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Roadmap

- Introduction
- Discriminatively Trained System
- Phonetically Inspired System
- Factor Analysis System
- Fusion Analysis
- Conclusions



Introduction

Introduction

- The IBM primary submission is a score-level linear combination of 6 core systems as follows:
 - 1 Phonetically inspired NAP-GMM system
 - 3 Discriminatively trained NAP-GMM systems
 - Varied by choice of front-end features, data subsets and system configurations.
 - 2 Factor analysis based systems
 - Trained on different frontend features: LPCCs and MFCCs
- The IBM alternate submission (a single system):
 - 1 Phonetically inspired NAP-GMM system (as above)

Discriminatively Trained System

Discriminative Regularization of Maximum Likelihood Estimation

- Estimating the UBM parameters using ML training is not related to reducing the speaker verification errors.
- We add a regularization term to the ML objective function to reduce the value of the imposter scores and increase the value of the target scores.
- The UBM parameters are updated using an EMlike algorithm.

The Objective Function



Discriminative Regularization System & Data System

- Classifier:
 - Dot-Product scoring of the GMM-based supervector representation of the enrollment and verification utterances developed from ASR front-end features.

Session Variability Compensation:

NAP and ZT score normalization.

Data

- UBM Training Database:
 - The NIST08 interview development data. (6 speakers, 6 sessions, 9 mics)
 - For NIST 2010 systems, also includes the NIST 2008 evaluation data.

NAP Training Database:

- 11400 conv-sides from SWB2, Dev08, NIST04, and NIST06 databases.
- For NIST 2010 systems, also includes the NIST 2008 evaluation data.
- ZT score normalization Database:
 - Same as NAP but divided into 2 gender-dependent subsets.



Interview Only Tasks of NIST08 – Discriminative Regularization

Table: (Norm. Min. DCF, EER (%)) for Baseline and Discriminative Regularization (DR) Systems

System	Int-Int-All	Int-Int-S	Int-Int-D
Baseline	(0.194, 4.2)	(0.029, 0.91)	(0.193, 4.1)
DR	(0.159, 2.7)	(0.019, 0.64)	(0.161, 2.8)

Task Description:

Int-Int-All: Interview speech in training and test.

Int-Int-S: Interview speech from the same (lapel) microphone in training and test.

Int-Int-D: Interview speech from different microphones in training and test.

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Telephone Tasks of NIST08 -Discriminative Regularization

Table: (Norm. Min. DCF, EER (%)) for Baseline and Discriminative Regularization (DR) Systems

System	Int-Tel	Tel-Mic	Tel-Eng	Tel-US
Baseline	(0.375, 7.9)	(0.323, 7.7)	(0.156, 3.5)	(0.164, 4.1)
DR	(0.329, 7.0)	(0.256, 7.2)	(0.140, 3.4)	(0.141, 4.1)

Task Description:

Int-Tel: Interview speech in training & telephone speech in test.

Tel-Mic: Telephone speech in training and telephone microphone speech in test.

Tel-Eng: English language telephone speech in training and test (any variety).

Tel-US: English language telephone speech spoken by a native US English Speaker in training and test.



Interview Tasks of NIST10 – Discriminative Regularization

Table: (Min. DCF, EER (%)) for Baseline and Discriminative Regularization (DR) Systems

System	Int-S	Int-D	Int-NTel	Int-NMic
Baseline	(0.39, 3.4)	(0.52, 5.1)	(0.38, 4.1)	(0.45, 3.4)
DR	(0.37, 3.1)	(0.43, 4.7)	(0.32, 3.4)	(0.42, 3.2)

Task Description:

Int-S: Interview speech (same-microphone) in training and test.

Int-D: Interview speech (different-microphones) in training and test.

Int-NTel: Interview speech in training and normal vocal effort telephone speech in test.

Int-NMic: Interview speech in training and normal vocal effort microphone telephone speech in test.



Phonetically Inspired System

UBM Construction From ASR Models





$$\begin{split} K \text{ Gaussian components, } U &= (u_1, u_2, \dots, u_K). \\ d(u_k, g_l) &= KL(u_k, g_l) + KL(g_l, u_k), \\ \text{and } KL(u_k, g_l) \text{ is the KL divergence between } u_k \text{ and } g_l. \end{split}$$



PIUBM System

Classifier:

System

 Dot-Product scoring of the GMM-based supervector representation of the enrollment and verification utterances based on ASR features.

Session Variability Compensation:

- NAP and ZT score normalization.

