

Eigenfaces

A Generative Model for Human Faces

Uncompressed Bitmaps

- What is a greyscale image?
- From Wikipedia's entry on Bitmap

Pixel storage [edit]

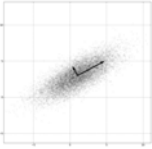
In typical **uncompressed bitmaps**, image pixels are generally stored with a **color depth** of 1, 4, 8, 16, 24, 32, 48, or 64 bits per pixel. Pixels of 8 bits and fewer can represent either **grayscale** or **indexed color**. An **alpha channel** (for transparency) may be stored in a separate bitmap, where it is similar to a **grayscale bitmap**, or in a fourth channel that, for example, converts 24-bit images to 32 bits per pixel.
- 8 bit: A matrix of numbers, 0 through 255.
- Octave
 - Assign a matrix to A
 - `imshow(A)`
- Linear Algebra
 - What is a matrix, but a vector? `reshape(A, numel(A), 1)`

Covariance

- <http://en.wikipedia.org/wiki/Covariance>
 - Covariance is a measure of how much two variables change together
- http://en.wikipedia.org/wiki/Covariance_matrix
 - Consider all the pixels. We want to know how pixel one changes with respect to pixel two...
 - The covariance matrix, Σ , will be 256x256 for a 16x16 image.
 - For a 256x256 image, the covariance matrix will have 65536 rows!
- How to compute the covariance matrix?
 - For 100 images, data = (65k x 100) matrix, with each image as a row.
 - Subtract the mean image from each to get A. Then...
 - $\Sigma = A A^T$
- But... Do we really need the whole thing?

PCA (Principle Component Analysis)

- http://en.wikipedia.org/wiki/Principal_component_analysis
 - PCA is mathematically defined as an orthogonal linear transformation that transforms the data to a new coordinate system such that the greatest variance by any projection of the data comes to lie on the first coordinate (called the first principal component)
- This amounts to an eigenvalue problem.
- Wanted: Eigenvalues of $\Sigma = A A^T$
 - Really, find eigenvalues of a 65k x 65k matrix?
 - Sort of...




Find Eigenvalues of AA^T

- Since AA^T is too hard, let's see if we can find a relationship between eigenvalues/eigenvectors of AA^T and $A^T A$.
- Suppose v is an eigenvector of $A^T A$, corresponding to λ .
- What does that mean about eigenvalues/eigenvectors of AA^T ?
- How is this helpful? Why is this easier to compute?

Result of PCA

- The higher the eigenvalue, the more variability is explained by that direction. Take the first 20. These are their eigenvectors.



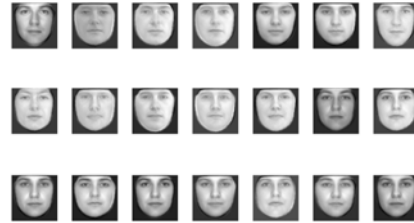
Eigenfaces

How Well Do the Top 20 Do?

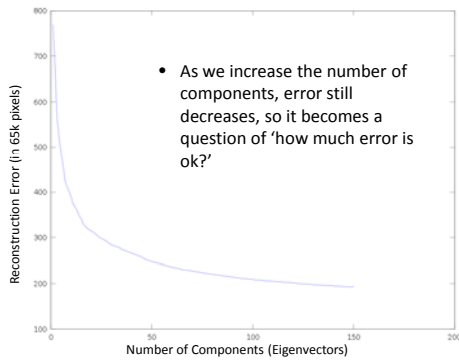
- Project a NEW image of a face onto those first 20 directions. We can display the face (and calculate per pixel error)
- How to project?
 - Observation 1: The eigenvectors will be orthogonal. Why?
 - Observation 2: We must normalize them. How?
 - Observation 3: We are in Euclidean space with the Euclidean inner product, dealing with orthonormal vectors. How can we calculate the projection?

Result of Projection

- We get faces, and they look different (potentially recognizable)!

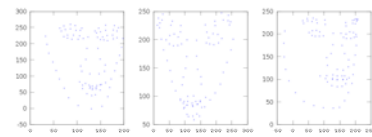


How Many Eigenvectors to Take?



Can We Do Better?

- Of course!
- Transform each image so that landmarks are aligned.



- Projections without Landmark Alignment



- Projections WITH Landmark Alignment



Want More Information?

- As always, wikipedia is a good starting place
 - <http://en.wikipedia.org/wiki/Eigenface>