Biometric Systems and Presentation Attacks

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CRISP
Center for Research in Security and Privacy

NTNU

What is a presentation attack?

What are Presentation Attacks?

We can learn from the James Bond movie

- 1971: Diamonds Are Forever …
  … and James Bond impersonates Peter Frank

Biometric Presentation Attacks

A new understanding of a

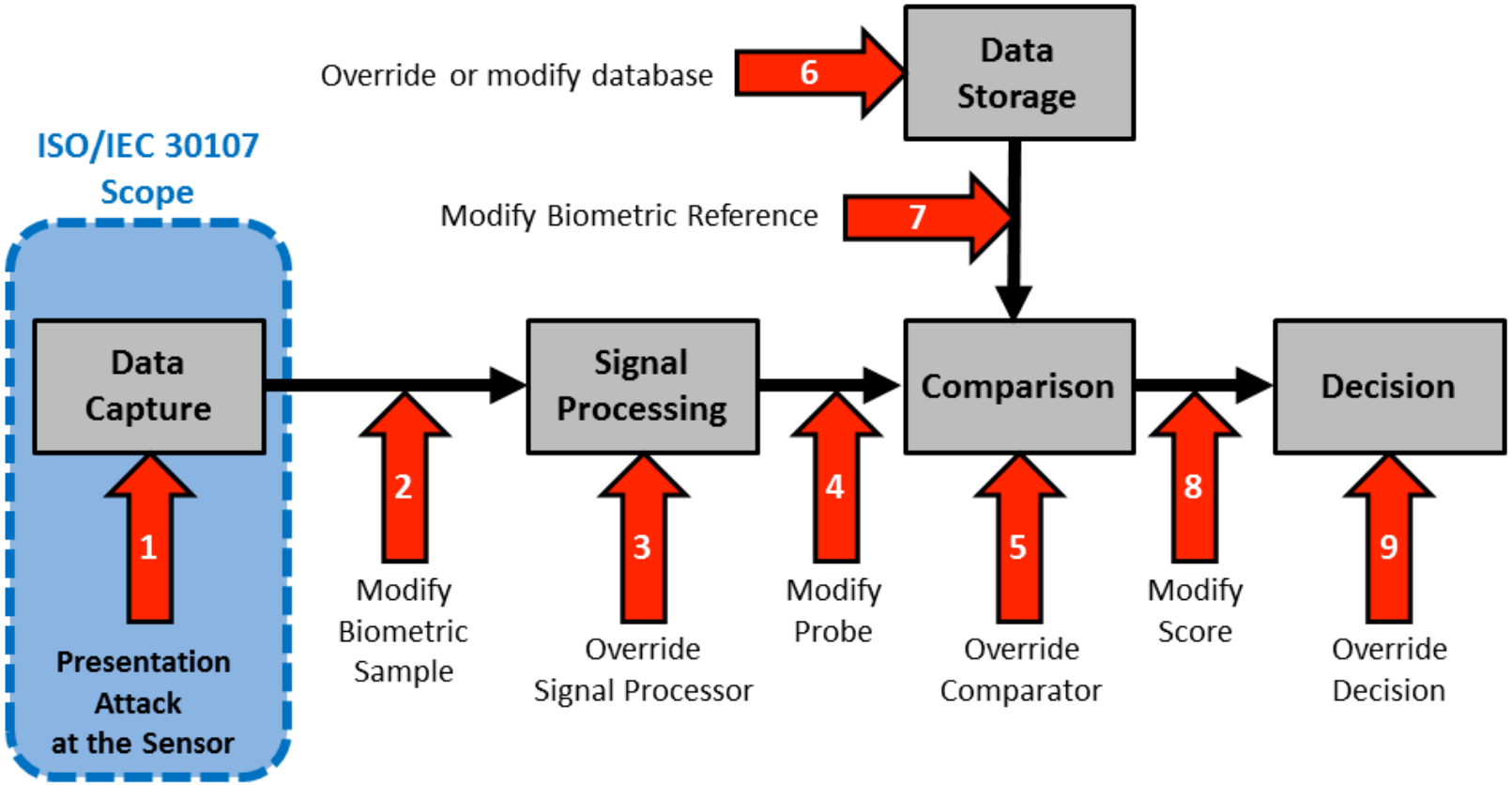
- Keyring - impersonating target victims that have the desired authorization