ASR System:

- ~2000 hours of SWB and Fisher databases.
- 13 PLP features compensated by CMVN+VTLN; 9-frame window + LDA 😿 40 dimensions.
- Speaker adaptation using FMLLR; FMPE transform estimated in the FMLLR feature space.

Data

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NAP Training Database:

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- For NIST 2010 system, also includes the NIST 2008 evaluation data.

ZT score normalization Database:

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Telephone Tasks of NIST08 - System Comparison

Table: (Norm. Min. DCF, EER (%)) for MFCC & ASR-frontend baselines, and PIUBM System

System	Int-Tel	Tel-Mic	Tel-Eng	Tel-US
Baseline	(0.375, 10.3)	(0.288, 7.4)	(0.156, 3.5)	(0.154, 4.4)
ASR frontend	(0.318, 7.6)	(0.218, 6.4)	(0.164, 3.4)	(0.154, 4.4)
PIUBM	(0.307, 8.6)	(0.221, 6.7)	(0.127, 2.7)	(0.116, 3.0)

Task Description:

Int-Tel: Interview speech in training & telephone speech in test.

Tel-Mic: Telephone speech in training and telephone microphone speech in test.

Tel-Eng: English language telephone speech in training and test (any variety of English).

Tel-US: English language telephone speech spoken by a native US English Speaker in training and test.

Telephone Tasks of NIST10 - System Comparison

Table: (Min. DCF, EER (%)) for Baseline and PIUBM Systems

System	NTel-NTel	NTel-HTel	NMic-HTel	NTel-LTel	NMic-LTel
Baseline	(0.67, 4.0)	(0.57, 2.9)	(0.43, 4.8)	(0.19, 1.1)	(0.28, 1.5)
PIUBM	(0.31, 2.9)	(0.58, 2.9)	(0.34, 5.8)	(0.25, 1.0)	(0.23, 1.5)

Task Description:

NTel-NTel: Normal vocal effort telephone speech in training & test.

NTel-HTel: Normal vocal effort telephone speech in training and high vocal effort telephone speech in test.

NMic-HTel: Normal vocal effort microphone telephone speech in training and high vocal effort telephone speech in test.

NTel-LTel: Normal vocal effort telephone speech in training and low vocal effort telephone speech in test.

NMic-LTel: Normal vocal effort microphone telephone speech in training and low vocal effort telephone speech in test.



Factor Analysis System

Factor Analysis

System

- Factor analysis point estimates used
- Log-likelihood ratio approximation to scoring
- ZT-Norm Applied + Symmetric scoring
- LPCC and MFCC features + feature warping

Data

FA Training Database:

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ZT score normalization Database:

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Factor Analysis – NIST 2008 Results

Task	LPCC (norm mDCF)	MFCC (norm mDCF)	Equal Weight Linear Fusion (norm mDCF)
1	0.098	0.143	0.077
2	0.019	0.029	0.017
3	0.101	0.149	0.079
4	0.223	0.225	0.171
5	0.175	0.166	0.128
6	0.326	0.306	0.303
7	0.114	0.079	0.080
8	0.125	0.093	0.088



Factor Analysis – MFCC system (NIST 2010 Result)



NIST 2010 Speaker Recognition Evaluation Workshop

Factor Analysis – MFCC system (NIST 2010 Result*)



*2010 Result using NIST 2008 min. DCF

Factor Analysis – LPCC system (NIST 2010 Result)





Fusion Analysis

NIST 2010 Speaker Recognition Evaluation Workshop

Data Fusion – NIST 2010 System Breakdown (min. DCF)

Task	Primary	Alternate	Best Single (of 6 systems)
1	0.498	0.398	0.371 (D10)
2	0.463	0.521	0.432 (D10)
3	0.311	0.383	0.325 (D10)
4	0.539	0.401	0.401 (Alt)
5	0.279	0.312	0.312 (Alt)
6	0.528	0.582	0.572 (D10)
7	0.422	0.342	0.342 (Alt)
8	0.259	0.258	0.198 (D10)
9	0.683	0.237	0.237 (Alt)

- D10 -Discriminatively Trained, 10 iteration system - Alt -IBM' s alternate system submission



Conclusions

Conclusions

- Discriminatively trained and Phonetically Inspired UBMs provided noteworthy contributions to overall system performance for NIST 2010.
- LPCC+MFCC Factor Analysis System components performed extremely well on NIST 2008 data. Performance did not carry across to the NIST 2010 data:
 - Suggests a possible implementation issue... additional follow-up required here
- The data set developed for fusion optimization and general development was not representative of the 2010 tasks.
 - A mixed bag of fusion results across the 9 conditions.