshape((5,30,30))
eigenface_titles = ["eigenface %d" % i for i in range(eigen_faces.shape[0])]
#plotting eigen faces
plt.figure(figsize=(4.2 * 3, 4.2 * 2))
plt.subplots adjust(bottom=0, left=.01, right=.99, top=.90, hspace=.20)
for i in range(5):
      plt.subplot(2, 3, i + 1)
      plt.imshow(eigen_faces[i], cmap=plt.cm.gray)
      plt.title(eigenface_titles[i])
      plt.xticks(())
      plt.yticks(())
```

# eigenface 0









# **Q2**

# In [ ]:

from scipy.io import loadmat

```
D = loadmat("/content/Code&Data/Abalone.mat")
In [ ]:
labels = D['label vector'] # labels = age values
data = D['instance matrix'].toarray() #input features
In [ ]:
from sklearn.model selection import train test split
X train, X test, y train, y test = train test split(data, labels, test size=0.5, random
_state=1)
X_test, X_val, y_test, y_val = train_test_split(X_test, y_test, test_size=0.4, random_st
ate=1) #splitting into train, test and validation set
In [ ]:
print("Length of train set:",len(X train))
print("Length of validation set:",len(X val))
print("Length of test set:",len(X test))
Length of train set: 2088
Length of validation set: 836
Length of test set: 1253
In [ ]:
from sklearn.linear model import LinearRegression,Ridge
from sklearn.metrics import mean_squared_error
# Using Linear regression to predict the values
L reg = LinearRegression()
L reg.fit(X train, y train)
print("Train Mean Squared Error:", mean squared error(y train, L reg.predict(X train)))
print("Test Mean Squared Error:: ", mean_squared_error(y_test, L_reg.predict(X_test)))
print("Validation Mean Squared Error:", mean squared error(y val, L reg.predict(X val)))
Train Mean Squared Error: 4.773808603984102
Test Mean Squared Error:: 5.358162932581064
Validation Mean Squared Error: 4.564933800320933
In [ ]:
print("Test Mean Squared Error:: ",mean_squared_error(y_test,L_reg.predict(X_test)))
Test Mean Squared Error:: 5.358162932581064
In [ ]:
val mse = []
all alphas = []
beta = []
\#starting value of alpha = 10^{(-4)}
alpha = 10**(-4)
#ending value of alpha = 10^4
while alpha < 10**5:
  all alphas.append(alpha)
  R_reg = Ridge(alpha = alpha)
 R reg.fit(X train, y train)
  # recording beta for different values of alpha
  beta.append(R reg.coef)
  # recording mean squared error for different values of alpha
  val mse.append([alpha, mean squared error(y val, R reg.predict(X val))])
  alpha*=10
```

```
sorted(val mse, key = lambda x:x[1]) # different validation mse for different values of
alpha
Out[]:
[[0.0001, 4.564951179057885],
 [0.001, 4.56510816340551],
 [0.01, 4.566731642341755],
 [0.1, 4.586012519920947],
 [1.0, 4.746072512688645],
 [10.0, 5.530356950917623],
 [100.0, 6.7723963289758755],
 [1000.0, 7.751449692871668],
 [10000.0, 9.397653761279683]]
In [ ]:
# alpha=0.0001 has the least mse
R reg = Ridge(alpha = 0.0001)
R_reg.fit(X_train,y_train)
print ("Test Mean Squared error for alpha: 0.0001: ", mean squared error (y test, L reg. pred
ict(X_test)))
Test Mean Squared error for alpha: 0.0001: 5.358162932581064
In [ ]:
# visualize different beta using bar graph when given different alpha
plt.figure(figsize=(4 * 4, 10 * 2))
plt.subplots_adjust(bottom=0, left=.01, right=.99, top=.90, hspace=.20)
for i in range(8):
      plt.subplot(4, 2, i + 1)
      lis = [j[0][i] for j in beta]
      plt.bar(lis,all alphas, color = 'maroon', width = 0.4)
      plt.xlabel("feature "+str(i))
      plt.ylabel("alpha value")
      plt.title("Features over multiple alpha values")
      plt.xticks(())
      plt.yticks(())
              Features over multiple alpha values
                                                                    Features over multiple alpha values
                      feature 0
                                                                            feature_1
              Features over multiple alpha values
                                                                    Features over multiple alpha values
              Features over multiple alpha values
                                                                    Features over multiple alpha values
```

