



Natural Speech Technology Programme Overview

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http://www.natural-speech-technology.org

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Natural Speech Technology Programme: Team



CSTR, University of Edinburgh:

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Speech Research Group, University of Cambridge:

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• Speech and Hearing Research Group, University of Sheffield:

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Overall aim



Significantly advancing the state-of-the-art in speech technology

- making it more natural
- applied to speech recognition and speech synthesis
- approaching human levels of

reliability

adaptability

fluency

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State of the art: Recognition & Synthesis



Learning from data – HMM/GMM framework

- Context-dependent modelling: divide and conquer using phonetic decision trees
- Speaker adaptation: MLLR and MAP families
- **Different training criteria**: maximum likelihood, minimum phone error, minimum generation error
- **Discriminative long-term features**: posteriograms, bottleneck features, deep networks
- Model combination: at the feature / distribution / state / model / utterance level

Key research themes



- Improving core speech technology
 - **Common modelling framework** for synthesis and recognition
 - Fluency
 - Capturing richer context
 - Personalisation
 - Expression and prosody

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Objectives



- I.Learning and Adaptation
- 2. Natural Transcription
- 3. Natural Synthesis
- 4. Exemplar **Applications** (driven by user group):
 - **homeService**: personalised speech technology to provide better interfaces (focus on older people & disabled people)
 - **lifeLog**: personalised wearable devices and transcribe/index all encountered audio
 - extracting structure from media archives

Learning and Adaptation



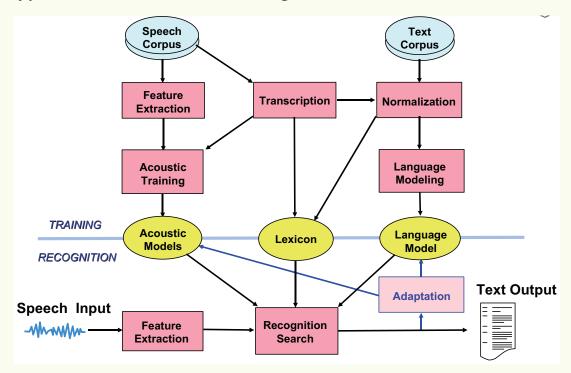
- Speech recognition and speech synthesis based on learning statistical models from data
- Current systems can adapt to the speaker or the domain automatically
- Challenges (for both recognition and synthesis)
 - Factoring models to different causes of variability
 - Almost instantaneous adaptation
 - Unsupervised training to take advantage of available data
 - Learning not to repeat mistakes

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Natural Transcription: Standard Systems



Typical architecture for training / test



Natural Transcription: Standard Systems



- **Acoustic models** (typically phone-context-dependent HMMs) trained from 100's to 1000's(+) hours of audio
- Language models (normally word N-grams) trained from large amounts of text data (up to billions of words) & lexicon normally fixed
- Training data is normally in-domain (including known language) with known correct transcripts (supervised).
- **Purpose-built** systems for different task domains e.g. meetings; broadcast news; voicemail etc
- Limited adaptation

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Current Transcription Systems: What's Wrong?



- Domain-specific systems (lots of point solutions)
 - All factors combined in one model & hence very data intensive (expensive, can't cover all situations)
- System performance isn't *natural*
 - slow to adapt
 - poor generalisation (accents, acoustic environment etc)
 - can't fully exploit all known context
 - doesn't produce fluent output

Natural Transcription



- Goal: Speech recognisers that
 - output "who spoke what, when, and how"
 - give high accuracy
 - have a wide coverage of speaker, environment etc
 - are flexible and minimise in-domain training data needs
 - can be personalised
 - produce fluent output

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Wide Domain Coverage



- Two approaches to transcription to cover many domains
 - Structure diverse data (from User Group + other sets) using advanced clustering techniques
 - Build canonical models (acoustic models and language models)
 which can be rapidly adapted to the particular sub-domain (incl
 speaker, environment etc)
- These overall approaches aim to allow
 - good performance on domains without any (or very limited) domain-specific training data
 - more robust speech recognition: works well for more situations

Factorisation

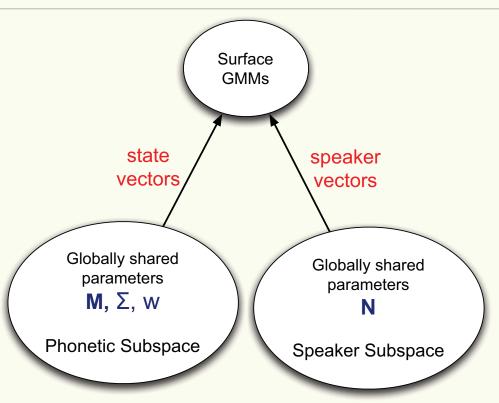


- Control effects of speaker, accent, style, acoustic environment, ...
- Factor different causes of variability (eg speaker, environment)
- Different approaches for both acoustic models & language models
- Allow use of speech knowledge in new situations
- Allows rapid adaptation to new situation or domain
- Can use subspace/canonical-state models for acoustic models
- Factorisation in recognition ⇔ control in synthesis

