# Eigenfaces

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#### • Key principle:

Turn image array into long vector Represent sample image (face) as weighted sum of eigenimages (eigenfaces)

## PCA-based Face Recognition

Eigenfaces (Turk & Pentland 1991)
Representation of faces using PCA directly on image intensities
One of most famous uses of PCA in computer vision
Seminal reference for face recognition (but would work better if we modeled shape variation rather than lightness variation)

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## **Eigenfaces**

- 1. Given set of K registered face images  $(R \times C)$  with varying capture conditions
- 2. Represent as  $R \times C$  long vectors
- 3. Do PCA (special trick for large matrices)

Eigenface Recognition Slide 5/8 Eigenface Recognition Slide 6/8



Mean face and subset of principle component axes/images [Morris '04]



4. Represent person i by projection weights  $\vec{w_i}$ 

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## Eigenface Results

 $2500\ 128 \times 128$  image database, varied lighting

- 96% successful recognition over lighting variations
- 85% over orientation variations
- 64% over size variations

## Eigenface Recognition

Given unknown face image  $F_n$ 

- 1. Subtract mean face and project onto eigenfaces  $\rightarrow \vec{w_u}$
- 2. Given database of projections  $\{\vec{w_i}\}_{i=1}^K$ , find class c with smallest Mahalanobis distance  $d_c$  to  $\vec{w}_u$
- 3. If  $d_c$  small enough, return c as identity

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#### Eigenface Discussion

- Variations in position, orientation, scale & occlusion cause problems
- Research topics
- 4-36% failure rate a problem at busy airports