

UNIVERSITI PUTRA MALAYSIA

DETECTION OF ROOF MATERIALS BASED ON AN OBJECT-ORIENTED APPROACH USING WORLDVIEW-2 SATELLITE IMAGERY

EBRAHIM TAHERZADEH MOBARAKEH

FK 2014 81



DETECTION OF ROOF MATERIALS BASED ON AN OBJECT-ORIENTED APPROACH USING WORLDVIEW-2 SATELLITE IMAGERY



By

EBRAHIM TAHERZADEH MOBARAKEH

 \bigcirc

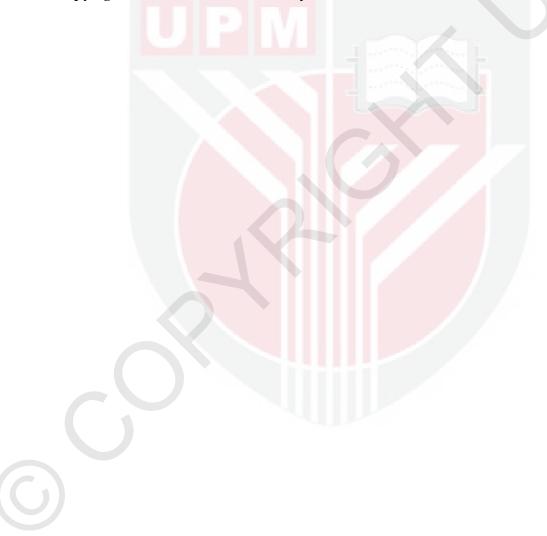
Thesis Submitted to the School of Graduate Studies, Universiti Putra Malaysia, in Fulfillment of the Requirement for the Degree of Doctor of Philosophy

February 2013

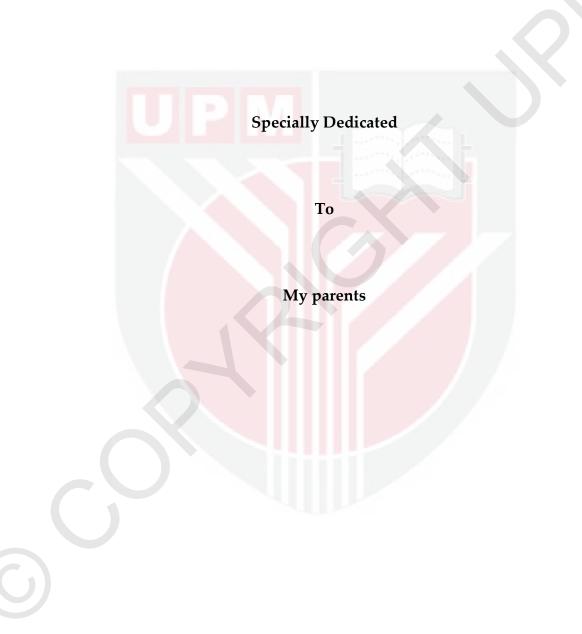
COPYRIGHT

All material contained within the thesis, including without limitation text, logos, icons, photographs and all other artwork, is copyright material of Universiti Putra Malaysia unless otherwise stated. Use may be made of any material contained within the thesis for non-commercial purposes from the copyright holder. Commercial use of material may only be made with the express, prior, written permission of Universiti Putra Malaysia.

Copyright © Universiti Putra Malaysia



In The Name of Allah, the Most Gracious and the Most Merciful



Abstract of the thesis presented to the Senate of Universiti Putra Malaysia in fulfillment of the requirement for the degree of Doctor of philosophy

DETECTION OF ROOF MATERIALS BASED ON AN OBJECT-ORIENTED APPROACH USING WORLDVIEW-2 SATELLITE IMAGERY

By

EBRAHIM TAHERZADEH MOBARAKEH

February 2014

Chairman: Assoc. Prof. Helmi Zulhaidi bin Mohd Shafri-PhD Faculty: Engineering

One of the most challenging tasks in urban remote sensing is detection of impervious surface (IS) which plays an important role in assessing urban environmental conditions. However, accurate impervious surface extraction is still a challenge.

New methods are needed due to the rapid expansion and development of urban centers which are able to do more frequent updating of existing databases instead of traditional methods. In this study, detection of the IS especially the roof surfaces based on their materials is proposed. Detection of roof types and conditions are important and the information on the roof material types can be useful for different applications such as disaster preparedness, urban heat island assessment and runoff quality.

Due to the limitations of airborne hyperspectral data in which data acquisition is normally expensive, the coverage area is limited, and the analysis can be too complex, very-high-resolution (VHR) imagery, such as Worldview-2 (WV-2) image was used. In order to do supervised classification and extract the IS at the parcel level, especially roof materials, adequate training data are needed, but lack of sufficient training data is one of the limitations, due to the security of buildings, permission to access the roof could not be obtained or access was impossible.

The Object-oriented (OO) approach was used in order to utilize the spectral, spatial and textural information which are inherent in VHR imagery. In order to define the objects based on OO approach, certain rules should be defined. This

is a difficult task due to the requirement of the prior knowledge about the objects. Lack of generic rule to extract the IS is another limitation in urban remote sensing.

The main goal of this research is to build the generic rules based on the spectral, spatial and textural information to predict roof materials in WV-2 images without using training data.

A generic model is proposed that is based on spectral, spatial and textural information which were extracted from available training data. Furthermore, discriminant analysis (DA) is used for dimensionality reduction and to discriminate between different spatial, spectral and textural attributes. The generic model consists of a discriminant function based on linear combinations of the predictor variables that provide the best discrimination between the groups. The DA result shows that of the 54 attributes extracted, only 13 attributes related to spatial, spectral and textural information are useful for discriminating different roof materials. Finally, this model was applied to different WV-2 images from different areas and proved that this model has a good potential to predict roof materials in different study areas with more than 81% accuracy. This is performed on WV-2 images without using training data.

Abstrak thesis yang diserahkan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk ijazah Doktor Falsafah

PEMBANGUNAN MODEL GENERIK UNTUK PENGESANAN BAHAN BUMBUNG BERDASARKAN PENDEKATAN BERORIENTASIKAN OBJEK MENGGUNKAN IMEJ SATELIT WORLDVIEW-2

Oleh

EBRAHIM TAHERZADEH MOBARAKEH

Februari 2014

Pengerusi: Prof. Madya Helmi Zulhaidi bin Mohd Shafri-PhD Fakulti: Kejuruteraan

Salah satu aplikasi yang amat mencabar di dalam penderiaan jauh bagi kawasan bandar adalah pengesanan permukaan kedap air yang mana memainkan peranan yang ber penting dalam menilai keadaan persekitaran bandar.

Kaedah baru untuk pengesanan permukaan tersebut adalah diperlukan bagi mengemaskini pangkalan data sedia dengan lebih cepat berbanding kaedah tradisional memandangkan perkembangan dan pembangunan pusat-pusat bandar yang pesat. Dalam kajian ini, pengesanan IS terutamanya permukaan bumbung berdasarkan bahan pembuatannya adalah dicadangkan. Pengesanan jenis dan keadaan bumbung adalah penting dan maklumat berkenaan jenis bahan binaan bumbung berguna untuk pelbagai aplikasi seperti persiapan menghadapi bencana, penilaian pulau haba bandar dan kualiti air larian.

Oleh kerana keterbatasan perolehan data hiperspektral udarayang kebiasaannya mahal, kawasan liputan yang terhad dan analisis boleh menjadi terlalu kompleks ; imej yang beresolusi tinggi seperti Worldview-2 (WV-2) telah digunakan. Dalam usaha untuk melakukan pengkelasan secara berselia dan mengekstrak maklumat IS data latihan yang mencukupi adalah diperlukan. Namun begitu, kekangan yang dihadapi adalah dari segi kekurangan data latihan kerana isu keselamatan bangunan dan kebenaran untuk mengakses bumbung bangunan agak sukar dan mustahil.

Pendekatan berorientasikan objek (OO) telah digunakan untuk menggunakan,. maklumat spektral, spatial dan tekstur yang sememangnya wujud dalam imej VHR. Untuk menentukan objek berdasarkan pendekatan OO, peraturan tertentu perlu ditakrifkan dan merupakan satu tugas yang sukar kerana memerlukan pengetahuan sebelum berkenaan objek. Kekurangan peraturan generik untuk mengesan IS adalah satu lagi limitasi penderiaan jauh bandar.

