

Volume 1

MACHINE VISION INSPECTION SYSTEMS

*Image Processing,
Concepts, Methodologies
and Applications*

Edited by

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Machine Vision Inspection Systems

Volume 1

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Preface

This edited book aims to bring together leading researchers, academic scientists and research scholars to put forward and share their experiences and research results on all aspects of an inspection system for detection analysis for various machine vision applications. It also provides a premier interdisciplinary platform for educators, practitioners and researchers to present and discuss the most recent innovations, trends, methodology, applications and concerns, as well as practical challenges encountered and solutions adopted in the inspection system in terms of image processing and analytics of machine vision for real and industrial application. The book is organized into ten chapters,

[Chapter 1](#) presents an overview of an automated methodology-based learning model classification technique for identifying the usage and coverage of land use in Sri Lanka by using satellite imagery data. This chapter also discusses the issue related to manual surveys and its limitations about the land-use of different regions.

[Chapter 2](#) focuses on the Indian sign language recognition using machine learning algorithm in machine vision and pattern recognition research areas. The work is to translate acquired images or videos either offline or online to corresponding words, numbers or sentences representing the meaning of the input sign. The Direct Pixel Value, Hierarchical Centroid, Local Histogram features of Image Processing techniques are used as a feature in the experimental analysis. The classifiers used here are k-Nearest Neighbour and Neural Network.

[Chapter 3](#) presents an unmanned aerial vehicle (UAV) assist the pest detection model to track pests in the stored grain (SG). This proposed model consists of four phases such as data acquisition, edge detection, feature extraction, and pest identification. In this model, the edge detection (ED) phase is focused on analyzing the data (pest in the SG images). Many standard edge detection (SED) methods such as Sobel, Prewitt, Roberts, Morphological, Laplacian of Gaussian (LoG), Canny etc. are used to track the shape, location, and quantity of pests in SG. The implementation of the methods are performed using MATLAB R2015b and evaluated using signal to noise ratio (SNR), peak signal to noise ratio (PSNR), and processing time (PT).

[Chapter 4](#) describes object selection as a trade-off between performance and accuracy. Particularly, in machine vision time versus precision for object selection plays a crucial role in image analysis is addressed. These regions are a group of segmented pixels that are used for processing. Such regions are often represented by numbers called “object descriptors”. Using such data, the authors compare and distinguish objects by matching the descriptors. Without loss of generality, these descriptors have certain properties like (a) invariance against geometrical transformations like translation, rotation and scaling, (b) Stability to noise and non-rigid local deformation, (c) Completeness.

[Chapter 5](#) explores flood control and disaster management technologies based on image processing and machine learning. The main objective of this chapter is to develop an understanding of the flood risks, explore the existing systems for managing the risks and devise a flood management model through machine vision. Furthermore, this chapter discusses the limitations of the current technologies and suggests a reliable model to overcome the problems. Finally, this chapter elaborates on the system of

how to detect flood-affected areas and determine rescue routes.

[Chapter 6](#) discusses the color changes on the avocado dip under microwave conditions through the machine vision approach. This chapter analyzes the change of color in $a^* - b^*$ space in terms of different treatments. Also, this chapter discusses real-time experimental analysis by various parameters.

[Chapter 7](#) deliberates the defect detection on defective roads and bridges through computer vision-based techniques. This chapter discusses the basic steps involved in defect detection using image processing along with existing systems and presents the pros and cons of the different existing methods in terms of performance. Also, this chapter applies multiple image processing techniques to solve the various types of defects.

[Chapter 8](#) presents the study and conducts experiments through machine vision techniques on diabetic retinopathy disease present in retinal fundus images. This chapter also discusses various factors of the disease that appears in the image and discusses the possible solutions in terms of image processing techniques. An effective analysis is shown for computer-aided solutions.

[Chapter 9](#) provides a robust method to solve the ambiguities in handwritten the OCR system. This has been resolved using the Convolutional Neural Network (CNN) based approach. This state-of-the-art of CNN-based approach for recognition of multiple handwritten numerals of various scripts is clearly shown here. It is also quite helpful to report the discriminate features of each individual and later lead to reporting a high recognition rate. At the simulation level we have listed the variance nature of the individual's images and through CNN we have reported a high recognition rate, which is quite

helpful in building the automatic recognition system for handwritten numerals to have the solution for real-time problems.

