

Presentation Attack Detection for Smartphone Finger Image Recognition

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Agenda

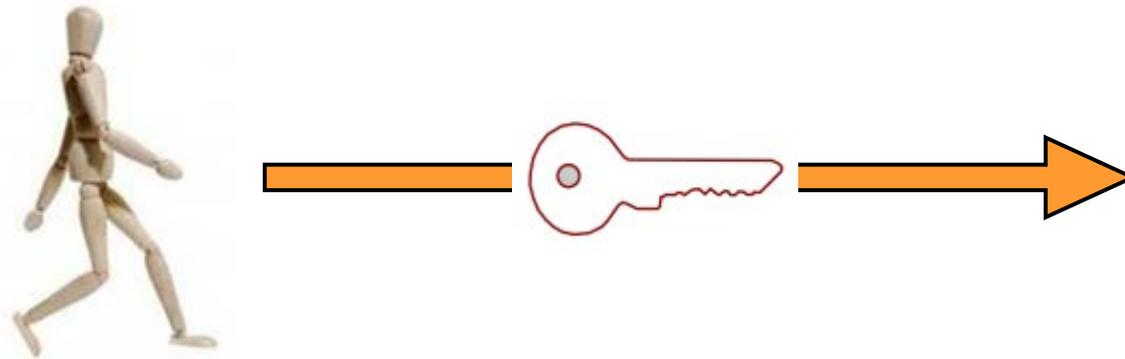
- Access Control
- Biometric authentication on Smartphones
- Presentation Attack Detection
- Are the metrics in 30107-3 applicable?

Access Control

Access Control

Traditionally we place between

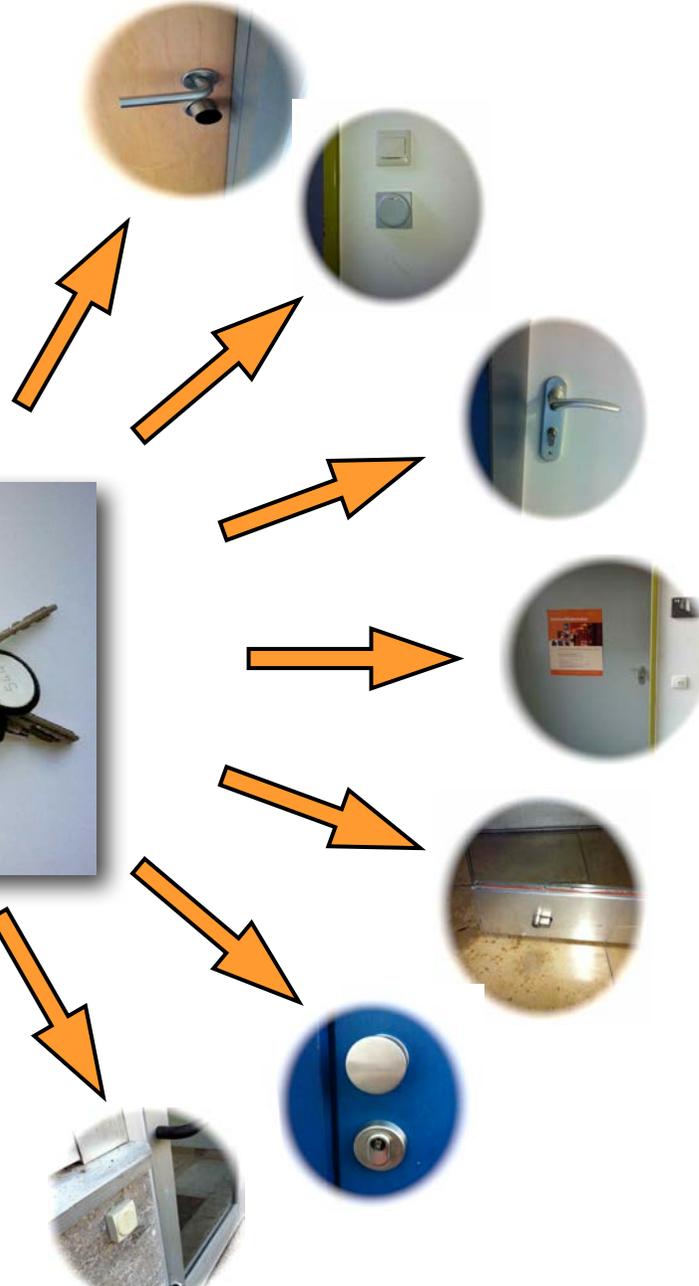
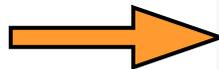
- individuals
- and objects
- a token (i.e. key)



