

Admin

- Lab 5 grades posted
- No lab on Nov. 5
- Prof Mathieson will cover Nov. 6 lecture

Dimensionality Reduction - used often for data visualization

- Principal Component Analysis (PCA) - algorithm that transforms high-dimensional data to usually 2 dimensional, or lower dimensional data
- Helps us see relationships that might not be obvious in high-dimensional form

Global Population structure example

- Similar ethnic groups can be seen in clusters based on principal component 1 and principal component 2 in 2 dimensions
- Populations splits separate populations
- Admixture merges populations
- Principal component 1 captures the biggest split in the data

PCA application: Eigenfaces

- Faces are inherently high-dimensional
 - You have all the pixels and their RGB values
- Can use PCA to lower the dimension to capture different features in the human face
- Can be combined with to rebuild the original image
- Compare how faces are represented by the basis set
 - Weighted representation of the number of eigenfaces
- All the features are distinct and represent different parts of the face

PCA for data visualization:

- Step 1: original matrix is $n \times p$
 - Where p is much larger than n
 - Goal: create $n \times 2$ matrix for visualization
- Step 2:
 - Subtract off column-wise mean

$$\begin{bmatrix} 2 & 1 \\ 3 & 3 \end{bmatrix}$$
$$\bar{x}_1 = 2.5 \quad \bar{x}_2 = 2$$

$$\begin{bmatrix} 2-2.5 & 1-2 \\ 3-2.5 & 3-2 \end{bmatrix}$$

$$\begin{bmatrix} -.5 & -1 \\ .5 & 1 \end{bmatrix}$$

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- Step 3:
 - Compute covariance matrix A
 - Same idea for multiple features

Step 3

f, g, h

$$A = \begin{bmatrix} \text{cov}(f, f) & \text{cov}(f, h) & \text{cov}(f, g) \\ \vdots & & \end{bmatrix}$$

$P \times P$

- - Runtime to calculate a pair of features is $O(n)$
 - To compute covariance matrix: $O(np^2)$
- Step 4:
 - Compute eigenvalues and eigenvectors of A
 - v is being scaled, but direction is not changing

Step 4

$$I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\lambda I = \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix}$$

- Principle components that we end up with are basically the eigenvectors
- Step 5:
 - Sort eigenvectors by eigenvalues (high->low)
 - Compute transformed data:
 - $T_{n \times r} = X_{n \times p} W_{p \times r}$