[Chapter 10](#) presents a detailed review of some of the attempts towards avoiding, detecting and preventing phishing in terms of visual methods. This chapter explains the frauds and criminal activities on phishing and, moreover, discusses the various solutions approached in recent years. Additionally, this chapter reviews the role of training and education on the reduction of phishing victims.

We have to start by thanking God Almighty for giving us the ability and opportunity to undertake to edit this book and to complete it satisfactorily. Completion of this book could not have been accomplished without the support of all editors starting from the “Call for Chapters” till their finalization. All the contributors have given their contributions amicably and is a positive sign of significant teamwork. The editors are sincerely thankful to all the members of Scrivener Publishing especially Martin Scrivener for providing constructive inputs and allowing an opportunity to edit this important book. We are equally thankful to all reviewers who hail from different places in and around the globe shared their support and stand firm towards quality chapter submissions. Finally, we are eternally grateful to the authors for contributing quality chapters.

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Land-Use Classification with Integrated Data

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Abstract

The identification of the usage and coverage of the land is a major part of regional development. Crowdsourced geographic information systems provide valuable information about the land use of different regions. Although these data sources lack reliability and possess some limitations, they are useful in deriving building blocks for the usage of the land, where the manual surveys are not up-to-date, costly, and time consuming. At present, in the context of Sri Lanka, there is a lack of reliable and updated land-use data. Moreover, there is a rapid growth in the construction industry, resulting in frequent changes in land-use and land-cover data. This paper presents a novel and an automated methodology based on learning models for identifying the usage and coverage of the land. The satellite imagery is used to identify the information related to land cover. They are integrated with Foursquare venue data, which is a popular crowdsourced geographic information, thus, enhancing the information level and the quality of land-use visualization. The proposed methodology has shown a kappa coefficient of 74.03%, showing an average land-use classification accuracy within a constrained environment.

Keywords: Geographic information system, land-cover identification, land-use classification, social computing, decision support system, satellite images, Foursquare data

1.1 Introduction

Regional planning and management are major concerns in the development strategy of a country. The information related to the coverage and usage of lands can be used to extract the features in an area and facilitate development activities. The land-use data are related to human activities, whereas the land-cover information represent the natural features and artificial constructions on the earth surface. Crowdsourced geographic information systems provide valuable information about the land use of different regions. At present, up-to-date data on land usage and coverage are not available for all the cities in Sri Lanka. This is due to the cost of labor, lack of the required technologies, and resources associated with the data surveys. Unavailability of a cost-effective way of obtaining such latest and reliable data is a bottleneck to the long-term planning and development of a region. This results in unplanned *ad hoc* developments, construction of unhealthy residential areas, deterioration of service and infrastructure, environmental pollution, increased traffic congestion, and so on [1], which can be widely seen in many urban areas in Sri Lanka. Therefore, up-to-date data on the usage and coverage of land are important to make strategic decisions on sustainable region planning.

The objective of this research is to design and develop a support system to classify the land-cover and land-use data using Google Satellite imagery [2] and Foursquare data, which is a type of volunteer geographic information (VGI), respectively [3]. The system produces a visualization of different types of land-use in each area (eg. residential, industrial, commercial, agriculture etc.) on a land-use map based on heterogeneous data sources including crowdsourced Foursquare data. Acquiring data on land cover and land use from different data types, which can be integrated into the classification system, will enhance the quality of the processed information [4].

Therefore, this research provides a novel way of identifying and classifying different forms of land-use data, specifically satellite imagery and Foursquare data, with the extensible features for other types of related data. The system refines the land-use mapping with the use of additional parameters, such as context-specific different data sources. Ultimately, the retrieved data can be used to monitor land-use changes in near real time [2].

Moreover, this study focuses on developing a common platform that enables the collaboration of heterogeneous data sources to produce enhanced land-use data. Further, this will increase the utility value of the retrieved information on land-cover and land-use, hence, widening the range of applicable applications from the results. Colombo district is selected as the study area considering the availability and sampling rates of different data sets and issues associated with data validation [4]. The proposed land-use visualization approach identifies and classifies different forms of land cover and land use in a selected area considering the satellite imagery and Foursquare data, respectively, and displays the classification on a land-use map.