Access Control

But in **reality** individuals

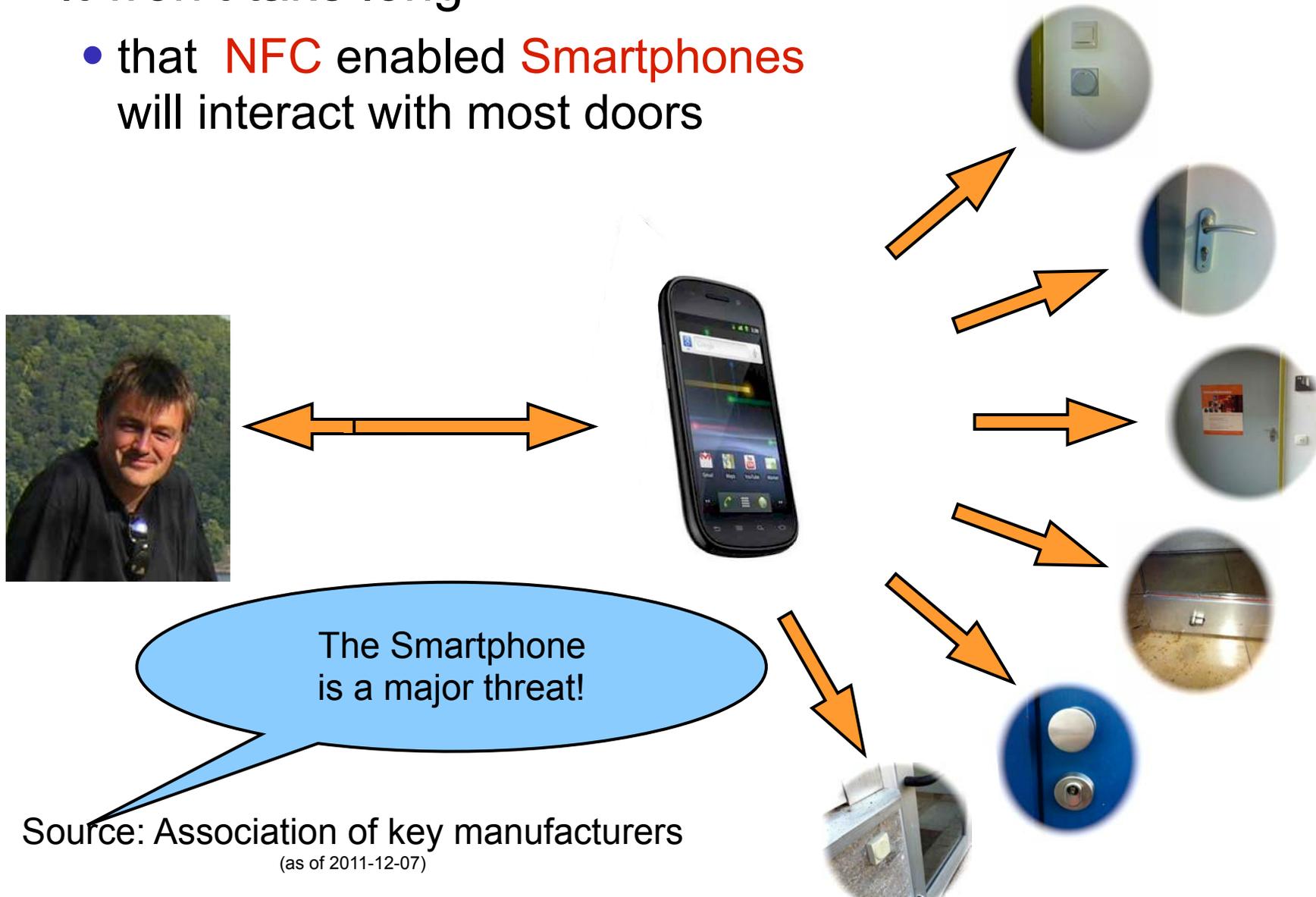
- do not have just one
- but **many** keys
- granting access to **many** doors



