# Q4. Eigenfaces - Face classification using PCA (40 classes)

- a) Use the following "face.csv" file to classify the faces of 40 different people.
- b) Do not use in-built function for implementing PCA.
- c) Use appropriate classifier taught in class (any classification algorithm taught in class like Bayes classifier, minimum distance classifier, and so on)
- d) Refer to the following link for a description of the dataset

#### In [2]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.naive_bayes import GaussianNB
from sklearn import metrics
```

#### In [3]:

```
df = pd.read_csv('face.csv')
df.drop(df.columns[-1], axis = 1, inplace = True)
df
```

#### Out[3]:

	0	1	2	3	4	5	6	7	8
0	0.309917	0.367769	0.417355	0.442149	0.528926	0.607438	0.657025	0.677686	0.690083
1	0.454545	0.471074	0.512397	0.557851	0.595041	0.640496	0.681818	0.702479	0.710744
2	0.318182	0.400826	0.491736	0.528926	0.586777	0.657025	0.681818	0.685950	0.702479
3	0.198347	0.194215	0.194215	0.194215	0.190083	0.190083	0.243802	0.404959	0.483471
4	0.500000	0.545455	0.582645	0.623967	0.648760	0.690083	0.694215	0.714876	0.723140
395	0.400826	0.495868	0.570248	0.632231	0.648760	0.640496	0.661157	0.636364	0.665289
396	0.367769	0.367769	0.351240	0.301653	0.247934	0.247934	0.367769	0.512397	0.574380
397	0.500000	0.533058	0.607438	0.628099	0.657025	0.632231	0.657025	0.669422	0.673554
398	0.214876	0.219008	0.219008	0.223140	0.210744	0.202479	0.276859	0.400826	0.487603
399	0.516529	0.462810	0.280992	0.252066	0.247934	0.367769	0.574380	0.615702	0.661157

400 rows × 4096 columns

```
In [5]:
```

```
train_data_set = df.iloc[2:10]

for i in range(10, 400, 10):
    train_data_set = train_data_set.append(df.iloc[i + 2:i + 10], ignore_index = True)
```

#### In [6]:

```
test_data_set = df.iloc[0:2]

for i in range(10, 400, 10):
    test_data_set = test_data_set.append(df.iloc[i:i + 2], ignore_index = True)
```

#### In [7]:

```
train_data_set = np.transpose(train_data_set)

test_data_set = np.transpose(test_data_set)

d = 4096
```

#### In [8]:

```
def mean_PCA(train_data_set):
    train_data_set_mean = np.mean(train_data_set, axis = 1)
    return np.array(train_data_set_mean)
```

#### In [9]:

```
train_data_set_mean = mean_PCA(train_data_set)
train_data_set_covariance_matrix = np.cov(train_data_set)
eigen_values, eigen_vectors = np.linalg.eigh(train_data_set_covariance_matrix)
```

#### In [10]:

```
eigen_vectors
```

#### Out[10]:

```
array([[ 0.00000000e+00,  0.00000000e+00,  0.00000000e+00, ...,  4.07541819e-04,  2.90708397e-02, -2.74754798e-03],  [ 1.41323302e-01,  1.47697747e-02, -3.20600723e-01, ...,  -1.51739158e-03,  3.42810801e-02, -5.43886813e-03],  [ 4.52142226e-01,  2.75796727e-01,  3.75162941e-01, ...,  -2.33185371e-03,  3.90819016e-02, -7.60074745e-03], ...,  [-1.08510497e-02, -2.50947201e-02, -6.58641110e-03, ...,  -1.30268290e-02, -2.99621494e-02, -2.94621031e-03],  [ 1.45648223e-03,  2.43254843e-02,  3.17920811e-03, ...,  -1.02592154e-02, -2.74033230e-02,  8.39296650e-04],  [-6.55733247e-03,  1.23748159e-02, -1.40301892e-02, ...,  -8.67573068e-03, -2.50393131e-02,  1.56090205e-03]])
```