The land-use data retrieved from the proposed methodology can be used to monitor land-use changes near real time. Analysis of these detailed snapshots of land-use enables authorities to detect a change and foresee its social and environmental consequences. This, in turn, will enable them to identify long-lasting sustainable solutions to urbanization issues in Sri Lanka.

The paper is structured as follows: Section II explores the related literature and Section III presents the design architectures of the system and Section IV describes the development process with the used techniques. Section V evaluates the validity of the proposed method, and finally, Section VI summarizes the research including future work.

1.2 Background Study

1.2.1 Overview of Land-Use and Land-Cover Information

The identification of the usage and coverage of the land is a major part of regional development. Land cover and land use are often interchangeably used in many information systems, despite the distinct difference between those two terms [1, 4]. Land cover refers to observable physical properties of land, such as the areas with trees, grounds, building, roads, water, and so on. On the other hand, land use refers to purposes for which lands are being used, such as residential, commercial, entertainment, and so on. It may be difficult to determine the actual purpose for which land is being used by solely using the information produced by a source of observation. For example, in the absence of additional information sources, it may be difficult to decide whether grasslands are used for agricultural activities or not. Moreover, there is a rapid growth in the construction industry, resulting in frequent changes in land-use and land-cover data.

As a summary, the land coverage and usage data are important to identify correctly and process timely manner in order to make decisions on regional development. However, it is challenging to obtain large-scale, latest data from reliable sources. The unplanned constructions may impact the region with unprogressive infrastructure, unhealthy residentials, and environment issues, such as traffic congestion and pollution.

1.2.2 Geographical Information Systems

A Geographical Information Systems (GIS) facilitates collection, store, process, analyze, and visualize data on the surface of the earth. Prior to the discovery of advanced technologies and concepts, a GIS primarily consisted of a database management system (DBMS) and a set of tools that allowed data to be retrieved from the DBMS. With the advent of the Internet and Web-based applications, as well as the increased utilization of mobile devices, the traditional notion of a GIS has been altered significantly. Particularly,

the user base of GISs has expanded from traditional users, such as relevant public and private sector to just about anyone who uses an application built on top of a GIS in their mobile or other electronic devices.

Even though the primary functions, such as data storage, retrieval, visualization, and so on, are common to all GISs, the nature of these functions depends largely on the underlying application. Based on the area of application, GIS can be classified into different information system types, such as cadastral, land and property, image-based, natural resource management, spatial information systems, and so on.

The GIS can be used to find solutions to a wide variety of problems, such as determining and predicting the features of a particular location, identifying locations with particular characteristics, monitoring change over periods, determining the best route between different locations, modeling environmental and societal evolution over time, crime location mapping, disaster management, transportation planning, management, and so on.

1.2.3 GIS-Related Data Types

1.2.3.1 Point Data Sets

Point data are used to represent discrete data points that have no adjacent features. For example, in a GIS, the location of a place can be represented by a point location. The GIS data are of two categories, the spatially referenced data and the attribute data table associated with it. The spatially referenced data are represented in vector and raster forms. Foursquare data [3, 5] and OpenStreetMap data [6] are two popular GIS point data sources that can be utilized for the identification of land-use utilization at a detailed level.

Foursquare [3, 5] is a mobile-based social networking service that facilitates to acquire user-location information using check-in and search history; and recommend places near the user's current location. These data contain the names, locations, and types of places. When providing recommendations, the application makes use of its citizen-contributed database, the places a user goes, and the advice and comments given by other users on a place. This is a volunteered/crowdsourced VGI, that uses geographic data provided by the users. Thus, Foursquare data are used to identify land-use information with proper quality validation and consumes a low-cost mechanism.

OpenStreetMap [6] is a popular GIS data source that provides geographic data. This presents the physical features within a given area, such as commercial buildings, roads, waterways, railways, highways, and amenities using tags, and each of those tags describes a geographical attribute. Moreover, OpenStreetMap releases map data under a free and open license, which makes them available for researchers and as a data validation source.

