

Air quality measurement and logging in taxi ranks and inside of taxis

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Report presented in partial fulfilment of the requirements of the module Project (E) 448 for the degree Baccalaureus in Engineering (Electrical and Electronic) in the Faculty of Engineering at Stellenbosch University.

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Abstract

English

The English abstract.

Afrikaans

Die Afrikaanse uittreksel.

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Nomenclature

Variables and functions

p(x) Probability density function with respect to variable x.

P(A) Probability of event A occurring.

 ε The Bayes error.

 ε_u The Bhattacharyya bound.

B The Bhattacharyya distance.

S An HMM state. A subscript is used to refer to a particular state, e.g. s_i

refers to the $i^{\rm th}$ state of an HMM.

S A set of HMM states.

F A set of frames.

Observation (feature) vector associated with frame f.

 $\gamma_s(\mathbf{o}_f)$ A posteriori probability of the observation vector \mathbf{o}_f being generated by

HMM state s.

 μ Statistical mean vector.

 Σ Statistical covariance matrix.

 $L(\mathbf{S})$ Log likelihood of the set of HMM states \mathbf{S} generating the training set

observation vectors assigned to the states in that set.

 $\mathcal{N}(\mathbf{x}|\mu,\Sigma)$ Multivariate Gaussian PDF with mean μ and covariance matrix Σ .

The probability of a transition from HMM state s_i to state s_j .

N Total number of frames or number of tokens, depending on the context.

D Number of deletion errors.

I Number of insertion errors.

S Number of substitution errors.

Nomenclature viii

Acronyms and abbreviations

AE Afrikaans English

AID accent identification

ASR automatic speech recognition

AST African Speech Technology

CE Cape Flats English

DCD dialect-context-dependent

DNN deep neural network

G2P grapheme-to-phoneme

GMM Gaussian mixture model

HMM hidden Markov model

HTK Hidden Markov Model Toolkit

IE Indian South African English

IPA International Phonetic Alphabet

LM language model

LMS language model scaling factor

MFCC Mel-frequency cepstral coefficient

MLLR maximum likelihood linear regression

OOV out-of-vocabulary

PD pronunciation dictionary

PDF probability density function

SAE South African English

SAMPA Speech Assessment Methods Phonetic Alphabet

Chapter 1

Introduction

The majority of South Africa's public sector uses taxis as a means of transport.

1.1. Section heading

This is some section with two table in it: Table 1.1 and Table 1.2.

Table 1.1: Performance of the unconstrained segmental Bayesian model on TIDigits1 over iterations in which the reference set is refined.

Metric	1	2	3	4	5
WER (%)	35.4	23.5	21.5	21.2	22.9
Average cluster purity (%)	86.5	89.7	89.2	88.5	86.6
Word boundary F -score (%)	70.6	72.2	71.8	70.9	69.4
Clusters covering 90% of data	20	13	13	13	13

Table 1.2: A table with an example of using multiple columns.

	Accurac		
Model	Intermediate	Output	Bitrate
Baseline	27.5	26.4	116
VQ-VAE	26.0	22.1	190
CatVAE	28.7	24.3	215

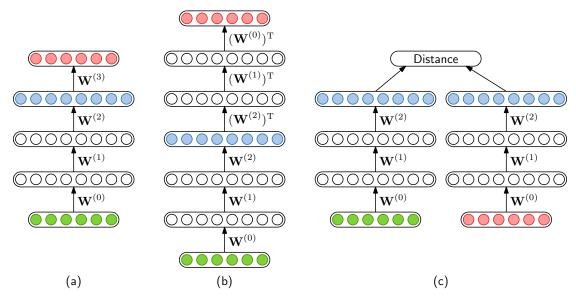


Figure 1.1: (a) The cAE as used in this chapter. The encoding layer (blue) is chosen based on performance on a development set. (b) The cAE with symmetrical tied weights. The encoding from the middle layer (blue) is always used. (c) The siamese DNN. The cosine distance between aligned frames (green and red) is either minimized or maximized depending on whether the frames belong to the same (discovered) word or not. A cAE can be seen as a type of DNN [1].

This is a new page, showing what the page headings looks like, and showing how to refer to a figure like Figure 1.1.

The following is an example of an equation:

$$P(\mathbf{z}|\boldsymbol{\alpha}) = \int_{\boldsymbol{\pi}} P(\mathbf{z}|\boldsymbol{\pi}) p(\boldsymbol{\pi}|\boldsymbol{\alpha}) d\boldsymbol{\pi} = \int_{\boldsymbol{\pi}} \prod_{k=1}^{K} \pi_k^{N_k} \frac{1}{B(\boldsymbol{\alpha})} \prod_{k=1}^{K} \pi_k^{\alpha_k - 1} d\boldsymbol{\pi}$$
(1.1)

which you can subsequently refer to as (1.1) or Equation 1.1. But make sure to consistently use the one or the other (and not mix the two ways of referring to equations).

Chapter 2 Summary and Conclusion

Bibliography

[1] G. E. Dahl, D. Yu, L. Deng, and A. Acero, "Context-dependent pre-trained deep neural networks for large-vocabulary speech recognition," *IEEE Trans. Audio, Speech, Language Process.*, vol. 20, no. 1, pp. 30–42, 2012.

Appendix A
 Project Planning Schedule

This is an appendix.

Appendix B Outcomes Compliance

This is another appendix.