Matlamat utama kajian ini adalah untuk membina kaedah-kaedah generik berdasarkan, maklumat spektral, spatial dan tekstur untuk meramalkan bahan bumbung didalam imej WV-2 tanpa menggunakan data latihan.

Satu model generik adalah dicadangkan berdasarkan maklumat spektral, spatial dan tekstur yang diekstrak daripada data latihan yang sedia ada. Tambahan pula, analisis diskriminan (discriminant analysis - DA) digunakan untuk mengurangkan dimensionaliti dan membezakan antara sifat-sifat spatial, spektral dan tekstur yang berbeza. Model generik terdiri daripada fungsi diskriminan berdasarkan gabungan linear bagi pemboleh ubah ramalan yang memberikan diskriminasi yang terbaik di antara kumpulan. Hasil DA menunjukkan bahawa daripada 54 sifat-sifat dikeluarkan, hanya 13 sifat-sifat yang berkaitan dengan maklumat spektral, spatial dan tekstur yang berguna untuk membezakan bahan-bahan bumbung yang berbeza. Akhir sekali, model ini telah digunakan untuk memberbezakan imej WV-2 dari daripada kawasan yang berlainan dan terbukti bahawa model ini berpotensi untuk menjangkakan bahan bumbung di kawasan kajian yang berlainan seperti Kuala Lumpur dan dua bahagian di Universiti Putra Malaysia dengan ketepatan melebihi 81% daripada kawasan-kawasan yang berbeza dan membuktikan bahawa model ini mempunyai potensi yang baik untuk meramalkan bahan-bahan bumbung dari WV-2 imej tanpa menggunakan data latihan.

ACKNOWLEDGEMENTS

IN THE NAME OF ALLAH

Praise and thanks are due to Allah who gave me strength and determination to complete my study. I would like to express my gratitude and sincere thanks to those who have helped me in preparing and conducting the research and finishing this thesis. Therefore, it pleases me to express my deep gratitude to them.

I would like to thank the chairman of my supervisory committee, Associate Professor Dr. Helmi Zulhaidi bin Mohd Shafri, for his supervision, kind guidance and advice on the completion of the thesis. I appreciate his patience and sincere approach to motivate, help, advice and guide me to finish my study. Thanks are also extended to my committee member, Professor Shattri Mansor and Professor Ravshan Ashurov.

Not forgetting my dearest friends, especially Farid Faghihinia, Hassan Khalifeh Soltani, Dr. Ramtin Ravanfar, Kaveh Shahi , Majid Ghasemi, Saraj & Mostafa Hosseini, Hamid Sattarifar, Pooya Kolivand, Mohammad Sadeghi, Ramin Azar and Hossein Attar. Thanks for your support and encouragement in times of need.

Last but not least, thanks to my beloved..., my father Ali, my mother Mohtaram, my brothers Mohammad and Mohsen, my sisters Soheila and Susan, my nephew and niece Sepideh, Roza, Nikta, Hossein, Reza and Kamand. I owe you everything. I certify that a Thesis Examination Committee has met on (7/02/2014) to conduct the final examination of Ebrahim Taherzadeh Mobarakeh on his thesis entitled "Dete" in accordance with the Universities and University Colleges Act 1971 and the Constitution of the Universiti Putra Malaysia [P.U.(A) 106] 15 March 1998. The Committee recommends that the student be awarded the Doctor of Philosophy.

Members of the Thesis Examination Committee were as follows:

Thamer Ahmed Mohamed, PhD Professor Engineering Faculty Universiti Putra Malaysia (Chairman)

Abdul Rashid b. Mohamed Shariff, PhD Associate Professor Engineering Faculty Universiti Putra Malaysia (Internal Examiner)

Abdul Rahman b. Ramli, PhD Associate Professor Engineering Faculty Universiti Putra Malaysia (Internal Examiner)

Mahesh Pal, PhD Professor Department of Civil Engineering, NIT Kurukshetra India (External Examiner)

> NORITAH OMAR, PhD Associate Professor and Deputy Dean School of Graduate Studies Universiti Putra Malaysia Date: 21 April 2014

This thesis submitted to the Senate of Universiti Putra Malaysia has been accepted as fulfillment of the requirement for the degree of Doctor of Philosophy. The members of the Supervisory Committee are as follows:

Helmi Z.M. Shafri, PhD Associate Professor Faculty of Engineering Universiti Putra Malaysia (Chairman)

Shattri Mansor, PhD

Professor Faculty of Engineering Universiti Putra Malaysia (Member)

Ravshan Ashurov, PhD

Professor Institute of mathematics National university of Uzbekistan (Member)

> **BUJANG BIN KIM MUAT, PhD** Professor and Dean School of Graduate Studies Universiti Putra Malaysia

Date:

Declaration By Graduate Student

I hereby confirm that:

- this thesis is my original work;
- quotations, illustrations and citations have been duly referenced;
- this thesis has not been submitted previously or concurrently for any other degree at any other institutions;
- intellectual property from the thesis and copyright of thesis are fully-owned by Universiti Putra Malaysia, as according to the Universiti Putra Malaysia (Research) Rules 2012;
- written permission must be obtained from supervisor and the office of Deputy Vice-Chancellor (Research and Innovation) before thesis is published (in the form of written, printed or in electronic form) including books, journals, modules, proceedings, popular writings, seminar papers, manuscripts, posters, reports, lecture notes, learning modules or any other materials as stated in the Universiti Putra Malaysia (Research) Rules 2012;
- there is no plagiarism or data falsification/fabrication in the thesis, and scholarly integrity is upheld as according to the Universiti Putra Malaysia (Graduate Studies) Rules 2003 (Revision 2012-2013) and the Universiti Putra Malaysia (Research) Rules 2012. The thesis has undergone plagiarism detection software.

Signature:

Date:

Name and Matric No.:

Declaration by Members of Supervisory Committee

This is to confirm that:

- the research conducted and the writing of this thesis was under our supervision;
- supervision responsibilities as stated in the Universiti Putra Malaysia (Graduate Studies) Rules 2003 (Revision 2012-2013) are adhered to.

Signature: Name of Chairman of Supervisory Committee:	Signature: Name of Member of Supervisory Committee:
	Committee:
Signature:	
Name of	
Member of	
Supervisory Committee:	

TABLE OF CONTENTS

	-	Page
DEDICATION		ii
ABSTRACT		iii
ABSTRAK		v
ACKNOWLEDGEMENTS		vii
APPROVAL		viii
DECLARATION		x
LIST OF FIGURES		xvi
LIST OF TABLES		xvii
LIST OF ABBREVIATIONS		xxiii

CHAPTER

 $\overline{\mathbb{G}}$

4	INTRODUCTION	1
1	INTRODUCTION	1
	1.1 Background	1
	1.2 Research Objectives	5
	1.3 Organization of Remaining Chapters	5
2	LITERATURE REVIEW	7
	2.1 Introduction	7
	2.2 Remote Sensing	7
	2.3 Urban Remote Sensing	7
	2.4 Impervious Surface	8
	2.4.1 Impervious Surface Impacts	11
	2.4.2 Roof Type Mapping	12
	2.5 Application of RS in Detection of Impervious	16
	Surface	
	2.5.1 Multispectral Remote Sensing	16
	2.5.2 Hyperspectral Remote Sensing	17
	2.6 Classification Methods	17
	2.6.1 Spectral-Based Approach	17
	2.6.2 Object-oriented Approach	20
	2.7 Discriminate Analysis	23
	2.8 Summary of Literature	24

MATERIALS AND METHODS 27 3.1 Introduction 27 3.2 Hyperspectral Remote Sensing 27 3.2.1 Data Acquisition 28 3.2.2 Preprocessing 30 3.2.3 Categorization of Urban Material 30 3.2.4 Endmember Extraction 31 3.2.5 SVM Classification 33 3.2.6 Parameters Selection For SVM 34 35 3.2.7 Using Adaptive Filters 35 3.2.8 Accuracy Assessment 3.3 Worldview-2 36 3.3.1 Data Acquisition 36 3.3.2 Categorization the Urban Materials 38 Spectral-Based Classification 3.3.3 38 3.3.4 **Object-oriented** (Mathematical 38 Morphology) Object-oriented (Generic model) 42 3.3.5 **RESULTS AND DISCUSSIONS** 48 4.1 Introduction 48 4.2 Hyperspectral Data 48 4.2.1 Ground-Truth Image 48 4.2.2 Significant bands 50 4.2.3 Classification 50 4.3 Worldview-2 56 4.3.1 Ground-Truth Image 56 Spectral-Based Classification 4.3.2 58 4.3.3 Object-oriented (Mathematical 62 Morphology) 4.3.4 Classification Based On The Generic 72 Model 4.4 Discussion 83

