

The 2016 KIT IWSLT Speech-to-Text Systems for English and German

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Outline



- IWSLT 2016 ASR Tasks
 - **English Talk**
 - **English and German MSLT**
- System Overview
- Evaluation Setups
 - Feature Extraction
 - Feed-forward and LSTM LM
 - GMM & DNN Systems
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- Results and Discussions
- Conclusion

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IWSLT 2016 ASR



- English Talk task
 - TED talks and talks from the QED Corpus.
 - Various topics, spontaneous speaking style
 - Not segmented
- English and German MSLT task
 - Conversations conducted via Skype
 - With provided segmentations
 - Unknown speakers



16-12-08



System Overview



Evaluation Setups

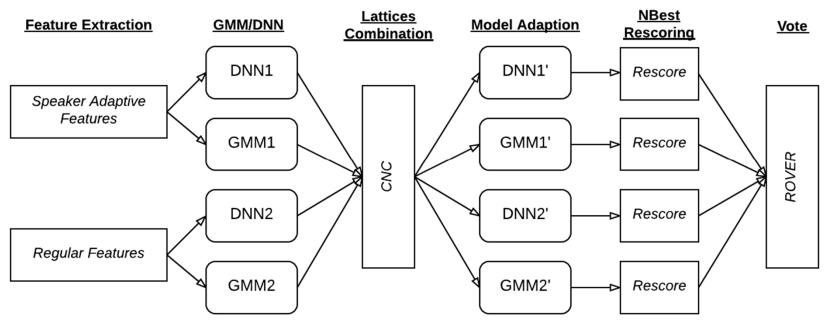


Results



System Overview





The pipeline involving the techniques to build final systems

IWSLT 2016 ASR \supset System Overview \supset Evaluation Setups \supset Results \supset Conclusion

Evaluation Setups



- Feature Extraction
 - Bottleneck features
 - Speaker adaptive feature (SAF)
- Language Models
 - Feed-forward
 - LSTM LM
- Systems
 - GMMs and DNNs using SAF
- Speaker Adaption Models



16-12-08



System Overview



Evaluation Setups

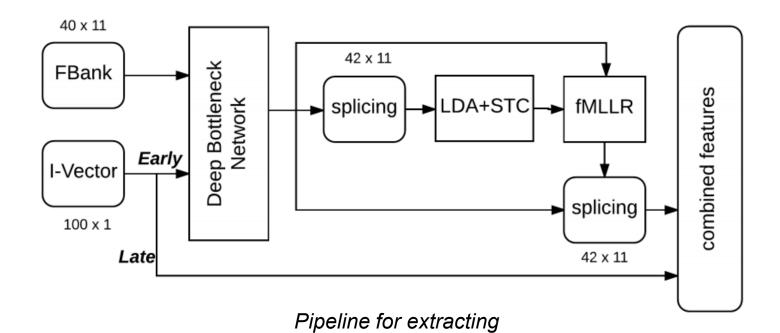


Results



Feature Extraction



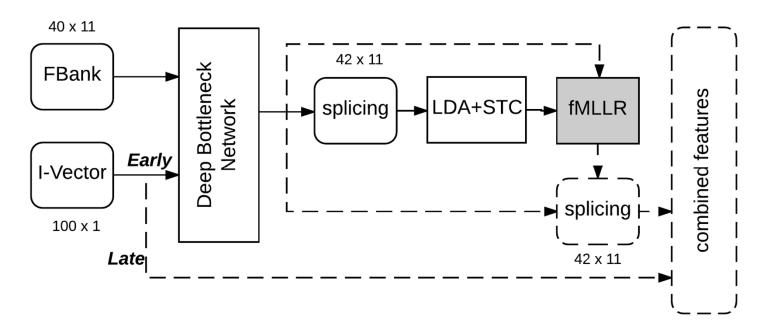


Speaker Adaptive Feature (SAF)

Input Features for GMMs



We used FBank and MVDR+MFCC+T (M2+T) features to build two GMMs

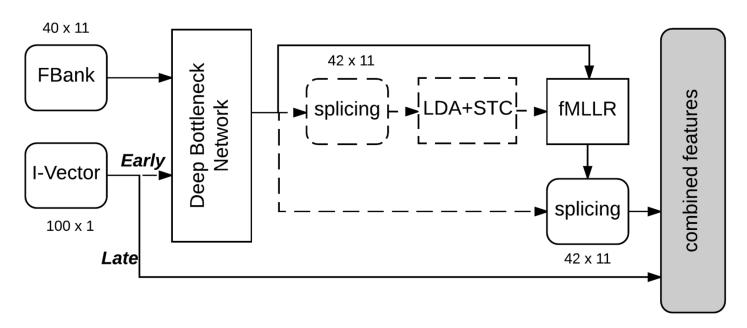


Feature Extraction for GMMs

Input Features for DNNs



Also FBank and MVDR+MFCC+T features for DNNs



Feature Extraction for DNNs

Language Models



- 4-gram LM from 150k words for English and 300k words for German
- Feed-forward Neural Network LM
 - 4 sigmoid layers of 600 units
 - 200-dimensional word embedding for the vocabulary size of 20k
 - To be used directly while decoding
- LSTM-RNN LM
 - 2 LSTM layers of 650 units
 - Vocabulary size of 50k
 - To rescore n-best lists

