Possible Attacks on Match-in-Database Fingerprint Authentication

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Introduction

Biometrics

Global Biometric market will exceed 70 billion by 2027

80% of Americans have used biometrics

- □ What are they?
- □ Used For?
- ☐ Types?



Source: Gait Recognition (2018)

Fingerprint Fun Facts

Unique to everyone

- ➤ Identical twins
- ➢ Formed from struggle in the womb
- ➢ Friction ridges

Loss of fingerprints

- ➢ Gene mutation
 - 4 families
- Bricklayers, lime workers, chemo drugs
- ➢ Burning off
- ➤ Will grow back
- Animals have them too
 - > Apes, Chimpanzees, Koalas





Possible Attacks on Match-In-Database Fingerprint Authentication Outline

- **Fingerprint Authentication** How it works/ modules Threshold variants Threat Model Proposed model Types of Attacks Spoofing Denial-Of-Service Replay Trojan Horse
- Conclusion



Source: PECB Insights (2018)

Match-in-Database Fingerprint Authentication System

- ✤ MiD Fingerprint Authentication
 - > Uses a remote database to store template
 - Template: digital representation of a fingerprint that has been encrypted
- Input Finger = Enrollment
- Live Finger = Authentication
- Verification vs. Identification
 - ➢ Fingerprint matches claimed registered user
 - > Unknown identity



Fingerprint Acquisition/ Image Enhancement

- Physical Factors
 - ≻ Sweat
 - > Pressure
 - ≻ Cut

Environmental Factors

- ≻ Humidity
- ➤ Temperature
- > Durability



Feature Extraction/ Template Generation

- Characteristic Features
 - Ridges/ Valleys
- ✤ Minutia

	Termination
A	Bifurcation
þ	Lake
	Independent ridge
-	Point or island
	Spur
Z	Crossover

Binary Strings

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00010111

Source: Characteristic Features (2014)

Comparison Module

- ✤ Threshold
- ✤ False Acceptance Rate (FAR)
- ✤ False Rejection Rate (FRR)
- Equal Error Rate (EER)



Outline

	gerprint Authenticati
Threat Model	
	Proposed model



Source: PECB Insights (2018)

Threat Model

- Process of identifying, and prioritizing potential security threats
- Diagram the system
- Identify where threats could occur (Attack Points: 16)
- ✤ Threat Level
- Mitigation techniques



Proposed by: Security Analysis (2020)

Outline

Types of Attacks Spoofing Denial-Of-Service Replay Trojan Horse



Source: PECB Insights (2018)

Types Of Attacks

- Direct: Attack to the scanner or cash dispenser itself
 AP 1, 2, 16
- Indirect: Attack to a component within the system



Spoofing Attack

- Direct attack
- ✤ AP 1
- Provides false
 biometric data



Spoofing Attack- spoofing methods

Cooperative

- > Direct Mold
- Non- Cooperative
- ✤ Latent Fingerprint
 - ≻ Invisible
 - Powder, Brush, Tape
- Fingerprint Reactivation
 - \succ From scanner itself
 - Heavy breathing, water filled bag, graphite powder
- Cadaver
 - Enrolled user dead
- Fingerprint Synthesis
 - \succ From template
 - Need access to database





Spoofing Attack- Anti-spoof methods -Hardware

Costly

- ✤ Add another problem
 - Leak confidential info through patterns of electromagnetic waves
 - > Power consumption



Spoofing Attack- Anti-spoof methods -Software

- Static
 - Pore-based
 - Quantity
 - Distribution
 - High resolution
 - > Perspiration
 - Shading
 - Depends on pressure applied
 - ≻ Texture
 - Coarseness
 - Multi-resolution texture analysis
 - Commonly used together
- Dynamic
 - > Snapshots
 - > Perspiration
 - Ridge Distortion



✤ High Risk

- Ease of creating fake fingerprint
- Not implementing antispoof methods

Denial-Of-Service (DOS) Attack

- Direct
- ✤ AP 1, AP 16
- Overload the system
- Unusable for everyone



Denial-Of-Service (DOS) Attack- Mitigation

- Eye on the scanner
 - Video cameras
 - ➤ Security guards
- Rugged devices
 - Created to withstand unusual circumstances
- ✤ Low threat
 - ≻ No info
 - > Annoyance



Replay Attack

- Indirect attack
- ✤ AP 3, AP 9
- Eavesdrops on the communication channels
- ✤ Intercepts
- Resends previous user input



Replay Attack- Mitigation

- Challenge/Response method
 - Query system
 - \succ Transaction server
 - Different pixel value sent every time
 - Scanner and Feature Extraction
 - Have to match
- Global Clock
 - ≻ Timestamps
- Liveliness Test
 - Active = swipe
 - > Passive
 - Texture, pore distribution, ridge distortion
- ✤ Medium Threat Level
 - Challenging to gain access to the channel
 - \succ Get all the info



Trojan Horse Attack

- Indirect
- ✤ AP 4, AP 11
- Trojan embedded
- Activated
 - > Modify, copy, delete



Trojan Horse- Mitigation

Trusted Biometric System (TBS)

- Mutual Authentication
 - Both sides of channel verify each other at the same time
- Increase computation power & time
- Code Signing
 - > Each module's digital signature
- Hardware
 - ≻ Tamper resistant
 - Communication channels
- ✤ Medium Threat Level
 - Hard to introduce Trojan
 - ➢ Easy to modify software



Conclusion



Spoofing Attack

Replay Attack

Trojan Horse Attack

DOS Attack

Increase in biometrics - Threshold

- ✤ Biometrics are increasing
- ✤ With more users
 - ➢ 0.1% FAR currently
 - 1 out of 1000
 - FBI: 6,000 out of 6 million
 - US pop: 332,000 out of 332 million ■



Questions?

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