

SUMMARY OF PROGRESS AT LIMSI

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LIMSI-CNRS

TLP Group

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MAJOR TASKS



- CTS English (J.L. Gauvain, L. Lamel, H. Schwenk, G. Adda, F. Lefevre)
- BN English (J.L. Gauvain, L. Lamel, H. Schwenk, G. Adda, F. Lefevre)
- BN Non English (J.L. Gauvain, L. Lamel, A. Messaoudi, L. Chen)
- Speaker diarization (C. Barras, S. Meignier, X.Zhu, J.L. Gauvain)



GENERAL IMPROVEMENTS

- Better acoustic modeling (speaker adaptive training, full-covariance for state-tying, semi-tied covariance models)
- Faster software and infrastructure for acoustic model training on large corpora (light supervision for BN)
- Improved acoustic model adaptation
- Updated language models and dictionaries for all tasks
- Speed up training and decoding for neural network language models
- Faster decoding for CTS and BN
- Large improvement on speaker diarization task



PROGRESS ON CTS ENGLISH

- Better acoustic models (1.7% absolute reduction)
- Improved acoustic and language models and lexicon using Fisher training data (2.5% absolute)
- Faster and better decoding with AM adaptation (0.4% absolute with a factor of 6 speed-up)
- Multiple phone set modeling (0.4 0.7% absolute)
- Cascade and Rover combination with BBN
- Overall relative error reduction without Rover combination of about 23%



PROGRESS ON ENGLISH BN

- Focused on cross-site combination
- Integrated feature optimization (MLLT) for diagonal Gaussians and speaker adaptive training (SAT)
- Improved unsupervised adaptation (CMLLR, MLLR, with variable number of adaptation classes)
- Training on 600 hours of BN data, with light supervision
- Neural network continuous space LM
- About 20% relative WER reduction since RT03



NEURAL NETWORK LM

- Training speed-up (better algorithms and code tuning)
 Factor of 30 speed-up
- Recognition speed-up (improved lattice rescoring)
 0.6xRT → 0.06xRT
 including lattice expansion and consensus decoding
- Word error reduction relative to the 4-gram backoff LM CTS dev04: 16.0% → 15.5% (B1-L1, 6xRT) BN dev04: 10.4% → 10.1% (B1-L1, 6xRT) Trained on about 27M words for BN and CTS

SYSTEM COMBINATION



- Two decoding strategies for cross-site system combination
- Lattice rescoring:
 - Faster (allows computation to be shared between systems)
 - Need to deal with compound words and OOVs
 - Reduces combination gain (due to sharing of components)
- Full search:
 - Requires fast decoding
 - Results in higher WER, but comes out ahead after combination
- Cross-site adaptation with fast full decode found to be more efficient for the BBN/LIMSI and SuperTeam combinations



PROGRESS ON NON-ENGLISH BN

Mandarin

• Improved acoustic models with lightly supervised training with TDT Word error reduction of over 10% on dev03 and eval03

Arabic

- Explicit modeling of short vowels (AMs, lexicon)
- 65k word class 4-gram language model regrouping all vowelized forms for each non-vowelized entry
- Dictionary has over 400k vowelized forms, obtained semi-automatically
- Acoustic model training on 150 hours of data
- RT-04F Primary system WER 18.5%, Contrast system 20.2%

PROGRAM-WIDE CONTRIBUTIONS



- LIMSI delivered aligned ASR transcripts with CC to LDC for fast manual correction
- Led design of the STT BN dev'04 set (targeting progress set WER)
- Collaborated with CU to produced MDE alignments for Eval03S and Eval03F data
- Alignment software packaged for NIST h4align, h5align, 150k dictionary
- STT website (slides of STT meetings, information exchange)
- Participation in weekly STT teleconferences

MAJOR ACHIEVEMENTS



- Participated in 4 RT-04F tasks (5 systems): CTS and BN English, BN Arabic, Speaker Diarization
- CTS English
 - Reduced WER by 23% and factor of 6 speed-up
 - Efficient BBN+LIMSI cascade/Rover system combination
 - BBN+LIMSI system 13.5% on progress test
- BN English
 - Concentrated effort on system combination
 - Reduced relative WER by about 20%
 - Efficient BBN/LIMSI combination: 9.5% on progress test
 - Also part of SuperTeam: 8.6% on progress test
- LIMSI STT partitioner output used by 4 of the 6 primary BN English RT-04 submissions

RT-04 TECHNICAL TALKS



- The RT04 BBN/LIMSI 20xRT English CTS System (J.L. Gauvain, R. Prasad)
- The RT04 BBN/LIMSI 10xRT BN English System (L. Nyugen, L. Lamel)
- Alternate phone models for CTS (L.Lamel)
- Using neural network language models for LVCSR (H. Schwenk)
- Improving Speaker Diarization (C. Barras)
- Towards using STT for Broadcast News Speaker Diarization (L.Lamel)
- The LIMSI RT04 BN Arabic System (J.L. Gauvain)

CONCLUSIONS



- Good progress on all tasks
- Strong teamwork with BBN (also SuperTeam participation)
- Participated in 4 RT-04F tasks (5 systems)
- Met program targets
- Still porting some CTS improvements to BN
- Training on large corpora: major effort, still in progress
- Some research did not make the RT-04 eval (not ready for integration, CPU time constraints) examples: dynamic LM, duration model, full covariance, syllable-position based phone modeling, discriminative LM training