1.2.3.2 Aerial Data Sets

An aerial data set is prepared by means of airborne methods and aerial photographs and Google satellite imagery some examples [2]. Satellite images of the Earth's surface captured by remote-sensing techniques have proven to be a useful data source for many research studies and applications in diverse fields [2, 7]. Satellite images enable wider geographical areas to be covered quickly with relative convenience. They provide a way of monitoring the Earth's surface, hence, eliminating the need to rely solely on labor-intensive processes, such as area frame surveys for maintaining up-to-date information on the Earth's surface, which tends to be a slow and a tedious process. Moreover, the digital format of satellite images enables to be directly processed digitally and integrated with other data sources with relative ease. Data retrieved from satellite images are used to

successfully extract the required information on the land usage for decision making and predicting systems.

High elevation aerial photographs of the Earth's surface are a remote-sensing data source. They can be used to identify the coverage of lands. These visual interpretations are also used in conjunction with satellite images, particularly to fill areas which are not clear on satellite images due to prevalent atmospheric conditions, such as cloud cover at the time of capturing those. Aerial photographs are of high spatial resolution. Therefore, these images comprise a high level of information which is useful for various analytical and reference purposes associated with land-cover classifications. The spectral range of aerial photography is narrow, and therefore, a considerable number of photographs will have to be patched together to cover wider geographical areas. Further, the cost per surface unit is higher with aerial photographs compared with satellite images.

1.2.4 Related Studies

Among the variety of research studies, the study by Quercia and Saez [5], has described the feasibility of acquiring data from the locations of social-media users, considering the mining urban deprivation information in London city. This study has addressed the reliable, latest and inferring free data retrieval, and shown the use of Foursquare data in monitoring fine-grain resolutions, which cannot be obtained from other land-use data used in practice. However, this gives an insight into the limitations of Foursquare data such as demographic and geographic biases and the Foursquare categories not being fully structured. For instance, these can be biased by the penetration rate of the Foursquare data and a given location can be named with different category types based on the thinking patterns of the users who entered the Foursquare data.

A multidata approach (MDA) is presented by Bareth [4] to generate data related to land use. They have combined the traditional types of data for land usage with the information obtained from the different remote-sensing analysis. As a first step, remote sensing data were classified using supervised classification techniques and then the quality of the classified data was assessed. The second step of the research was to import the classified data into a GIS environment and overlay them with relevant 3rd-party data sources, such as official topographic or land-use data. The importance of this approach is that it enables useful and high-quality land-use information in various spatial databases, such as spatial biotope or biodiversity databases, spatial databases of national parks, and so on to be integrated with the results of remote-sensing analysis. Further, land-use data retrieved from official data sources can be integrated to MDA for cross-checking the results of remote-sensing analysis. Also, by incorporating the results of land-cover change models to MDA, they have simulated the change scenarios for the usage and coverage of the land.

According to the literature, satellite images are mainly used to identify urban land usage that covers a large area without frequent changes. However, it is challenging to obtain high-resolution images. Thus, there is a research direction toward the integration of these remote-sensing data with the social knowledge data for better analysis of the usage and coverage patterns of the land.

1.3 System Design

The primary focus of this study is to classify land usage into several categories, such as residential, hotels, restaurants, industrial, and so on. Two data sources google satellite images for the coverage and foursquare point data for usage of the land are used for this study. First, the google satellite imagery is used to classify land-cover data that represent primitive categories, such as water, built-up, vegetation, and bare soil. Next, this output would be further differentiated using foursquare point data into the abovementioned land usage classes. [Figure 1.1](#) shows the overall view of the proposed system.

System design mainly consists of four components. The data preprocessing component removes and purifies noises in the input data. The data classification component applies classification and clustering techniques on the preprocessed data. These results of heterogeneous data will be integrated into the data integration component. Then, data analysis component will evaluate the accuracy of the classified data. The main output of the system is a Shape file that adds a layer to Quantum Geographic Information System (QGIS). The QGIS is a free desktop application that supports different platforms. It has features to view, edit, and analyze geospatial data together with quality measures. The final system is expected to be used for urban planning and government officials who want to collect data, and so on.

