

PCA: Part II

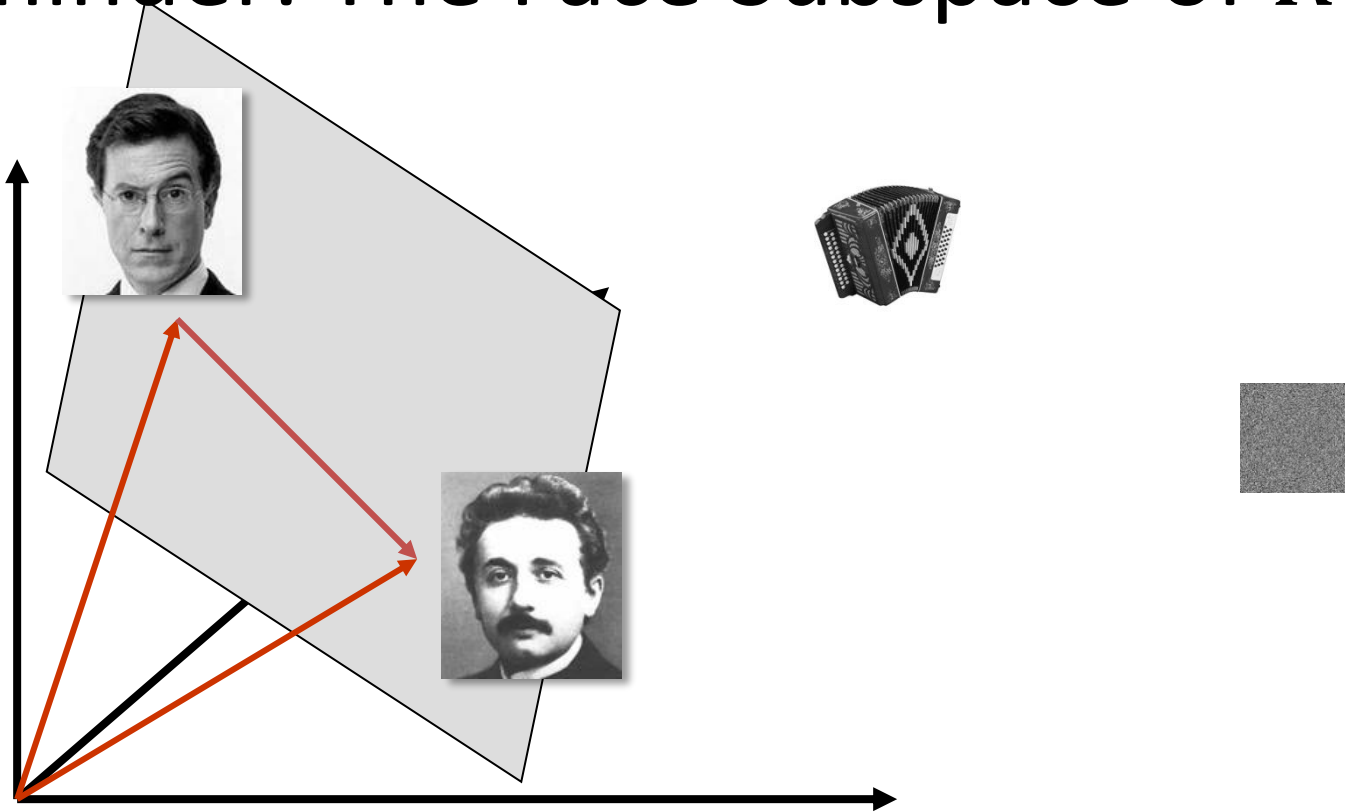


Pablo Picasso, "The Kitchen"

