Recognition & Organization of Speech and Audio

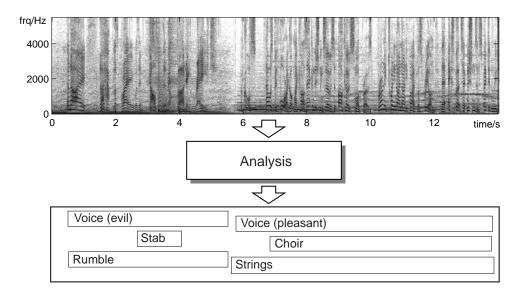
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Outline

- 1 Introducing LabROSA
 - Tandem modeling for robust ASR
- **3** Other current projects
- 4 Future projects
- **5** Summary & conclusions





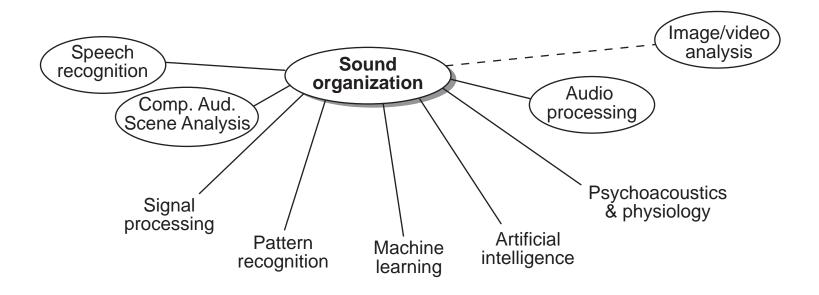


• Core operation:

Converting continuous, scalar signal into discrete, symbolic representation



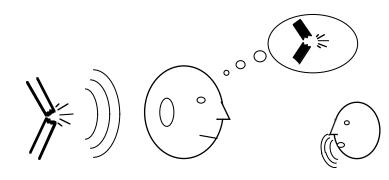
Positioning sound organization



- Draws on many techniques
- Abuts/overlaps various areas



About auditory perception



- Received waveform is a mixture
 - two sensors, N signals ...
 - need knowledge-based constraints
- Psychoacoustics: the study of human sound organization
 - 'auditory scene analysis' (Bregman'90)
- Auditory perception is ecologically grounded
 - scene analysis is preconscious (\rightarrow illusions)
 - perceived organization:
 real-world objects + events (transient)
 - subjective not canonical (ambiguity)



Key themes for LabROSA

http://www.ee.columbia.edu/~dpwe/LabROSA/

- Sound organization: construct hierarchy
 - at an instant (sources)
 - along time (segmentation)

• Scene analysis

- find attributes according to objects
- use attributes to form objects
- ... plus constraints of knowledge
- Exploiting large data sets (the ASR lesson)
 - supervised/labeled: pattern recognition
 - unsupervised: structure discovery, clustering
- Special cases:
 - speech recognition
 - other source-specific recognizers
- ... within a 'complete explanation'



Outline





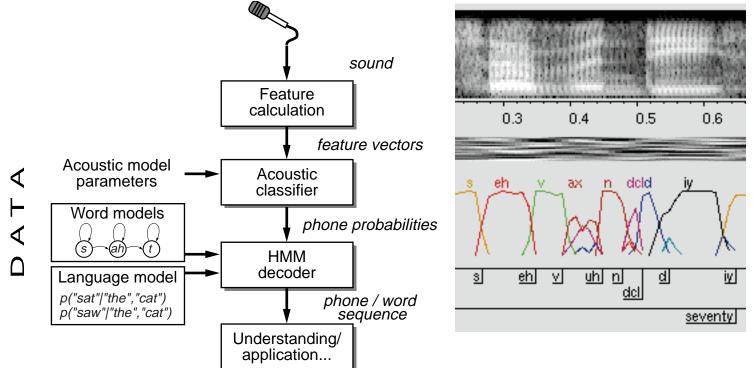
2 Tandem modeling for robust ASR

- ASR overview
- Tandem modeling
- Investigating the benefits
- Other current projects 3
- **Future projects**
- **Summary & conclusions** 5



