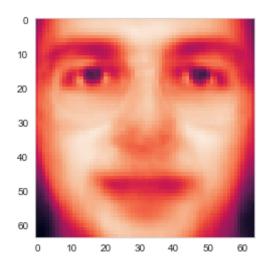
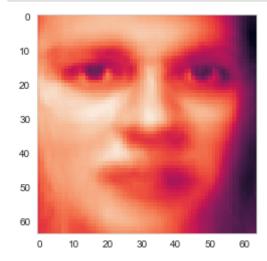
```
In [170]: #Submitted by : Philani Mpofu (1848751)
          #Submitted by : Matthew Kruger (1669326)
          #Submitted by : Chloe Smith (1877342)
          #Submitted by : Jesse Bristow (1875955)
In [171]: import numpy as np
          from numpy import linalg
          import pandas as pd
          import matplotlib.pyplot as plt
In [172]: # Load data
          dataset = datasets.fetch olivetti faces()
          images = dataset['data']
          target = dataset['target']
          print('Images Shape: ', images.shape)
          print(images)
          #print(target)
          #print(images)
          Images Shape: (400, 4096)
          [[0.30991736 0.3677686 0.41735536 ... 0.15289256 0.16115703 0.1570248
           [0.45454547 0.47107437 0.5123967 ... 0.15289256 0.15289256 0.1528925
           [0.3181818    0.40082645    0.49173555    ...    0.14049587    0.14876033    0.1528925
          6]
                       0.53305787 0.607438 ... 0.17768595 0.14876033 0.1900826
           [0.5
           [0.21487603 0.21900827 0.21900827 ... 0.57438016 0.59090906 0.6033057
          61
           [0.5165289  0.46280992  0.28099173  ...  0.35950413  0.3553719  0.3842975
          2]]
```

```
In [173]: # Normalize the data, then get the Covariance Matrix
          # targetNew = target.reshape(400,1) # Need to rearrange target vector i
          n order to make a matrix to find covariant matrix
          covMatrix = (1/images.shape[0])*(np.dot(images.T,images))
          print('Covariance Matrix:\n', covMatrix)
          print('Covariance Matrix Shape:',covMatrix.shape)
          Covariance Matrix:
           [[0.19267644 0.2059697 0.21917908 ... 0.12233528 0.11950444 0.1195607
          71
           [0.2059697  0.22438237  0.24084835  ...  0.1306454  0.12774651  0.1280675
          11
           [0.21917908 0.24084835 0.26467338 ... 0.14092542 0.13807441 0.1383975
          91
           [0.12233528 0.1306454 0.14092542 ... 0.13850415 0.13217118 0.1277516
          21
           [0.11950444 0.12774651 0.13807441 ... 0.13217118 0.13200496 0.1286022
          21
           [0.11956077 0.12806751 0.13839759 ... 0.12775162 0.12860222 0.1289292
          911
          Covariance Matrix Shape: (4096, 4096)
In [174]: # Compute eigenvalues/vectors and convert values to their real number r
          epresentations
          eigvalues, eigvectors = np.linalg.eig(covMatrix)
          eigvalues = eigvalues.real
          eigvectors = eigvectors.real
In [187]: print('Eigenvalues Shape:',eigvalues.shape)
          print('Eigenvectors Shape:',eigvectors.shape)
          # # Make a list of (eigenvalue, eigenvector) tuples
          # eig pairs = [(np.abs(eigvalues[i]), eigvectors[:,i]) for i in range(l
          en(eigvalues))]
          # # Sort the (eigenvalue, eigenvector) tuples from high to low
          # eig pairs.sort(key=lambda x: x[0], reverse=True)
```

```
eigvalues sum = np.sum(eigvalues) # Sum up all of the eigenvalues
         # Use variation matrix to store the variation of each eigenvalue
         var matrix = ([])
         for i in range(len(eigvalues)):
             var matrix = np.append(var matrix,(eigvalues[i]/eigvalues sum))
         X = eigvectors[:5] # Store the eigenvectors with the 5 lowest variation
         projection = np.dot(images, X.T)
         print('Projection Matrix: \n', projection)
         print('Projection Matrix Shape: ', projection.shape)
         Eigenvalues Shape: (4096,)
         Eigenvectors Shape: (4096, 4096)
         Projection Matrix:
          [ 1.0470986 -0.17621446 0.02594492 -0.1250127 0.1380943 ]
          [ 0.8913089 -0.25974685 -0.10215159 -0.21585543
                                                         0.118418221
          [ 0.6764893 -0.5245818 -0.11842033 -0.12979957
                                                         0.3205065 1
          [ 1.0825269  -0.5509006  -0.12855414  -0.14033079   0.35880226]
          [ 1.0696241  -0.45040724  -0.01769611  0.01406448  0.4974962  ]]
         Projection Matrix Shape: (400, 5)
In [189]: projection = np.dot(images, eigvectors[:,:5])
         image = np.sum(projection[0]*eigvectors[:,:5],axis=1)
         plt.imshow(image.reshape((64,64)))
         plt.show()
```

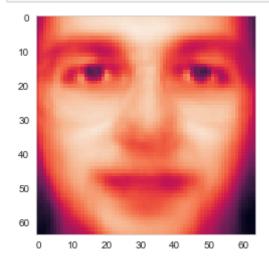


```
In [190]: projection = np.dot(images,eigvectors[:,:5])
  image = np.sum(projection[1]*eigvectors[:,:5],axis=1)
  plt.imshow(image.reshape((64,64)))
  plt.show()
```

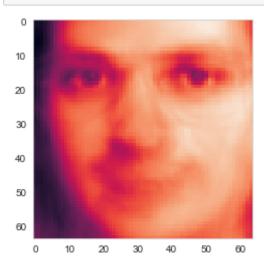


```
In [191]: projection = np.dot(images,eigvectors[:,:5])
image = np.sum(projection[2]*eigvectors[:,:5],axis=1)
```

## plt.imshow(image.reshape((64,64))) plt.show()



```
In [192]: projection = np.dot(images,eigvectors[:,:5])
  image = np.sum(projection[3]*eigvectors[:,:5],axis=1)
  plt.imshow(image.reshape((64,64)))
  plt.show()
```



```
In [193]: | projection = np.dot(images,eigvectors[:,:5])
          image = np.sum(projection[4]*eigvectors[:,:5],axis=1)
          plt.imshow(image.reshape((64,64)))
          plt.show()
           10
           30
           50
           60
  In [ ]:
```