



Effectiveness of a Biometric Patient Identification System

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BSc in ICT (Hons.) (Software Development) with the University of Malta

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Scientific Areas of Interest:

e-Health

Internet of Things

Knowledge management for people with disabilities

Web and Mobile Development

Structure



1. Introduction

Around <u>2.6</u> <u>million deaths¹</u> per year, due to medical errors <u>684</u> patient misidentification events in the US led to <u>patient</u> <u>harm</u>, and in some cases, <u>death</u> (32 months span)

<u>Missing</u> patient wristbands or <u>incorrect</u> <u>information</u> on them $\frac{\$42 \text{ billion}^1}{\&}$ $\frac{\$1.2 \text{ million}^2}{\text{in costs each year}}$

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¹ In low to middle-income countries (WHO)

² for the average healthcare organization in the US ("2016 National Patient Misidentification Report" independently conducted by Ponemon Institute LLC Sponsored by Imprivata)

Objective - Effectiveness of using biometric technology for identifying patients

Research

- Patient Identification Process
- Existing problems
- Identification methods and solutions
- Security and privacy issues

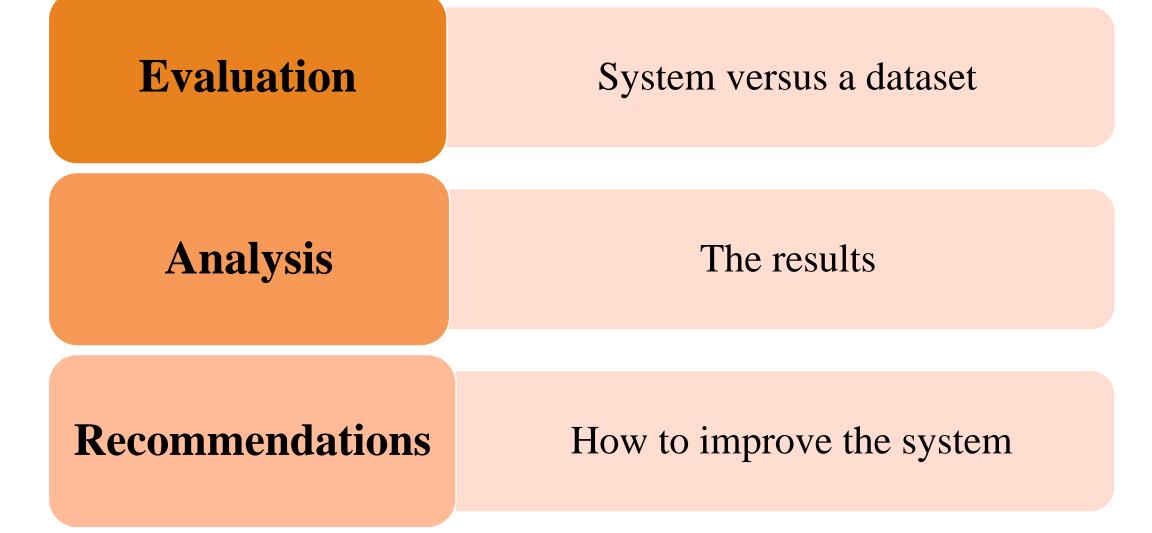
System Requirements

- Developing a list of system requirements
- By means of a questionnaire to healthcare professionals

