

Karlsruhe Institute of Technology Communications Engineering Lab



TBS

Master's Thesis

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Declaration

With this statement I declare that I have independently completed the above master's thesis. The thoughts taken directly or indirectly from external sources are properly marked as such. This thesis was not previously submitted to another academic institution and has also not yet been published.

Karlsruhe, 06.12.2017

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Abstract

CEL thesis rules require it to be about 3-5 pages. It is a summary of what you do in your thesis. Use around 5 pictures and outline whatever you did. And now a few lines of information.

This is a todo example

Polar codes are the first codes to asymptotically achieve channel capacity with low complexity encoders and decoders. They were first introduced by Erdal Arikan in 2009 [Ari09]. Channel coding has always been a challenging task because it draws a lot of resources, especially in software implementations. Software Radio is getting more prominent because it offers several advantages among which are higher flexibility and better maintainability. Future radio systems are aimed at being run on virtualized servers instead of dedicated hardware in base stations [RMP+15]. Polar codes may be a promising candidate for future radio systems if they can be implemented efficiently in software.

In this thesis the theory behind polar codes and a polar code implementation in GNU Radio is presented. This implementation is then evaluated regarding parameterization options and their impact on error correction performance. The evaluation includes a comparison to state-of-the-art Low-Density Parity-Check (LDPC) codes.

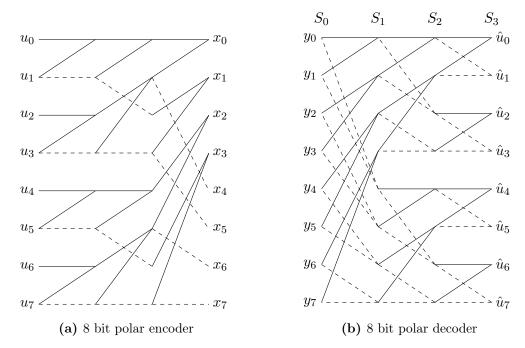


Figure 0.1.: Polar code encoding and decoding

The polar encoder is shown in Fig. 0.1a.

Contents

1.	Intr	oducti	ion	3
2.	Art	ificial]	Intelligence - Introduction	5
	2.1.	Overvi	riew	 5
	2.2.	Machi	ine Learning	 5
		2.2.1.	Training Models	 5
		2.2.2.	Testing Models	 5
		2.2.3.	Model Evaluation	 5
		2.2.4.	Supervised Learning	 5
		2.2.5.	Feature Engineering	 5
		2.2.6.	Unsupervised Learning	 5
		2.2.7.	Reinforcement Learning	 5
	2.3.	Deep 1	Learning	 5
		2.3.1.	Neural Networks	 5
		2.3.2.	Convolutional Neural Networks	 5
		2.3.3.	Optimization of Cost Functions	 5
3.	Test	tbed In	mplementation	7
			are Defined Radio approach	 7
		3.1.1.		7
		3.1.2.	Universal Software Radio Peripheral	7
	3.2.		ine Learning models in Python and Jupyter	7
		3.2.1.	scikit-learn	7
		3.2.2.	keras	7
	3.3.	Data s	set Generation	7
		3.3.1.	Measure Campaign	7
		3.3.2.	Feature Engineering	7
		3.3.3.	Spectrograms generation	7
4.	Eva	luatior	n and Results	9
	4.1.	Scenar	rio Classification	 9
	4.2.	Perfor	rmance metrics	 9
			an setup comparison	9
5 .	Live	e imple	ementation in GNURadio	11
6.	Con	clusio	n	13
Δ	∆ հե	reviat	tions	15

1. Introduction

Back in 1999, Joseph Mitola III coined the term Cognitive Radio [MM99] as a way to enhance the software defined radio capabilities by the means of a dynamic model that, based on human intervention, improved the flexibility of devices by making them fully configurable and capable of adapting to the user needs and react to the changes it the surrounding environment. Give a short introduction about cognitive radio and overlay systems description of the problem Describe the Dyspan spectrum challenge and its setup briefly descirbe our participataion and things done

W

This is the introductory chapter. It is usually a page or two. Tell a story about the objectives, explain them briefly and outline the structure of your thesis.

And don't forget 10.0815 GB of data is quite a lot. This is an example how to use the siunity package.

2. Artificial Intelligence - Introduction

HERE I PLAN TO DESCRIBE THE THEORY BEHIND THE SCENES THAT HAS BEEN USED FOR THE DEVELOPMENT OF THIS THESIS

2.1. Overview

2.2. Machine Learning

- 2.2.1. Training Models
- 2.2.2. Testing Models
- 2.2.3. Model Evaluation

Overfitting

Underfitting

- 2.2.4. Supervised Learning
- 2.2.5. Feature Engineering

K-nearest Neighbors

Support Vector Machines

Binary trees

- 2.2.6. Unsupervised Learning
- 2.2.7. Reinforcement Learning

2.3. Deep Learning

- 2.3.1. Neural Networks
- 2.3.2. Convolutional Neural Networks
- 2.3.3. Optimization of Cost Functions

3. Testbed Implementation

- 3.1. Software Defined Radio approach
- 3.1.1. GNURadio
- 3.1.2. Universal Software Radio Peripheral
- 3.2. Machine Learning models in Python and Jupyter
- 3.2.1. scikit-learn
- 3.2.2. keras
- 3.3. Data set Generation
- 3.3.1. Measure Campaign
- 3.3.2. Feature Engineering
- 3.3.3. Spectrograms generation

4. Evaluation and Results

- 4.1. Scenario Classification
- 4.2. Performance metrics
- 4.3. Dyspan setup comparison

5. Live implementation in GNURadio

6. Conclusion

So you made it! This is the last part of your thesis. Tell everyone what happened. You did something... and you could show that ... followed.

In the end make a personal statement. Why would one consider this thesis to be useful?

A. Abbreviations

LDPC Low-Density Parity-Check

Bibliography

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- [RMP⁺15] P. Rost, A. Maeder, H. Paul, D. Wübben, A. Dekorsy, G. Fettweis, Ignacio Berberana, Vinay Suryaprakash, and Matthew Valenti. Benefits and challenges of virtualization in 5g radio access networks. accepted for publications in Special Issue Advanced Cloud & Virtualization Techniques for 5G Networks of the IEEE Communications Magazine, Dec 2015.