3

5	CONCLUSIONS AND RECOMMENDATIONS FOR FUTURE RESEARCH	87
	5.1 Introduction	87
	5.2 Conclusion	87
	5.3 Recommendations for Future Work	90
REFERENCES		91
APPENDICES		107
BIODATA OF ST	TUDENT	119
		120

LIST OF FIGURES

Figure		Page
2.1	Urban Heat Island Profile	9
2.2	Effect of Imperviousness on Penetration of Runoff	10
2.3	Relationship of Watershed Imperviousness to Runoff Coefficient Levels	11
2.4	Relationship Between Watershed Imperviousness and Base Flow Water Temperature	12
2.5	Effect of Percent of IS to Stream Health	12
2.6	Effect of Different Roof Material in Energy Consumption	13
2.7	Concrete Tiles versus Clay Tile Compared on the Days of Matching Outdoor Temperature	14
2.8	Concrete Tile Roof Affected by Algae	14
3.1	Materials and Methods Chapter Overview	27
3.2	Hyperspectral Image From a part of Kuala Lumpur (false color)	29
3.3	Steps Within the Spectral Hourglass Wizard	32
3.4	Classification Using SVM	34
3.5	The Locations and Images of the Study Areas A) WV-2 Image Of Part of KL; B, C) WV-2 Images of Parts of the Universiti Putra Malaysia (UPM) -UPM_Eng and UPM_Itma Respectively	37
3.6	Structuring Elements: (A) Cross, (B) Square and (C) Line	39
3.7	Flowchart of the Analytical Procedures to Extract the Generic Model Based on the OO Approach	43
4.1	Urban spectra from AISA sensor for an Urban Scene in	49
	Kuala Lumpur, Malaysia	
4.2	SVM Classification Result (Without Filtering)	51
4.3	SVM Classification Using Lee Filter	53
4.4	SVM Classification Result After Using Enhanced Lee Filter Used	55
4.5	Study area of part of Kuala Lumpur A) WV-2 image of part of KL b) Training area c) Testing area	57
4.6	Maximum Likelihood Classification Result of a Part of KL WV-2 Image	59
4.7	Spectral-Based Classification Using SVM Method	61

4.8	Different SE Sizes With Disk Shape Using Erosion	63
	Operator	
4.9	Result of Applying Dilation Using Different SE Size And Disk Shape	64
4.10	Result of Applying An Opening Operator (Column A) And Opening By Reconstruction Using Different SE Size (Column B)	66
4.11	Closing Operator (Column A) And Closing By Reconstruction (Column B) Using Different SE Sizes	67
4.12	MP for Band One	69
4.13	Derivative Morphological Profiles (DMP)	70
4.14	Spatial +Spectral (MM) Based Classification Using SVM Classifier	71
4.15	A) Segmentation Preview B) Merging Preview C) Result Of Merging	73
4.16	Segmentation Layer	75
4.17	Spatial Distribution of Roof Materials in KL Image Based on the Gene <mark>ric</mark> Model	79
4.18	Figure Spa <mark>tial Distribution</mark> of Roof Types Based on the Generic Model on UPM_Eng	81
4.19	Figure Sp <mark>atial Distribution</mark> of Roof Types Based on the Generic Model on UPM_Itma	82

 \bigcirc

LIST OF TABLES

Table		Page
2.1	Metal concentration in storm water runoff from direct surface in Australia (Concentration in µg/l)	16
4.1	Training and testing pixel	50
4.2	Classification accuracy without applying the filter	52
4.3	Classification accuracy after applying Lee filter	54
4.4	Classification accuracy, using an Enhanced Lee filter	56
4.5	Number of training and testing pixels	58
4.6	MLH classification accuracy based on confusion matrix	59
4.7	Spectral-based classification accuracy, using SVM method	62
4.8	MM approach classification accuracy	72
4.9	Spectral attributes definition which is extracted for each band	74
4.10	Textural attribute definition which is extracted from the image	74
4.11	Color and band ratio attribute definition	75
4.12	Number of attributes which were computed based on spatial, spectral, texture and color	76
4.13	Number of training in a segment and testing in Pixels	76
4.14	Attribute definitions (ENVI ZOOM Tutorial)	77
4.15	Discriminant function coefficients	78
4.16	Functions at group centroids	79
4.17	Accuracy of the model in predicting roof materials in different areas	83

LIST OF ABBREVIATIONS

°C	Celsius degree
ASTER	Advanced space-borne thermal emission and reflection
	radiometer
CASI	Compact airborne spectrographic imager
CV	Cross-validation
DA	Discriminate analysis
et al.	et alia
etc.	et cetera
ETM+	Landsat enhanced thematic mapper plus
g/l	Gram per liter
HSI	Hue, saturation, and intensity
KL	Kuala Lumpur
LDA	linear discriminant analysis
LST	land surface temperature
М	Meter
Mg	Milligram
mg/l	milligram per liter
MLH	Maximum likelihood
MM	Mathematical morphology
MNF	Minimum noise fraction
MP	Morphological profile
MTMF	Mixture tuned matched filtering
NIR	Near infrared
NN	Neural Network

- OO Object oriented
- PCA Principle component analysis
- PPI Pixel purity index
- QUAC Quick atmospheric correction
- RBF Radial basis function
- ROI Region of interest
- RS Remote sensing
- SAM Spectral angle mapper
- SWIR Shortwave infrared
- TIR Thermal infrared
- TM Thematic mapper
- UHI Urban heat island
- UPM University Putra Malaysia
- VHR Very high resolution
- V-I-S Vegetation-Impervious surface–Soil
- WV-2 Worldview-2

CHAPTER 1

INTRODUCTION

1.1 Background

Almost half of the world population lives in urban areas and this proportion will be increased to 60% by 2030 (United Nations, 2001). Over the last few decades, urban area has been growing very fast, although this area occupies the small fraction of the earth's land surface (5% of the earth's surface). The large diversity of man-made and natural surface materials in the urban area can affect the ecological (Arnold and Gibbons, 1996), climatic and energetic (Oke, 1987) conditions. The centers of human activity are cities, and because of the boom of cities, urban mapping is becoming more significant. Due to the rapid expansion and development of urban centers and cities, new methods are needed in order to update the databases instead of traditional methods, which are typically based on field investigations and the visual interpretation of aerial photographs.

Remote sensing (RS) refers to the technology that is able to collect the data from the object without direct contact with it. Using RS technology and RS data for urban areas applications is relatively new topic and becoming one of the interesting topics in the RS community. With the advent of high resolution imagery and new sensors the new opportunities are opened to the RS community for different urban application.

Impervious surface (IS) is defined as the entirety of surfaces through where the water cannot penetrate the soil, including roads, sidewalks, parking lots, rooftops, and so on (Arnold and Gibbons, 1996). It has been known as an important indicator in urban environmental conditions (Conway, 2007; Hu and Weng, 2009; Weber and Bannerman, 2004). One of the sources of pollution in the rivers and lakes is runoff in the urban area that mostly created through the IS (Arnold and Gibbons, 1996; Booth and Jackson, 1997).

Detection and assessment of the percentage of IS in the heterogeneous urban area are one of the challenging and important tasks in urban RS. Due to the impact of the IS on the environment the concentration in this field of study has been growing (Weng, 2001; Civco *et al.*, 2002; Dougherty *et al.*, 2004; Wang and Zhang, 2004).

In this study, detection of the IS especially the roof of buildings based on their materials using multispectral RS data is proposed. Detection of the roof types and conditions are important; knowledge about the roof material types can assist applications such as disaster preparedness (Bhaskaran *et al.*, 2001), urban heat island assessment (Ben-Dor *et al.*, 2001) runoff quality (Paul and Meyer 2001; Birs and Robert- Sainte 2009,; Clark *et al.*, 2008).

