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In [1]: import pandas as pd
In [2]: import numpy as np
         import matplotlib.pyplot as plt
         from PIL import Image
         from glob import glob
         %matplotlib inline
In [3]: # load the data
         data = np.load('./dataa/data_10000_norm.npz')
In [4]: data.files
Out[4]: ['arr_0', 'arr_1']
In [5]: X = data['arr_0'] # Independent features
         y = data['arr_1'] # dependent features
In [6]: X.shape, y.shape
Out[6]: ((5458, 10000), (5458,))
         Eigen Images
In [7]: X1 = X - X.mean(axis=0)
In [8]: from sklearn.decomposition import PCA
In [9]: pca = PCA(n_components=None, whiten=True, svd_solver='auto')
In [10]: x_pca = pca.fit_transform(X1)
In [11]: x_pca.shape
Out[11]: (5458, 5458)
In [12]: eigen_ratio = pca.explained_variance_ratio_
         eigen_ratio_cum = np.cumsum(eigen_ratio)
In [13]: plt.figure(figsize=(10,4))
         plt.subplot(1,2,1)
         plt.plot(eigen_ratio[:200], 'r>--')
         plt.xlabel('Number of componenets')
         plt.ylabel('Eigen ratio')
         plt.subplot(1,2,2)
         plt.plot(eigen_ratio_cum[:200], 'r>--')
         plt.xlabel('Number of componenets')
         plt.ylabel('Cumulative Eigen ratio sum')
Out[13]: Text(0, 0.5, 'Cumulative Eigen ratio sum')
            0.25
                                                   0.9
                                                 0.8
N
            0.20
                                                   0.7
          Eigen 0.10
            0.05
                                                   0.3
            0.00
                              100
                                      150
                                                                            150
                                                                                   200
                                                                    100
                        Number of componenets
                                                              Number of componenets
         Conclusion: Using elbow method, consider number of components between 25-30
                since if I consider component between 25-30 the explained variance is around 75% so, inorder to get min 80%
                variance I am considering 50 components
In [14]: pca_50 = PCA(n_components=50, whiten=True, svd_solver='auto')
         x_pca_50 = pca_50.fit_transform(X1)
In [15]: # saving pca
         import pickle
         pickle.dump(pca_50,open('./dataa/pca_50.pickle','wb'))
In [16]: # consider 50 components and inverse transform
         x_pca_inv = pca_50.inverse_transform(x_pca_50)
In [17]: x_pca_inv.shape
Out[17]: (5458, 10000)
In [20]: #checking one image
         eigen_img = x_pca_inv[0,:]
         eigen_img = eigen_img.reshape((100,100))
         plt.imshow(eigen_img, cmap='gray')
Out[20]: <matplotlib.image.AxesImage at 0x2801aaf6108>
          20
          40
          60
          80
                       40
                  20
In [22]: def label(y):
              if y==0:
                  return 'Male'
              else:
                  return 'Female'
          np.random.randint(1001)
         pics = np.random.randint(0,5458,40)
          plt.figure(figsize=(15,8))
          for i,pic in enumerate(pics):
              plt.subplot(4,10,i+1)
              img = X[pic:pic+1].reshape(100,100)
             plt.imshow(img,cmap='gray')
             plt.title('{}'.format(label(y[pic])))
             plt.xticks([])
              plt.yticks([])
         plt.show()
         print("="*20+'Eigen Images'+"="*20)
         plt.figure(figsize=(15,8))
         for i, pic in enumerate(pics):
              plt.subplot(4,10,i+1)
              img = x_pca_inv[pic:pic+1].reshape(100,100)
              plt.imshow(img, cmap='gray')
              plt.title('{}'.format(label(y[pic])))
              plt.xticks([])
              plt.yticks([])
         plt.show()
             Male
                                Female
                                          Female
                                                   Female
                                                              Male
                                                                                Female
            Female
             Male
         ======Eigen Images========
             Male
                      Female
                                Female
                                         Female
                                                   Female
                                                              Male
                                                                       Male
                                                                                Female
                                                                                          Female
                                                                                                    Female
                                                                                 Male
                       Male
                                Male
                                                   Female
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                       Male
                                                                       Male
                                                                                 Male
            Female
                                Female
                                          Male
                                                    Male
                                                             Female
                                                                                          Female
                                                                                                    Female
In [23]: # saving
          np.savez('./dataa/Data_pca_mean_50.pickle',x_pca_50,y,X.mean())
In [ ]:
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