



UNIVERSITI PUTRA MALAYSIA

***LIVENESS DETECTION IN FACIAL BIOMETRICS USING COMPLETE
DYNAMIC LOCAL TERNARY PATTERN TECHNIQUE***

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**LIVENESS DETECTION IN FACIAL BIOMETRICS USING COMPLETE
DYNAMIC LOCAL TERNARY PATTERN TECHNIQUE**

By

SAJIDA PARVEEN

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

September 2016



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DEDICATION

To my encouraging parents, my siblings

My beloved husband, my children and all family members



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

LIVENESS DETECTION IN FACIAL BIOMETRICS USING COMPLETE DYNAMIC LOCAL TERNARY PATTERN TECHNIQUE

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September 2016

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Facial biometric systems have recently received increased deployment in various applications such as surveillance, access control and forensic investigations. However, facial biometrics facing various tangible threats, one of them is spoofing attacks. A spoofing attack occurs when a person tries to masquerade as someone else by non-real faces such as photograph, video clips or dummy faces and thereby gaining advantages from applications. In order to identify the spoofing attacks on such biometric systems, face liveness detection countermeasure have been developed.

There are numerous ways to detect the liveness of face such as through motion analysis, texture analysis, identify life sign clues and thermal sensors. Recently, texture analysis has received more attention because of its non-intrusiveness; high efficiency and accuracy to discriminate face skin texture from spoof attacks. For this purpose, a numbers of texture descriptors have been proposed in the literature for face liveness detection. However, they exhibit some limitations in terms of noise with center pixel, manual setting of threshold (τ) value and ignorance of global intensity in the image.

Thus, a robust face liveness detection method based on Complete Dynamic Local Ternary Pattern (CDLTP) has been proposed in this thesis. The CDLTP was designed to overcome the limitations of reported texture descriptors. Weber's law was used to automatically set the threshold value in ternary pattern by considering the *sign*, *magnitude* and global *intensity* of the image. Its effectiveness has been tested and benchmarked against other existing texture descriptors (i.e. Local Binary Pattern (LBP), Local Ternary Pattern (LTP), Dynamic Local Ternary Pattern (DLTP), Complete Local Binary Pattern

(CLBP) and Complete Local Ternary Pattern (CLTP)) on self collected database and other publically available databases (i.e. NUAA, CASIA and REPLAY-ATTACK).

The evaluations have been carried out via statistical hypothesis testing and through liveness detection itself. The results have consistently demonstrated that CDLTP outperforms other techniques across various types of spoof mediums. The comparison analysis of CDLTP with other texture descriptors were also carried out on self collected and public domain face spoof databases. In all these experiments, the results obtained with CDLTP exceeds from the state-of-art.

Various score level fusion strategies have been adopted to evaluate the performances of the overall system which comprises both face liveness detection and face recognition systems. The achieved decisions from scores level fusion strategies proved that CDLTP based face liveness detection reduces 89% of vulnerability of face verification system against spoof attacks. The measured result of serial method in which the face liveness detection performed before face recognition system was found to be the most effective methods among other score level fusion strategies that were analyzed.

Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia
sebagai memenuhi keperluan untuk ijazah Doktor Falsafah

**PENGESANAN RUPA MUKA YANG HIDUP UNTUK PENGESAHAN RUPA
MUKA DENGAN MENGGUNAKAN CORAK TEMPATAN KETIGA YANG
MENYELURUH DAN DINAMIK**

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Sistem biometrik berdasarkan rupa muka kini telah banyak digunakan di dalam pelbagai aplikasi seperti pengawasan, kawalan masuk dan penyiasatan forensik. Namun sistem biometrik berdasarkan rupa muka meghadapi pelbagai ancaman ketara, salah satu daripadanya ialah serangan penyamaran. Serangan penyamaran berlaku apabila seseorang cuba menyamar sebagai orang lain dengan menggunakan rupa muka seperti gambar, klip video atau muka tiruan dan sekali gus mendapat kelebihan daripada aplikasi tersebut. Untuk mengenalpasti serangan penyamaran pada sistem biometrik itu, pengesanan rupa muka yang hidup telah dibangunkan.