System Proposal

• Propose a system based on biometric technology





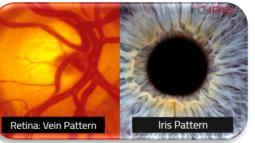
2. Research Identification methods



ii. Palm Vein Pattern Recognition









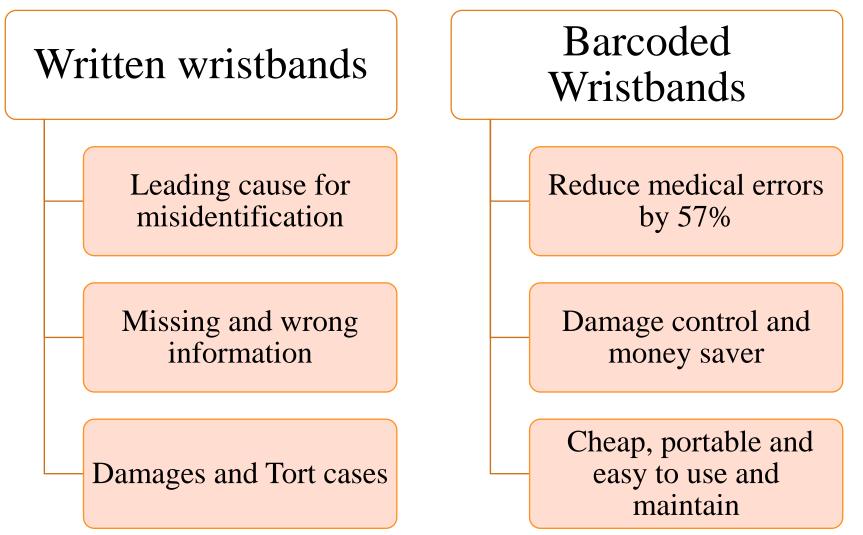
iv. Face Recognition

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i. Wristbands

iii. Ocular Based Identification





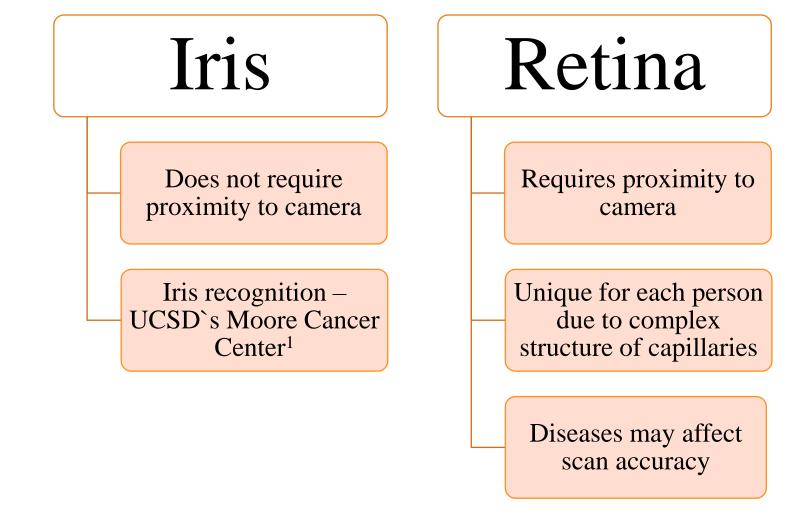
ii. Palm Vein Pattern Recognition

Uniqueness of palm vein pattern Impossible to reproduce with fake palms

More accurate – More costly and Intrusive

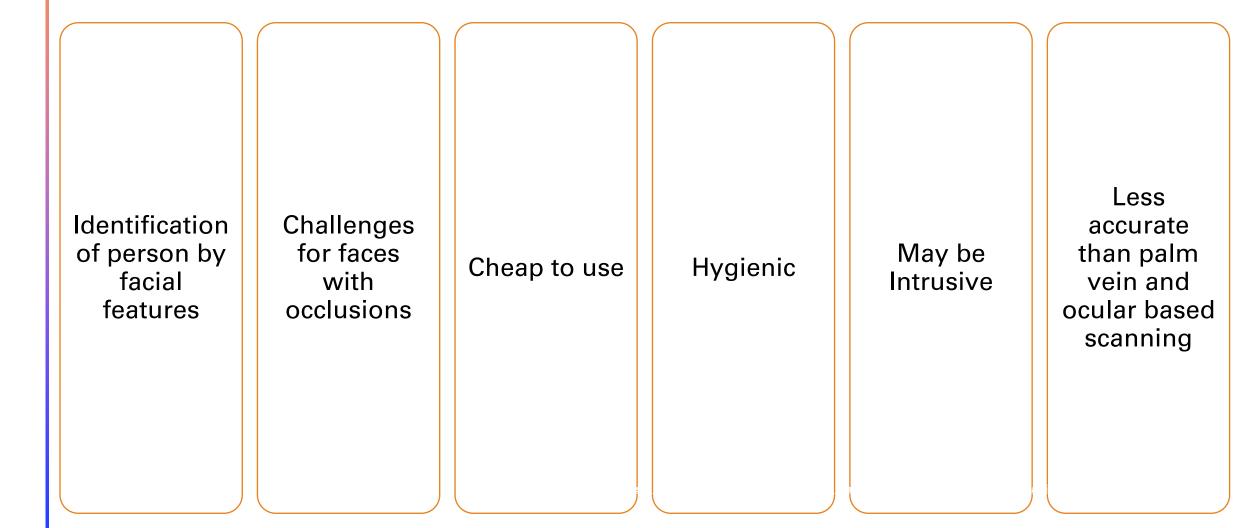
Less Hygienic

iii. Ocular Based Identification



¹B. N. HAILE, "THE EYES HAVE IT : IRIS BIOMETRICS SAFELY IDENTIFY UCSD PATIENTS FOR RADIATION ONCOLOGY TREATMENT," 2010.