```
In [12]:
eigen_pairs = [(np.abs(eigen_values[i]), eigen_vectors[:, i]) for i in range(len(eigen_values[i]))
eigen_pairs.sort(key = lambda x : x[0], reverse = lambda x : x[0])
eigen_pairs
Out[12]:
[(19.105796095321377,
  array([-0.00274755, -0.00543887, -0.00760075, ..., -0.00294621,
          0.0008393 , 0.0015609 ])),
 (12.207087351051666,
  array([ 0.02907084, 0.03428108, 0.0390819 , ..., -0.02996215,
         -0.02740332, -0.02503931])),
 (6.166659790621014,
  array([ 0.00040754, -0.00151739, -0.00233185, ..., -0.01302683,
         -0.01025922, -0.00867573])),
 (3.766708963990769,
  array([-0.00253901, -0.00561563, -0.01419596, ..., 0.031047 ,
          0.02984297, 0.02726496])),
 (2.730287928495426,
  array([0.02352622, 0.02307565, 0.02036678, ..., 0.03885212, 0.03756989,
         0.0392973 ])),
 (2.491505137016736,
  array([-0.02104077, -0.0211751, -0.01811579, ..., 0.0269678,
          0.02674315. 0.026402751)).
In [13]:
def get_reduced_dimension(eigen_pairs, eigen_values, percent):
    eigen_values_sum = np.sum(eigen_values)
    eigen_values_sum_reduced_dimension = percent * eigen_values_sum / 100
    sum = 0.0
```

```
def get_reduced_dimension(eigen_pairs, eigen_values, percent):
    eigen_values_sum = np.sum(eigen_values)
    eigen_values_sum_reduced_dimension = percent * eigen_values_sum / 100
    sum = 0.0
    count = 0
    for i in eigen_pairs:
        if(sum < eigen_values_sum_reduced_dimension):
            sum += i[0]
            count += 1
        else:
            break
    return count</pre>
```

#### In [14]:

```
d_prime = get_reduced_dimension(eigen_pairs, eigen_values, 95)
d_prime
```

#### Out[14]:

111

#### In [15]:

```
matrix_w = eigen_pairs[0][1].reshape(d, 1)
```

#### In [16]:

```
for i in range(1, d_prime):
    matrix_w = np.hstack((matrix_w, eigen_pairs[i][1].reshape(d, 1)))
```