Traditional methods which are based on field survey is very time consuming and costly, additionally in some part of the urban area collection of the data is very difficult tasks due to the security of the building. RS data can play the critical role in order to provide the information about the spatial distribution of IS in the urban area.

Still, there is a lack of suitable methods to quantify and assess IS in an accurate, quick and economical way. In order to update the information about the cities new methods are needed instead of the existing and traditional methods.

In order to assess and classify the urban land cover at the material level, very high resolution (VHR) imagery is required (Zhou and Troy, 2008). For supervised classification of IS at material level, such as roofing materials, the adequate training data is needed. Lack of sufficient training data is one of the limitations, due to the security of buildings, permission to access the roof could not be obtained, or access was impossible. Based on the literature, several studies have been done to illustrate the potential of hyperspectral data to discriminate the urban surface materials (Ben-Dor *et al.*, 2001; Heiden *et al.*, 2001; Heiden *et al.*, 2007; Herold *et al.*, 2004) because of its high spectral resolution able to detect, classifying, and disseminate materials on the earth more than the traditional multispectral imagery with only several wideband spectral channels (Platt and Goetz, 2004).

In this research, two types of RS data were employed such as hyperspectral and multispectral data to detect the different roof types in Malaysia. Despite the availability of space-borne hyperspectral systems (Shafri *et al.*, 2012) data acquisition is normally expensive, the coverage area is limited and the analysis can be too complex compared to the multispectral imagery. In this study VHR imagery such as Worldview-2 (WV-2) image was used which, unlike other commercial sensors, contains 8 spectral bands high spatial resolution (0.5 m pan

sharpened). Some studies employed Landsat or SPOT image, but these images are not recommended for an urban study because of the low spatial resolution. It should be mentioned that despite the fact that VHR imagery provides more information about an object, but once spatial resolution increases, discrimination between classes is reduced because of increased internal variability of objects (Thomas *et al.*, 2003; Kumar and Castro, 2001). Some studies have used the VHR satellite image such as Ikonos or Quickbird; they are relatively limited in their spectral resolution and these types of data are employed to discriminate between significant land cover in urban areas such as road, building and water body and still there are some misclassification in order to discriminate and classifying the residential and non-built up materials (Herold *et al.*, 2002).

Studies show that the traditional RS classification methods such as Maximum Likelihood (MLH) and Parallelepiped which are only based on the spectral information are insufficient for classifying the VHR imagery in urban area (Cushnie, 1987; Thomas *et al.*, 2003, Chen *et al.*, 2004). Recent research shows that using the spectral information only is not sufficient for urban area mapping up to materials level and integration of high spectral and spatial information is necessary to achieve accurate mapping output.

Spatial information is valuable and significant when analyzing very high spatial resolution RS data and utilizing this type of information can be directly exploited for modeling the objects in the scene, especially to detect and extract the objects and in addition this can be increased the discriminability between different classes in classification tasks (Dalla Mira *et al.*, 2010). It is very important to combine the spectral information with another source of information which is inherent in an image such as spatial information (Gong *et al.*, 1992; Pesaresi and Benediktsson, 2001; Shackelford and Davis, 2003a; Chen *et al.*, 2004) and texture information for land use and land cover classification that can improve the classification accuracy.

One of the successful techniques in order to utilize the spectral and spatial information by Pesaresi and Benediktsson (2001) was presented. They presented a technique based on morphological profiles (MPs), which was carried out with opening and closing transformations that are defined in mathematical morphology (MM) based on the geodesic reconstruction.

With the aim of extracting and integrating spatial, spectral and textural information, the object-oriented (OO) approach is used. Studies show the

discrimination between different land cover in the urban area is increased with spectral similarity when these types of information are employed (Wang *et al.*, 2007; Gong *et al.*, 1992; Shackelford and Davis, 2003b; Wang *et al.*, 2012; Goetz *et al.*, 2003). The OO classification approaches, in general, show better results compared with pixel-based approaches when mapping individual landscape features (Lu and Weng, 2007).

Nevertheless, to define objects in terms of certain rules based on extracted information, a priori knowledge is needed which is a difficult task. Generally, the relevant information is not well formalized, and it is difficult to directly obtain such implicit knowledge, which is ordinarily held only by domain experts (Sheeren *et al.*, 2006). Previous studies, such as (Hamedianfar and Shafri, 2013) have defined the optimal rule sets for a limited coverage urban area, but the rules are not transferable and generic for other different areas.

According to some limitations of urban RS in order to extract the IS at material level, it should be mentioned that most of the studies employed airborne hyperspectral data due to high spectral resolution of this data which enable us to extract more information about the IS up to the material level. However, there is a lack of studies that used multispectral satellite data to extract the IS at the material level. In addition, collecting adequate training data is very time consuming and costly, thus inadequate training data and lack of generic and transferable model that utilized the spectral, spatial and textural information inherent in the VHR data are other challenging tasks in urban RS context.

Thus the main goal of this research is to build a generic model to map roof materials based on the OO approach and available training data and finally apply this generic model to predict and detect the spatial distribution of the roof materials in different areas without using training data using WV-2 images. The proposed technique has the potential to be economically and technically superior; for example there is no need to use the training data and utilize all the information that are inherent in the WV-2 image thus leads to savings in time and money.

As an added motivation of this research, determination of housing assessment in Malaysia is now a controversial issue. The use of the proposed technique can help with not only identifying the roof material, but also helps to determine the illegal extension or renovation of housing blocks.

1.2 Research Objectives

In general, this study will investigate the applicability of VHR satellite RS data in mapping urban features, especially roof materials and build a generic model based on spectral, spatial and texture information.

The specific objectives of this study are:

- To investigate the performance of hyperspectral and high spatial resolution multispectral data for their suitability and effectiveness in urban mapping.
- To design a new approach based on generic rules for extracting the roof materials based on spectral, spatial, texture and color combination in order to improve urban area classification accuracy and better discrimination of urban roof materials.
- To validate the generic model in different study areas in order to test the transferability of the proposed technique.

1.2 Organization of Remaining Chapters

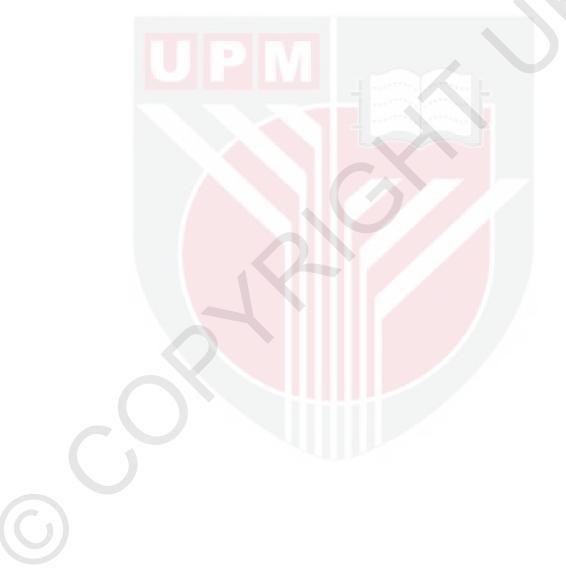
Chapter 2 reviews the literatures related to definition of RS, urban RS, IS and later on the application of RS including the multispectral and hyperspectral data in order to detect the IS. The next part of this chapter, some important types of classification approaches such as spectral-based and OO will be reviewed. In the last part of this chapter the summary and the gaps of the previous researches will be explained.

Chapter 3 includes the materials and methods utilized in this research to achieve the objectives of this study. The first of this chapter is related to hyperspectral data analysis in order to extract the roof materials. The next part of this chapter assigns to the procedure and methods applied to WV-2 image such as the spectral-based classification approaches (MLH, Support vector machine) and the developed techniques to extract the roof materials that include MM and generic model.

Chapter 4 includes the results of this research after applying the different methods. As mentioned before, two types of data were used in this study thus the first part of this chapter is related to the experimental result of hyperspectral

data using spectral-based approach and the second part is related to multispectral data which is divided to three parts including the spectral-based, OO using MM and finally the generic model result. At the end of this chapter the findings and results will be interpreted.

