# DETECTING HEALTH MISINFORMATION IN WEB PAGE TEXT USING DEEP LEARNING METHODS

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## ACKNOWLEDGMENTS

I would like to acknowledge  $\dots$ 

STATEMENT OF CANDIDATE

I, (insert name here), declare that this report, submitted as part of the require-

ment for the award of Bachelor of Engineering in the School of Engineering,

Macquarie University, is entirely my own work unless otherwise referenced or

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Date:

#### ABSTRACT

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# Contents

A	cknov	vledgments	i
$\mathbf{A}$	bstra	${f ct}$	i
Ta	able o	of Contents in	X
Li	st of	Figures	i
Li	st of	Tables	i
1	<b>Intr</b> 1.1	Project Overview	1 1
2			3
	2.1	Ÿ	3
			3
			4
		·	4
		v	4
	2.2	11	4
		v	5
		11	5
			5
	2.2		5
	2.3	1 0	5
		1 0	6
	2.4		6
	2.4	Conclusion	6
3	-		7
	3.1		7
	3.2	v	7
	3.3	Study Data	7

<u>x</u> CONTENTS

	3.4 3.5 3.6	System Model	7
4		clusions and Future Work	9
		Conclusions	
5	Abb	previations	11
$\mathbf{A}$	A.1	ne of appendix A  Overview	
В	B.1	ne of appendix B Overview	
Bi	bliog	raphy	15

# List of Figures

# List of Tables

# Introduction

One of the key components required to minimize the propagation of misinformation online is to have the ability of automatically evaluating and quantifying the credibility of articles. However, traditional automated methods - such as shallow learning-based techniques, still require the domain knowledge of experts to be able to develop the features required by the model. Thus, this project aims to investigate the performance of Deep Learning-based (DL) techniques in evaluating the credibility of information within domain-specific articles via the classification of set criteria that have deemed to be highly correlated with articles that have low credibility. Specifically, this project will focus on evaluating the credibility of online health articles related to vaccination due to the commonly misinformed and controversial views associated with its effects [3].

## 1.1 Project Overview

This section details the scope of the project and its associated outcomes outlining the various tasks that must be accomplished to successfully complete the project.

## 1.1.1 Project Scope

With the primary objective of this project being the evaluation on the effectiveness of deep learning models in determining the credibility of online health-related articles. Due to the complexity of this project, a set of activities - divided into main goals and stretch goals, have been defined to ensure that the completion of this project remains feasible in the given time frame. The completion of all activities categorized as main goals will signal the realization of the primary objective and the completion of the project. Stretch goals are activities of interest that have been identified as non-essential to the completion of the primary objective but (talk about the overarching goal that all stretch goals have in common e.g. understand the model, utilize the model etc.) and will be worked on after the completion of the project.

#### Main Goals

- Develop a set of criteria that will be used to determine the credibility of online articles.
- Evaluate the performance of common ML-based methods on the classification of online vaccine-related articles, based on the criteria developed.
- Evaluate the performance of the proposed DL model on the classification of online vaccine-related articles, based on the criteria developed.
- Evaluate the effect of transfer learning methods in the performance of the proposed DL method (assuming that the chosen method doesn't rely on transfer learning)
- Evaluate the effectiveness of various transfer learning methods for the classification task

#### **Stretch Goals**

• Utilize attention mechanisms to understand how the aforementioned DL model classifies the criteria for credibility.

# Background and Related Work

A literature review has been conducted to develop an understanding on the research that has been done in the assessment of the credibility of information, specifically in the context of information related to health and the limitations and capabilities of shallow learning techniques and how it differs from deep learning-based methods for the task of text classification.

## 2.1 Assessing Credibility of Information

To have the capability of automating the process of evaluating the quality or credibility of online information, a definition that outlines of what is required by an article to be considered as a credible source of information must be developed. While there has been a significant amount of research that has been done on the development of tools and frameworks that aim to assess the credibility of online health information, there is currently no standardized method or benchmark that is universally used. Tools and frameworks that have been identified to be applicable for this project are (*Note: add citation for each one*): DISCERN [1], HealthNewsReview [2], Quality Index for health-related Media Reports (QIMR) [5] and (*Wiley 2017 - update this when you find the source*)

Talk about the work done in establishing the measurement of quality in online health information e.g. DISCERN, QIMR and the sources from that document in the slack channel

Try to understand the following for each source:

- How the scoring system works
- How credibility is defined

#### 2.1.1 DISCERN

DISCERN is a questionnaire designed to assess the reliability of a publication, it consists of 16 questions each with a Likert scale, ranging from 1 (no) to 5 (yes) and is divided into 3

sections. The first section (questions 1 - 8) investigate the reliability of the information and is comprised of questions such as "Are the aims clear?", "Is it balanced and unbiased?" and "Does it refer to areas of uncertainty?". The second section (questions 9 - 15) assesses the quality of information provided by the publication for treatment choices and is composed of questions such as "Does it describe the risks/benefits of each treatment?" and "Does it provide support for shared decision-making?". The final section (question 16) assesses the overall rating of the publication ("Based on the answers to all of the above questions, rate the overall quality of the publication as a source of information about treatment choices").

