Speech Biometric

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May 2007

Information in Speech

- **non-linguistic**, (*who said it*) gender, emotional states, speaker name
- linguistic (what he said) Language name and what was said (written language)
- paralinguistic (how well said manner, clarity or accent, aspects related to quality) deliberately added by the speaker, and not inferable from the written text.

Goal: Automatically extract information in speech signal

Speaker Recognition Vs Verification

```
Gallery /Edna/
/Sunil/
/Akhilesh/ /????/
/Dipti/
/Devanuj/
```

Response: Sunil

Speaker Recognition Vs Verification

/????/ Is this Sunil?

/????// Is this Sunil?

Response: Yes or No

Will concentrate on Speaker Verification

Speaker Verification: Overview

- Is the process of verifying the claimed identity of a registered speaker using his voice characteristics.
- The speaker needs to enroll before using the system.
- During enrollment, the speaker speaks a given set of utterances, using which the systems builds statistical (or deterministic) models characterizing the speaker's voice.
- A user claims he is X. Speaks a pass-phrase. The system gives a binary output YES (accept claimed identity) | NO.

Need for threshold to be able to say Yes or No.

Types of Speaker Verification

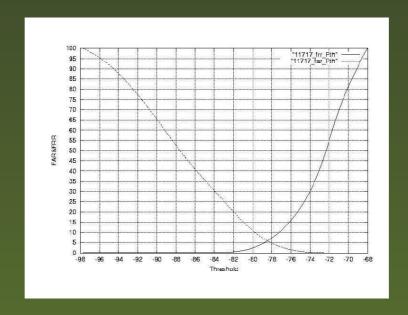
- 1. **Fixed Phrase** pre-determined phrase used for verification
- 2. **Fixed Vocabulary** verification more flexible and practical; training and testing materials for a speaker are generated based on words of a fixed vocabulary
- 3. Flexible Vocabulary a general set of sub-word phone models is created during speaker model training
- 4. **Text-Independent** user is not constrained to say fixed or prompted phrases

Clearly, both complexity and security increases as we go from fixed phrase to text-independent.

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Speaker Verification Terms

- FAR False Acceptance Ratio (The percentage of incorrect successful verifications)
- FRR False Rejection Ratio (The percentage of incorrect failed verifications)
- EER Equal Error Rate (The value at which FAR equals FRR)



Our Objective

- Utilize the speech signal processing knowledge to research and build in-house a state of the art speech recognition, synthesis, verification engine
- Understand and explore aspects of speech recognition technology which can be used to enhance the performance in actual use
- Utilize in-house engine to custom build applications, rather than use an off the shelf speech recognition system from a third party vendor as a black-box.

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- Functional for attendance monitoring at several locations of TCS

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- Select speech feature set which captures the identity of the speaker makes it very robust with very low FRR and FAR

Performance

	T_1	T_2
FAR	6.47%	1.75%
FRR	0.95%	10.90%

- Threshold (T_1) was chosen to be such that the FRR was close to 0% (pass all) and T_2 was chosen so that FRR was approximately 10%.
- Experiments carried out on a set of 15 speakers. Imposter's aware of the pass phrase (i.e. skilled forgery)
- 25 Parameters; Continuous HMM models used
- Ported and tested on BREW SDK2.0 simulator

Thank You