Automatic Speech Recognition (ASR)

• Standard speech recognition structure:



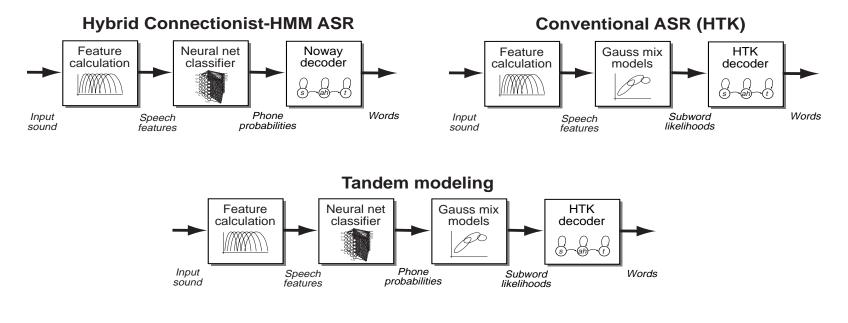
- 'State of the art' word-error rates (WERs):
 - 2% (dictation) 30% (telephone conversations)
- Can use multiple streams...



Tandem speech recognition

(with Hermansky, Sharma & Sivadas/OGI, Singh/CMU)

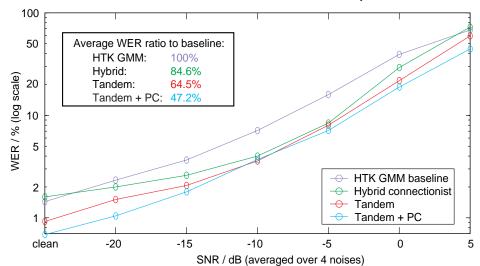
- Neural net estimates phone posteriors; but Gaussian mixtures model finer detail
- Combine them!



- Train net, then train GMM on net output
 - GMM is ignorant of net output 'meaning'

Tandem system results

• It works very well ('Aurora' noisy digits):



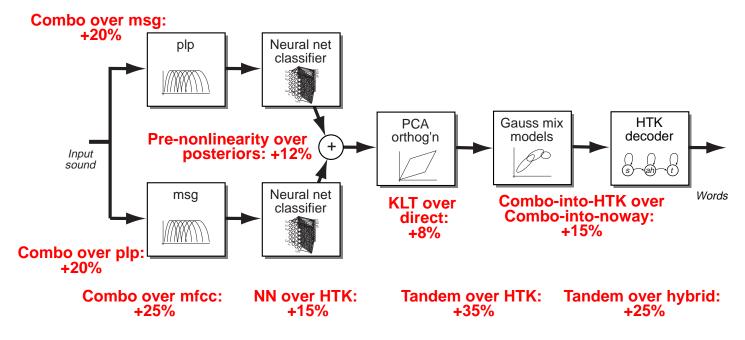
WER as a function of SNR for various Aurora99 systems

System-features	Avg. WER 20-0 dB	Baseline WER ratio
HTK-mfcc	13.7%	100%
Neural net-mfcc	9.3%	84.5%
Tandem-mfcc	7.4%	64.5%
Tandem-msg+plp	6.4%	47.2%

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Relative contributions

• Approx relative impact on baseline WER ratio for different component:

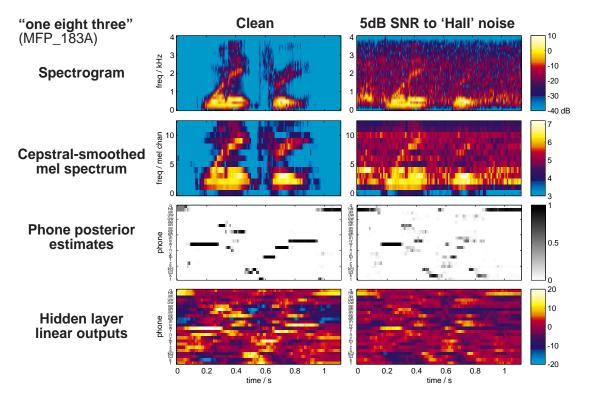


Tandem combo over HTK mfcc baseline: +53%



Inside Tandem systems: What's going on?

