

Facial Recognition

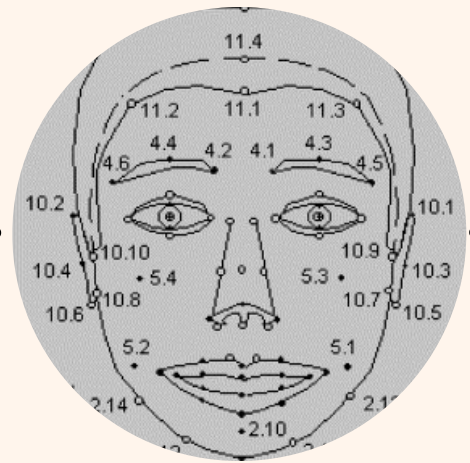
Using Eigenfaces and Eigen Vector

START

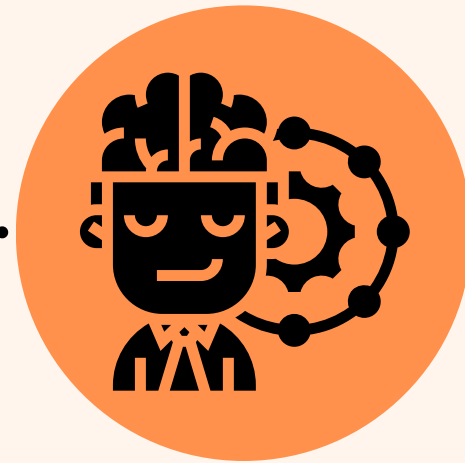
Process



**Data
PreProcessing**



**Eigen Face
Extraction**



**Training
the SVM**



**Testing and
Evaluation**



**Qualitative
Evaluation**

SVM Classification Model

Support Vector Machines (SVM) is a powerful supervised machine learning algorithm used for classification tasks.

SVM aims to find a hyperplane that best separates data points of different classes.

In this project, SVM is employed as a classifier to recognize and classify faces based on the extracted eigenfaces.

PCA?

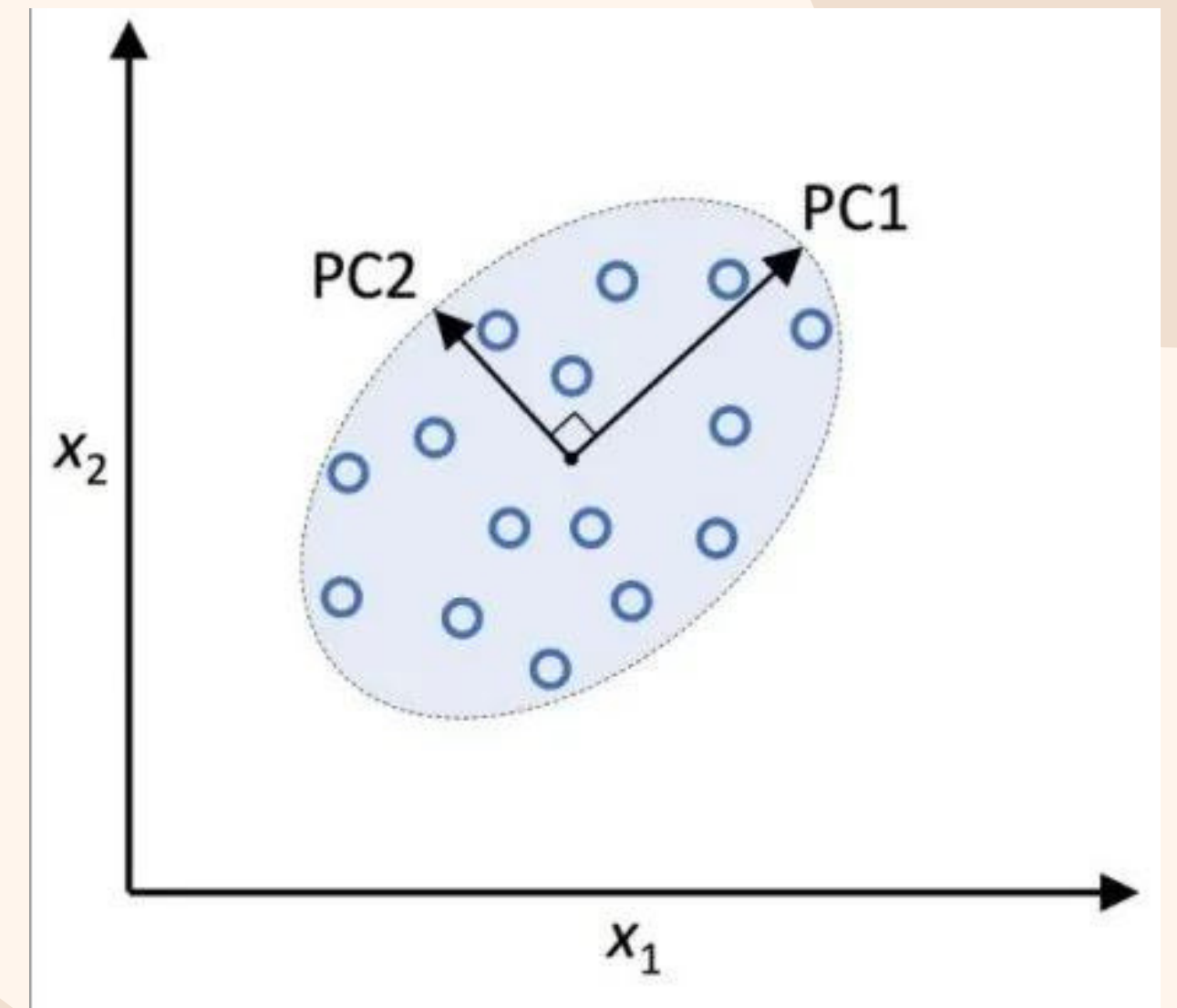
- Dimension reduction
- Feature Extraction
- In the figure, x_1 and x_2 are the original feature axes, and PC1 and PC2 are the principal components
- Why PCA in face recognition?