Smartphone Based Access Control

It won't take long

- that **NFC** enabled **Smartphones** will interact with most doors



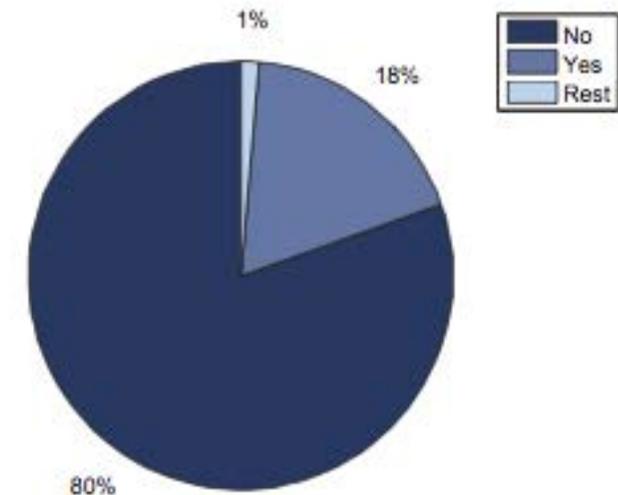
Source: Association of key manufacturers
(as of 2011-12-07)

Do we use Access Control
before we unlock our Smartphone?

End-User Survey

Data in **mobile devices** is often insufficiently **protected**

- No PIN-authentication required after stand-by phase
 - Survey-result with 962 users : **only 18%** use PIN code or visual pattern to unlock
- All **data** on the phone is **freely** available
 - Emails, addresses, appointments, photos
 - PINs etc.



Reason for this:

- PIN-authentication is too much effort (30%)
- People are self-responsible for their phones

[Ni12] C. Nickel: „Accelerometer-based Biometric Gait Recognition for Authentication on Smartphones“, PhD-thesis, TUD, 2012

Biometrics on Smartphones

Is the integration of fingerprint sensors in Smartphones a security gain?

- Chaos Computer Club: NO
- cb: YES - it motivated many users to activate access control in the first place



Image Source: Apple 2013



Image Source: Samsung 2013

Preliminary assessment:

- Apples introduction of iPhone 5s offers a **convenience solution** that satisfies the security requirements for authentication for low volume transaction.
- For the experienced attacker the sensor has shown weaknesses

Smartphone Access Control

Foreground authentication (user **interaction**)

- Deliberate decision to capture (wilful act)
- **Camera**-Sensor
 - **Fingerprint** recognition
 - Apples iPhone 5S
 - Finger**photo** analysis
 - Face recognition
 - Iris recognition



Background authentication (**observation** of the user)

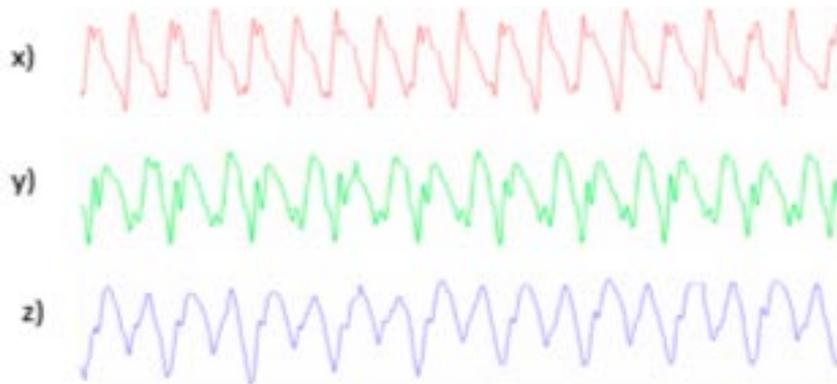
- Microphone
 - **Speaker** recognition
- Accelerometer
 - **Gait** recognition
 - concurrent - unobtrusive