In chapter 5, conclusions and recommendations are presented. Overview of the findings, limitations and the benefits of this research are explained and furthermore some recommendations for future studies are given in the recommendation part of this chapter.



REFERENCES

- A.A.V.V., "Council Directive 76/769/EEC of 27 July 1976 on the approximation of the laws, regulations and administrative provisions of the Member States relating to restrictions on the marketing and use of certain dangerous substances and preparations (76/769/EEC)", Official Journal of the European Communities, L 262/023, pp. 1-3, 1976.
- A.A.V.V., "Legge 27 marzo 1992, n. 257 Norme relative alla cessazione dell'impiego dell'amianto", Supplemento Ordinario alla Gazzetta Ufficiale, 87, pp. 1-13, 13 Aprile 1992.
- Arnold, C.L.J., and Gibbons, C.J. (1996). Impervious surface coverage: The emergence of a key environmental indicator. *Journal of American Planning Association*. 62(2): 243–258.
- Ayuga, R., Gregorio, J., Marin, R.M., Velasco, M., Antonio, J, Hyperspectral Remote Sensing Application for Semi-Urban Areas Monitoring, Urban Remote Sensing Joint Event, Paris, France, April. 11-13, 2007.
- Bähr, H.P., Lemp, D., Weidner, U, Hyperspectral Meets Laser scanning: Image Analysis of Roof Surfaces, In: ISPRS Workshop High Resolution Earth Imaging for Geospatial Information, Vol. XXXVI Part I/W3, 2005.
- Baltsavias, E.P. (2004). Object extraction and revision by image analysis using existing geodata and knowledge: current status and steps towards operational systems. *Journal of Photogrammetry and Remote Sensing*. 58(3): 129-151.
- Bandos, T.V., Bruzzone, L., and Camps-Valls, G. (2009). Classification of hyperspectral images with regularized linear discriminant analysis. *IEEE Transactionon Geoscience and Remote Sensing*, 47(3): 862-873.
- Bannerman, R. (1994). Sources of urban storm water pollutants defined in Wisconsin. *Watershed Protection Techniques*, 1(1): 30-31.
- Ben-Dor, E., Levin, N., Saaroni, H. (2001). A spectral based recognition of the urban environment using the visible and near-infrared spectral region (0.4–1.1 nm) a case study over Tel-Aviv. *International Journal of Remote Sensing*, 22(11): 2193–2218.

- Benediktsson, J.A., Palmason, J.A., and Sveinsson, J. (2005). Classification of hyperspectral data from urban areas based on extended morphological profiles. *IEEE Transaction Geosciences and Remote Sensing*, 43(3): 480-491.
- Bhaskaran , S., Datt, B., Neal, T., and Forster, B,*Hail storm vulnerability assessment by using hyperspectral remote sensing and GIS techniques*, Proceedings of the IGARSS symposium,Sydney, Australia, July. 9-13, 2001.
- Birs, L.A., and Robert-Sainte. P, *Classification of roof materials for rainwater pollution modelization*, Proceeding of International Archives of Photogrammetry and Remote Sensing, Hanover, Germany, June, 2009.
- Blaschke, T. (2010). Object based image analysis for remote sensing. *Journal of Photogrammetry and Remote Sensing*, 65(1): 2-16.
- Boller, M., (1997). Tracking heavy metals reveals sustainability deficits of urban drainage systems. *Water Science and Technology*, 35(9): 77-87.
- Booth, D.B. and Jackson, C.R. (1997).Urbanization of aquatic system degradation thresholds, stromwater detection, and the limits of mitigation. *Journal of the American Water Resources Association*, 33(5): 1077-1090.
- Burns, R. P., and Burns, R. (2008). Business research methods and statistics using SPSS.Sage.
- Cablk, M.E. and Minor, T.B. (2003). Detecting and discriminating impervious cover with high-resolution IKONOS data using principal component analysis and morphological operators. *International Journal of Remote Sensing*, 24(23): 4627-4645.
- Casals-Carrasco, P., Kubo, S., and Madhavan, B.B. (2000). Application of spectral mixture analysis for terrain evaluation studies. *International Journal of Remote Sensing*, 21(16): 3039-3055.
- Cavalli, R.M., Fusilli, L., Pascucci, S., Pignatti, S., and Santini, F. (2008). Hyperspectral sensor data capability for retrieving complex urban land cover in comparison with multispectral data: Venice city case study (Italy). *Sensors*, 8(5): 3299-3320.
- Chen, D., Stow, D.A., and Gong, P. (2004).Examining the effect of spatial resolution and texture window size on classification accuracy: an urban

environment case. International Journal of Remote Sensing, 25(11): 2177–2192.

- Chen, Y., Shi, P., Fung, T., Wang, J., and Li, X. (2007). Object-oriented classification for urban land cover mapping with ASTER imagery. *International Journal of Remote Sensing*, 28(20): 4645-4651.
- Civco, D. L., Hurd, J. D., Wilson, E. H., Arnold, C. L., and Prisloe, S. (2002). Quantifying and describing urbanizing landscapes in the northeast United States. *Photogrammetric Engineering and Remote Sensing*, 68(10): 1083-1090.
- Clark, M. L., Roberts, D. L., and Clark, D. B. (2005). Hyperspectral discrimination of tropical rain forest tree species at leaf to crown scales. *Remote Sensing of Environment*, 96(3-4): 375–398.
- Clark, S. E., Steele, K. A., Spicher, J., Siu, C. Y. S., Lalor, M. M., Pitt, R., Kirby, J. T. (2008). Roofing materials' contributions to storm-water runoff pollution. *Journal of Irrigation and Drainage Engineering*, 134 (5): 638-645.
- Congalton, R. G. (1991). A review of assessing the accuracy of classifications of remotely sensed data. *Remote Sensing of Environment*, 37(1): 35–46.
- Conway, T. M. (2007). Impervious surface as an indicator of pH and specific conductance in the urbanizing coastal zone of New Jersey, USA. *Journal of Environmntal Management*, 85(2): 308-16.
- Crabill, C., Donald, R., Snelling, J., Foust, R. and Southam, G. (1999). The impact of sediment fecal coliform reservoirs on seasonal water quality in Oak Creek, Arizona. *Water Research*, 33(9): 2163-2171.
- Cushnie, J. L. (1987). The interactive effects of spatial resolution and degree of internal variability within land-cover type on classification accuracies. *International Journal of Remote Sensing*. 8: 15–22.
- Dalla Mura, M., Benediktsson, J. A., Waske, B., and Bruzzone, L. (2010).
 Extended profiles with morphological attribute filters for the analysis of hyperspectral data. *International Journal of Remote Sensing*, 31(22): 5975-5991.

- Davis, M.P., Nordin, N.A., Ghazali, M. (2006). Thermal comfort Honeycomb Housing. Institute of Advanced Technology, Universiti Putra Malaysia, 2006 ISBN 983-3455-38-7.
- Debusk, K.M., Hunt, W.F., Osmond, D.L., Cope, G.W. (2009). Water quality of rooftop runoff: Implications for residential water harvesting systems. Published byNorth Carolina cooperative extension.<u>www.bae.ncsu.edu/stormwater/PublicationFiles/RooftopRunof</u> <u>f2009.pdf</u>
- Dell'Acqua, F., Gamba, P., Ferari, A., Palmason, J. A., Benediktsson J. A., and Arnason, K. (2004). Exploiting spectral and spatial information in hyperspectral urban data with high resolution. *IEEE Geoscience and Remote Sensing Letter*, 1(4): 322-326.
- Dougherty, M., Dymond, R. L., Goetz, S. J., Jantz, C. A., and Goulet, N. (2004). Evaluation of impervious surface estimates in a rapidly urbanizing watershed. *Photogrammetric Engineering and Remote Sensing*, 70(11): 1275-1284.
- Douglas, D. H., and Peucker. T. K. (1973). Algorithms for the reduction of the number of points required to represent a digitized line or its caricature. *Cartographica*, 10(2): 112-122.
- Driscoll, M.O., Clinton, S., Jefferson, A., Manda A., and Mcmillan, S. (2010).Urbanization effects on watershed hydrology and in-stream processes in the southern United States. *Water*, 2(3): 605-648.
- Du, Q., and Chang, C. I. (2001). A linear constrained distance-based discriminant analysis for hyperspectral image classification. *Pattern Recognition*, 34(2): 361-373.