Algorithm:

$$x = [x_1, x_2, \dots, x_d], \quad x \in \mathbb{R}^d$$

$$\downarrow xW, W \in \mathbb{R}^{d \times k}$$

$$z = [z_1, z_2, \dots, z_k], \quad z \in \mathbb{R}^k$$



Continuing with algorithm

2. Construct the covariance matrix:

$$\Sigma = \begin{bmatrix} \sigma_1^2 & \sigma_{12} & \sigma_{13} \\ \sigma_{21} & \sigma_2^2 & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & \sigma_3^2 \end{bmatrix}$$

3. Decompose the covariance matrix into its eigenvectors and eigenvalues.\

$$\det(A - \lambda I) = 0$$

Continuing with algorithm

4. Eigen values in decreasing order

5. Select k eigenvectors which correspond to the k largest eigenvalues, where k is the dimensionality of the new feature subspace ($k \leq d$).

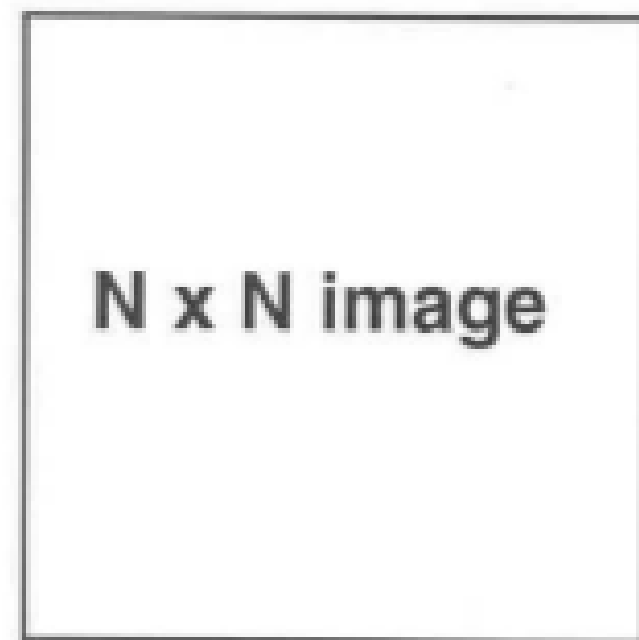
6. Transform the d -dimensional input dataset X using the projection matrix W to obtain the new k -dimensional feature subspace:

In this step, we use the matrix W that we just computed to transform our samples onto the new subspace via the equation $y = W' \times x$ where W' is the transpose of the matrix W .