DNN & GMM Systems



DNNs

- 8k states of CD-Phone for English systems, 18k states for German systems
- SAF-IMEL and SAF-M2+T

■ GMMs

- The same number of CD-Phone states
- The same front-ends





System Overview



Evaluation Setups



Results



System Training



- 480 hours for English, 360 hours for German
- Deep feed-forward neural network
 - Input layer of 11-15 stacked frames
 - 5-6 hidden layers with 2000 units per layer
 - Pre-training with denoising auto-encoders
 - Fine-tuning with cross-entropy loss function
 - Newbob training schedule
- Deep bottleneck network
 - Have the same architecture as the DNNs
 - Except a bottleneck layer of 42 units



Speaker Adaption



- Use transcriptions from the CNC system
- Align and eliminate the frames with confidence score less than 0.7
- GMMs
 - fMLLR and MLLR
- DNNs
 - One adapted DNN per speaker
 - Training one more epoch on the adaption data with a small learning rate



N-best List Rescoring



- Janus-based systems
 - Use single or combined system with feed-forward LM to generate 1000-best list
 - Then rescore with LSTM-RNN LM
- Kaldi-based EN system (s5 recipe)
 - Use 3-gram LM to generate 1000-best list
 - Then apply LSTM-RNN LM to rescore



Evaluation Setups



Results



English Talk Task



| System | tst2013 | Gain |
|-----------------------|---------|------|
| GMM(BN-M2+T) | 14.4 | - |
| DNN(IMEL) | 14.9 | _ |
| GMM(SAF-M2+T) | 13.4 | 1.0 |
| DNN(SAF-IMEL) | 12.0 | 2.9 |
| CNC-4-sys | 10.5 | 1.5 |
| GMM(SAF-M2+T) adapted | 10.5 | 2.9 |
| DNN(SAF-IMEL) adapted | 9.8 | 2.2 |
| Kaldi-s5 RNN rescored | 11.8 | - |
| ROVER-5-sys | 9.4 | 0.4 |

Results for TED Talk task on tst2013

English MSLT Task



| System | dev2016 | Gain |
|-------------------|---------|------|
| GMM(BN-lMEL+T) | 26.7 | _ |
| GMM(BN-IMEL+IVec) | 26.6 | _ |
| DNN(lMEL+T) | 27.1 | _ |
| DNN(lMEL+IVec) | 27.6 | - |
| DNN(BN-IMEL) | 26.6 | _ |
| DNN(BN-M2+T) | 26.7 | _ |
| CNC | 22.9 | 3.7 |
| CNC rescored | 21.6 | 1.3 |

Results for English MSLT task on dev2016



German MSLT Task



| System | dev2016 | Gain |
|--------------------|---------|------|
| DNN(BN-IMEL+T) | 33.7 | _ |
| DNN(BN-IMEL+T+bsv) | 33.8 | _ |
| DNN(BN-M2+T) | 33.0 | _ |
| DNN(BN-M2+lMEL+T) | 32.7 | _ |
| DNN(Mod-M2+lMel+T) | 32.3 | - |
| CNC | 30.8 | 1.5 |
| CNC rescored | 28.7 | 2.1 |

Results for German MSLT task on dev2016



- Our used techniques and systems
 - Speaker Adaptive Feature
 - Feed-forward & LSTM-RNN LM
 - Model Adaption
 - System Combinations
- WER results on the official tst2016 set:
 - 8.5% on English Talk
 - 22.3% on English MSLT
 - 25.5% on German MSLT

Training Data



About 483 hours and 364 hours for acoustic modeling of English and German systems

| Source | # Amount |
|----------------------------|-----------|
| Quaero from 2010 to 2012 | 200 hours |
| Broadcast news [8] | 80 hours |
| TED-LIUM v2 [9] | |
| excluding disallowed talks | 203 hours |
| Total | 483 hours |

English acoustic modeling data

| Source | # Amount |
|------------------------------|-----------|
| Quaero from 2009 to 2012 | 180 hours |
| Broadcast news | 24 hours |
| Baden-Württemberg parliament | 160 hours |
| Total | 364 hours |

German acoustic modeling data

Results – Talk Task



| System | tst2013 | tst2014 |
|-----------------------|---------|---------|
| GMM(SAF-IMEL) | 13.5 | 11.0 |
| GMM(SAF-M2+T) | 13.4 | 10.9 |
| DNN(SAF-IMEL) | 12.0 | 10.4 |
| DNN(SAF-M2+T) | 12.3 | 10.0 |
| CNC | 10.5 | 8.6 |
| GMM(SAF-IMEL) adapted | 10.7 | 8.5 |
| GMM(SAF-M2+T) adapted | 10.5 | 8.6 |
| DNN(SAF-IMEL) adapted | 9.8 | 8.6 |
| DNN(SAF-M2+T) adapted | 10.2 | 8.8 |
| Kaldi-s5 RNN rescored | 11.8 | 8.6 |
| ROVER | 9.4 | 7.8 |

Results for English talk task on tst2013 and tst2014

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