• Visualizations of the net outputs

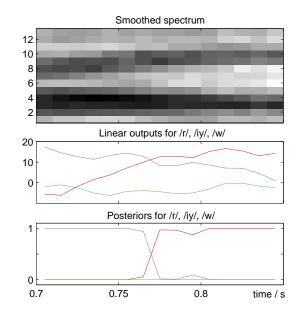


• Neural net normalizes away noise

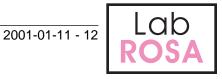


Tandem feature space 'magnification'

• Neural net performs a nonlinear remapping of the feature space



- small changes across critical boundaries result in large output changes



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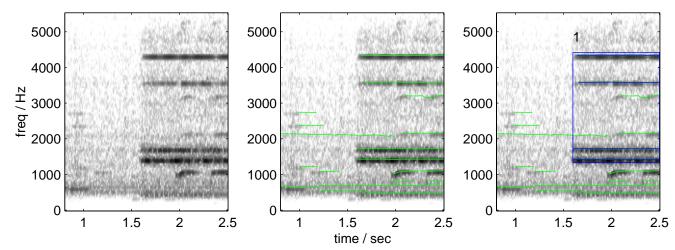
3 Other current projects

- Alarm sound detection
- Computational Auditory Scene Analysis
- Multi-source and missing-data recognition
- The Meeting Recorder project
- 4 Future projects
- **5** Summary & conclusions



Alarm sound detection

- Alarm sounds have particular structure
 - people 'know them when they hear them'
 - build a generic detector?

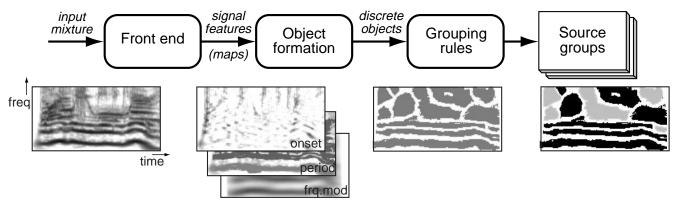


Isolate alarms in sound mixtures

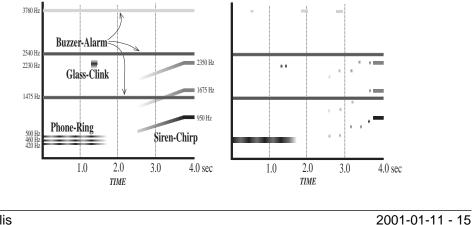
- representation of energy in time-frequency
- formation of atomic elements
- grouping by common properties (onset &c.)
- classify by attributes...

Computational Auditory Scene Analysis (CASA)

• Implement psychoacoustic theory? (Brown'92)



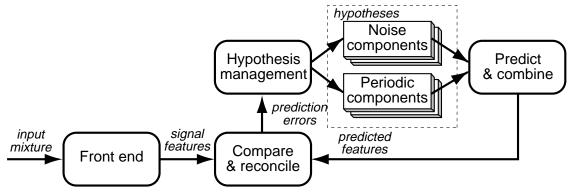
- what are the features? how are they used?
- Additional 'knowledge' needed (Klassner'96)