ENVI Help, (2006). ENVI 4.3 Help. ITT Industries Inc., Boulder, CO, USA.

- Fauvel, M., Benediktsson, J. A., Chanussot, J., and Sveinsson, J. R, Spectral and spatial classification of hyperspectral data using SVMs and morphological profiles, Proceeding on IGARSS, Barcelona, Spain, July. 2007.
- Fauvel, M., Chanussot, J. and Benediktsson, J. A, Evaluation of Kernels for Multiclass Classification of Hyperspectral Remote Sensing Data, Proceeding in IEEE International Acoustics, Speech and Signal Processing (ICASSP), May. 14-19, 2006a.

- Fauvel, M., Chanussot, J., Benediktsson, J.A, A combined support vector machines classification based on decision fusion, Proceeding in IGARSS, Denver, CO, USA, July. 31- Aug. 4, 2006.
- Fauvel, M., Palmason, J. A., Benediktsson, J. A., Chanussot, J., and Sveinsson, J. R, *Classification of remote sensing imagery with high spatial resolution*, Image and Signal Processing for Remote Sensing XI, SPIE ,Bellingham, WA, USA, October, 2005.
- Flanagan, M., and Civco, D. L, *Subpixel impervious surface mapping*, Proceedings of the ASPRS Annual Convention, St. Louis, MO, April,2001.
- Galli, J., (1990). Thermal impacts associated with urbanization and stormwater management best management practices. Publication 91701.metropolitan Washington council of governments. Washington D.C. 172 p.
- Göbel, P., Dierkes, C., and Coldewey, W. G. (2007). Storm water runoff concentration matrix for urban areas. *Journal of contaminant hydrology*, 91(1), 26-42.
- Goetz, S. J., Wright, R. K., Smith, A. J., Zinecker, E., and Schaub, E. (2003).
 IKONOS imagery for resource management: tree cover, impervious surfaces and riparian buffer analyses in the mid-Atlantic region. *Journal of Remote Sensing of Environment*, 88(1): 195–208.
- Gong, P., Marceau, D.J., and Howarth, P. J. (1992). A comparison of spatial feature extraction algorithms for land-use classification with SPOT HRV data. *Remote Sensing of Environment*, 40(2): 137–151.
- Good, J. C. (1993). Roof runoff as a diffuse source of metals and aquatic toxicity in storm water. *Water Science Technology*, 28 (3-5): 317-322.
- Govender, M., Chetty, K., and Bulcock, H. (2007). A review of hyperspectral remote sensing and its application in vegetation and water resource studies. *Water*, 33(2): 145-152.
- Hahn, C., Wijaya, A. and Gloaguen, R, *Application of Support Vector Machine fo r Complex Land Cover Classification using Aster and Landsat Data*, Proc eedings of Gemeinsame Jahrestagung der SGPBF, DGPF und OVG, Pubications of DGPF, Muttenz/Basel, Switzerland, 2007.

- Hamedianfar, A., and Shafri, H. Z. M. (2013). Development of fuzzy rule-based parameters for urban object-oriented classification using very high resolution imagery. *Geocarto*, 1-25.
- Hay, G. J., Niemann , K. O., and Mclean, G. (1996). An object-specific image texture analysis of high-resolution forest imagery. *Remote Sensing of Environment*. 55(2): 108–122.
- He, L. M., Kong, F. S., and Shen, Z. Q, *Multiclass SVM based land cover classification with multisource data*, In ternational Conference on Machine Learning and Cybernetics(ICMLC), Guangzhou, China, Aug. 18-21, 2005.
- Heiden, U., Roessner, S., Segl, K., and Kaufmann, H,(2001) Analysis of spectral signatures of urban surfaces for their identification using hyperspectral HyMap data, Proceeding Remote Sensing and Data Fusion over Urban Areas, IEEE/ISPRS Joint Workshop, Rome, Italy, NOV. 8.
- Heiden, U., Segl, K., Roessner, S., and Kaufmann, H. (2007). Determination of robust spectral features for identification of urban surface materials in hyperspectral remote sensing data. *Journal of Remote Sensing of Environment*, 111(4): 537-552.
- Herold, M., Gardner, M., Hadley, B., and Roberts, D,(2002) The spectral dimension in urban land cover mapping from high-resolution optical remote sensing data, Proceedings of the 3rd Symposium on Remote Sensing of Urban Areas, Istanbul, Turkey, June. 2002.
- Herold, M., Liu, X., and Clarke, K. C. (2003). Spatial metrics and image texture for mapping urban land use. *Photogrammetric Engineering and Remote Sensing*, 69(9): 991–1001.
- Herold, M., Roberts, D. A., Gardner, M.E., and Dennison, P.E. (2004).Spectrometry for urban area remote sensing-development and analysis of a spectral library from 350 to 2400 nm. *Journal of Remote Sensing of Environment*, 91(3): 304–319.
- Hodgson, M. E., Jensen, J. R., Tullis, J. A., Riordan, K. D., and Archer, C. M. (2003).Synergistic use of lidar and color aerial photography for mapping urban parcel imperviousness. *Photogrammetric Engineering and Remote Sensing*, 69(9): 973-980.

- Hu, X. and Weng, Q. (2009). Estimating impervious surfaces from medium spatial resolution imagery using the self-organizing map and multi-layer perceptron neural networks. *Remote Sensing of Environment*, 113(10): 2089–2102.
- Hu, X., and Weng, Q. (2011). Impervious surface area extraction from IKONOS imagery using an object-based fuzzy method. *Geocarto*, 26(1): 3-20.
- ITT VIS (Visual Information Solutions). 2008. Chapter 3: Finding Objects. ENVI feature extraction module 4.6 User's Guide. pp. 1–78
- Jeng, H. A. C., Englande, A. J., Bakeer, R. M., and Bradford, H. B. (2005). Impact of urban stormwater runoff on estuarine environmental quality. *Estuarine Coastal and Shelf Science*, 63(4): 513-526.
- Jensen, J. R. (2007). *Remote Sensing of the Environment: An Earth Resource Perspective*, 2nd ed. Prentice Hall, Upper Saddle River, NJ, USA.
- Jensen, J.R.(1996). *Introductory Digital Image Processing*: A remote Sensing Perspective. Prentice Hall, Inc.,Old Tappan, NJ.
- Jong, S. M. D., Meer, F. D. V., and Clevers, J. G. P. W. (2004). Remote sensing image analysis: Including the spatial domain. *Basic of Remote Sensing* (pp. 1-15). Springer.
- Kärdi, T. (2007). Remote sensing of urban areas: linear spectral unmixing of landsat thematic mapper images acquired over tartu (Estonia). *Estonian Academy of Science, Biology and Ecology*, 56: 19–32.
- Karydas, C. G., Gitas, I. Z. (2011). Development of an IKONOS image classification rule-set for multi-scale mapping of Mediterranean rural landscapes. *International Journal of Remote Sensing*, 32(24): 9261-9277.
- Keshava, N., and Mustard, J. F. (2002). Spectral unmixing. *Signal Processing Magazine*, 19(1): 44-57.
- Kumar, M., and Castro, O.T, *Practical aspects of IKONOS imagery for mapping*, In Proceeding the 22nd Asian Conference on Remote Sensing, Singapore, 2001.

