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**REINFORCEMENT LEARNING-BASED SECURITY  
VULNERABILITY DETECTION FOR MICROSERVICES**

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MSc thesis submitted in partial fulfilment of the requirement for the degree Master of  
Science.

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## DECLARATION

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The above candidate has carried out research for the Masters' thesis under my supervision.

Name of the supervisor: Prof. Indika Perera

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02/07/2025  
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## ABSTRACT

Microservices architecture (MSA) is the main architectural model for contemporary software systems due to its scalability, flexibility, and maintainability. Yet, there are major security concerns that emerge from the distributed and dynamic nature of microservices. Most of these risks and vulnerabilities are difficult to identify due to the complexity and the evolving nature of microservice based systems. Due to this reason, conventional security vulnerability detection methods designed for monolithic systems are inadequate and ineffective in microservice based systems. To address this shortcoming, this thesis investigates the use of Reinforcement Learning (RL) and offers a Proximal Policy Optimization (PPO)-based framework as an automated and adaptive tool for finding security vulnerabilities in microservices. “Online Boutique” application, which built on microservices architecture is utilized as a testbed for assessing the performance of the developed RL framework in finding security vulnerabilities. The PPO agent learns to interact with the system, simulate attacks, and find real-time security vulnerabilities. The study intends to mimic DoS attack situations targeting the Online Boutique application. The suggested approach is a scalable, consistent security testing tool able to evolve to identify developing threats and adapt to new security vulnerabilities. This study proposes a promising substitute for conventional manual testing for DoS attacks. The results illustrate the efficacy of the PPO framework in detecting vulnerabilities in the microservices environment, with implications for improving the security of microservices-based applications.

Keywords: Microservices Architecture, Reinforcement Learning, Proximal Policy Optimization, Vulnerability Detection, Security Testing, Automated Security, Penetration Testing, DoS Simulation, Machine Learning in Cybersecurity.

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## LIST OF ABBREVIATIONS

MSA	Microservice Architecture
RL	Reinforcement Learning
PPO	Proximal Policy Optimization
DoS	Denial of Service
SQL	Structured Query Language
URL	Uniform Resource Locator
CPU	Central Processing Unit
API	Application Programming Interface
TLS	Transport Layer Security
DQN	Deep Q-Networks
IDS	Intrusion Detection Systems
TRPO	Trust Region Policy Optimization
CI	Continuous Integration
CD	Continuous Deployment
XSS	Cross-Site Scripting
A3C	Asynchronous Advantage Actor-Critic
RAM	Random Access Memory
GB	Giga Byte
MB	Mega Byte
KB	Kilo Byte
BIOS	Basic Input/Output System
UEFI	Unified Extensible Firmware Interface
LTS	Long Term Support
WSL	Windows Subsystem for Linux
CLI	Command Line Interface

IP	Internet Protocol
HTTP	Hyper Text Transfer Protocol
ID	Identifier
JSON	JavaScript Object Notation
QPS	Queries Per Second
TCP	Transmission Control Protocol
GAE	Generalized Advantage Estimator
UI	User Interface