Prediction-driven CASA

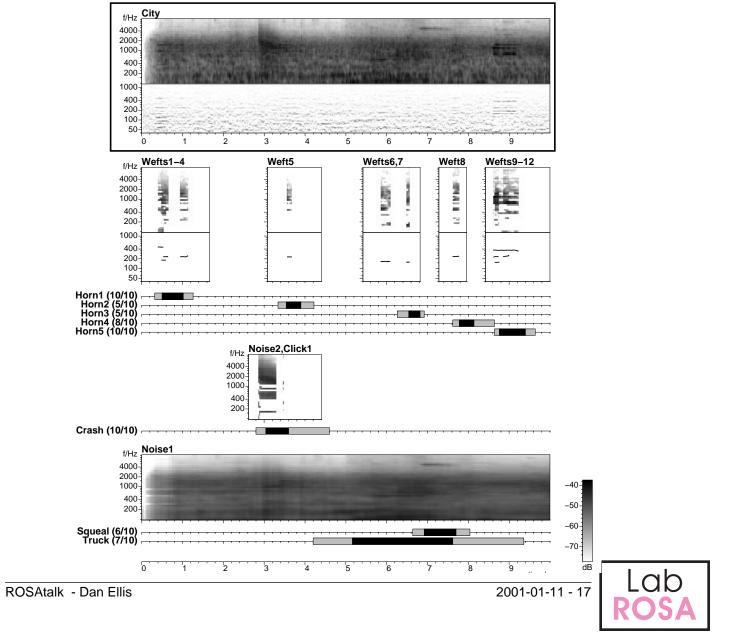
- Data-driven (bottom-up) fails for noisy, ambiguous sounds (most mixtures!)
- Need top-down constraints:



- fit vocabulary of generic elements to sound
 - ... bottom of a hierarchy?
- account for entire scene
- driven by prediction failures
- pursue alternative hypotheses



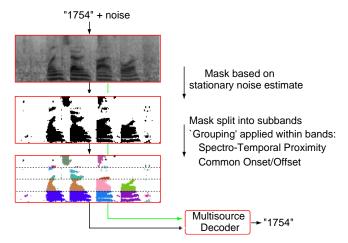
PDCASA example



Missing data recognition & CASA

(with Barker, Cooke, Green/Sheffield)

- Missing-data recognition
 - integrate across 'don't-know' values
 - 'perfect' mask \rightarrow excellent performance in noise
- Multi-source decoder
 - Viterbi search of sound-fragment interpretations



- CASA for masks/fragments
 - larger fragments \rightarrow quicker search



Meeting recorder

(with ICSI, UW, SRI, IBM)

• Microphones in conventional meetings

- for transcription/summarization/retrieval
- informal, overlapped speech
- Data collection (ICSI and ...):



- 10s of hours collected, ongoing
- now being transcribed



Meeting recorder: Research issues

• Preliminary analysis

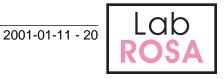
- transcription & forced alignment
- ground truth in turns/overlaps
- preliminary distant-mic recordings

• Research areas

- meeting dialog: overlaps, turns etc.
- language modeling for meetings
- feature design for distant acoustics

Applications

- information retrieval from meetings
- 'mapping' meeting content
- sociological analysis of meeting behavior



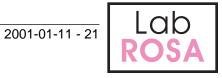
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- Audio Content-Based Retrieval
- A 'machine listener'
- Audio-video-text content analysis

5 Summary & conclusions

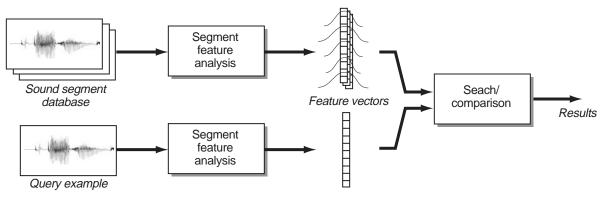


Audio Information Retrieval

- Searching in a database of audio
 - speech .. use ASR
 - text annotations .. search them
 - sound effects library?