Note: Need to talk about advantages/disadvantages, limitations, relevance to this project etc.

#### 2.1.2 HealthNewsReview

HealthNewsReview provides a set of 10 criteria designed to act as a framework for evaluating the credibility of health-related media. The criteria is based on the various elements that all health-related media should consist of, the criteria is composed of criterion such as "Does the story compare the new approach with existing alternatives?", "Does the story use independent sources and identify conflicts of interest?" and "Does the story appear to rely solely or largely on a news release?".

Note: Need to talk about advantages/disadvantages, limitations, relevance to this project etc.

#### 2.1.3 QIMR

#### 2.1.4 Wiley 2017

## 2.2 Prior Approaches

Discuss the prior work that has been done in terms of text classification e.g. spam, sentiment, topic

For each model, talk about the following:

- How it works and the mechanisms involved
- Advantages
- Limitations

Another thing to consider is why I'm looking at ML models in the first place, which is because I plan to compare the performance of ml models against dl models using typical features that don't require domain-specific knowledge to create e.g. BoW, language models, ngrams etc.

- 2.2.1 Naive Bayes
- 2.2.2 Support Vector Machines
- 2.2.3 Random Forests
- 2.2.4 Artificial Neural Networks

## 2.3 Deep Learning

Deep learning models are a class of machine learning models that have the capability of automatically learning a hierarchical representation of data. These hierarchical representations are constructed through the use of artificial neural networks, the main underlying mechanism of deep learning models. Typically, large amounts of training data is required to train a model in learning the language model required to attain state of the art results, in the task of text classification for instance, the size of commonly used non-domain specific datasets range from hundreds of thousands of training examples to millions [4] [6] (note: look into the datasets used by state of the art approaches). Due to these constraints, it is not feasible to procure a dataset for the domain specific task of this project due to the aforementioned knowledge expertise and time requirements to manually label the articles required. Hence, (Talk about transfer learning/N-shot learning/domain adaptation here) will be used to overcome this issue.

Introduce the state-of-the-art DL based approaches for text classification and try to compare it performance with state-of-the-art ML approaches

For each model, talk about the following:

# NOTE: REMEMBER WHEN WRITING THIS SECTION TO ALWAYS CONSIDER HOW IT DIFFERS TO ML TECHNIQUES

- How it works and the mechanisms involved
- Advantages
- Limitations

Introduce the typical architectures used for text classification e.g. RNNs, LSTMs, CNNs, GRUs?

#### 2.3.1 Deep Learning Models

Recurrent Neural Networks

Gated Recurrent Unit Networks

Long Short-Term Memory Networks

Convolutional Neural Networks

#### 2.3.2 GET PROPER NAME FOR THIS SECTION

#### Transfer Learning

Talk about transfer learning and how it works and how it is applicable to this project.

#### N-Shot Learning

Talk about zero/few/etc-shot learning and how it works and how it is applicable to this project.

#### 2.4 Conclusion

Summarize lit review and describe why DL-based approaches should be preferred over ML-based for this type of problem. Also talk about Transfer/N-Shot learning and describe which one will be feasible given the project's time constraints

# Proposed Approach

#### 3.1 Rationale

Introduce and discuss the factors that led to me choosing the proposed approach

## 3.2 Credibility Criteria

Introduce and discuss the 7 criteria that will be classified and describe how the criteria was determined

## 3.3 Study Data

Talk about the data I'll be using, how we got it, its characteristics etc.

## 3.4 System Model

Describe the architecture of the model

## 3.5 Experiments

Describe the experiments that I'm planning to do (in such a way that they are easily reproducible)

### 3.6 Outcome Measures

Talk about the type of analyses that I'll be doing to determine the performance of my proposed model

# Conclusions and Future Work

# 4.1 Conclusions

The end

## 4.2 Future Work

# Abbreviations

AWGN Additive White Gaussian Noise

BC Broadcast Channel

BS Base Station

CSI Channel State Information

CSIR Channel State Information at Receiver
CSIT Channel State Information at Transmitter

dB Decibels

DPC Dirty Paper Coding GS Gram-Schmidt

RVQ Random Vector Quantisation SISO Single Input Single Output SNR Signal to Noise Ratio

SINR Signal to Interference plus Noise Ratio

MISO Multiple Input Single Output
SIMO Single Input Multiple Output
MIMO Multiple Input Multiple Output
MMSE Minimum Mean Square Error
MRC Maximum Ratio Combining

QoS Quality of Service
TDD Time Division Duplex
FDD Frequency Division Duplex

ZF Zero-Forcing

ZFBF Zero-Forcing Beamforming

ZMCSCG Zero Mean Circularly Symmetric Complex Gaussian

# Appendix A<br/>name of appendix A

## A.1 Overview

here is the Overview of appendix A  $\dots$ 

### A.2 Name of this section

here is the content of this section ...

# Appendix B<br/>name of appendix B

## B.1 Overview

here is the Overview of appendix B  $\dots$ 

## B.2 Name of this section

here is the content of this section ...

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