

**A LOW MEMORY CONTINUAL LEARNING CLASSIFICATION ALGORITHM
FOR LOW RESOURCE HARDWARE**

by
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A Thesis

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Dedications

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Acknowledgements

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Abstract

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LOW RESOURCE HARDWARE

2021-2022

Example Professor Name, Ph.D.
Master of Science in Computer Science

Continual learning (CL) is a branch of machine learning which aims create algorithms that are able to learn from real-life data sources which are naturally changing over time without forgetting what it has already learned [1]. This issue of Catastrophic Forgetting, has plagued machine learning problems for for years and many new algorithmic approaches have been proposed to solve this issue. In this paper, we will approach this issue from a perspective of low resource utilization such as in the case with micro controllers and embedded systems.

The algorithm proposed relies on the process of weighted choices, multiple networks, and a client server connection for full training, inheriting design features from both the Regularization and Architectural sides of Continual learning.[2] The algorithm proposed follows similar patterns as other recent continual learning algorithms with following the concept of memory retention during a sleep period while also creating and merging network outputs to solve the catastrophic forgetting problem.

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List of Schemes

Scheme

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Chapter 1

Introduction

- 1.1 Continual Learning**
- 1.2 The Failure of Remembering**
- 1.3 Premise**
- 1.4 Edge Device Learning Approaches**
- 1.5 Contributions to Research**
- 1.6 Scope**

Chapter 2

Literature Review

2.1 Continual Learning Early Days

2.2 Continual Learning Current Research

Chapter 3

Technical Approach Overview

3.1 Low Memory Continual Learning

Chapter 4

Micro Continual Learning

4.1 Algorithm Design

Appendix A
Experimental Equipment

Table A1

Survey Data

Headings	Data
Yes	50%
No	50%

Table A2

More Survey Data

Headings	Data
Yes	20%
No	80%

Appendix B
The Second Appendix

Figure A1

Scattergraph Chart