Terdapat pelbagai cara untuk mengesan rupa muka yang hidup seperti melalui analisis pergerakan, analisis tekstur, mengenalpasti petunjuk tanda hidup dan pengesan haba. Terbaru, analisis tekstur telah mendapat perhatian yang lebih kerana ia bebas dari sifat mengganggu; mempunyai kecekapan dan ketepatan yang tinggi untuk membezakan tekstur kulit muka daripada serangan penyamaran. Bagi tujuan ini, sejumlah penerang tekstur telah dicadangkan di dalam kajian untuk mengesan rupa muka yang hidup. Walau bagaimanapun, teknik tersebut mempunyai beberapa kekurangan dari segi gangguan terhadap piksel pusat, menetapkan nilai ambang (τ) secara manual dan ketidakpekaan terhadap intensiti global di dalam imej.

Oleh itu, kaedah pengesanan rupa muka yang hidup berasaskan pada corak ketiga tempatan yang lengkap dan dinamik (CDLTP) telah dicadangkan dalam tesis ini. CDLTP direka untuk mengatasi kekangan penerang tekstur yang dilaporkan terdahulu. Hukum Weber telah digunakan untuk menetapkan

secara automatik nilai ambang dalam corak ketiga dengan mengambil kira tanda, magnitud dan keamatan global di dalam imej. Keberkesannya telah diuji dan telah dibandingkan dengan penerang tekstur terdahulu yang sedia ada (iaitu corak perduaan tempatan (LBP), corak ketiga tempatan (LTP), corak ketiga tempatan yang dinamik (DLTP), corak perduaan tempatan yang lengkap (CLBP) dan corak ketiga tempatan yang lengkap (CLTP)) dengan menggunakan pangkalan data yang dikumpulkan sendiri serta pangkalan data lain yang sedia ada (iaitu NUAA, CASIA dan REPLAY-ATTACK).

Penilaian yang telah dijalankan melalui pengujian hipotesis statistik dan melalui pengesanan rupa muka yang hidup itu sendiri. Keputusan secara konsisten telah menunjukkan bahawa prestasi CDLTP telah melebihi teknik-teknik lain merentasi pelbagai jenis media penyamaran. Analisis perbandingan CDLTP dengan penerang tekstur lain juga telah dijalankan pada pangkalan data yang dikumpul sendiri dan pangkalan data awam yang lain. Dalam semua ujikaji ini, keputusan yang diperolehi menunjukkan bahawa CDLTP melebihi dari teknik-teknik yang lain.

Pelbagai strategi gabungan skor telah diguna pakai untuk menilai prestasi keseluruhan sistem yang terdiri daripada kedua-dua sistem pengesanan rupa muka yang hidup dan dan sistem pengesanan rupa muka. Keputusan yang dicapai dari skor gabungan strategi telah membuktikan bahawa sistem pengesanan rupa muka yang hidup berasaskan CDLTP telah berjaya mengurangkan 89% daripada kelemahan sistem pengesahan rupa muka terhadap serangan penyamaran. Keputusan yang diukur daripada kaedah bersiri di mana sistem pengesanan rupa muka yang hidup dilakukan sebelum sistem pengesanan muka didapati merupakan kaedah yang paling berkesan jika dibandingkan dengan strategi gabungan skor yang telah dianalisis.

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I certify that a Thesis Examination Committee has met on 28 September 2016 to conduct the final examination of Sajida Parveen on her thesis entitled "Liveness Detection in Facial Biometrics using Complete Dynamic Local Ternary Pattern Technique" 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.