PCA in python

In Python, you can perform Principal Component Analysis (PCA) using the scikit-learn library. The scikit-learn library provides a PCA class that implements the PCA algorithm and provides various methods and functionalities for performing PCA on datasets.

Calculation



N^2
N x 1 vector



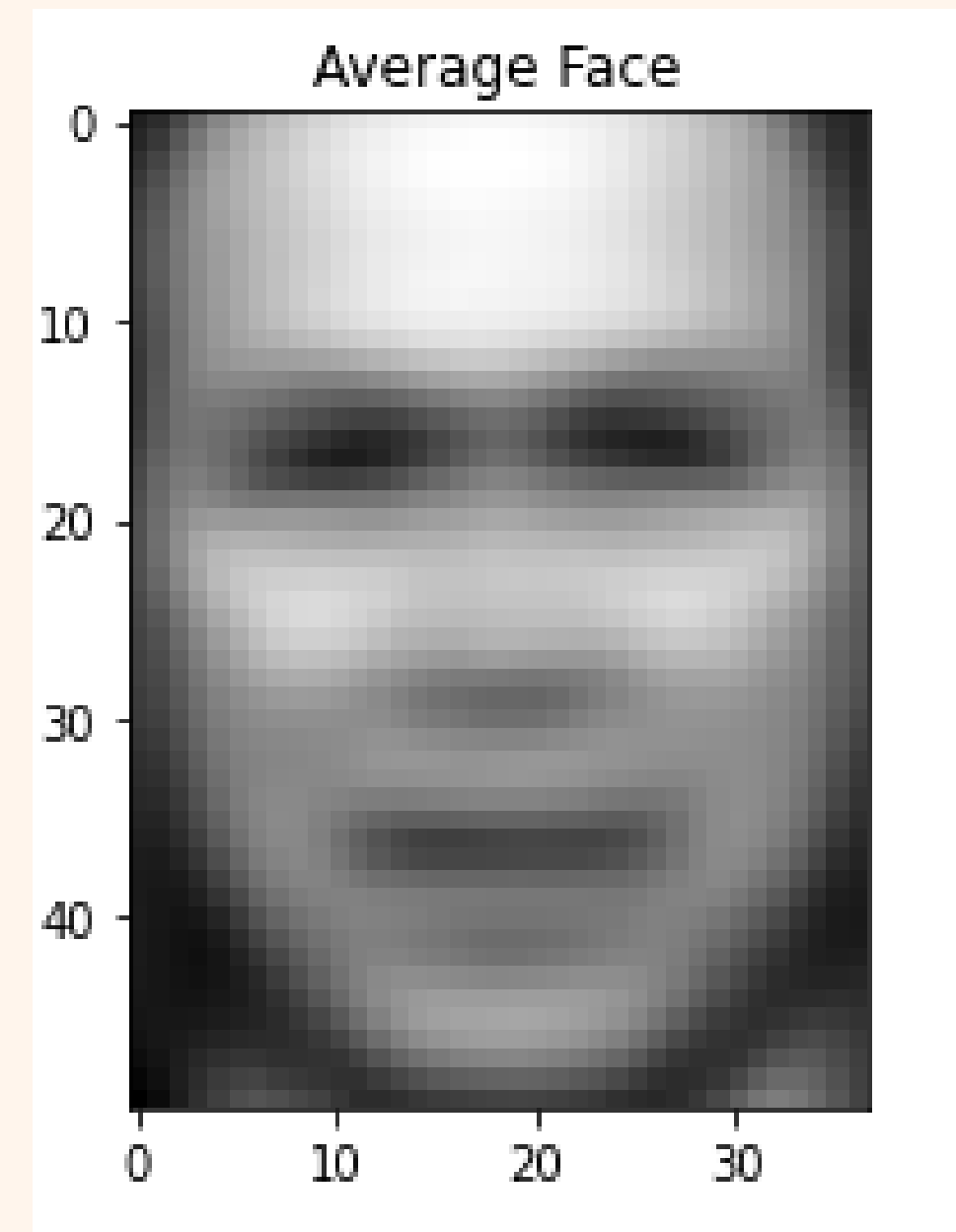
$N * N$



M number of images

We calculate the average of all these face vectors and subtract it from each vector