Biometric Gait Recognition

Offer an **unobtrusive** authentication method

- Use **accelerometers** - already embedded in mobile devices to record the gait
 - Many phones contain **accelerometers**
 - No extra hardware is necessary
 - Acceleration measured in 3-directions



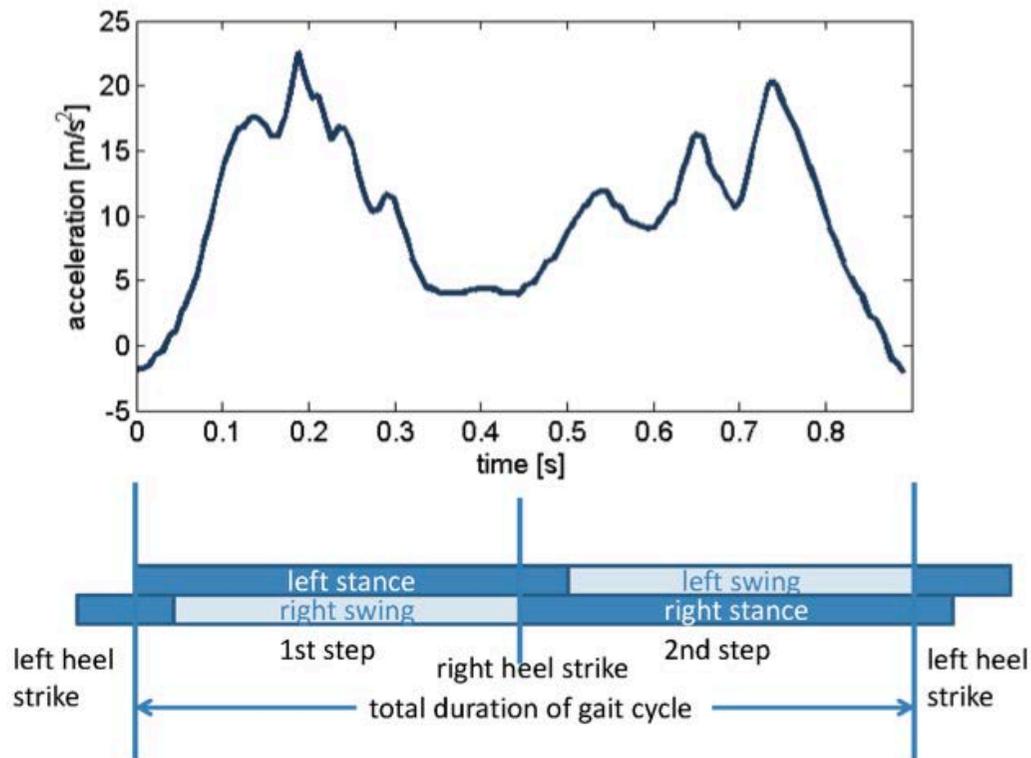
- First paper on this topic:
[DNBB12] M. Derawi, C. Nickel, P. Bours, C. Busch: „Unobtrusive User-Authentication on Mobile Phones using Biometric Gait Recognition“, Sixth International Conference on Intelligent Information Hiding and Multimedia Signal Processing (IHMSP 2010)
- EER 20% at that time