```
In [17]:
train_data_set_transformed = np.transpose(matrix_w.T.dot(train_data_set))
test_data_set_transformed = np.transpose(matrix_w.T.dot(test_data_set))
In [18]:
train_data_set_transformed
Out[18]:
array([[-3.84278500e+01, 1.05496914e+01,
                                          2.76257385e+00, ...,
        -1.49311834e-01, 2.64169294e-01,
                                          4.94872391e-01],
       [-3.80958302e+01, -2.36530844e+00,
                                          2.77395344e+00, ...,
         1.07242465e-01, -2.50675502e-01,
                                          2.44819087e-01],
       [-3.70319559e+01, 1.22432369e+01, 6.31392191e+00, ...,
       -1.36547366e-01, -4.01670809e-02, 3.49316185e-01],
       [-3.11444304e+01, 1.01462209e+01, 5.96225995e+00, ...,
       -6.70871573e-02, -1.75148272e-01, 3.32349524e-01],
       [-3.88338294e+01, 9.01930153e-01, 1.98477303e+00, ...,
        -2.30756990e-02, -3.69142681e-01, 3.00644319e-01],
       [-3.45162687e+01, 6.32423245e+00, 3.80672052e+00, ...,
        -3.53049751e-01, 2.48143830e-01, 2.12767436e-01]])
In [19]:
test_data_set_transformed
Out[19]:
array([[-3.95811272e+01, 9.73374957e+00,
                                          1.16472961e+00, ...,
         1.14704260e-02, 2.30335788e-02, 1.73677045e-01],
       [-3.39897567e+01, 1.54595402e+01, 4.52888350e+00, ...,
       -4.59040975e-01, -1.37985001e-01, 4.52161510e-01],
```

1.71149373e+00, ...,

4.28431021e+00, ...,

5.43861931e-01],

# **Bayes Classifier Code**

**Bayes Classifier with dimesion = 4096** 

[-2.30611479e+01, 9.42401369e+00,

-8.64847903e-02, -4.13596812e-01,

[-3.73080804e+01, 6.39686729e+00,

[-3.42602930e+01, 9.76087593e+00, -1.34650656e+00, ..., -1.48750983e-01, 2.39329589e-01, 3.29394425e-01],

[-3.18795206e+01, 1.02691521e+01, 5.97999216e+00, ..., -2.37681112e-02, 6.24611657e-02, 2.73599124e-01],

-2.95259371e-01, -4.42425011e-01, 2.70357075e-01]])

## In [20]:

train\_data\_set

## Out[20]:

	0	1	2	3	4	5	6	7	8
0	0.318182	0.198347	0.500000	0.549587	0.330578	0.128099	0.243802	0.380165	0.657025
1	0.400826	0.194215	0.545455	0.545455	0.305785	0.185950	0.297521	0.442149	0.677686
2	0.491736	0.194215	0.582645	0.541322	0.330578	0.247934	0.367769	0.483471	0.698347
3	0.528926	0.194215	0.623967	0.537190	0.351240	0.314050	0.454545	0.545455	0.706612
4	0.586777	0.190083	0.648760	0.537190	0.425620	0.388430	0.495868	0.582645	0.702479
4091	0.148760	0.743802	0.177686	0.661157	0.487603	0.157025	0.177686	0.173554	0.115702
4092	0.144628	0.764463	0.177686	0.690083	0.190083	0.152893	0.190083	0.173554	0.115702
4093	0.140496	0.752066	0.177686	0.714876	0.144628	0.152893	0.190083	0.173554	0.115702
4094	0.148760	0.752066	0.173554	0.706612	0.152893	0.173554	0.181818	0.173554	0.115702
4095	0.152893	0.739669	0.173554	0.702479	0.152893	0.173554	0.190083	0.173554	0.103306

4096 rows × 320 columns

In [21]:

test\_data\_set

Out[21]:

	0	1	2	3	4	5	6	7	8
0	0.309917	0.454545	0.541322	0.644628	0.578512	0.628099	0.169422	0.561983	0.454545
1	0.367769	0.471074	0.586777	0.690083	0.603306	0.665289	0.264463	0.665289	0.429752
2	0.417355	0.512397	0.640496	0.702479	0.632231	0.685950	0.219008	0.698347	0.537190
3	0.442149	0.557851	0.661157	0.702479	0.665289	0.694215	0.280992	0.698347	0.611570
4	0.528926	0.595041	0.685950	0.706612	0.677686	0.719008	0.421488	0.731405	0.652893
4091	0.132231	0.152893	0.363636	0.111570	0.194215	0.177686	0.363636	0.231405	0.376033
4092	0.148760	0.152893	0.111570	0.115702	0.165289	0.161157	0.198347	0.214876	0.409091
4093	0.152893	0.152893	0.095041	0.111570	0.177686	0.185950	0.210744	0.219008	0.409091
4094	0.161157	0.152893	0.111570	0.107438	0.161157	0.161157	0.235537	0.219008	0.384298
4095	0.157025	0.152893	0.111570	0.119835	0.152893	0.190083	0.214876	0.227273	0.384298

4096 rows × 80 columns

4

```
In [22]:
```

```
train_data_set = np.transpose(train_data_set)
test_data_set = np.transpose(test_data_set)

X_train = train_data_set
y_train = []

for i in range(40):
    for j in range(int(len(train_data_set)/ 40)):
        y_train.append(str(i))
```

#### In [23]:

```
X_test = test_data_set
y_test = []

for i in range(40):
    for j in range(int(len(test_data_set)/ 40)):
        y_test.append(str(i))
```

#### In [24]:

