#### Resource development and experiments in automatic South African broadcast news transcription

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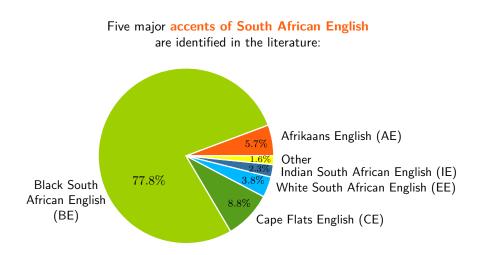
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#### Motivation

Report on **baseline results** of a straight-forward system:

- Use resources collected at Stellenbosch University (2000 present)
- Aim is to use baseline for comparative/interesting further studies

#### Accents of English in South Africa



## South African broadcast news data



20 hours SAFM broadcasts from 1996 to 2006:

- RD: Newsreader speech, prepared 27 speakers, 12.9 hours (BE, EE, IE)
- SI: Studio interview speech, fairly spont. 61 speakers, 0.6 hours
- NST: Non-studio telephone speech, spont. 262 speakers, 2.07 hours
- NS: Non-studio wideband speech, noisy 208 speakers, 1.54 hours

Accent annotated for each sentence-level segment. Test set similar in composition to training set  ${\sim}2.7$  hours.

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- Solution Acoustic model for p(X|W) 20h SABN corpus (previous slide)

#### Language modelling

- 109M word corpus from **South African newspapers**, collected 2000 2005: The Financial Mail, Business Day, The Sunday Times, The Times, Sunday World, The Sowetan, The Herald, The Algoa Sun and The Daily Dispatch
- SRILM toolkit used to train trigram language models on above text as well as on the transcriptions of acoustic training set (185k words)
- Also considered interpolation of the two language models



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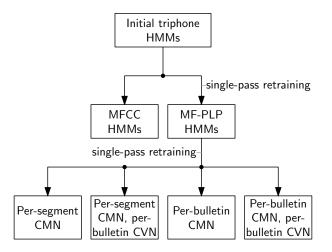
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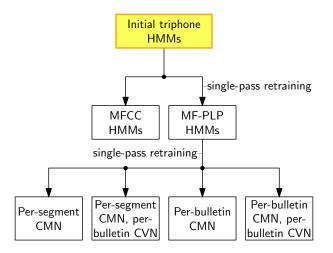


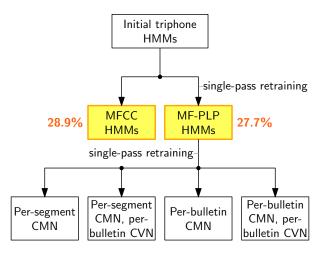
Language model	Perplexity	
Trained on 109M newspaper corpus	162.9	
Trained on acoustic training set	328.9	
Interpolation of the above two	139.9	

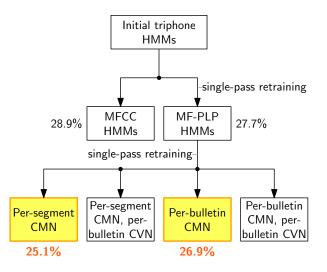
## **Pronunciation dictionary**

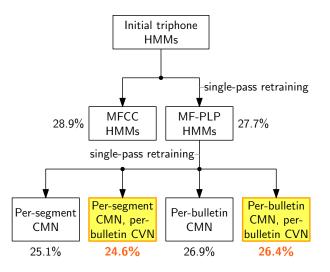
- Pronunciation dictionaries developed by a phonetic expert
- Reflect typical EE pronunciation
- Phone set: 45 **ARPABET** phones
- Training pronunciation dictionary: 15k words
- Recognition pronunciation dictionary: 60k words
- Average number of pronunciations per word: 1.25
- $\bullet$  Out-of-vocabulary rate on test set: 1.02%

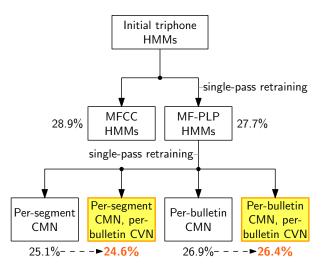


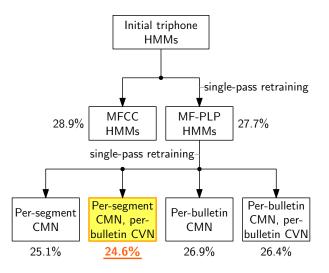












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#### Final system

- Acoustic model set: 2624 states
- Features: mel-frequency perceptual linear prediction (MF-PLP)
- Normalisation: per-segment CMN, per-bulletin CVN

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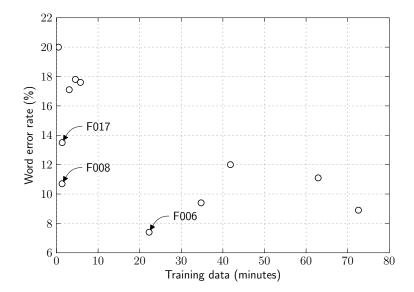
#### Evaluation

- Used the first-best output from HTK HDecode decoder
- Measured WERs separately for each accent and channel condition

# System performance

Accent	RD	SI	NST	NS	Overall
AE	-	-	60.7	67.0	63.3
BE	13.7	19.6	64.3	56.9	29.4
CE	-	-	61.7	-	61.7
EE	14.1	-	54.1	41.6	17.2
IE	12.7	-	59.2	-	16.6
UKE	-	17.7	22.7	32.2	23.8
USE	-	39.3	-	50.5	48.0
Other	-	-	63.0	66.7	65.3
Overall	13.6	19.5	57.3	52.0	24.6

# System performance



# MP3 audio compression

MP3 bit-rate	RD	SI	NST	NS	Overall
128 kbps	13.6	18.9	57.0	51.9	24.6
64 kbps	13.4	18.8	57.8	52.3	24.6
32 kbps	14.3	20.8	58.7	50.7	25.3

# **Summary and conclusions**

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- Described compilation of resources and subsequent language, pronunciation and acoustic modelling
- Compared MFCC and MF-PLP parametrisation
- Normalisation: compared CMN and CVN
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#### Main findings

- Final system: MF-PLP, per-segment CMN, per-bulletin CVN
- WER of 24.6%, poor performance on spontaneous and telephone speech
- MP3 compression: system maintains performance except at very low bit-rates

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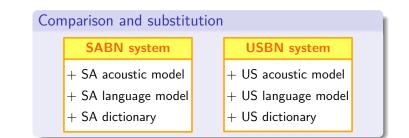
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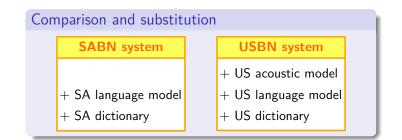
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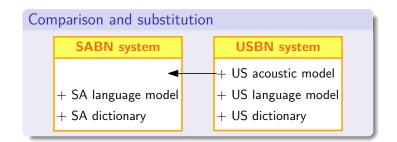
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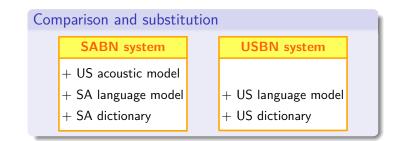
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