Biometric Gait Recognition

Data capture process

- periodical pattern in the recorded signal



Best result

- now at **6.1%** EER

The following is
prehistoric work

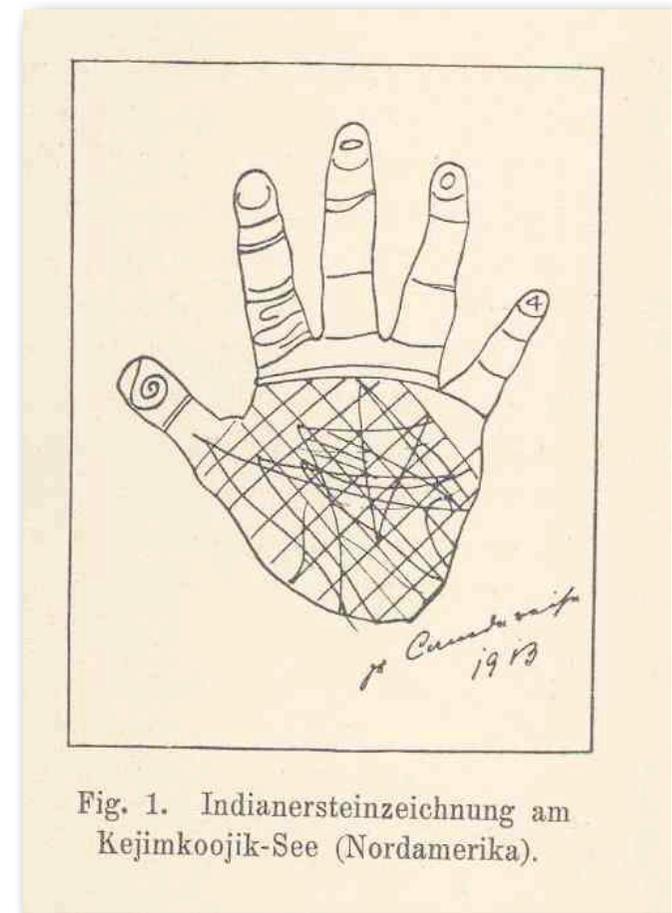


Fig. 1. Indianersteinzeichnung am
Kejimkoojik-See (Nordamerika).

Image Source: Heindl 1927

The following is
prehistoric work (before the Apple iPhone5 arrived)
but as always:
we can learn from history

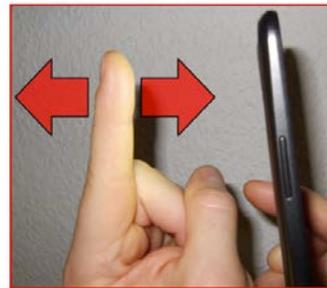
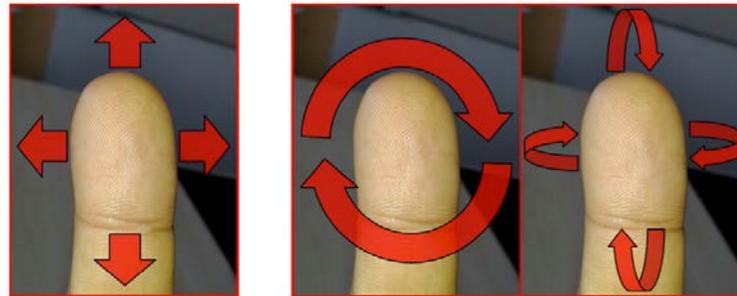
Smartphone Access Contol

Master Thesis Chris Stein 2012: Finger recognition

- Smartphone **camera** as sensor
- Authentication based on photo of the finger

Challenges

- Translation and rotation
- Distance finger to camera
- uncontrolled background and illumination



Smartphone Access Control

Capture process

- Camera operating in **macro** modus



Preview image of the camera with LED on (left) and LED off (right)

- LED permanent on



Finger illuminated

[SNB12] C. Stein, C. Nickel, C. Busch, „Fingerphoto Recognition with Smartphone Cameras“, Proceedings 11th Intern. Conference of the Biometrics Special Interest Group (BIOSIG 2012)

Smartphone Access Control

Fingerprint recognition

- Preprocessing, minutiae extraction and comparison are performed on the phone
- Results of 18% EER are based on DigitalPersona FingerJetFX OSE (Open Source Edition) and home-made-minutiae comparator



