

# A Study on Image Change Detection Methods for Multiple Images of the Same Scene Acquired by a Mobile Camera

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### **NOMENCLATURE**

#### **Acronyms**

**CCTVs Close-Circuit Televisions** 

IR Infra-Red

GPS Global Position System

VMD Video Motion Detection

3DOF Three Degrees of Freedom

PZT Pan-Zoom-Tilt

FFT Fast Fourier Transform
MP Morphological Pyramid

GA Genetic Algorithm

RGB Red, Green and Blue

HSV Hue, Saturation and Value

AIR Automatic Image Registration

**UCHR** Unimportant Changes Removal

**RoiD** Region of Interest of the Subtracted Image

SIFT Scale Invariant Feature Transform

IdII Indoor Input Image

NCC Normalized Cross-Correlation

TP True PositiveFP False PositiveFN False Negative

TPR True Positive Rate
FNR False Negative Rate

**CPs** Control Points

Template RI Human-Made Template

AIO1\_RI Area of Interest of Reference Image

**AOI1 II** Area of Interest of Input Image

**3DOF** Three Degrees of Freedom

II Input Image for Outdoor Experiment

ROI2\_RI New Region of Interest of the AOI1\_RI

ROI2\_II New Region of Interest of the RAOI1\_II

RAOI1\_II Registered AOI1\_II

ZKA The Zitnick and Kanade Algorithm

OMI Occlusion Map Image

CMI Confidence Map Image

OMI II Occlusion Map Image of Input Image

CMI\_II Confidence Map Image of Input Image

MSS Landsat Multispectral Scanner

TM Landsat Thematic Mapper

MR Magnetic Resonance

PET Positron Emission Tomography

HDS1\_IIs First Hybrid Decision-Making System Input Images

FIS1 First Fuzzy Inference System

CHM1 II Changed Mask Generated by the First Algorithm

E II-1(i, j) The Edged Image of the ROI2 II-1

BW\_II-1(x, y) The Binary Image of the E\_II-1(i, j)

HDS2 Second Hybrid Decision-Making System

FIS2 Second Fuzzy Inference System

CHM2\_II Changed Mask Produced by the Second Algorithm

#### **ABSTRACT**

Detecting regions of change while reducing unimportant changes in multiple outdoor images of the same scene containing fence wires (i.e., a chain-link mesh fence) acquired by a mobile camera from slightly different viewing positions, angles and at different times is a very difficult problem. Regions of change include appearing of new objects and/or disappearing of old objects behind fence wires, breaches in the integrity of fence wires and attached objects in front of fence wires. Unimportant changes are mainly caused by camera movement, considerable background clutter, illumination variation, tiny sizes of fence wires and non-uniform illumination that occurs across fence wires. There are several issues that arise from these kinds of multiple outdoor images. The issues are: (1) parallax (the apparent displacement of an object as seen from two different positions that are not on a line with the object) among objects in the scene, (2) changing in size of same objects as a result of camera movement in forward or backward direction, (3) background clutter of outdoor scenes, (4) thinness of fence wires and (5) significant illumination variation that occurs in outdoor scenes and across fence wires. In this dissertation, an automated change detection method is proposed for these kinds of multiple outdoor images.

The change detection method is composed of two distinct modules, which are a module for detecting object presence and/or absence behind fence wires and another module for detecting breaches in the integrity of fence wires and/or attached objects in front of fence wires. The first module consist of five main steps: (1) automated image registration, (2) confidence map image production by the Zitnick and Kanade algorithm, (3) occlusion map image generation, (4) significant or unimportant changes decision by the first hybrid decision-making system and (5) false positives reduction by the template subtraction approach. The second module integrates: (1) the Sobel edge detector combined with an adaptive thresholding technique in extracting edges of fence wires, (2) an area-based measuring in separating small and

big objects based on their average areas determined once in the calibration process and (3) the second hybrid decision-making system in classifying objects as significant or unimportant changes.

Experimental results demonstrate that the change detection method can identify and indicate approximate locations and possible percentages of significant changes whilst reducing unimportant changes in these kinds of multiple outdoor images. The study has utilized occluded regions in a confidence map image produced by the Zitnick and Kanade algorithm as potential significant changes in the image change detection research. Moreover, the study proves that the use of the Sobel edge detector combined with an adaptive thresholding technique is applicable in extracting edges of outdoor fence wires. In the future, the method could be integrated into patrol robots in order to provide assistance to human guards in protecting outdoor perimeter security.

### STATEMENT OF ORIGINALITY

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Guntur Tanjung			
Date	•	•	·

### **PUBLICATIONS**

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