# Multiple-Sensor Based Approach for Road Terrain Classification

## **Shifeng Wang**

Submitted in fulfilment of the requirements for the degree of

**Doctor of Philosophy** 

University of Technology, Sydney

Faculty of Engineering and Information Technology

Centre for Autonomous Systems

**Intelligent Mechatronic Systems Group** 

Student : Shifeng Wang

Supervisor : Dr. Sarath Kodagoda

## **Certificate of Original Authorship**

I, Shifeng Wang, certify that the work in this thesis has not previously beer
submitted for a degree nor has it been submitted as part of the requirements for a degree
except as fully acknowledged within the text.
I also certify that the thesis has been written by me. Any help that received in my
research work and the preparation of the thesis itself has been acknowledged. In addition

I certify that all information sources and literature used are indicated in the thesis.

Signed:			
Date:			

#### **Acknowledgment**

I would like to express my heartfelt gratitude to my supervisor Dr. Sarath Kodagoda for his helpful insights, intelligent ideas and supporting. From the moment I met him when I came to UTS, he continually assisted me throughout these years resulting in this thesis. I really appreciate his patience and the directions he showed to me when I met difficulties in my research. I would like to say that it is my honour being your student.

Thanks to the people who have offered me great helps and cooperation. Firstly my colleague Dr. Stephan Sehestedt who was my neighbour and friend, he was my role model being a good researcher with a professional attitude. Dr. Zhan Wang and Dr. Rami Khushaba, they advised me a lot regarding mathematics and Matlab coding. My colleague and friend Lei Shi, he always offered me great help on programming and some suggests at anytime. Dr. Alen Alempijevic and Dr. Stephan Sehestedt who developed and set up the experiment platform CRUISE where my research work carried out. And Mr. Marian Himstedt, Dr. Shan Lynn, Dr. Jack Wang, Mr. Xiang Luo, Mr. Zulkarnain Zainudin, Mr. Xiang Ren, Miss Yan Li, and Mr. Kanzhi Wu, thanks for your jobs during the data collection experiments.

My sincere thanks go to the University of Technology Sydney who offered me scholarships all along my candidature. It would be a mission impossible without financial support for me.

Special thanks to my wife Lei Chen and lovely daughter Yifei Wang. I cannot image if I can finish this PhD without your understanding and supporting. I hope I can bring our family a better future. Thanks to my parents for their constant support in my life. Hope you are proud of me.

#### **Contents**

List of Abbreviations	IX
List of Figures	. X
List of TablesX	Ш
Abstract	۲V
Chapter I	. 1
Introduction	. 1
1.1 Background	. 1
1.2 Motivation	. 2
1.3 Contributions	. 5
1.4 Publications	. 5
1.5 Thesis Overview	. 6
Chapter II	. 8
Review of Related Work	. 8
2.1 Accelerometer Applications	. 8
2.1.1 Small Sized Rover Platform Using Accelerometer	. 9
2.1.2 Road Vehicle Using Accelerometer	12
2.2 Camera Applications	13
2.2.1 Small Sized Rover Using Camera	13

2.2.2 Road Vehicle Using Camera	15
2.3 LRF Applications	16
2.3.1 Small Sized Rover Using LRF	16
2.3.2 Road Vehicle Using LRF	18
2.4 Mutiple-Sensor Applications	20
2.4.1 Small Sized Rover Using Multiple Sensors	20
2.4.2 Road Vehicle Using Multiple Sensors	23
2.5 Conclusion	24
Chapter III	25
Acceleration Based Road Terrain Classification	25
3.1 Road Profile Estimation	26
3.1.1 Acceleration (acc-t)	26
3.1.2 Quarter Vehicle Model (acc-t to y-t)	27
3.1.3 Vertical Displacement (y-t)	29
3.1.4 Speed (v-t)	31
3.1.5 Speed to Displacement (v-t to x-t)	32
3.1.6 Road Profile (y-x)	32
3.2 Features Extraction	36
3.2.1 FFT feature extracted from road profile (y-x)	37