• e.g. Muscle Fish "SoundFisher" browser

- define multiple 'perceptual' feature dimensions
- search by proximity in (weighted) feature space

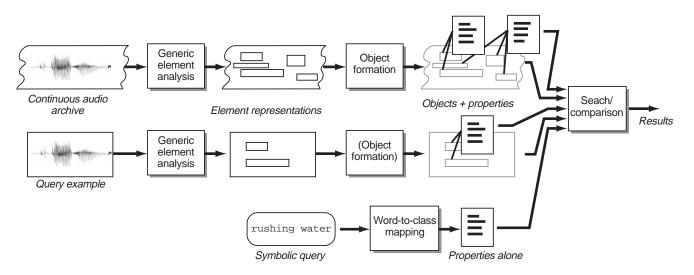


- features are 'global' for each soundfile, no attempt to separate mixtures



CASA for audio retrieval

- When audio material contains mixtures, global features are insufficient
- Retrieval based on element/object analysis:

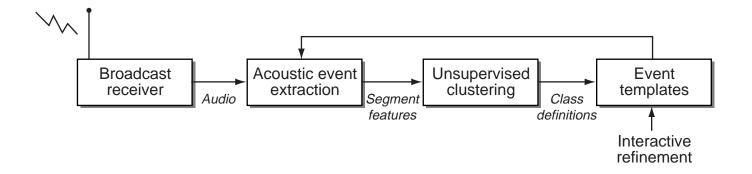


- features are calculated over grouped subsets



A 'machine listener'

• Goal: Unsupervised structure discovery

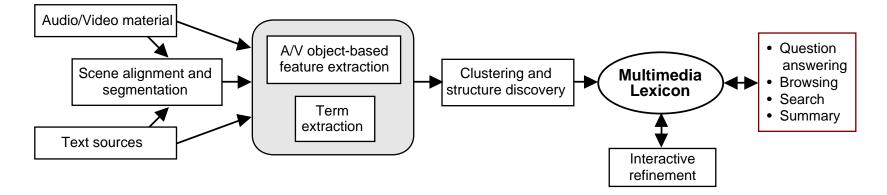


- What can you do with a large unlabeled training set (e.g. broadcast)?
 - bootstrap learning: look for common patterns
 - have to learn generalizations in parallel:
 e.g. self-organizing maps, EM HMMs
 - post-filtering by humans may find 'meaning' in clusters



Audio-video-text content analysis

(with Shih-Fu Chang, Kathleen McKeown)



- Audio and video provide complementary info
 - correlate object features to define templates?
- Associated text annotations provide a very small amount of labeling
 - .. but for a very large number of examples
 - sufficient to obtain purchase?
 - build a 'multimedia lexicon' for questionanswering



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Applications for sound organization

What do people do with their ears?

- Human-computer interface
 - .. includes knowing when (& why) you've failed
- Robots
 - intelligence requires perceptual awareness
 - Sony's AIBO: dog-hearing
- Archive indexing & retrieval
 - pure audio archives
 - true multimedia content analysis
- Content 'understanding'
 - intelligent classification & summarization
- Autonomous monitoring
- Broader 'structure discovery' algorithms



Summary

SOMAINS

- Broadcast
- Movies
- Lectures

- Meetings
- Personal recordings
- Location monitoring

ROSA

- Object-based structure discovery & learning
- Speech recognition
- Speech characterization
- Nonspeech recognition
- Scene analysis
 - Audio-visual integration
 - Music analysis

APPLICATIONS

- Structuring
- Search
- Summarization
- Awareness
- Understanding



Conclusions

- New classification schemes for ASR
 - ... combining multiple approaches/sources
- But sound is more than just speech!
 - speech is a special case
 - need to deal with the 'other stuff'

• Object-based analysis

- it's what people do
- the world presents acoustic mixtures
- Whole-scene representation
 - it's what people do
 - provides mutual constraints of overlap
- Broad range of approaches for a broad range of phenomena