CSC320: Introduction to Visual Computing
Michael Guerzhoy

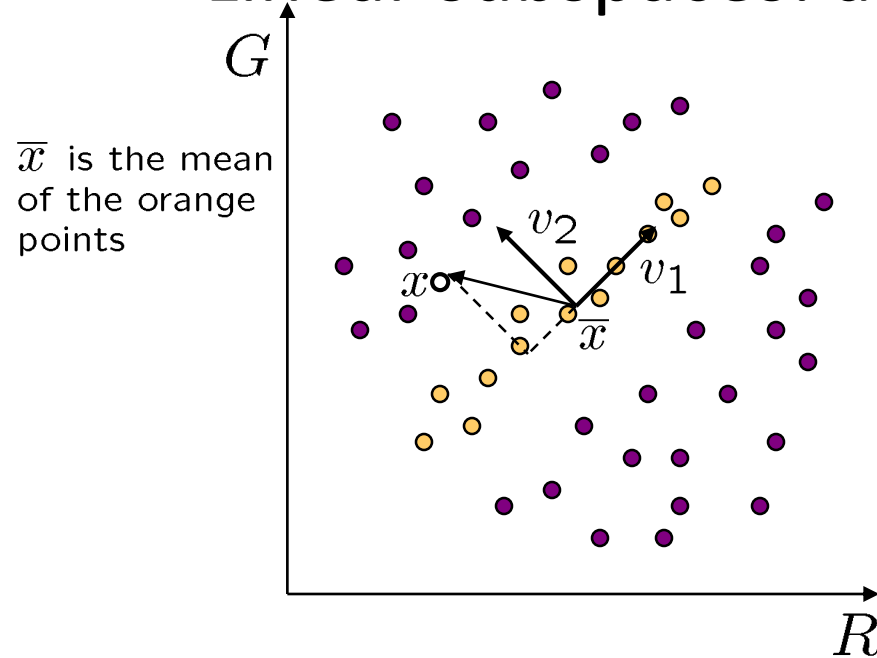
Many slides from
Noah Snavey, Derek Hoeim, Robert Collins

Reminder: The Face Subspace of R^n



- All the images of size n pixels are vectors in R^n
 - Basis: $\{(1,0,0,0,\dots)'$, $(0,1,0,0,0,\dots)'$, $(0,0,1,0,0,\dots)$, ...}
- Hope/assumption/model: images of (centered) faces of size n pixels are all (approximately) lying in a k -dimensional *subspace* of R^n
 - We can find the best subspace (i.e., k vectors of size n) using PCA

Linear subspaces: a 1-D subspace of a Plane



convert \mathbf{x} into $\mathbf{v}_1, \mathbf{v}_2$ coordinates

$$\mathbf{x} \rightarrow ((\mathbf{x} - \bar{\mathbf{x}}) \cdot \mathbf{v}_1, (\mathbf{x} - \bar{\mathbf{x}}) \cdot \mathbf{v}_2)$$

What does the \mathbf{v}_2 coordinate measure?

- distance to line
- use it for classification—near 0 for orange pts

What does the \mathbf{v}_1 coordinate measure?

- position along line
- use it to specify which orange point it is

Aside: Why are the Faces Centered?

- Centering: if the faces are the columns of matrix X , the centered faces are $X - \bar{X}$, where \bar{X} is the average column of X
 - The average column of $(X - \bar{X})$ is 0
- All linear spaces must contain 0

Representation and reconstruction

- Face \mathbf{x} in “face space” coordinates:



$$\mathbf{x} \rightarrow [\mathbf{u}_1^T (\mathbf{x} - \mu), \dots, \mathbf{u}_k^T (\mathbf{x} - \mu)]$$
$$= w_1, \dots, w_k$$

- Reconstruction:



=



+



$$\hat{\mathbf{x}} = \mu + w_1 \mathbf{u}_1 + w_2 \mathbf{u}_2 + w_3 \mathbf{u}_3 + w_4 \mathbf{u}_4 + \dots$$

Face Detection using PCA

- For each (centered) window x and for a set of principal components V , compute the Euclidean distance $|VV^T x - x|$
- That is the distance between the reconstruction of x and x . The reconstruction of x is similar to x if x lies in the face subspace

