# Learning the parts and features of speech?

#### Lawrence Saul

## Computer & Information Science University of Pennsylvania

#### **Message for experts**

## This talk is not for you. (But stay anyway for the Karaoke.)

## **Overview**

#### Learning

What can automatic algorithms infer from large data sets of sensory inputs?

#### Advances

Learning the parts of objects. Sparse nonnegative decompositions.

#### Applications

Multiple f<sub>0</sub> estimation. Robust feature detection?

# **Exploratory data analysis**

Dimensionality reduction

How to discover low dimensional structure in high dimensional data?

Subspace methods

Compute maximum variance subspace. Project data into subspace.



# Learning parts of objects

#### Objects

Represent objects (such as images or speech) by high dimensional vectors.

#### • Parts

Represent parts by basis vectors of maximum variance subspace.

#### Model

Model objects as weighted sums of parts. Is this meaningful?

#### **Parts of faces?**



eigenfaces from 7562 images

(MIT Media Lab)

Hard to interpret as parts...

# Nonnegativity

- What if data is nonnegative?
   Ex: pixels of images, power spectra of speech.
- Dimensionality reduction

Data lives in high dimensional orthant. Project into low dimensional cone.

#### Constraints

Nonnegative basis vectors. Only constructive combinations.

## Nonnegative parts of faces



Parts resemble eyes, mouth, mustache, ... Basis yields sparse, distributed encodings.

("Mr. Potato Head" Model)

# **Nonnegative parts of digits**



#### Parts resemble cursive strokes. Digits are modeled as sums of strokes.

# **Analysis by synthesis**

#### • To recognize a face:

Which eyes, mouth, nose yield best match to observed face?

#### • To recognize a digit:

Which cursive strokes are composed to draw the digit?

#### Nonnegative deconvolution

Infer parts from object and basis vectors. Robust to missing parts?

# **Auditory scene analysis**

#### Parts as sources

View auditory scene as complex object. Constituent parts are individual sources.

Start simple: periodic sources

Basis vectors are harmonic stacks. Mixed signal is nonnegative superposition.



# Nonnegative deconvolution for estimating multiple f<sub>0</sub>

# Polyphonic music Smaragdis & Brown (2003) Abdallah & Plumbley (2003)

# Overlapping voices Goto (2000), Sha & Saul (2004)



# **Real-time applications**



Voic	e2Midi	
Input	Output	
Signal O	Mute Piano	)
Detection Thresholds	O MIDI Acoustic Grand Piano	).

## **Robust features for ASR**

Parts as features

face = eyes + nose + mouth
digit = sum of strokes
phoneme = ? + ? + ?

Parts of speech

Will nonnegative decompositions yield localized patterns in time-frequency?

Robustness

Glimpses, missing data, binary masks, multiband ASR: compatible with above?

# Conclusion

Dimensionality reduction

Learn parts of objects by projecting to lower dimensional spaces.

Nonnegativity constraints

Nonnegative decompositions can lead to more interpretable parts.

Other work

Nonlinear projections, manifolds, and continuous modes of variability.