3.2.2 FFT feature extracted from acceleration (acc-t)	42
3.2.3 Fast wavelet transform feature extracted from acceleration (a	<i>icc-t)</i> and
road profile ( <i>y-x</i> )	43
3.3 Normalization	44
3.4 Principal Component Analysis	45
3.5 K-Fold Cross Validation	45
3.6 Alternative Classifiers	46
3.6.1 Naïve Bayes Classifier	47
3.6.2 Neural Network Classifier	48
3.6.3 Support Vector Machines Classifier	50
3.7 Experiment	52
3.7.1 Experiment Platform	52
3.7.2 Acceleration Based Experiments	55
3.8 Experiment Results	58
3.8.1 Feature Selection	58
3.8.2 Speed Dependency	60
3.8.3 Classifiers Selection	64
3.8.4 Acceleration Based Experiment Result	66
3.9 Conclusion	73

hapter IV	. 74
nage Based Road Terrain Classification	. 74
4.1 Texture Features from Image	. 75
4.2 Image Feature Matrix Establishment	. 76
4.2.1 Gray-Level Co-occurrence Matrix	. 76
4.2.2 Feature Extraction and Feature Matrix Formation	. 79
4.3 Experiment	. 83
4.3.1 Experimental Platform	. 83
4.3.2 Image Based Experiments	. 84
4.4 Experiment Results	. 84
4.5 Conclusion	. 94
hapter V	. 95
RF Based Road Terrain Classification	. 95
5.1 Geometric Arrangement of the LRF	. 95
5.2 Reconstruction of the Road Surface	. 96
5.2.1 Range Data Processing	. 97
5.2.2 Speed Data Processing	. 98
5.2.3 Road Surface	. 99
5.3 Feature Matrix	100

5.	.4 Experiment	. 101
	5.4.1 Experimental Platform	. 101
	5.4.2 LRF Based Experiments	. 102
5.	.5 Experiment Results	. 102
	5.5.1 Speed Independency	. 103
	5.5.2 LRF Based Experiment	. 104
5.	.6 Conclusion	. 109
Cha <sub>l</sub>	pter VI Multiple-Sensor Based Road Terrain Classification	. 110
6.	.1 Predicting LRF Based Probe	. 111
6.	.2 Markov Random Field	. 113
	6.2.1 Conditional Independence Properties	. 114
	6.2.2 Factorization Properties	. 116
6.	.3 Establishment of MRF Application	. 119
	6.3.1 Nodes in MRF	. 119
	6.3.2 Variable Values of Nodes in MRF	. 120
	6.3.3 Clique Potentials in MRF	. 121
	6.3.4 Values of Clique Potentials in MRF	. 122
	6.3.5 Energy Function	. 122
	6.3.6 Optimization	. 124

6.4 Experiment	124
6.4.1 Experimental Platform	124
6.4.2 Multiple-Sensor Fusion Based Experiment	125
6.5 Experiment Results	125
6.6 Conclusion	131
Chapter VII	133
Conclusion and Future Direction	133
7.1 Conclusion	133
7.2 Future Direction	135
Deferences	126

## **LIST OF ABBREVIATIONS**

CRUISE CAS Research Ute for Intelligence, Safety and Exploration

c/m cycles/meter

DARPA Defense Advanced Research Projects Agency

DMU Differential Measurement Unit

FFT Fast Fourier Transform

FPS Frame per Second

FWT Fast Wavelet Transform

GLCM Grey-Level Co-occurrence Matrix

GPS Global Positioning System

IMU Inertial Measurement Unit

LRF Laser Range Finder

MSE Mean Squared Errors

MV Majority Vote

PCA Principal Component Analysis

PSD Power Spectral Density

RPM Revolutions per Minute

SVM Support Vector Machine

USB Universal Serial Bus

# **LIST OF FIGURES**

Figure 1.1 A vehicle on mud terrain	3
Figure 1.2 A vehicle on rocky terrain	4
Figure 1.3 A vehicle on snow terrain	4
Figure 3.1 Illustration of road profile estimation	26
Figure 3.2 the quarter car model of the experimental vehicle	27
Figure 3.3 Even sampling and even profile with constant speed	30
Figure 3.4 Even sampling and uneven profile with varying speed	30
Figure 3.5 Profiles of four road terrain types	33
Figure 3.6 Histograms of four road terrain types	35
Figure 3.7 Illustration of features extraction	36
Figure 3.8 Procedure of forming the FFT feature matrix	37
Figure 3.9 Road profile (y-x) – special frequency feature extraction	37
Figure 3.10 Spatial frequency distributions of four road types	39
Figure 3.11 Spatial frequency domain of four road types	40
Figure 3.12 Acceleration (acc-t) – time frequency feature extraction	42
Figure 3.13 Acceleration/road profile data – FWT feature extraction	43
Figure 3.14 Linear SVM	50