$$\psi = \frac{1}{m} \sum_{i=1}^m x_i$$
$$a_i = x_i - \psi$$



Then we take all face vectors so that we get a matrix of size of $N2 * M$.

$$A = [a_1 \quad a_2 \quad a_3 \quad \dots \quad a_m]$$

Covariance matrix $Cov = A^T A$

To calculate eigen values and eigenvectors of above covariance matrix use the formula:

where, $C' = AA^T$ and $u_i = A\nu_i$

$$A^T A\nu_i = \lambda_i \nu_i$$

$$AA^T A\nu_i = \lambda_i A\nu_i$$

$$C' u_i = \lambda_i u_i$$

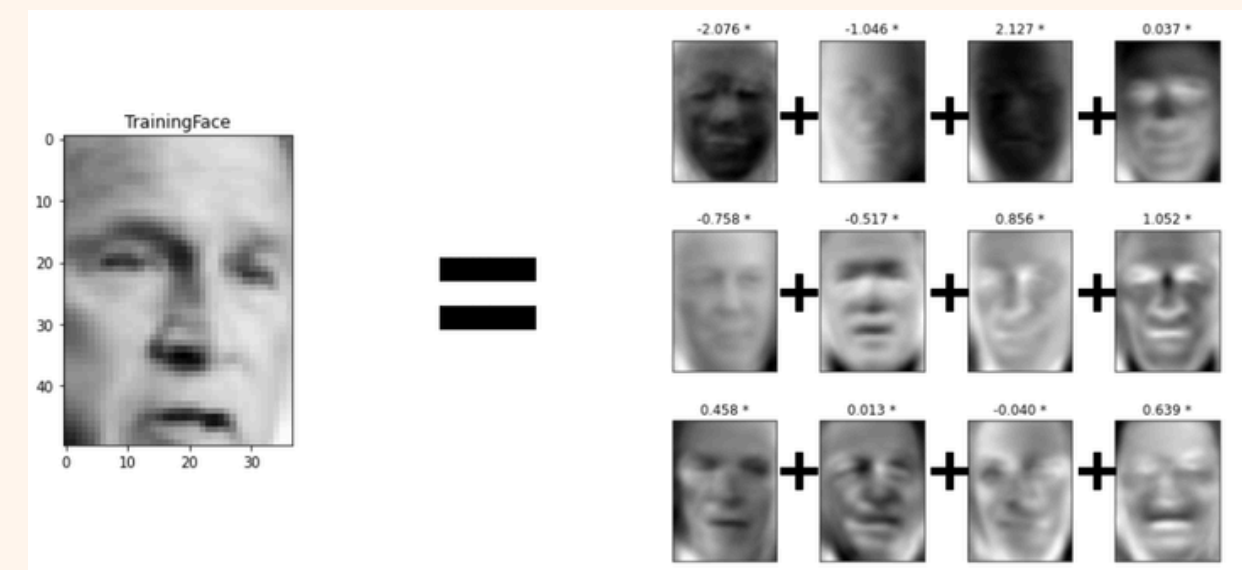
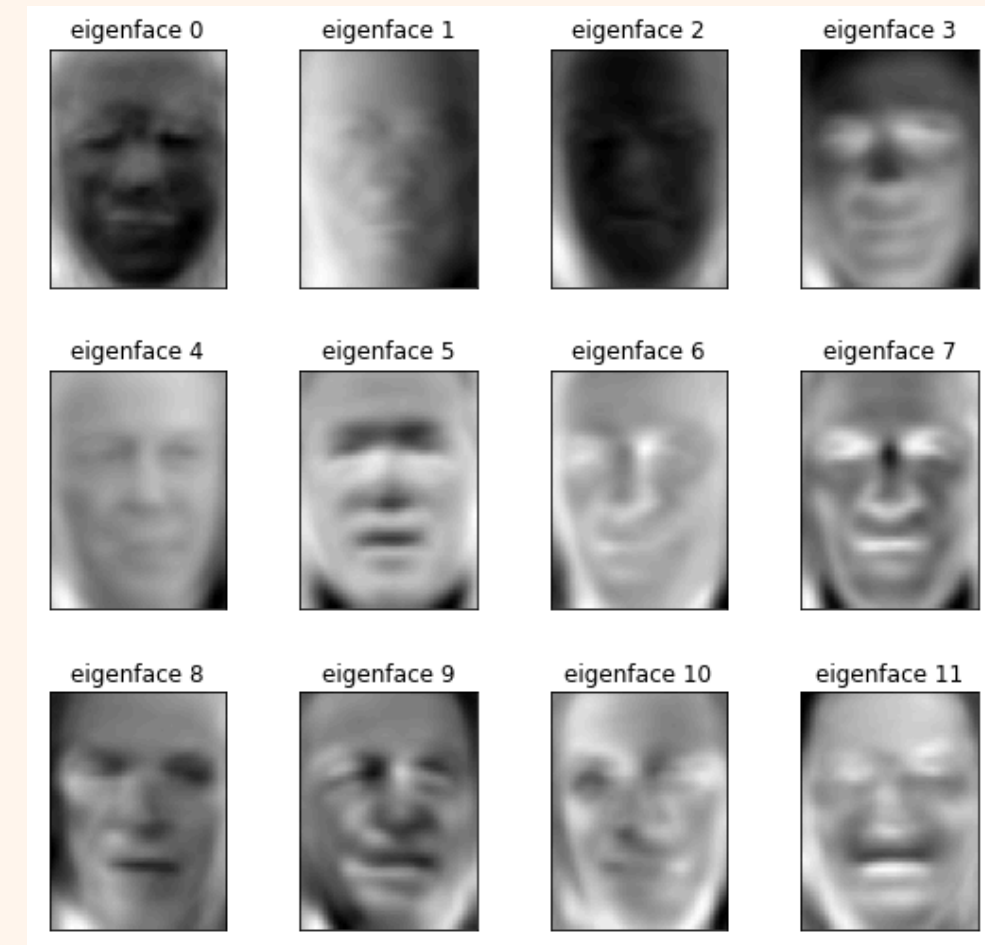
Select the K eigenvectors of C' corresponding to the K largest eigenvalues (where K < M). These eigenvectors has size N^2 .

