

**VISION-BASED REAL-TIME  
TRAFFIC CONTROL USING  
ARTIFICIAL NEURAL  
NETWORK ON  
GENERAL-PURPOSE  
EMBEDDED HARDWARE**

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## DECLARATION

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I declare that this is my own work and this dissertation does not incorporate without acknowledgement any material previously submitted for a Degree or Diploma in any other University or institute of higher learning and to the best of my knowledge and belief it does not contain any material previously published or written by another person except where the acknowledgement is made in the text.

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

Name of the supervisor: Prof. Rohan Munasinghe

Signature of the Supervisor:

Date:

## Abstract

In urban cities, traffic management of intersections is a substantially challenging problem. Inappropriate traffic control leads to waste of fuel, time, and productivity of nations. Though the traffic signals are used to control traffic, it often causes problems due to the pre-programmed timing being not appropriate for the actual traffic intensity at the intersection. Traffic intensity determination based on statistical methods only gives the average intensities expected at any given time. However, to control traffic effectively, the knowledge of real-time traffic intensity is a must-have. In this project, vision-based technology and artificial intelligence (AI) are used to estimate traffic in real-time and control the traffic in order to reduce the traffic congestion. General-purpose electronic hardware has been used for in-situ image processing based on edge-detection methods. A Neural Network (NN) was trained to infer traffic intensity in each image in real-time using a scale of 1(very low) to 5 (very high). A Trained AI unit, which takes approximately 4 seconds to process each image and estimate traffic intensity was tested on the road where it recorded a 90% acceptance rate. In order to control the traffic, a ratio-based method and a reinforcement learning (RL)-based method was used. The performance of these methods are compared with a pre-programmed traffic controller.

***Keywords***-traffic sensing, traffic control, neural network, reinforcement learning

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## LIST OF ABBREVIATIONS

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Abbreviation	Description
AI	Artificial Intelligence
NN	Neural Network
RL	Reinforcement Learning
GPS	Global Positioning System
YOLO	you-only-look-once
ReLU	Rectified Linear Unit
SUMO	Simulation of Urban Mobility
TraCI	Traffic Control Interface

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