Figure 3.15 Research platform: CRUISE	52
Figure 3.16 The hardware structure of the system	53
Figure 3.17 The mounted accelerometer on the suspension of CRUISE	54
Figure 3.18 Data Collection Routines of Four Road Types	55
Figure 3.19 Acceleration based experiment label sequence	67
Figure 4.1 Four types of road terrain surfaces	74
Figure 4.2 Angular relationships with nearest neighbour	76
Figure 4.3 The mounted camera on top of the frame	82
Figure 4.4 Image based experiment label sequence	85
Figure 4.5 (a) Image sequence of an asphalt terrain segment	88
Figure 4.5 (b) Image sequence of a grave terrain segment	89
Figure 4.6 Poor quality images	92
Figure 5.1 The geometric arrangement of the LRF	95
Figure 5.2 Reconstruction of the road surface	96
Figure 5.3 Mounting geometry of the LRF	96
Figure 5.4 Three-dimensional surface data of four different road types	98
Figure 5.5 The mounted downward-looking LRF	100
Figure 5.6 LRF based experiment label sequence	104
Figure 6.1 The mounted forward-looking LRFs	110

Figure 6.2 An example of an undirected graph	113
Figure 6.3 A Markov blanket of a node in an undirected graph	115
Figure 6.4 A four-node undirected graph	116
Figure 6.5 MRF model for road terrain identification	118
Figure 6.6 MRF experiment label sequence	126

# **LIST OF TABLES**

Table 3.1 Parameters for the quarter car model28
Table 3.2 Parameters of height of four road terrain types34
Table 3.3 Locations of the data collection experiments56
Table 3.4 Classification with different feature selection58
Table 3.5 Classification of same road type with different speeds as different classes60
Table 3.6 FFT feature on acceleration and road profile based classification accuracy
training and testing with different speeds61
Table 3.7 FFT feature on acceleration and road profile based classification results training
and testing with different speeds62
Table 3.8 Comparison of different classifiers usage64
Table 3.9 Acceleration data for training and testing65
Table 3.10 The experimental results using acceleration data without PCA process69
Table 3.11 The experimental results using acceleration data with PCA process71
Table 4.1 Image data for training and testing84
Table 4.2 The experimental results using image data without PCA process90
Table 4.3 The experimental results using image data with PCA process91
Table 5.1 LRF data for training and testing101
Table 5.2 Classification at different speeds103

Table 5.3 The experimental results using downward LRF data without PCA process	.106
Table 5.4 The experimental results using downward LRF data with PCA process	.107
Table 6.1 The experimental results using forward LRF data	.110
Table 6.2 The values of variables	.119
Table 6.3 The numbers of samples before and after the MV	.124
Table 6.4 The comparison after using MRF multiple-sensor fusion	.129

## **ABSTRACT**

Complete perception of the environment around a vehicle plays a crucial role in safe driving. Those include other stationary or dynamic objects, drivable regions, road signs and terrain types. In this thesis, classification of terrain types is investigated. Knowledge of the road terrain is useful to improve passengers' safety and comfort in road vehicles. It is also useful for bounding safe navigational routes for autonomous vehicles. Therefore, the aim of this thesis is to develop a methodology to identify impending road terrain types by using on-board sensors.

Two kinds of sensors are used in this research. The first kind of sensors measure parameters of the vehicle in terms of vibration and speed. Therefore, these sensors can only produce measurements while the vehicle is navigating on a particular terrain. The second type of sensors measure properties of the terrain in terms of structure and visual cues, for example cameras and Laser Range Finders (LRFs). These sensors can even produce measurements of impending terrain types. However, all those kinds of sensors have their own advantages and disadvantages. In this thesis, it is proposed to fuse them to improve the terrain type classification results.

The sensor fusion is achieved using Markov Random Field (MRF). The MRF model is designed to contain five nodes and five cliques which describe the relationships between the classification results of the accelerometer, camera, and two LRFs. The MRF model's energy function is appropriately synthesized to improve the classification accuracies of the impending terrains. Experiments carried out on a real vehicle test-bed, CRUISE (CAS Research Ute for Intelligence, Safety and Exploration) on different types of roads with

various speeds show that the MRF based fusion algorithm lead to significant improvements (approximate 30%) of the road terrain classification accuracies.

**Keywords**—road terrain classification, Markov Random Field, sensor data fusion, acceleration, image, Laser Range Finder