Image Source: c't magazine

Weakness of Biometric Systems

Overview of attacks on a Biometric System

- Capture Device (1): Camera, CMOS-Chip, optical- / capacitive sensor



Source: ISO/IEC 30107-1
Inspired by N.K. Ratha, J.H. Connell, R.M. Bolle, "Enhancing security and privacy in biometrics-based authentication systems," IBM Systems Journal, Vol 40. NO 3, 2001.

Gummy Finger Production in 2000 !

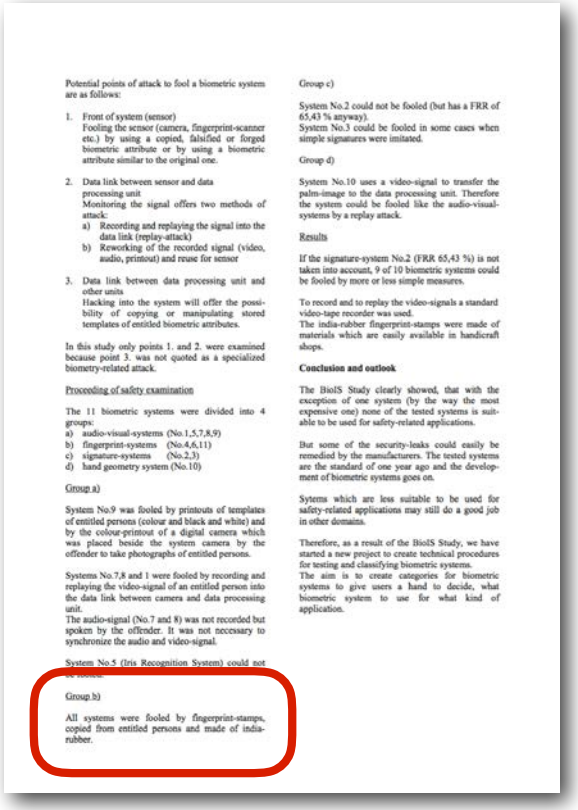
Attack without support of the target victim

- Recording of a latent fingerprint from flat surface material
  - z.B. glass, CD-cover, etc.
    with iron powder and tape
- Scanning and post processing:
  - Correction of scanning errors
  - Closing of ridge lines (as needed)
  - Image inversion
- Print on transparent slide
- Photochemical production of a circuit board
- Artefact with silicon, which will have flexibility and humidity

Gummy Finger Production in 2000 !

Reported in a publication by the German Federal Police

- Findings:
  - "*All systems were fooled by fingerprint-stamps, copied from entitled persons and made of india- rubber.*"



Potential points of attack to fool a biometric system are as follows:

1. Front of system (sensor)
   Fooling the sensor (camera, fingerprint-scanner etc.) by using a copied, falsified or forged biometric attribute or by using a biometric attribute similar to the original one.
2. Data link between sensor and data processing unit
   Monitoring the signal offers two methods of attack:
   a) Recording and replaying the signal into the data link (replay-attack)
   b) Reworking of the recorded signal (video, audio, printout) and reuse for sensor
3. Data link between data processing unit and other units
   Hacking into the system will offer the possibility of copying or manipulating stored templates of entitled biometric attributes.

In this study only points 1. and 2. were examined because point 3. was not quoted as a specialized biometry-related attack.