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LIST OF ABBREVIATIONS

LBP	Local Binary Pattern
LTP	Local Ternary Pattern
DLTP	Dynamic Local Ternary Pattern
CLBP	Complete Local Binary Pattern
CLTP	Complete Local Ternary Pattern
CDLTP	Complete Dynamic Local Ternary Pattern
ILTP	Improved Local Ternary Pattern
ELTP	Enhanced Local Ternary Pattern
PDF	Probability Density Function
2-D	Two Dimensions
ROC	Receiver Operating Characteristic
S_CDLTP	Sign- Complete Dynamic Local Ternary Pattern
M_CDLTP	Magnitude- Complete Dynamic Local Ternary Pattern
C_CDLTP	Center- Complete Dynamic Local Ternary Pattern
WLD	Weber Local Descriptor
LPQ	Local Phase Quantization
BSIF	Binarized Statistical Image Features
LCPD	Local Contrast-Phase Descriptor

CHAPTER 1

INTRODUCTION

1.1 Background

Biometrics refers to technologies designed either to verify or recognize the identity of a person based on one or more physical or behavioral characteristics (Rejman-Greene, 2005) such as faces, fingerprints, irises, voices and gates. The biometric technologies are widely used in different areas ranging from governmental to private sectors and from personal computers to commercial purposes. Biometrics makes the system more secure, private and convenient to use (Jain and Ross, 2007). Nevertheless, similar to any other system, biometric is also vulnerable to malicious attacks particularly spoofing attacks.

A spoof attack is defined as an attempt by a person to masquerade as an authentic biometric user by using his reproduced biometric traits for gaining illegitimate access to the biometric system (Edrognus and Marcel, 2014). This can be done through an artificial finger, a contact lens on an eye or a mask over a face because our faces are visible, voices can be recordable, and fingerprints are left on mostly everything a person touches.

There are a numbers of different computer companies currently operating in the market that provide embedded face biometric solution with built-in webcams that authenticate users by scanning their faces such as Lenovo, Dell, Asus, Apple and Toshiba. However, spoofing attacks (or copy attacks) are still serious threats to these solutions despite high performances for biometric accuracy for verification and identification.

Face spoof attacks are very cheap and easy to capture and reproduce as compared to the other biometrics. Particularly with the advancement of digital camera devices and availability of social media, facial images can be captured from far distance or can be downloaded from social websites. The reproduction of fake facial specimen is also a simple task whereby faces can easily be shown on different mediums such as on printed photos, or playback images / videos on portable electronic devices.

Face biometrics based systems are very vulnerable in the real world and may encounter various spoof attacks. The trust of users in biometric based applications can be increased by securing the system with an additional layer of liveness detection which enables the system to differentiate between live and spoof face specimen.

Face liveness detection is an effective countermeasure to face spoofing attacks. In the Face liveness detection system as shown in the Figure 1.1. According to that the user needs to show his or her face in front of the system's camera. The camera captures a facial image and passes that image to the feature extraction module. The result obtained from this stage is used to differentiate between original facial samples from fake samples. Hence, original samples will further be treated in order to get verified and fake one are to be rejected automatically by the system.

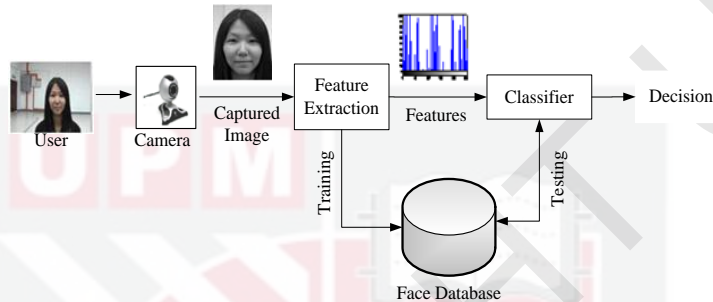


Figure 1.1: Block diagram of face liveness detection system

The appearance of a fake face can be identified through several clues including detection of several life signs such as motions (i.e. eye blink, head rotation, lip movement etc) and comparison of textures of skin. The former requires a user to be cooperative with the system for producing spontaneous facial movement. Thus, such type of systems become powerless when someone spoofs with video attacks or if the user could not match the perfect desired movement because of health issues (i.e. which lead to a high level of false positive in spoofing detection).