After that we used the eigenvectors that we got in previous step. We take the normalized training faces (face – average face) x and represent each face vectors in the linear of combination of the best K eigenvectors (as shown in the diagram).

$$x_i - \psi = \sum_{j=1}^K w_j u_j$$

After that we take the coefficient of eigenfaces and represent the training faces in the form of a vector of those coefficients.

$$x_i = \begin{bmatrix} w_1^i \\ w_2^i \\ w_3^i \\ \vdots \\ w_k^i \end{bmatrix}$$



Testing

- Given an unknown face y , we need to first preprocess the face to make it centered in the image and have the same dimensions as the training face
- Now, we subtract the face from the average face ψ

$$\phi = y - \psi$$

- Now, we project the normalized vector into eigenspace to obtain the linear combination of eigenfaces. $\phi = \sum_{i=1}^k w_i u_i$
- From the above projection, we generate the vector of the coefficient such that

$$\Omega = \begin{bmatrix} w_1 \\ w_2 \\ w_3 \\ \cdot \\ \cdot \\ w_k \end{bmatrix}$$

- We take the vector generated in the above step and subtract it from the training image to get the minimum distance between the training vectors and testing vectors

$$e_r = \min_l \|\Omega - \Omega_l\|$$

- If this e_r is below tolerance level Tr , then it is recognised with l face from training image else the face is not matched from any faces in training set.