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Subspace GMMs

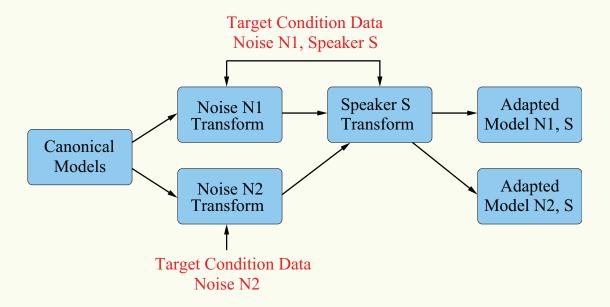




http://kaldi.sourceforge.net/

Canonical models for acoustic factorisation





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Use of Rich Contexts



- Extend the notion of context in speech recognition
 - beyond phone context in acoustic models
 - beyond word N-grams in language models
- More detailed context
 - Include extra information into models cf synthesis models
 - Model general situation highly specialised domains (speaker, location etc)
 - Adapt structured acoustic and language models

Metadata Use/Generation



- Generate additional information to augment / modify output.
- Some metadata may be given and is part of rich context.
 - Use metadata to generate appropriate recognition models
 - Factorised form allows rapid model generation
- Generate information on
 - Speakers & acoustic conditions
 - Emotional state
 - Sentence boundaries (slash units), disfluencies etc
 - Genre/style (eg meeting, comedy show etc etc)

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Diverse Accents/Dialects



- Structured accent models
 - Build on canonical models to capture accents & dialect variations
- Cross-lingual recognition
 - Minority language transcription with limited linguistic resources
 - Build on language independent knowledge of speakers, environments etc
 - Use models to help with areas normally requiring expert knowledge (e.g. pronunciation dictionaries)
 - Use factorised multi-level models

Environment Models



- Models that represent the physical acoustic environment. This includes representations of speaker and sound source location.
- Objective Robustness
 - Reverberation
 - Noise
 - Several moving speakers
- Recording facilities
 - Multichannel recordings (audio/video) with speaker location
 - Digital MEMS microphone array



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Generating Fluent Output



- Speech recognition systems normally give literal output
- For many applications need a more *natural*, *readable* output need to define naturalness ... (grammatical, acceptable)
- Transform/translate output as required for application
- Create suitable models of fluent data
 - Initial study based on use of N-grams and Combinatory Categorical Grammars (CCG): generation allowing substitution, permutation, insertion, deletion from ASR output
 - Promising initial results but hard to formally evaluate!

Applications: HomeService



- Personalised, interactive speech technology which can interact with environmental control systems and home monitoring device
- Integration of synthesis and recognition, to provide better interfaces to assistive technology
- Focus on older people and disabled people (& deal with dysarthric speech)
- Closely linked to work on voice restoration and voice banking

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HomeService



PC "box"; will run immediate audio collection and processing



Infrared transmitter for controlling TV, set-top box, curtains etc.



Microcone; can provide speaker diarization information.



Android tablet; mounted on wheel chair. Interface between system and user.





homeService - use case





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Initial Results on Dysarthric Speech



UASPEECH corpus

- 18 speakers
- 1000 different words spoken several times
- Speakers are assessed in terms of medical speech quality level covering a range from 2% - 95%.

Best results so far

- General triphone HMMs MAP adapted to domain and speakers
- Word correctness by speaker is ranging from 95% to 5% inverse to speech quality assessment, average 55%.

Application: Transcribing Media Archives



- BBC providing large quantities of data (aim to make all archives publicly available with searchable transcripts by BBC centenary in 2022)
 - Many genres and styles of broadcast audio data
 - radio incl. news, interviews/discussions, radio dramas
 - TV incl dramas, comedy, chat shows etc
 - Meta-data exists at various levels/types/completeness/accuracy: speakers, topics, genres & styles, subtitles, ...
 - Aim to structure models to produce targeted models given all available context
 - Pilot systems with wide range of error rates (!) from well under 20% (discussions, news) to over 60% (TV drama)

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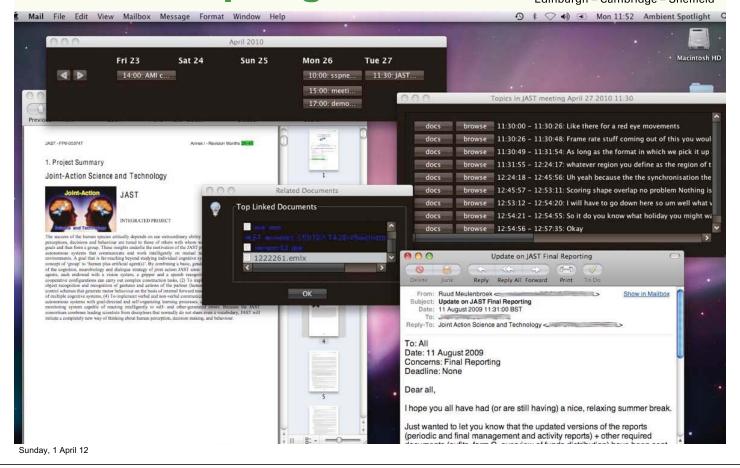
Application: Personalised Transcription



- Aim to create personalised transcription devices
 - Analyse data over extended period from particular speaker
 - High levels of adaptation/specialisation at all levels
 - Access to rich context about user and also those with which the user interacts
- Examples
 - *lifeLog:* Wearable, gathers information on user and the world with which the user commonly interacts
 - Ambient Spotlight: Recording, transcribing tutorials, linking to other related material

Ambient spotlight





Natural Synthesis: Standard Systems



• State-of-the-art: Intelligible synthetic voices, but not perceived as natural, very limited expressivity.