Motion based methods does not get much attention in the research field because of its intrusiveness (Anjos et al. 2014). This limit of work, leads toward the cause of dependency on user's movement for liveness detecting such as head and mouth movement. While, the non-intrusive based face liveness detection systems are getting much more attention and a number of methods have been proposed for such purposes.

The second approach is often based on analysis of facial appearance properties, such as texture and reflectance (Anjos et al. 2014). Most of the methods in this category differentiate the live face from the spoof by using only one image, which make them cheaper in processing. The main advantage over motion based attack is this; no user cooperation is needed in analysis of facial appearance properties.

Most recently, texture analysis for face liveness detection has become widely used because of the tremendous advantages such as simple implementation, lower complexity in calculation and non-intrusiveness in terms of user involvement. Texture analysis mainly depends on the texture feature extraction and classifiers. For calculating the texture features, there are a number of texture descriptors which are used in the feature extraction module. A common descriptor includes Local Binary Pattern (Ojala et al. 1994) and its various extended versions.

1.2 Problem Statement

- **Current limitation of existing databases**

Spoof specimens can breach the security of face verification systems in a number of ways by using various possible attacks such as mask, printed photograph, 3D model or video display device and etc. Currently, in publicly available databases such as NUAA face imposter database (Tan et al. 2010), REPLAY-ATTACK and CASIA face anti-spoofing database (Zhang et al. 2012) the available face spoof attacks are limited in variations and still rather limited as compared to the other possible measures that can be used for spoofing attempts. In order to conduct a more robust research in face liveness detection, a challenging database is required to collect that should cover more types of display media and paper material for face spoof attacks for the texture based approaches.

- **Limitation in commonly used texture descriptors**

Ideally for texture analysis, the common texture descriptor such as Local Binary Pattern (LBP) (Ojala et al. 1994) should be powerful enough to provide reliable features to classify a particular texture; which is in this case, to identify live skin texture. Mostly, in LBP and its all other extended descriptors utilized two values i.e. central and neighbor pixel in each pattern for calculating local binary difference. In these texture descriptors, the limitation in terms of noise with central pixel is appeared in many cases, which are observed in similar micro-structures showed in Figure 1.2. The solution of this limitation is proposed in Local Ternary Pattern (Tan and Triggs, 2010), which is basically robust to the central pixel noise because of its three level quantization. Later on, this technique also encounters the limitation in terms of setting threshold value manually for every application.

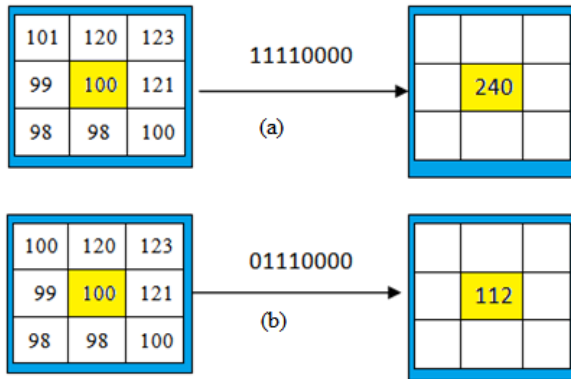


Figure 1.2: Similar LBP patch coding limitation

To overcome this problem, Improved Local Ternary Pattern (ILTP) (Yang and Sun, 2011), Dynamic Local Ternary Pattern (DLTP) (Ibrahim et al. 2014) and Extended Local Ternary Pattern (ELTP) (Zhenyu et al. 2014) have been proposed. These all texture descriptors also poses another limitation which is the sharing of the same code but actually both have different gray value distribution as shown in Figure 1.3 that $LBP_{8,1} = 219$, in positive LTP =145 and in negative LTP =32. The reason is only to concentrate on local features where as the global feature of the image is neglected in these texture descriptors.