- see video at: <http://www.dasec.h-da.de/research/biometrics/mbassy/>

Smart Phone Access Control

Finger recognition study - 2012/2013

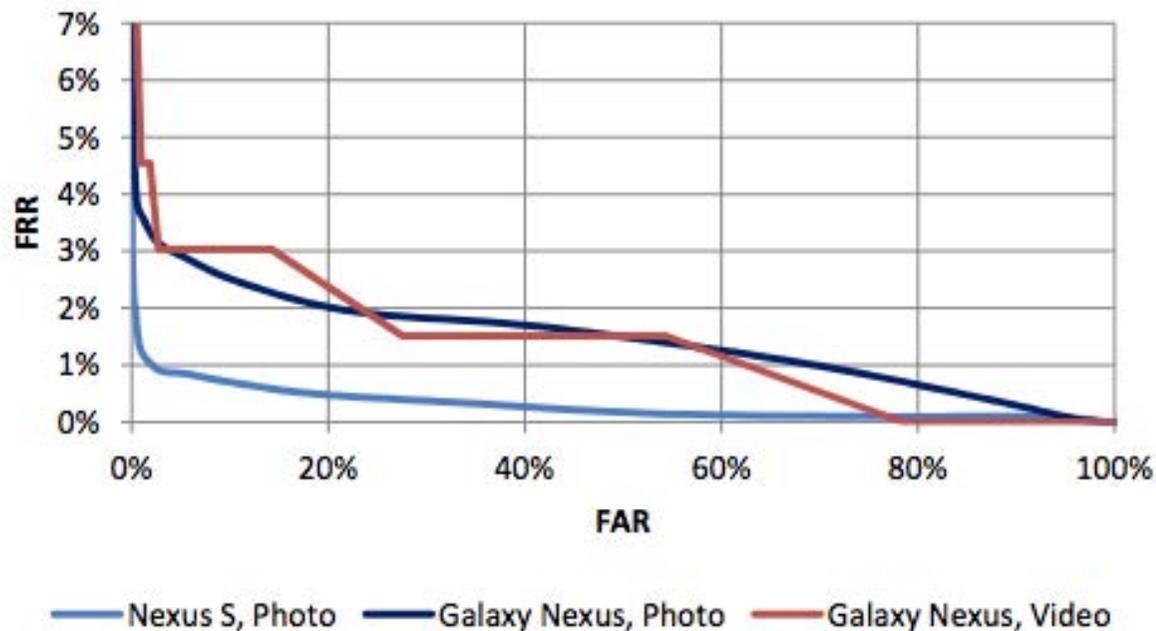
- Objectives:
 - Replace home-made comparator (and the Digital Persona extractor) by COTS standard technology to **increase performance**
 - Investigate **Presentation Attack Detection** capabilities with reflection analysis and video recordings

Smart Phone Access Control

Finger recognition study - 2012/2013

- Results: **biometric performance** at 1.2% EER

DET Curve



Capture Method and Device	EER from [SC-2012]	EER	FRR (FAR=0.1%)
Photo, Nexus S	22.3%	1.2%	2.7%
Photo, Galaxy Nexus	19.1%	3.1%	6.7%
Video, Galaxy Nexus	-	3.0%	12.1%

[SBB13] C. Stein, V. Bouatou, C. Busch, „Video-based Fingerphoto Recognition with Anti-spoofing Techniques with Smartphone Cameras“, Proceedings 12th Intern. Conference of the Biometrics Special Interest Group (BIOSIG 2013)

Smart Phone Access Contol

Finger recognition study - 2012/2013

- Presentation Attacks
 - 1: replay from Smartphone display (simple)
 - 2: self generated print-outs (not critical to detect)
 - 3: Ralph Breithaupt's / BSI best artefacts (very challenging)



Replay attack



Simple artefacts



Challenging artefacts

Smart Phone Access Control

Finger recognition study - 2012/2013

- Observation
 - significant strong **light reflection** near the fingertip
 - from the cameras LED
- Reflection depends on
 - **Shape** of the finger
 - **Consistency** of the finger
 - **Angle** of the finger to the camera



