



SANTA CLARA UNIVERSITY

School of Engineering

COEN 140L Lab 9 Intro

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Dataset

- A face image database of 10 subjects, (in the att_faces_10.zip file).
- Each subject has 10 gray-scale images of 112×92 pixels
- use face images 1,3,4,5,7,9 as the training images, and face images 2,6,8,10 as the test images.



PCA for face recognition

- C subjects/Classes
- Training set (each subject has N training images)
 - Subject 1: $\mathbf{x}_{11}, \dots, \mathbf{x}_{1N}$ (i.e. $\mathbf{x}_{1n}, n = 1, \dots, N$)
 - Subject 2: $\mathbf{x}_{21}, \dots, \mathbf{x}_{2N}$
 - ...
 - Subject C : $\mathbf{x}_{C1}, \dots, \mathbf{x}_{CN}$



PCA for face recognition

- SVD for dimensionality reduction

- $\mathbf{X} = [\mathbf{x}_{11}, \dots, \mathbf{x}_{1N}, \mathbf{x}_{21}, \dots, \mathbf{x}_{2N}, \dots, \mathbf{x}_{C1}, \dots, \mathbf{x}_{CN}]$

$$\mathbf{X} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$$

- Find the **top- d left singular vectors**

$$\mathbf{U}_{:, [1:d]} = [\mathbf{u}_1, \dots, \mathbf{u}_d]$$

- Project each training image onto these vectors

- The n th training image of the c th subject: \mathbf{x}_{cn}

$$\mathbf{y}_{cn} = \begin{bmatrix} y_{cn,1} \\ \vdots \\ y_{cn,d} \end{bmatrix} = \mathbf{U}_{:, [1:d]}^T \mathbf{x}_{cn} = \begin{bmatrix} \mathbf{u}_1^T \mathbf{x}_{cn} \\ \vdots \\ \mathbf{u}_d^T \mathbf{x}_{cn} \end{bmatrix}$$

- **Nearest-Neighbor Classifier**

- For a test image \mathbf{x}

- Project the test image onto vectors

$$\mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_d \end{bmatrix} = \begin{bmatrix} \mathbf{u}_1^T \mathbf{x} \\ \vdots \\ \mathbf{u}_d^T \mathbf{x} \end{bmatrix} = \mathbf{U}_{:, [1:d]}^T \mathbf{x}$$

- Determine its class label by

$$\hat{c} = \arg \min_{\substack{c=1,2,\dots,C \\ n=1,2,\dots,N}} \|\mathbf{y} - \mathbf{y}_{cn}\|_2$$



SVD Algorithm

`full_matrices : bool, optional`

If True (default), u and vh have the shapes $(..., M, M)$ and $(..., N, N)$, respectively. Otherwise, the shapes are $(..., M, K)$ and $(..., K, N)$, respectively, where $K = \min(M, N)$.

`U, Sigma, V = numpy.linalg.svd(L, full_matrices=False)` **# must set full_matrices=False**

use this link to check how to use numpy.linalg.svd

<https://numpy.org/doc/stable/reference/generated/numpy.linalg.svd.html>

check whether the singular values in Sigma are sorted **from the largest to smallest**. If not, need to sort them from the largest to smallest, and also sort the left singular vectors in U correspondingly.



Basic idea of this assignment

loading images as train and test dataset

- Convert each image to a vector of length $D=112 \times 92=10304$. Stack 6 training images of all 10 subjects to form a matrix of size 10304×60 .

Apply SVD by `numpy.linalg.svd(L,full_matrices=False)`

- Apply singular-value decomposition (SVD) for dimensionality reduction. Find the top-K left singular vectors ($K=1,2,3,6,10,20, 30$, and 50) corresponding to the K largest singular values of the data matrix.

calculate the distance between the test image and each of the training image

- Project the face images to the top-K left singular vectors and apply the nearest-neighbor classifier in the reduced dimensional space.



Include in the report

- Plot the recognition accuracy rate, versus different K values.
 - *$\frac{\text{number of correct classification}}{\text{total number of test images}}$*
- Analyze the results you observe.



Lab Tasks

- **Need demo** for week 9 assignment(10% points).
- Submit to Camino a **pdf report with answers**(60% points), the report contains some **results** which required by lab document, you also need to add some **observations** for the questions.
- Submit **all the source code** needed to generate these answers to Camino(30% points).