

# Fast Prototyping Exercise 1

Exercises 5, 6, 7

CSC872

Pattern Analysis and Machine Intelligence

[https://bidal.sfsu.edu/~kazokada/csc872/  
FaceRecognition\\_Data.zip](https://bidal.sfsu.edu/~kazokada/csc872/FaceRecognition_Data.zip)

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## Fast Prototyping Exercise

- **Fast Prototyping**
  - Learn how to do a quick proof of concept by building a “prototype” (from papers you read, no public codes)
  - **Correctness** matters (no sloppy algorithm!)
  - **Speed** matters (no beautification!)
  - No perfect SE necessary
  - **No copying of codes online (but use base Matlab functions).**
  - **When Done: Parameterization/Visualization/Experimentation**
    - Find out what are **free parameters** in your algorithm whose value must be hand-picked by you
    - Learn how to view internal variable’s current values
    - Learn how to visualize your prototype’s results in plots/images etc
    - Tweak the parameter values and study your prototype’s behavior **quantitatively** to understand the how algorithm works
- **Group Work**
  - **You are encouraged to freely exchange ideas and codes**
  - **Contributions to others are as valuable as making your own work**

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## Fast Prototyping Exercise

- **Please upload your matlab codes thru iLearn forum for my grading and your playing!**
  - First two exercises: **Due on midnight of the day** (just what you did during the exercise)
  - Third last exercise: **Due on midnight next day** (complete version with some doc/screen shots of running the code)
- **Your grade on FP exercise will be partly based on these submitted codes and what I observe during the in-class exercises.**
- **If received helps from others and/or used codes from others, please credit the person who helped you.**

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## Platforms

- **MATLAB**
  - MathWorks: <http://www.mathworks.com/>
  - <http://en.wikipedia.org/wiki/MATLAB>
- **MATLAB @ SFSU**
  - <https://at.sfsu.edu/at-mathworks-matlab>
- **Various tutorials available online**
  - [https://matlabacademy.mathworks.com/?s\\_tid=acb\\_tut](https://matlabacademy.mathworks.com/?s_tid=acb_tut)

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## Public Libraries

- OpenCV (Computer Vision)
  - <http://www.intel.com/technology/computing/opencv/overview.htm>
- ITK (Medical Imaging)
  - <http://www.itk.org/>
- WEKA (Machine Learning)
  - <http://www.cs.waikato.ac.nz/~ml/weka/index.html>

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## Face Recognition by Eigenface

- Let's create a face recognition system using one of the most basic algorithm called "**Eigenface**".
  - You have not studied this in the lecture yet but
  - You read a paper on this (Turk & Pentland)
- You will need to implement 3 components
  - 1) Image I/O + visualization
  - 2) PCA for learning
  - 3) Recognition by nearest neighbor classification

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## Paper 1

- M. Turk, A. Pentland,
- *Eigenfaces for Recognition*, Journal of Cognitive Neuroscience, 3(1): 71-86 (1991)
- <http://portal.acm.org/citation.cfm?id=1326887.1326894&coll=&dl=>
- <http://en.wikipedia.org/wiki/Eigenface>

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## Data

- I provide a set of facial images
- [https://bidal.sfsu.edu/~kazokada/csc872/FaceRecognition\\_Data.zip](https://bidal.sfsu.edu/~kazokada/csc872/FaceRecognition_Data.zip)
- Images are organized in 3 folders
- ALL = FA+FB (for **Training**)
- FA: 12 32x32 8bit facial images (for **Known faces DB**)
- FB: 23 facial images (for **Test Set**)

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# Principal Component Analysis

- Conceptual Steps

- 1) Collect  $M$  Training Images (must be aligned,  $N_x$  by  $N_y$  matrix)
- 2) Vectorize the Images:  $X = \{x_1, \dots, x_M\}$  Each of  $M$  images is a column vector with  $N$  coefficients where  $N = N_x \text{ times } N_y$
- 3) Compute mean image:  $\mu = \text{mean}(X)$ ; a vector of  $N$  coeffs
- 4) Construct Covariance Matrix:  $C = (X - \mu^T)(X - \mu^T)^T$   $N$  by  $N$  mat
- 5) Solve Eigenvalue Problem:  $Cv_i = \lambda_i v_i$
- 6) Sort resulting eigen vectors in decreasing order of corresponding eigen values.
- 7) Select the top  $K$  Eigenvectors  $W = \{v_1, \dots, v_K\}$ , resulting in a face model  $\{\mu, W\}$

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# Nearest Neighbor Recognition

- Learning & Database Construction

- 1) Do PCA, yielding a face model  $\{\mu, W\}$
- 2) Construct DB of known faces with codes  $y_j = W^T(x_j - \mu^T)$  for all known faces  $\{x_j\}$  in  $\mathcal{F}$

- Face Recognition by NN Classification

- 1) Test face  $z$  is also projected to the model  $W^T(z - \mu^T) = y_z$
- 2) Nearest neighbor classification of  $y_z$  with  $\{y_i\}$  by picking the index " $i$ " that best match to  $y_z$  according to Euclidean distance

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TRAINING	TESTING
<p><math>X \leftarrow \{\text{all images in ALL}\}</math></p> <p>① DO PCA on <math>X</math></p> <p> <math>\swarrow</math> Eigen Face space.    <math>\searrow</math> mean face  <math>W = \begin{Bmatrix}   &amp;   &amp; \dots &amp;   \\ e_{v1} &amp; e_{v2} &amp; \dots &amp; e_{vk} \\   &amp;   &amp; \dots &amp;   \end{Bmatrix}</math>    <math>\mu = 0</math> </p> <p>② Choose <math>k = \frac{\log 20}{\log 2}</math></p> <p> <math>\rightarrow W_k = \begin{Bmatrix}   &amp; \dots &amp;   \\ e_{v1} &amp; \dots &amp; e_{vk} \\   &amp; \dots &amp;   \end{Bmatrix}</math> </p>	<p>① Prepare Known Face DB. <math>M=12</math></p> <p>For all images in <math>FA = \{x_1, \dots, x_M\}</math></p> <p>compute <math>C_i = W_k^T (x_i - \mu)</math></p> <p>resulting in <math>\{C_1, \dots, C_M\}</math></p> <p>② Testing! (FR!).</p> <p>Given a test image <math>Z \in FB</math>,</p> <p>compute <math>d_z = W_k^T (Z - \mu)</math></p> <p>compute distances from <math>d_z</math> to <math>\{C_i\}</math></p> <p>Find index <math>i^* \in 1, \dots, M</math> with min dist.</p>

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## Useful MATLAB Codes

For PCA

- Set  $X$  as a matrix with each row is a vectorized face
- $m = \text{mean}(X)$ : sample mean of  $X$ , pay attention to dim.
- $M = \text{repmat}(\mu', 1, N)$ ; create a matrix by repeating a column matrix  $\mu'$   $N$  times ( $M$  will be length of  $\mu \times N$ )
- $S = \text{cov}(X)$ : covariance matrix (mean removed)
- $[V D] = \text{eig}(S)$ : eigen value decomposition of a matrix  $S$ 
  - Each column of  $V$  is an eigen vector.
  - $D$  is a diagonal matrix of eigen values.
  - Columns of  $V$  and  $D$  are corresponding to each other
- $d = \text{diag}(D)$ ; vectorize the diagonal component of a matrix
- Use for-loop to get cumulative distribution of eigen values then divide it by the total variance ( $\text{sum}(\text{diag}(D))$ )
- $\text{Plot}(\text{cumulative distribution of eigen values})$

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