# Speaker Verification Based on the German VeriDat Database

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#### Abstract

Here we introduce

- the new German speaker verification (SV) database VeriDat of telephone speech,
- the baseline performance of our experimental speaker verification system
- and several experiments with different data splittings according to the network type and the acoustical environment.

#### **Motivation**

"Real world" telephone data show several difficulties, namely

- varying acoustical environments and telephone channels;
- linguistic and meta-linguistic "noise";
- erratic speaking style for some speakers ("goats" have unusually high false-rejection rate);
- sparse material for building models.
- ⇒ Introducing some knowledge about certain properties of the data might give a better performance in the verification task.
- ⇒ Further analyzing of the data might give clues to goat-like behaviour and how it could be detected a priori.

# The VeriDat database

- extension of standardized specification for SV databases in the SpeechDat project.
- 150 speakers covering the whole range of German dialects.
- Speech items used here: number triples (e.g. "21 35 76").
- different networks: fixed and cellular
- different background noises: quiet and noisy.

# System Design

| Setup & Experimental Protocol |               |  |  |
|-------------------------------|---------------|--|--|
| speaker set                   | # of speakers |  |  |
| client                        | 30            |  |  |
| world                         | 30            |  |  |
| impostor                      | 60            |  |  |
| development                   | 30            |  |  |

#### Features, Modelling and Score Normalization

- standard HTK features (LPCC).
- word based HMMs, left-right topology.
- temporal segmentation and normalization with world model.
- EERs are gender-balanced using a posteriori speaker-dependent thresholds.

#### **Experiments**

- same amount of training data for world and client models.
- cheating experiments in the domain "Fixed/GSM" (F/G) and "Quiet/Noisy" (Q/N).

| Exp.    | World Model(s)    | Client<br>Model(s) |
|---------|-------------------|--------------------|
| Base    | one model:        | one model:         |
|         | all cond.         | all cond.,         |
|         | 1260 recs.        | 20 recs.           |
| FQ      | one model:        | one model:         |
|         | FixedQuiet        | FixedQuiet         |
|         | 1260 recs.        | 20 recs.           |
| F/G 2/2 | two models:       | two models:        |
|         | Fixed, 1260 recs. | Fixed, 20 recs.    |
|         | GSM, 1260 recs.   | GSM, 20 recs.      |
| F/G 1/2 | one model         | two models         |
| F/G 2/1 | two models        | one model          |
| Q/N 2/2 | two models:       | two models:        |
|         | Quiet, 1260 recs. | Quiet, 20 recs.    |
|         | Noisy, 1260 recs. | Noisy, 20 recs.    |
| Q/N 1/2 | one model         | two models         |
| Q/N 2/1 | two models        | one model          |

Test parameters for base line and cheating experiments.

#### Results

| Experiment | EER mean % |
|------------|------------|
| Base       | 3.93       |
| FQ         | 2.30       |

Results for the baseline experiments.

| # World Models | # Client Models |      |
|----------------|-----------------|------|
|                | 1               | 2    |
| 1              | 3.93            | 4.45 |
| 2              | 3.93<br>3.89    | 4.22 |

Results (EER mean %) for the cheating experiment F/G.

| # World Models | # Client Models |      |  |
|----------------|-----------------|------|--|
|                | 1               | 2    |  |
| 1              | 3.93            | 3.66 |  |
| 2              | 4.47            | 3.93 |  |

Results (EER mean %) for the cheating experiments Q/N.

#### "Goat-like" behaviour



Histogram of LLH scores per frame of the world model using client's enrollment data.

# Conclusion

- Base line performance deteriorates when including all acoustical variations in the training material.
- Splitting the training material according to the network ("Fixed/Quiet") does not result in any improvement.
- Splitting the training material according to the background noise ("Quiet/Noisy") gives a slight gain in performance.
- A few client speakers show extremly low verification performance.
- Detection of "goat-like" behaviour is a crucial point for a real world SV system.

# **Online-Resource**

http://www.bas.uni-muenchen.de/Bas/SV/