Thus for calculating global features, Complete Local Binary Pattern (CLBP) (Guo et al. 2010) and Complete Local Ternary Pattern (CLTP) (Rassem and Khoo, 2014) were introduced. Such limitations on current texture descriptors i.e. LBP, LTP, DLTP etc degraded the performances, thus there is a need to design a more robust descriptor.

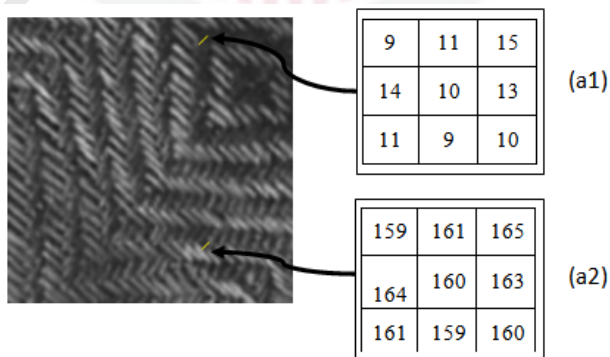


Figure 1.3: Micro structure of different gray values with similar code (a1) light (a2) dark regions

- **Study on the statistical differentiating properties of texture descriptors across various spoof techniques is not reported**

Study on the effectiveness of the face descriptor across various spoof medium has been little reported. Most treat them as generic. Hence, there is a need to analyze the effects of various spoof attacks on face liveness detection in order to define the level of attack affected according to the different spoof medium.

- **Study on the effect of various decision fusion techniques from liveness detection and face verification modules is rarely reported**

Face liveness detection does not suppose to serve as a standalone system. The combined operation of face verification and face liveness detection is rarely reported and limited fusion strategies are being evaluated (Wild et al. 2016). There is need to incorporate the face descriptor in face liveness detection and investigate the effectiveness with different score fusion method with face verification.

1.3 Research Objectives

The primary aim of this work is to propose a robust non-intrusive face liveness detection method based on skin texture analysis. In order to achieve this aim, following objectives have been set:

1. To design a robust face texture descriptor and analyze its statistical discriminating properties across various types of spoof attacks
2. To design a non-intrusive face liveness detection method based on proposed feature descriptor
3. To improve the effectiveness of score fusion techniques between face liveness detection and face verification modules.

1.4 Scope of Research Work

The scope of this research work is to design a face liveness detection method for facial biometrics. The proposed method is based on texture analysis of static images. The performance of proposed method is to be analyzed in terms of accuracy of the system. In order to evaluate the performance of a face liveness detection module with joint operation of face verification by using proposed texture descriptor, the serial and parallel score fusion strategies are employed. Three publicly available facial spoof databases e.g. NUAA face imposter database (Tan et al. 2010), REPLAY-ATTACK and CASIA (Zhang et al. 2012) are utilized for benchmarking.

1.5 Organization of Thesis

This thesis is organized into five chapters. The summary of each chapter is given below:

Chapter 1 provides a general introduction to the research area and identifies the current problems in designing of texture descriptor for face liveness detection system that motivated this research. It also introduces the objectives and scope of research as well.

Chapter 2 presents a thorough literature review and theoretical background about face spoofing attacks and face liveness detection schemes. It also covers currently used face spoof databases along with the limitations. The background and history of texture analysis descriptor and its usage in face liveness detection is also provided in this chapter. The recent face verification techniques are also discussed in this chapter.

Chapter 3 provides database collection and compilation of face spoof database and face verification database. It also describes the design aspects of Complete Dynamic Local Ternary Pattern descriptor. The models for decision fusion strategies of face liveness detection and face verification systems are also elaborated in this chapter.

The testing results of proposed mathematical model on collected face spoof database with different texture analysis of spoof attacks, face liveness detection and score fusion model of embedded system are presented in chapter 4. Finally the conclusion is provided in chapter 5.

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