iv. Face Recognition



3. Methodology

Questionnaire

- Stakeholder information
- Their awareness of the problem, if any
- Their **process** of patient identification
- Their **preferences** of **solutions**

Participants

- Nurses
- Doctors
- Physiotherapists
- Surgeon
- Speech language pathologist

Results

- 67% think that their current system works moderately well, with low-cost being the main reason behind this
- Security was the biggest concern

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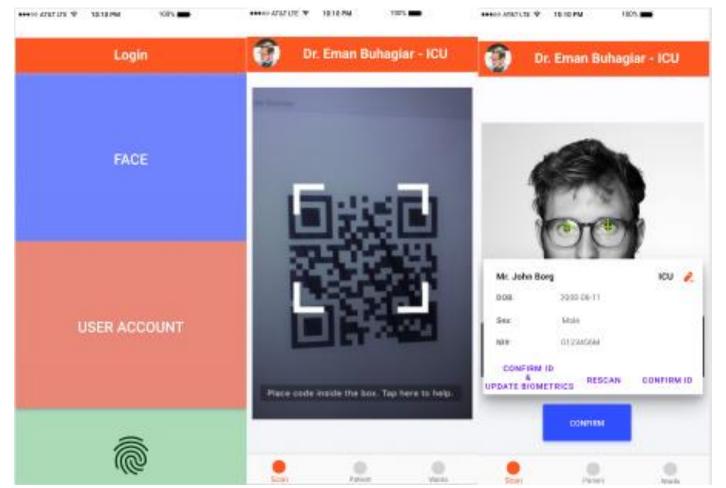
House of Quality (HoQ) Matrix for System Requirements

4. Design

- Mobile app with user authentication and authorisation.
- User can identify an already registered patient by:
 - Scanning the patient's barcode
 - Taking a **photo** of the patient's face
 - Confirming the patient details
- Patient crucial information is then displayed on the phone, which can be adjusted to the organization's needs.

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4. Design



Proposed app system designs. Authentication, Barcode/QR code scanning, Identification confirmation

5. Evaluation

Implementation using Microsoft Cognitive Services and their Face API

- Seamless, secure and easy to integrate and operate
- Face images are **not stored** on their servers
- Configurable confidence thresholds
- Relatively cheap

Accuracy Evaluation

- **Dataset** of faces¹ to be evaluate against
- All faces were registered with the API
- Identification was tested

Performance Evaluation

- A proof-of-concept mobile application was developed
- Scanning of a barcode and a person's face
- Biometric information sent to Microsoft Face API for identification.
- Database call to fetch patient's fake records.

1S. Milborrow, J. Morkel, and F. Nicolls, "The MUCT Landmarked Face Database," Proc. Pattern Recognit. Assoc. South Africa, pp. 32–34, 2010.

TABLE I. CASE 1 RESULTS

6. Results

- Accuracy
 - 1. Different **angles**:
 - 88%-93% with 0.97 confidence
 - 100% with 0.94 confidence
 - 2. Different lighting:
 - **40%-44%** with 0.97 confidence
 - 93%-97% with 0.94 confidence
 - 3. After **training** the dataset:
 - **65%** with 0.97 confidence
 - 97% with 0.94 confidence
- Performance
 - **5 to 7 seconds** with full-bar Wi-Fi connection to detect and identify the patient and get their records from a database.

Thr. Sc.	0.97	0.96	0.95	0.94	
b	88.04	98.55	99.28	100.00	
c	88.04	98.55	99.28	100.00	
d	99.28	100.00	100.00	100.00	
•	93.48	97.83	98.91	100.00	

n = 276

TABLE II. CASE 2 RESULTS

Thr. Sc.	0.97	0.96	0.95	0.94	0.93	0.92
rb	43.96	73.63	90.11	96.70	100.00	100.00
sb	39.56	64.84	82.42	93.41	98.54	100.00

n = 91 (photos 000 - 090)

TABLE III. CASE 3 RESULTS'

Thr. Sc	0.97	0.96	0.95	0.94	0.93	
sb	64.84	82.42	92.31	96.70	100.00	

n = 91 (photos 000 - 090)

Conclusion and Future Works

Patient Misidentification

- Known global problem in the Health Sector
- Complications Patient and Organisation

Face Recognition

- This was the main focal point of study for identification of patients
- Most biometric preferred method chosen by questionnaire participants
- Over 80% accuracy

Conclusion and Future Works

Future Works

- System needs to be evaluated against a larger dataset
- More face occlusions and real-case scenarios

Security Aspects

- Minimise risks of malicious attacks
- Gain more confidence from end users.

Thank you for your attention!

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