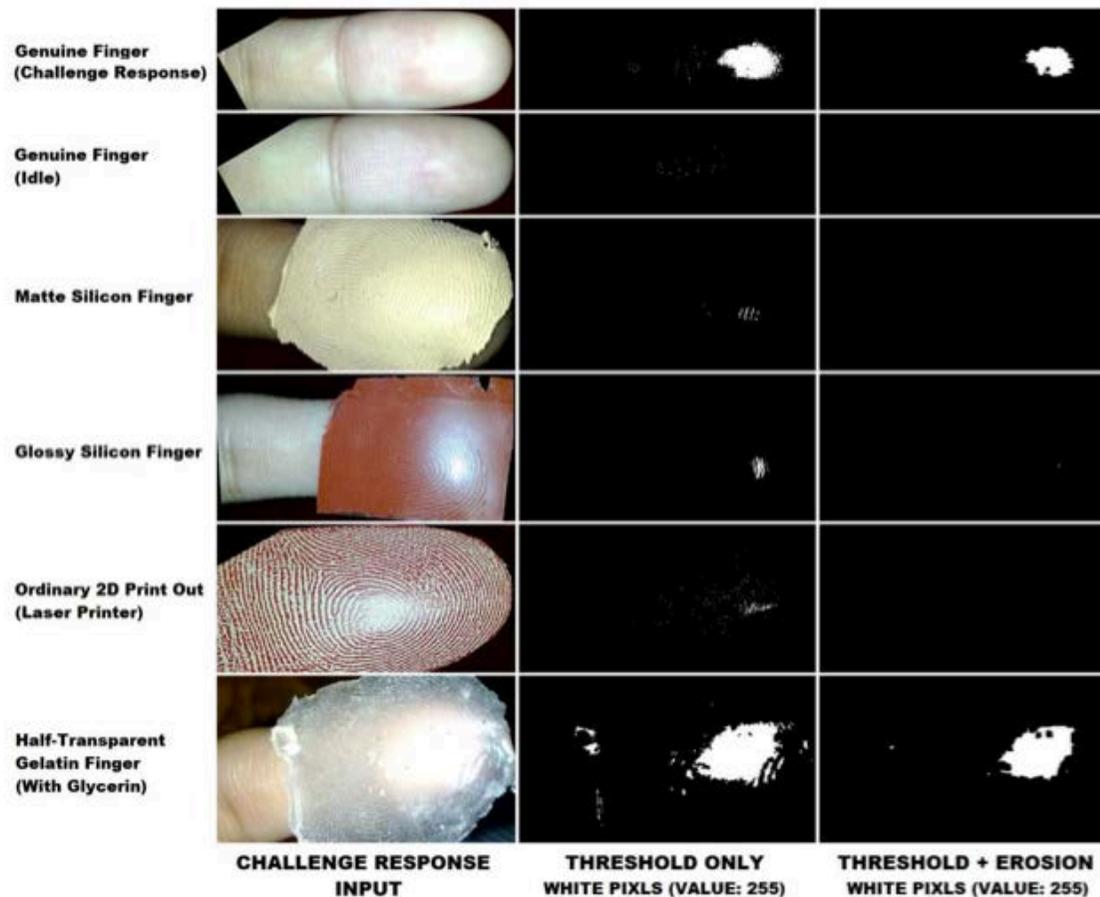
- Attack detection, as light reflection differs from artefacts to genuine fingers

- [SBB13] C. Stein, V. Bouatou, C. Busch, „Video-based Fingerphoto Recognition with Anti-spoofing Techniques with Smartphone Cameras“, Proceedings 12th Intern. Conference of the Biometrics Special Interest Group (BIOSIG 2013)

Smart Phone Access Control

Finger recognition study - 2012/2013

- Results: Presentation Attack Detection



- Conclusion:
better **Presentation Attack Detection** than capacitive sensors

Reporting about the PAD using ISO/IEC WD 30107

PAD-Standard

Definitions in ISO/IEC 30107 PAD - Part 1: Framework

- **artefact**

*artificial object or representation presenting a **copy** of biometric characteristics or synthetic biometric patterns.*

- **artefact species**

*artefacts based on sources whose biometric characteristics differ but which are otherwise identical (e.g. based on a **common medium and production method** but with different biometric characteristic sources)*

- **attack potential** *(this definition is from CC terminology)*

*attribute of a biometric presentation attack **expressing** the **effort** expended in the preparation and execution of the attack in terms of elapsed time, expertise, knowledge about the capture device being attacked, window of opportunity and equipment, graded as “no rating“, “minimal“, “basic“, “enhanced-basic,“ “moderate” or “high.*

PAD-Standard

Metrics in ISO/IEC 30107 PAD - Part 3: Testing and reporting and classification of attacks

- **Attack presentation classification error rate (APCER)**
proportion of attack presentations incorrectly classified as normal presentations in at the component level a specific scenario
- **Normal presentation classification error rate (NPCER)**
proportion of normal presentations incorrectly classified as attack presentations at the component level in a specific scenario

Applying ISO/IEC 30107-3 Metrics

Do the metrics currently in ISO/IEC 30107 PAD - Part 3: serve to provide a meaningful report?

