Q2.1 PCA of Yale Face Database **Importing Libraries** In [1]: import numpy as np from matplotlib.image import imread import matplotlib.pyplot as plt import scipy.io import copy plt.rcParams['figure.figsize'] = [8,4] Importing Yale Faces Database from .mat file scipy.io.loadmat() function imports .mat file as dictonary In [2]: data = scipy.io.loadmat('./Yale_64x64.mat') print(type(data)) <class 'dict'> Dictonary to Numpy Array In [3]: A = np.array(data['fea']).T In [4]: print(A.shape) (4096, 165)Sample Image/Face from database In [5]: img = plt.imshow(A[:,1].reshape(64,64).transpose()) img.set_cmap('gray')
plt.axis('off') plt.show() Centering Matrix in terms of Columns and Rows 1. Subtracting Column Mean In [6]: Amean = A.mean(axis=1, keepdims=True) Am = A - AmeanIn [7]: img = plt.imshow(Am[:,1].reshape(64,64).transpose()) img.set_cmap('gray')
plt.axis('off') plt.show() 2. Subtracting Row Mean In [8]: Amean = Amean.mean(axis=0, keepdims=True) Am = Am - AmeanIn [9]: img = plt.imshow(Am[:,1].reshape(64,64).transpose()) img.set_cmap('gray') plt.axis('off') plt.show() Calculating SVD In [10]: U, D, Vt = np.linalg.svd(Am)# Complete SVD i.e. calculation corresponding to ze # U,D,Vt = np.linalg.svd(Am, full_matrices=False) # Economy SVD i.e. Calculations D = np.diag(D)In [11]: print(U.shape, D.shape, Vt.shape) (4096, 4096) (165, 165) (165, 165) Finding number of eigen values with least significance In [12]: i = 0n = len(D)while(i<n):</pre> if abs(D[i][i]) <10:</pre> i += 1 $eig_vals_with_least_significance = n - i$ In [13]: print(eig_vals_with_least_significance) 2 Visualizing singular values by plotting graph 1. Singular values vs Count 2. (Cumulative sum/Total sum) vs Count In [14]: d = D[150:, 150:]plt.figure(1) plt.semilogy(np.diag(d)) plt.title('Singular Values') plt.show() plt.figure(2) plt.plot(np.cumsum(np.diag(d))/np.sum(np.diag(d))) plt.title('Singular Values: Cumulative Sum') plt.show() Singular Values 10² 10^{0} 10^{-2} 10^{-4} 10^{-6} 10^{-8} 10^{-10} 10^{-12} 2 4 6 8 10 12 0 14 Singular Values: Cumulative Sum 1.0 -8.0 0.6 0.4 0.2 2 6 12 10 14 In Sample Projection and Prediction In [15]: sample_size = 150 def InSampleProjectionAndReconstruction(image_number): j **=** 0 for r in (50, 100, 200, 500, 800, 2000, 4096, 4096-eig_vals_with_least_significance # Construct approximate image u = U[:,:r]# Projection A_train_model = np.matmul(u.T, A[:,:sample_size]) # Reconstruction A_train_pred = np.matmul(u, A_train_model) $Fimg = A_train_pred$ plt.figure(j+1) j **+=** 1 plot1 = plt.subplot(121) img = plt.imshow(A[:,image_number].reshape(64,64).transpose()) img.set_cmap('gray') plt.title(f'Original Image') plt.axis('off') plot2 = plt.subplot(122) img2 = plt.imshow(Fimg[:,image_number].reshape(64,64).transpose()) img2.set_cmap('gray') plt.axis('off') plt.title(f'Approximate Image (r = {r})') plt.show() In [16]: InSampleProjectionAndReconstruction(0) Original Image Approximate Image (r = 50)Approximate Image (r = 100)Original Image Approximate Image (r = 200)Original Image Original Image Approximate Image (r = 500)Original Image Approximate Image (r = 800)Original Image Approximate Image (r = 2000)Original Image Approximate Image (r = 4096)Original Image Approximate Image (r = 4094)Out off Sample Projection and Prediction $\textbf{def} \ \ \textbf{outOffSampleProjectionAndReconstruction(image_number):}$ if(image_number>=sample_size): j **=** 0 for r in (50, 100, 200, 500, 800, 2000,4096, 4096-eig_vals_with_least_signific # Construct approximate image u = U[:,:r]# Projection A_test_model = np.matmul(u.T,A[:,image_number]) # Reconstruction A_test_pred = np.matmul(u,A_test_model) $Fimg = A_test_pred$ plt.figure(j+1) j **+=** 1 plt.subplot(121) img = plt.imshow(A[:,image_number].reshape(64,64).transpose()) img.set_cmap('gray') plt.axis('off') plt.title(f'Original Image (r = {r})') plt.subplot(122) img2 = plt.imshow(Fimg.reshape(64,64).transpose()) img2.set_cmap('gray')
plt.axis('off') plt.title($f'Approximate Image (r = {r})'$) plt.show() else: print("Object Belongs to Sample") In [18]: outOffSampleProjectionAndReconstruction(155) Approximate Image (r = 50)Original Image (r = 50)Original Image (r = 100)Approximate Image (r = 100)Original Image (r = 200)Approximate Image (r = 200)Original Image (r = 500)Approximate Image (r = 500)Original Image (r = 800)Approximate Image (r = 800)Original Image (r = 2000)Approximate Image (r = 2000)Approximate Image (r = 4096)Original Image (r = 4096)Original Image (r = 4094)Approximate Image (r = 4094)