- Landgrebe, D. (1999). Some fundamentals and methods for hyperspectral image data analysis. SPIE Vol 3603,systems and technologies for clinical diagnostics and drug discovery II, 3603, Gerald E. Cohn; John C. Owicki; Eds
- Lee, J. H., and Bang, K. W. (2000).Characterization of urban stormwater runoff. *Water Research*, 34(6): 1773-1780.
- Lee, J. S. (1980). Digital Image Enhancement and Noise Filtering by Use of Local Statistics. *IEEE Transactions on Pattern Analysis and Machine Intelligence*, 2(2): 165-168.
- Li, H., Sharkey, L. J., Hunt, W. F., and Davis, A. P. (2009). Mitigation of impervious surface hydrology using bioretention in North Carolina and Maryland. *Journal of Hydrologic Engineering*, 14(4): 407-415.
- Liao, W., Pizurica, A., Philips, W., and Pi, Y, *Feature extraction for hyperspectral images based on semi-supervised local discriminant analysis*, In urban remote sensing event (JURSE), Munich, Germany, April. 11-13, 2011.
- Lillesand, T.M., and Kiefer, R. (1994). *Remote sensing and image interpretation*. John Wiley and Sons, Inc., New York.
- Liu, J., Pattey, E., Nolin, M. C., Miller, J. R., and Ka, O. (2008). Mapping withinfield soil drainage using remote sensing, DEM and apparent soil electrical conductivity. *Geoderma*, 143(3): 261-272.
- Lobo, A. (1997). Image segmentation and discriminant analysis for the identification of land cover units in ecology. *IEEE Transcation Geosciences Remote Sensing*, 35(5): 1136–1145.
- Lohani, V., Kibler, D. F., and Chanat, J. (2002).Constructing a problem solving environment tool for hydrologic assessment of land use change. *Journal of the American Water Resources Association*, 38(2): 439-452.
- Lopes, A., Touzi, R., and Nezry, E. (1990). Adaptive speckle filters and scene heterogeneity. *IEEE Transactions on Geoscience and Remote Sensing*. 28(6): 992-1000.
- Lu, D., and Weng, Q. (2006). Use of impervious surface in urban land-use classification. *Remote Sensing of Environment*, 102(1): 146-160.

- Lu, D., and Weng, Q. (2007). A survey of image classification methods and techniques for improving classification performance. *International Journal* of Remote Sensing, 28(5): 823–870.
- Lu, D., and Weng, Q. (2009).Extraction of urban impervious surfaces from IKONOS imagery. International Journal of Remote Sensing, 30(5): 1297– 1311.
- Lu, D., Batistella, M., Li, G., Moran, E., Hetrick, S., Freitas, C. D. C., and Sant'Anna, S. J. S. (2012). Land use/cover classification in the Brazilian Amazon using satellite images. *Pesquisa Agropecuária Brasileira*, 47(9): 1185-1208.
- Lu, D., Hetrick, S. and Moran, E. (2010). Land cover classification in a complex urban-rural landscape with Quickbird imagery. *Photogrammetric Engineering and Remote Sensing*, 76(10): 1159–1168
- Mabbott, A. (1998). About roof mold, lichen, algae.<u>http://www.squeegeepro.com/our-blog/466-about-roof-mold-lichen-algae.html</u>
- Maktav, D. and Erbek, F.S. (2005). Analysis of urban growth using multitemporal satellite data in Istanbul, Turkey. *International Journal of Remote Sensing*. 26(4): 797–810.
- Maktav, D., Erbek, F. S., Jurgens, C. (2005a). Remote sensing of urban areas. International Journal of Remote Sensing, 26(4): 655–659.
- Marino, C.M., Panigada, C., Busetto, L., Galli A., and Boschetti, M, Environmental applications of airborne hyperspectral remote sensing: Asbestos concrete sheeting identification and mapping, Proceedings of the 14th International Conference Workshops Applied Geologic Remote Sensing, Rome, Italy, Nov. 2001.
- Matheron, G. (1967). Eléments pour une Théorie des Milieux Poreux, Masson, Paris, France.
- MDNR (Maryland department of natural resources). Impacts of impervious land cover on Maryland. Manta publication, 2007 <u>http://www.dnr.state.md.us/streams/pdfs/ImperviousFactSheet.pdf</u>

- MDNR (Maryland Department of Natural Resources).(2012).HowImperviousSurfaceImpactsHealth.http://www.streamhealth.maryland.gov/impervious.asp
- Mercier, G., and Lennon, M. Support vector machines for hyperspectral image classification with spectral-based kernels, Proceeding in Geoscience and Remote Sensing Symposium, IGARSS'03, July. 2003.
- Mika, S., Rätsch, G., Weston, J., Schölkopf, B., Smola, A. J., and Müller, K.R. (2000).Invariant feature extraction and classification in kernel spaces. *Advances in Neural Information Processing Systems*, 12: 526-532.
- Mohan, A., Sapiro, G., Bosch, E. (2007). Spatially coherent nonlinear dimensionality reduction and segmentation of hyperspectral images. *IEEE Geoscience and Remote Sensing Letters*. 4(2): 206–210.
- Myint, S. W., Gober P., Brazel, A., Grossman-Clarke, S., Weng, Q. (2011). Perpixel vs. object-based classification of urban land cover extraction using high spatial resolution imagery. *Remote Sensing of Environment*, 115(5): 1145-1161.
- NASA Satellite maps provide better urban sprawl insight; NASA News Release, June ,2001.
- Netzband, M., and Jürgens, C. (2010). Urban and suburban areas as a research topic for remote sensing. *Remote Sensing of Urban and Suburban Areas* (pp. 1-9). Springer.
- Oke, T. R. (1987). *Boundary layer climates (2 Edn.)*. New York, Methuen and Co. Ltd., Routledge.
- Pal, M., and Mather, P.M., (2005). Support vector machines for classification in r emote sensing. *International Journal of Remote Sensing*. 26(5): 1007-1011.
- Paul, M.J., and Meyer, J.L. (2001). Streams in the urban landscape. *Annual Review* of Ecology and Systematic. 32: 333–365.
- Pesaresi, M., and Benediktsson, J. A. (2001). A new approach for the morphological segmentation of high-resolution satellite imagery. *IEEE Transaction on Geoscience and Remote Sensing*, 39(2): 309–320.

- Phinn, S., Stanford, M., Scarth, P., Murray, A. T., and Shyy, P. T. (2002). Monitoring the composition of urban environments based on the vegetation-impervious surface-soil (VIS) model by sub-pixel analysis techniques. *International Journal of Remote Sensing*, 23(20): 4131-4153.
- Platt, R. V., and Goetz, A. F. H. (2004). A comparison of AVIRIS and Landsat for land use classification at the urban fringe. *Photogrammetric Engineering* and Remote Sensing, 70(7): 813-819.
- Plaza, A., Benediktsson, J. A., Boardman, J. W., Brazile, J., Bruzzone, L., Camps-Valls, G., Chanussot, J., Fauvel, M., Gamba, P., Gualtieri, A., Marconcini, M., Tilton, J. C., and Trianni, G. (2009). Recent advances in techniques for hyperspectral image processing. *Remote Sensing of Environment*, 113(1): 110-122.
- Poole, G. C., and Berman, C. H. (2001). An ecological perspective on in-stream temperature: natural heat dynamics and mechanisms of human-caused thermal degradation. *Environmental Management*, 27(6): 787-802.
- Rajasekar, U., and Weng, Q. (2009). Urban heat island monitoring and analysis using nonparametric model: a case of Indianapolis. *ISPRS Journal of Remote Sensing*. 64(1): 86–96.
- Ramakrishna, G., Sundararajan, T., and Kothandaraman, S. (2011). Strength of corrugations of roofing sheets reinforced with sisal fibers. *Journal of Engineering and Applied Sciences*, 6(12):24-32.
- Ray, S.S., Singah, J.P., and Panigrahy, S. (2010). Use of hyperspectral data for crop stress detection: ground-based studieds. *International Archives of The Photogrammetry, Remote Sensing And Spatial Information Science*. 38: 562-567.
- Ren, H., and Chang, Y. L, Feature extraction with modified Fisher's linear discriminant analysis, Proceedings in Signal Processing for Standoff Detection of SPIE 5995, Boston, USA, NOV.4, 2005.
- Ridd, M.K. (1995). Exploring a V-I-S (vegetation-impervious surface-soil) model for urban ecosystem analysis through remote-sensing-comparative anatomy for cities.*International Journal of Remote Sensing*. 16(12): 2165-2185.