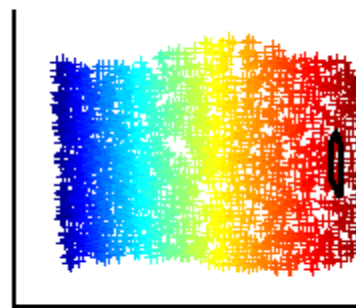
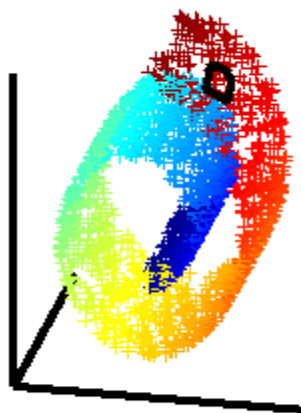
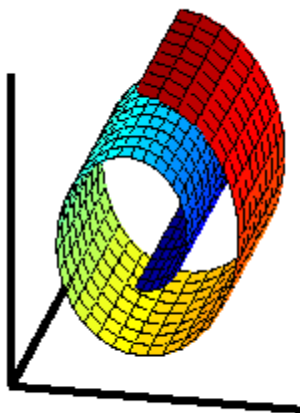
Note: the reconstruction is *always* in the face subspace

- Win: instead of comparing x to a large dataset of faces, we are only comparing x to the columns of V
 - $V^T x$ is just a vector of the dot products $v_i \cdot x$ for every i
 - That still works, since V contains (we hope) all the information about the appearance of faces that there is

Issues: dimensionality

What if your space isn't *flat*?

- PCA may not help



Nonlinear methods

LLE, MDS, etc.

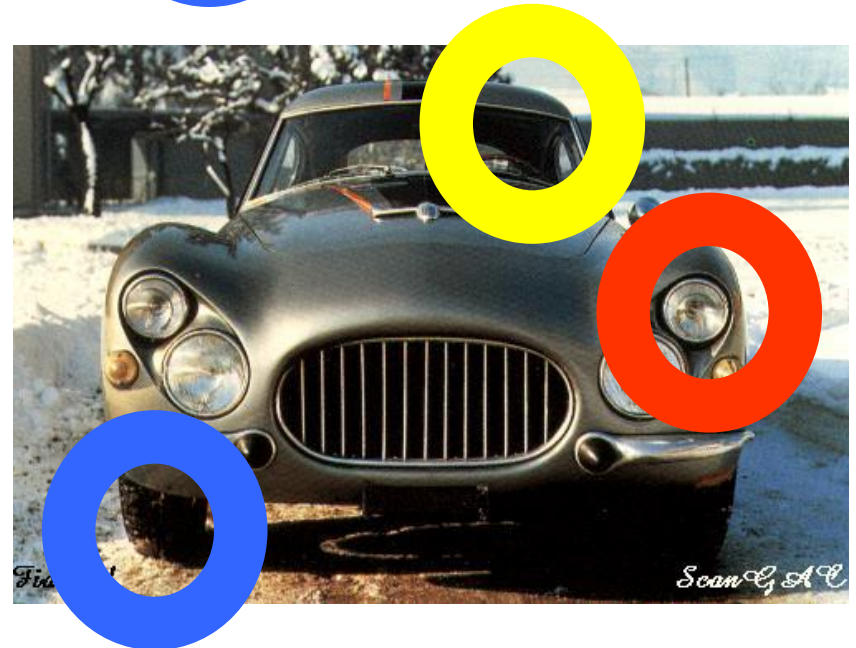
Moving forward

Faces are pretty well-behaved

- Mostly the same basic shape
- Lie close to a low-dimensional subspace

Not all objects are as nice

Different appearance, similar parts



Idea: Denoising images of the letter “a” with PCA

- Denoising: taking an image corrupted by some noise process, and recovering the original
 - Recall: convolving with a Gaussian filter worked pretty well for Gaussian noise, median filtering worked pretty well for salt-and-pepper noise
- Idea: take the noisy image x , and reconstruct it using PCA
 - In other words, project x onto the subspace of “a”s