Unit selection

- large, carefully segmented inventory of speech from single speaker
- inflexible method based on cut-and-paste of phone-sized units
- widespread commercial use

• Statistical parametric ("HMM-based")

- flexible method based on a statistical model of the speech signal
- amenable to various modifications
- can be learned from speech of multiple speakers ('average voice model')
- can be adapted to new speakers, etc.

Natural Synthesis



- Long term vision: Fully controllable speech synthesis, indistinguishable from a human voice, with high intelligibility in all acoustic conditions.
- Goals within NST
 - Statistical parametric synthesis,
 - controllable in terms of speech parameters,
 - adaptable without new data,
 - personalisable with minimal data,
 - high degree of expressivity if required.

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Examples of synthesis work



- Creating voices from existing data, instead of high-quality studio recordings
- Speech-to-speech translation
- New models and extensions of HMM-based speech synthesis
- Clinical applications voice output communication aids
- Voice cloning

Voice cloning from normal speech



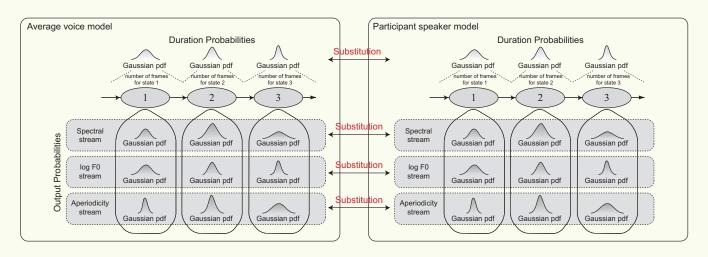
- Example voices
 - only 100 recorded sentences
 - non-professional speakers (volunteer 'voice donors')
 - average voice model adapted to this data

(examples)

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Reconstructing disordered speech: clinical applications





- "Mix-and-match" model components
 - some **copied** from an average voice model
 - some **learned** from the target speaker (patient)
 - some adapted from the average voice model to the target speaker

Reconstructed voices - Motor Neurone Disease





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A personalised voice output communication aid





Intelligibility



- Intelligibility is almost a solved problem, in good listening conditions
- in challenging conditions, or for hearing-impaired listeners, much work still to be done
 - synthesisers that **adapt** to the environment and the listener
 - initial work has been on speech in additive noise
 - can already get intelligibility gains in some conditions, without altering signalto-noise-ratio (SNR)
- What we plan to do next: adaptation to new listening environments without requiring additional data

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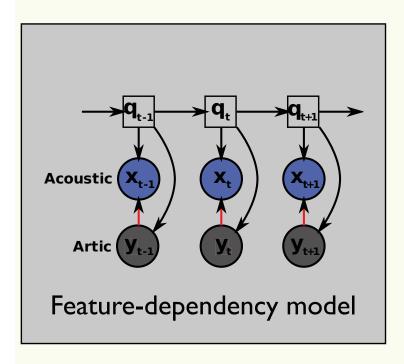
Naturalness and quality



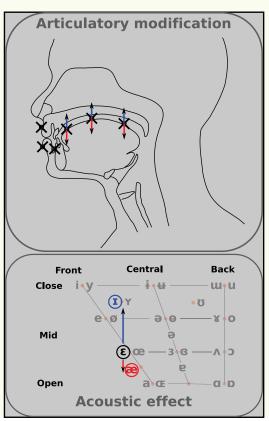
- Naturalness is far from being solved
 - Can achieve reasonable read-text speaking style using concatenative techniques, but these offer no flexibility or control
 - HMM-based method currently slightly less natural, but much more flexible
- Quality of HMM-based synthetic speech slightly behind the best unitselection, but catching up fast
- What we plan to do next: factored models that explicitly represent the structure of speech processes, the input text, and the listening situation

Speech knowledge Acoustic-Articulatory TTS





Ling, Richmond, Yamagishi & Wang, 2009



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Example: Controlling tongue height



Tongue height (cm)	+1.5	
	+1.0	
	+0.5	
	default	peck
	-0.5	
	-1.0	
	-1.5	

Summary



- High profile UK funded programme on speech technology: aiming to improve naturalness for both recognition and synthesis
- Many aspects being investigated including
 - Use of rich context information to build highly adapted / focused systems from limited training data
 - Use of factorised models
 - Personalisation
 - Healthcare applications including voice restoration
- Large user group

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Thanks.

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