Proceeding of safety examination

The 11 biometric systems were divided into 4 groups:
a) audio-visual-systems (No.1,5,7,8,9)
b) fingerprint-systems (No.4,6,11)
c) signature-systems (No.2,3)
d) hand geometry system (No.10)

Group a)

System No.9 was fooled by printouts of templates of entitled persons (colour and black and white) and by the colour-printout of a digital camera which was placed beside the system camera by the offender to take photographs of entitled persons.

Systems No.7,8 and 1 were fooled by recording and replaying the video-signal of an entitled person into the data link between camera and data processing unit.
The audio-signal (No.7 and 8) was not recorded but spoken by the offender. It was not necessary to synchronize the audio and video-signal.

System No.5 (Iris Recognition System) could not

Group b)

All systems were fooled by fingerprint-stamps, copied from entitled persons and made of india-rubber.

Group c)

System No.2 could not be fooled (but has a FRR of 65,43 % anyway).
System No.3 could be fooled in some cases when simple signatures were imitated.

Group d)

System No.10 uses a video-signal to transfer the palm-image to the data processing unit. Therefore the system could be fooled like the audio-visual-systems by a replay attack.

Results

If the signature-system No.2 (FRR 65,43 %) is not taken into account, 9 of 10 biometric systems could be fooled by more or less simple measures.

To record and to replay the video-signals a standard video-tape recorder was used.
The india-rubber fingerprint-stamps were made of materials which are easily available in handicraft shops.

Conclusion and outlook

The BioIS Study clearly showed, that with the exception of one system (by the way the most expensive one) none of the tested systems is suitable to be used for safety-related applications.

But some of the security-leaks could easily be remedied by the manufacturers. The tested systems are the standard of one year ago and the development of biometric systems goes on.

Syterns which are less suitable to be used for safety-related applications may still do a good job in other domains.

Therefore, as a result of the BioIS Study, we have started a new project to create technical procedures for testing and classifying biometric systems.
The aim is to create categories for biometric systems to give users a hand to decide, what biometric system to use for what kind of application.

[Zwiesele2000] A. Zwiesele et al. „BioIS Study - Comparative Study of Biometric Identification Systems", In: 34th Annual 2000 IEEE International Carnahan Conference on Security Technology, Ottawa, (2000)

Presentation Attack Detection

Impostor

- impersonation attack
  - positive access 1:1 (two factor application)
  - positive access 1:N (single factor application)
- finding a look-a-like
- making appearance similar to the reference
- artefact presentation

Image Source: http://upshout.net/game-of-thrones-make-up

For fingerprint recognition:
e.g. silicon artefact production

For face recognition:
e.g. find a look-a-like first and then consult a make-up-artist

Presentation Attack Detection

Impostor

- impersonation attack
  - positive access 1:1 (two factor application)
  - positive access 1:N (single factor application)
- finding a look-a-like
- making appearance similar to the reference
- artefact presentation

Image Source: http://upshout.net/game-of-thrones-make-up

Concealer

- evasion from recognition
  - negative 1:N identification (watchlist application)
- depart from standard pose