- Robert-Sainte, P., Gromaire, M. C., Gouvello, B., Saad, M., and Chebbo, G. (2009). Annual metallic flows in roof runoff from different materials: Test-bed scale in Paris Conurbation. *Environmental science and technology*, 43(15): 5612-5618.
- Russ, J. C. (2002). The Image Processing Handbook, Fourth Edition. Boca Raton, FL: CRC Press.
- Sarkar, A., Biswas, M. K., Kartikeyan, B., Kumar, V., Majumdar, K. L., Pal, D. K. (2002). A MRF model based segmentation approach to classification for multispectral imagery. *Transaction on Geosciences and Remote Sensing*, 40(5): 1102-1113.
- Schueler, T. R. (1987). *Controlling Urban Runoff: A Practical Manual for Planning and Designing Urban BMPs.* Paper presented at Metropolitan Washington Council of Governments, Washington, DC.
- Schueler, T. R. (1994). The importance of imperviousness. *Watershed Protection Techniques*, 1(3): 100-111.
- Serra, J. (1982). Image Analysis and Mathematical Morphology, Academic Press, London.
- Shackelford, A. K., and Davis, C. H. (2003b). A combined fuzzy pixel-based and object-based approach for classification of high-resolution multispectral data over urban areas. *IEEE Transactions Geoscience and Remote Sensing*, 41(10): 2354–2363.
- Shackelford, A. K., and Davis, C. H. (2003a). A hierarchical fuzzy classification approach for high-resolution multispectral data over urban areas. *IEEE Transactions on Geoscience and Remote Sensing*, 41(9): 1920-1932.
- Shafri, H.Z.M., Taherzadeh, E., Mansor, S., and Ashurov, A. (2012). Hyperspectral remote sensing of urban areas: an overview of techniques and applications. *Research Journal of Applied Sciences Engineering and Technology*. 4(11): 1557-1565.
- Shariff, A. R. M., Aik, Y. Y., Hong, W. T., Mansor, S., and Mispan, R. Automated Identification And Counting Of Pests In The Paddy Fields Using Image Analysis, Proceeding in 4th Congress, Computers in agriculture and natural resources, Orlando, Florida, USA, 24 July 2006.

- Shaw, G., and Burke, H. H. K. (2003). Spectral imaging for remote sensing. *Lincoln Laboratory journal*, 14(1): 3-28.
- Sheeren, D., Quirin, A., Puissant, A., Gançarski, P., and Weber, C, Discovering Rules with Genetic Algorithms to Classify Urban Remotely Sensed Data, Proceeding Geosciences and Remote Sensing Symposium, Denver, CO, USA, July. 31- Aug. 4, 2006.
- Smith, R.B. (2001). Introduction to hyperspectral imaging.<u>www.microimages.com</u>(Accessed 11/03/2006).
- Soille, P., and Pesaresi, M. (2002). Advances in mathematical morphology applied to geosciences and remote sensing. *IEEE Transactions* on Geoscience and Remote Sensing, 40(9): 2042-2055.
- Szykier, A., (2008). Extraction of roof surface for solar analysis. Maps capital management, retrieved from: <u>http://www.mapscapital.com/schoolpowers/media/pdf/RoofSurfaceExtraction.p</u> <u>df</u>
- Taebi, A., and Droste, R. L. (2004). First flush pollution load of urban stormwater runoff. *Journal of Environmental Engineering and Science*, 3(4): 301-309.
- Tarabalka, Y., Benediktsson, J. A., and Chanussot, J. (2009). Spectral-spatial classification of hyperspectral imagery based on partitional clustering techniques. *IEEE Transaction Geosciences and Remote Sensing*, 47(8): 2973-2987.
- Thomas, N., Hendrix, C., and Congalton, R. G. (2003). A comparison of urban mapping methods using high-resolution digital imagery. *Photogrammetric Engineering and Remote Sensing*, 69(9): 963-972.
- Thomas, P.R., and Greene, G.R. (1993).Rainwater quality from different roof catchments.*Water Science Technology*. 28 (3-5): 291-297.
- Tseng, M. H., Chen, S.J., Hwang, G. H., and Shen, M. Y. (2008). A genetic algorithm rule-based approach for land-cover classification. *Journal of Photogrammetry and Remote Sensing*, 63(2): 202-212.
- *United Nations ,The state of the world cities;* United Nations Centre for Human Settlements: Nairobi, Kenya, 2001.

- US agency for toxic Substances and Disease Registry. Toxicological Profile for Asbestos. September 2001. Retrieved April 10, 2009, from: <u>http://www.atsdr.cdc.gov/toxprofiles/tp61.pdf</u>.
- Vapnik, V. N. (2000). The nature of statistical learning theory. Springer Verlag.
- Wang, L., Dai, Q., Hong, L., and Liu, G. (2012). Adaptive regional feature extraction for very high spatial resolution image classification. *Journal of Applied Remote Sensing*, 6(1): 063506.
- Wang, P., Feng, X., Zhao, S., Xiao, P., and Xu, C, Comparison of object-oriented with pixel-based classification techniques on urban classification using TM and IKONOS imagery, Proceeding SPIE 6752, Geoinformatics: Remotely Sensed Data and Information, Nanjing, China, May. 25, 2007.
- Wang, Y., and Zhang, X. (2004). A SPLIT model for extraction of sub-pixel impervious surface information. *Photogrammetric Engineering and Remote Sensing*, 70(7): 821–828.
- Warwick, J. J., and Tadepalli, P. (1991). Efficacy of SWMM application. *Journal of Water Resources Planning and Management*, 117(3): 352-366.
- Weber, D. N., and Bannerman, R. (2004). Relationships between impervious surfaces within a watershed and measures of reproduction in fathead minnows (Pimephales promelas). *Hydrobiology*, 525(1-3): 215-228.
- Weng, Q. (2001). Modeling urban growth effect on surface runoff with the integration of remote sensing and GIS. *Environmental Management*, 28(6): 737–748.
- Weng, Q., Lu, D. and Schubring, J. (2004). Estimation of land surface temperature vegetation abundance relationship for urban heat island studies. *Remote Sensing of Environment*, 89(4): 467–483.
- Wentz, E. A., Song, Y., Anderson, S., Roy, S. S., Myint, S. W., and Stefanov, W.L. Discriminant Analysis with Spatial Weights for Urban Land Cover Classification (No. 1034).Geoda Center for Geospatial Analysis and Computation.2010.
- Wijaya, A. (2010). Complex land cover classifications and physical properties retrieval of tropical forests using multi-source remote sensing. Doctoral dissertation, Technische Universität Bergakademie Freiberg, Germany.

- Wonorahardjo, S., Koerniawan, M.D., Tedja S., Benedictus, E, The Influence of Building Material to Urban Thermal Environment, A Case Study of Caringin Market, Bandung, Indonesia, Proceedings of The Eight SENVAR, Petra University, Surabaya, Indonesia, 2007.
- Wu, C. and Murray, A. T. (2003). Estimating impervious surface distribution by spectral mixture analysis. *Remote Sensing of Environment*, 84(4): 493-505.
- Wu, C., (2009). Quantifying high-resolution impervious surfaces using spectral mixture analysis. *International Journal of Remote Sensing*, 30(11): 2915–2932.
- Yang, L., Xian, G., Klaver, J. M. and Deal, B. (2003). Urban land-cover change detection through sub-pixel imperviousness mapping using remotely sensed data. *Photogrammetric Engineering and Remote Sensing*, 69(9): 1003-1010.
- Young, R. G., and Huryn, A. D. (1999). Effects of land use on stream metabolism and organic matter turnover. *Ecological Applications*, 9(4): 1359-1376.
- Yu, Q., Gong, P., Clinton, N., Biging, G., Kelly, M., and Schirokauer, D. (2006).Object-based detailed vegetation classification with airborne high spatial resolution remote sensing imagery. *Photogrammetric Engineering* and Remote Sensing, 72(7): 799–811.
- Yuan, F., and Bauer, M. E. (2007).Comparison of impervious surface area and normalized difference vegetation index as indicators of surface urban heat island effects in Landsat imagery. *Remote Sensing of Environment*, 106(3): 375–386.
- Zhang, L., and Huang, X. (2010). Object oriented subspace analysis for airborne hyperspectral remote sensing imagery. *Neurocomputing*, 73(4): 927–936.
- Zhao, G., and Maclean, A. L. (2000). A comparison of canonical discriminant analysis and principal component analysis for spectral transformation. *Photogrammetric Engineering and Remote Sensing*, 66(7): 841-847.
- Zhou, W., and Troy, A. (2008). An object-oriented approach for analysing and characterizing urban landscape at the parcel level. *International Journal of Remote Sensing*, 29 (11): 3119–3135.

Zug, M., Phan, L., Bellefeur, D., and Scrivener, O. (1999). Pollution wash- off modeling on impervious surfaces: calibration, validation, transposition. *Water Science and Technology*, 39(2): 17-24