- [SBB12] - Publication:
The reported number of attack presentations incorrectly classified as normal presentations was **one** out of **five** artefacts
- Thus the APCER to be reported is

$$APCER = \frac{1}{5} = 0.2$$

- but there were in fact **27 artefact species**, that were used in the background but **not reported** as they are not challenging

$$APCER = \frac{1}{27} = 0.04$$

Refining ISO/IEC 30107-3 Metrics

Findings

- The **size** of the corpus with the artefact species is essential
- The APCER should be based on **presentation attack instrument (PAI)** and not only on artefacts, which includes both artefacts and **lifeless** biometric **characteristics** (i.e. stemming from dead bodies)
 - 30107-1: **PAI** - *biometric trait or object used in a presentation attack.*
- The CC-related **attack potential** should be included in the definition
 - 30107-1: **attack potential** - *attribute of a biometric presentation attack **expressing the effort** expended in the preparation and execution of the attack in terms of elapsed time, expertise, knowledge about the capture device being attacked, window of opportunity and equipment, graded as “no rating“, “minimal”, “basic”, “enhanced-basic,” “moderate” or “high.*
- The known **success rate** of an artefact species is relevant

Refining ISO/IEC 30107-3 Metrics

Suggested **augmented** metric for ISO/IEC 30107-3

- **Attack presentation classification error rate (APCER)**
proportion of attack presentations incorrectly classified as normal presentations at the component level a specific scenario - taking the **attack potential** and the known **artefact species success rate** into account.
- **Attack potential (AP)** = {0.2 for “minimal”, 0.4 for “basic”, 0.6 for “enhanced-basic,” 0.8 for “moderate” . 1.0 for “high.”}
- **Presentation attack instrument success rate (PAISR)**
Proportion of evaluated capture devices that could be spoofed by the specific PAI (i.e. artefact).
 - would start with a value of 1 for a new discovered artefact species and could be reduced over time (as more sensors become robust)

Refining ISO/IEC 30107-3 Metrics

Suggested refined metrics for ISO/IEC 30107-3

- The APCER could thus be expressed as

$$APCER = \frac{\sum_{i=1}^{N_{AS}} RES_i * AP_i * PAISR_i}{N_{AS}}$$

N_{AS} number of presentation attack instruments (PAI) (i.e. artefact species) in the corpus

RES_i result of attack with i^{th} PAI
{0 for detected attack, 1 for successful attack}

AP_i attack potential of the i^{th} PAI
(close to zero, if artefact is easy to produce)

$PAISR_i$ presentation attack instrument success rate
(close to zero, if all sensor can detect this artefact)

Open Question

To be clarified

- Should there be a fixed-size of the corpus, such that all labs use a minimum number of artefact species
- Can one expect that a testing lab has access to non-artefact PAI (from dead bodies)?
- What happens with the new sensor?
The success rates starts with 1 and is decrease as robust sensor do appear
- How can evaluation labs have an equivalent set of PAI with all the same attack potential?

Conclusion

- **Smartphones** without biometric access control are a risk today and will be a **critical factor** tomorrow
 - once they will open doors via NFC
- The iPhone5 has changed this
- Biometric sensors are available in Smartphones at **zero** cost
 - even though they were built-in for other purposes
- Gait recognition shows reasonable biometric performance
- Currently defined metrics in ISO/IEC 30107-3 deserves refinement

Credits

Thanks to all that supported this work

This talk is based:

- on the **work** of my group members
 - Claudia Nickel (survey on Phone security) and
 - Chris Stein (Presentation Attack Detection)
- on Ralph Breithaupt - providing the **artefacts** / presentation attack instruments used in this study
- Morpho (Safran Group) - funding the study
- Elaine Newton, Olaf Henniger and Michael Thieme serving as **editor** for ISO/IEC 30107
- all ISO/IEC JTC1 SC37 WG3 members providing the **energy** and **patience** to generate a good PAD-multipart standard

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