# Evaluation of Machine Learning Algorithms for Speech Prioritisation in Noisy Environments

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MInf Project (Part 1) Report
Master of Informatics
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2024

# Abstract

TODO: Write Abstract

# Research Ethics Approval

This project was planned in accordance with the Informatics Research Ethics policy. It did not involve any aspects that required approval from the Informatics Research Ethics committee.

# Declaration

I declare that this thesis was composed by myself, that the work contained herein is my own except where explicitly stated otherwise in the text, and that this work has not been submitted for any other degree or professional qualification except as specified.

(*Nikodem Bieniek*)

# Acknowledgements

TODO: Write Acknowledgement

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# Chapter 1

# Introduction

## 1.1 Motivations

Hearing loss is a prevalent condition affecting as much as 430 million people - or 1 in 18 people. This is expected to rise to 1 in 10 by 2050 [1]. The most common treatment for hearing loss is the provision of hearing technology - such as Hearing Aid (HA) or cochlear implants. There are many types of HAs, but the most common type is the behind-the-ear (BTE) hearing aid [2]. However, HA users often report that they struggle to hear speech in noisy environments. For example, in a study by Kochkin [2], 42% of HA users reported that wind noise was a significant issue for them. This project aims to evaluate the effectiveness of machine learning algorithms in prioritising the speech in various environments (such as windy environments).

Modern hearing aids now apply a wide range of techniques to achieve better speech prioritisation. For wind noise reduction, this can be achieved from mechanical solutions - product design to covers that reduce wind noise - to signal processing techniques to compensate for mechanical limitations. However, current techniques are still not perfect as shown by the study from Kochkin.

Wind noise reduction and indeed, noise reduction in general, is a challenging problem when paired with speech. This is because you have to strike a balance between reducing background noise and preserving speech. Korhen's paper [3] outlines various techniques that could be used to reduce the wind noise in hearing aids - from modulation-based noise reduction algorithms (Wiener filtering), adaptive filtering algorithms, to machine learning techniques. The paper mentions that the the proposed ML technique: Long Short-Term Memory (LSTM) neural networks provided modest improvements in wind noise reduction, however, it did highlight that ML techniques may still have utility through further research and careful algorithmic choices.

This project aims to pair the proposed machine learning techniques with signal processing techniques to evaluate the effectiveness of speech prioritisation in noisy environments. The idea is to first perform acoustic scene analysis (ASA) to classify the environment based on the audio signal. Afterwards, speech enhancement techniques will be applied to the signal to prioritise speech. ASA can be done using a dataset of varying environments, and using machine learning techniques to classify the environment.

In this project, we will be using a novel dataset proposed by Huwel et al. [4] This dataset (called HEAR-DS) is unique because it is specially tailored for HA signal processing and contains various environments. Normally, voice activity detection (VAD) is used to detect speech, however, the dataset helpfully contains labels which samples contain speech. This can be used to implicitly train the machine learning model to classify the environment and whether speech is present. Additionally, the paper presents an elementary example of how the dataset can be used: to classify the environment - it showcases the use of a convolutional neural network (CNN) to classify the environment. This project will be extending the paper by actually using the dataset and comparing various machine learning techniques to evaluate the effectiveness of the proposed machine learning techniques.

This project will investigate how recurrent neural networks (RNN) and their variants, such as LSTMs, can be used to classify the environment. Based on the environment classified, the system will then apply signal processing techniques to enhance the speech. The project however has to be mindful in its algorithmic choices - as the computational power required in HA is limited. It is difficult to pinpoint the exact computational power of a HA due to the proprietary nature of the devices. In August 2024, Phonak (Sonova Holding AG) released a new HA which is their first AI equipped HA. The device is said to be capable of handling 7,700 Million Operations Per Second to accommmodate its neural network with 4.5 million parameters. Contrast this with a paper from 2021 investigating techniques in VAD for hearing aids quotes that it ‘rarely exceeds 5 million instructions per second (MIPS)‘ [5] Moreover, Apple’s release of a FDA approved hearing aid in which was previously a mainstream earphone wearable, Apple Airpods, also shows that there’s more interest in this area. So suffice to say, the computational power of hearing aids is accelerating and is most likely going to continue to grow.

Delay constraints are also important in HA, as the user needs to hear the speech in real-time. For example, so long as the speech production is no higher than 30 milliseconds (ms), and ideally less than 20ms, the user is unaffected by the delay [6].

The project is predominantly aimed at the hearing aid industry. If successful, the project could advance the state-of-the-art techniques in hearing aids. Which in turn could improve the quality of life for hearing aid users. A secondary goal is to make hearing aid research more accessible to the computer science community. The project also hopes to benefit other fields that deal with audio signal processing, such as mainstream wearables like headphones or microphones.

# Chapter 2

# Background

## 2.1 Hearing Loss

### 2.1.1 Ear Anatomy

TODO

### 2.1.2 Congenital Hearing Loss

TODO

### 2.1.3 Exposure to Loud Noises

TODO

### 2.1.4 Ageing

TODO

## 2.2 Hearing Loss Treatment

### 2.2.1 Hearing Technology

As mentioned in the Introduction, the most common treatment for hearing loss is the provision of hearing technology. It should be pointed out that there is an active area of research into curing congenitial hearing loss - individuals who have been diagnosed with a hearing loss since birth.

#### 2.2.1.1 Hearing Aids

TODO Microphone info, the reason for multiple mics etc There is a surge into rechargable hearing aids. Smaller model = higher battery life

#### 2.2.1.2 Cohlear Implants

TODO

### 2.2.2 New Approaches to Treatment

#### 2.2.2.1 Gene-Based Therapy

Genitic mutations account for 70-75% of congenital hearing loss, and with the advent of gene-based therapy, it is not surprising to see that this avenue is explored with hearing loss. However, besides it being an active area of research, it will not be applicable to individuals that have

## 2.3 Acoustic Scene Analysis (ASA)

### 2.3.1 Approach 1: Voice Activation Detection (VAD)

TODO: Write

### 2.3.2 Approach 2: Machine Learning

TODO: Write

## 2.4 Speech Processing

### 2.4.1 Waveform

TODO

### 2.4.2 Feature Engineering

## 2.5 Related Work

HEAR-DS...

# Bibliography

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