```
gnb = GaussianNB()
gnb.fit(X_train, y_train)
y_pred = gnb.predict(X_test)
overall_accuracy = metrics.accuracy_score(y_test, y_pred)
```

#### In [25]:

```
print("Bayes Classifier with dimension = " + str(d))
print("\nOverall Accuracy = " + str(overall_accuracy))
```

```
Bayes Classifier with dimension = 4096
Overall Accuracy = 0.9
```

# **Bayes Classifier with dimesion = 111**

```
In [26]:
X_train_reduced_dimension = train_data_set_transformed
y_train_reduced_dimension = []
for i in range(40):
    for j in range(int(len(train_data_set_transformed)/ 40)):
        y_train_reduced_dimension.append(str(i))
X_train_reduced_dimension
Out[26]:
array([[-3.84278500e+01, 1.05496914e+01, 2.76257385e+00, ...,
        -1.49311834e-01, 2.64169294e-01,
                                          4.94872391e-01],
       [-3.80958302e+01, -2.36530844e+00, 2.77395344e+00, ...,
         1.07242465e-01, -2.50675502e-01, 2.44819087e-01],
       [-3.70319559e+01, 1.22432369e+01, 6.31392191e+00, ...,
        -1.36547366e-01, -4.01670809e-02, 3.49316185e-01],
       . . . ,
       [-3.11444304e+01, 1.01462209e+01, 5.96225995e+00, ...,
        -6.70871573e-02, -1.75148272e-01,
                                           3.32349524e-01],
       [-3.88338294e+01, 9.01930153e-01, 1.98477303e+00, ...,
       -2.30756990e-02, -3.69142681e-01, 3.00644319e-01],
       [-3.45162687e+01, 6.32423245e+00, 3.80672052e+00, ...,
        -3.53049751e-01, 2.48143830e-01, 2.12767436e-01]])
In [27]:
X_train_reduced_dimension.shape
Out[27]:
(320, 111)
In [28]:
X_test_reduced_dimension = test_data_set_transformed
y_test_reduced_dimension = []
for i in range(40):
    for j in range(int(len(test_data_set_transformed)/ 40)):
        y test reduced dimension.append(str(i))
```

```
X test reduced dimension
```

#### Out[28]:

```
1.16472961e+00, ...,
array([[-3.95811272e+01, 9.73374957e+00,
        1.14704260e-02, 2.30335788e-02,
                                          1.73677045e-01],
       [-3.39897567e+01, 1.54595402e+01, 4.52888350e+00, ...,
       -4.59040975e-01, -1.37985001e-01, 4.52161510e-01],
      [-3.42602930e+01, 9.76087593e+00, -1.34650656e+00, ...,
       -1.48750983e-01, 2.39329589e-01,
                                         3.29394425e-01],
      [-2.30611479e+01, 9.42401369e+00,
                                          1.71149373e+00, ...,
       -8.64847903e-02, -4.13596812e-01,
                                          5.43861931e-01],
                                          5.97999216e+00, ...,
       [-3.18795206e+01, 1.02691521e+01,
       -2.37681112e-02, 6.24611657e-02, 2.73599124e-01],
                                          4.28431021e+00, ...,
      [-3.73080804e+01, 6.39686729e+00,
        -2.95259371e-01, -4.42425011e-01,
                                          2.70357075e-01]])
```

```
In [29]:

X_test_reduced_dimension.shape

Out[29]:
(80, 111)

In [30]:

gnb_reduced_dimension = GaussianNB()
gnb_reduced_dimension.fit(X_train_reduced_dimension, y_train_reduced_dimension)

y_pred_reduced_dimension = gnb_reduced_dimension.predict(X_test_reduced_dimension)

overall_accuracy_reduced_dimension = metrics.accuracy_score(y_test_reduced_dimension, y_pred_reduced_dimension)

print("Bayes Classifier with dimension = " + str(d_prime))
print("\noverall Accuracy = " + str(overall_accuracy_reduced_dimension))

Bayes Classifier with dimension = 111

Overall Accuracy = 0.975

In []:
```