[Figure 1.2](#) shows the sequence diagram of the proposed Web application. It shows how such a user can easily retrieve the results produced by the system so that he can use this information in his respective work, such as sustainable urban development planning. First, the user will access a Web interface where he is provided with a google map. He can zoom in the map and select a rectangular area using the mouse pointer. The analyst can highlight the area from the map, which he wants the land-use classification. The common interface will send the boundaries (latitudes and longitudes) of the area to the land-cover classification module, which will return the stitched image. Next, the common interface will send the land-cover classified map along with the boundaries to the land usage classification module. This module will get the Foursquare locations and their respective classes from the Foursquare database, perform the land-use classification, and return the final land usage classified map to the analyst. In addition, a legend is also provided with the different categories of land usage classes, such as residential, restaurants, hotels, and so on, and their assigned colors will also appear with the final map.

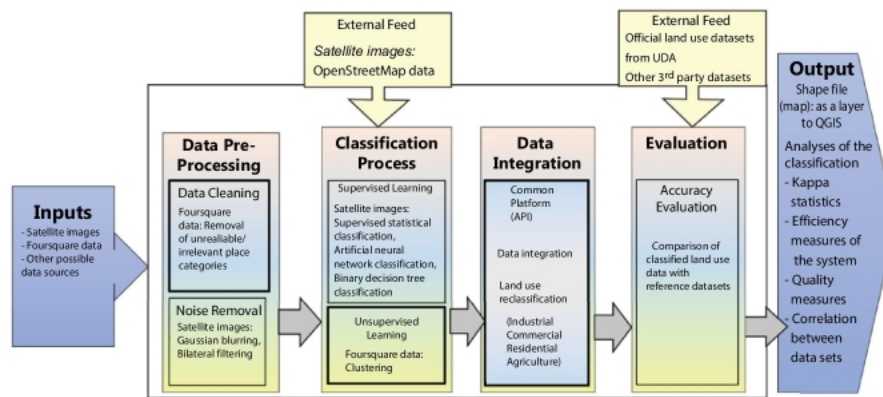


Figure 1.1 High-level design diagram.

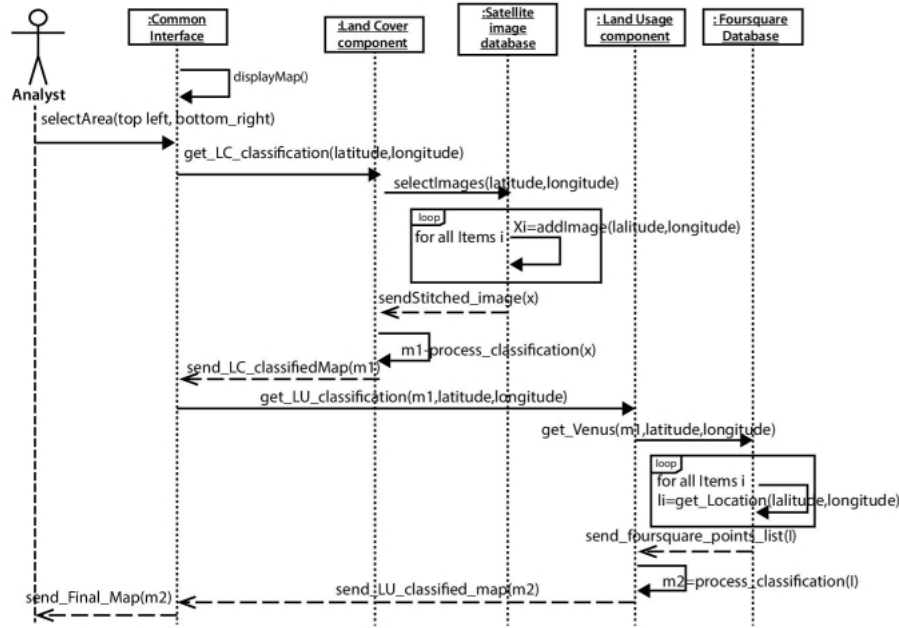


Figure 1.2 Process view of the land-use visualization Web application.

[Figure 1.3](#) shows the architectural view of the proposed system, which complies with the overall view, and the components are described with the workflow given in [Figure 1.4](#). The module “Input Data Manager” handles the input data types to the system. The current system uses satellite images, Foursquare data, and official land-use data. The architecture is designed in an extensible feature to incorporate other data types, such as OpenStreetMap. The module “Visualization Manager” visualizes the processed Foursquare data and point location data in terms of land-use classification with a color-coded layer based on the predefined categories and described in detail in the chapter. The logical layer comprises three main modules responsible to preprocess data, integrate data sources, and the learning process.