- evade face detection

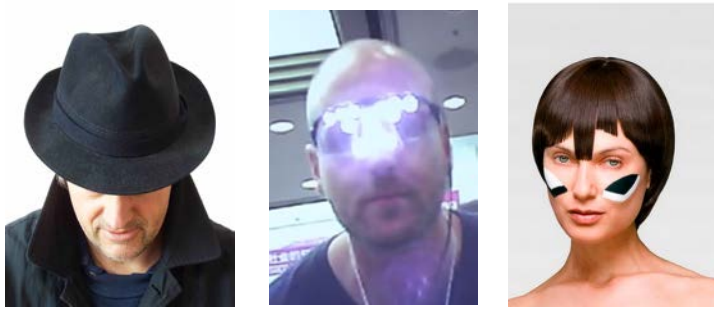


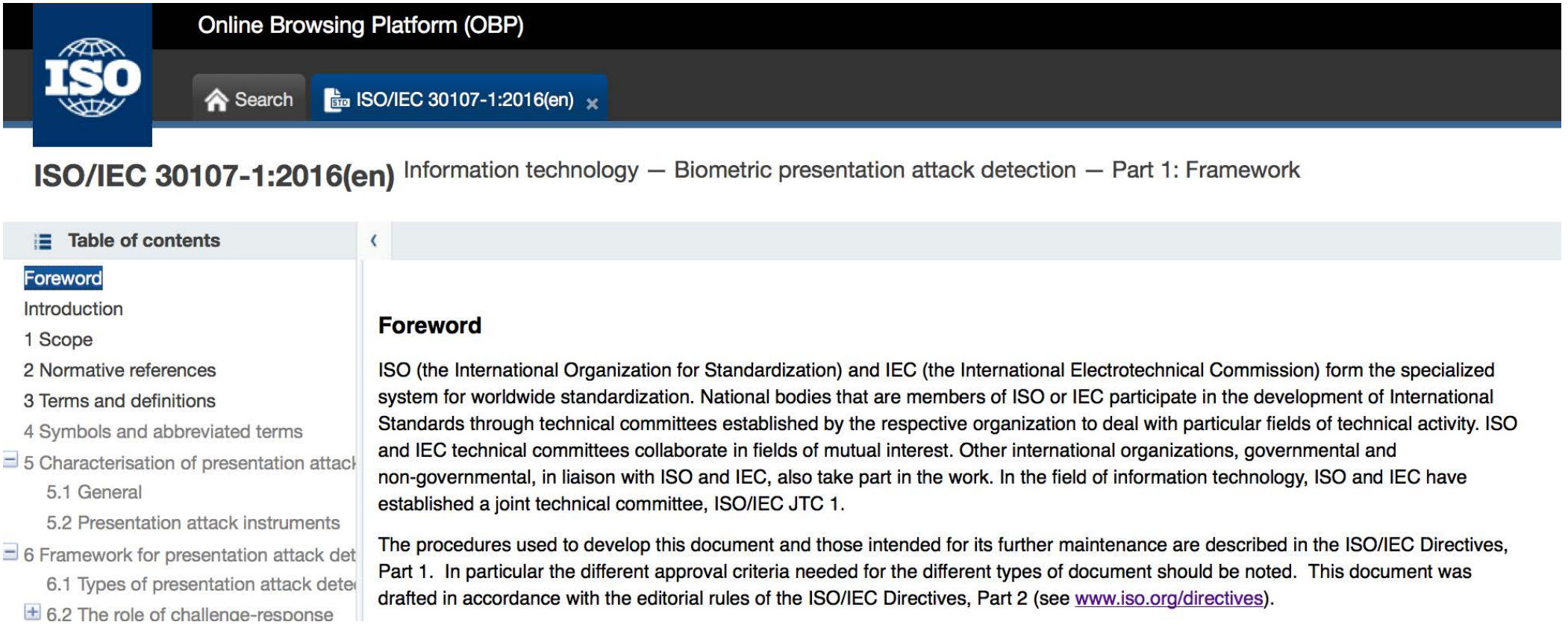
Image Source: https://www.youtube.com/watch?v=LRj8whKmN1M

Image Source: https://cvdazzle.com

Presentation Attack Detection - Framework

The international standard ISO/IEC 30107-1

- freely available in the ISO-Portal
  http://standards.iso.org/ittf/PubliclyAvailableStandards/c053227_ISO_IEC_30107-1_2016.zip



Presentation Attack Detection

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

- **presentation attack**
  presentation to the biometric capture subsystem with the goal of interfering with the operation of the biometric system
- **presentation attack detection (PAD)**
  automated determination of a presentation attack

Definitions in ISO/IEC 2382-37: Vocabulary
http://www.christoph-busch.de/standards.html

- **impostor**
  subversive biometric capture subject who attempts to being matched to someone else's biometric reference
- **identity concealer**
  subversive biometric capture subject who attempts to avoid being matched to their own biometric reference

Presentation Attack Detection

ISO/IEC 30107-1 - Definitions

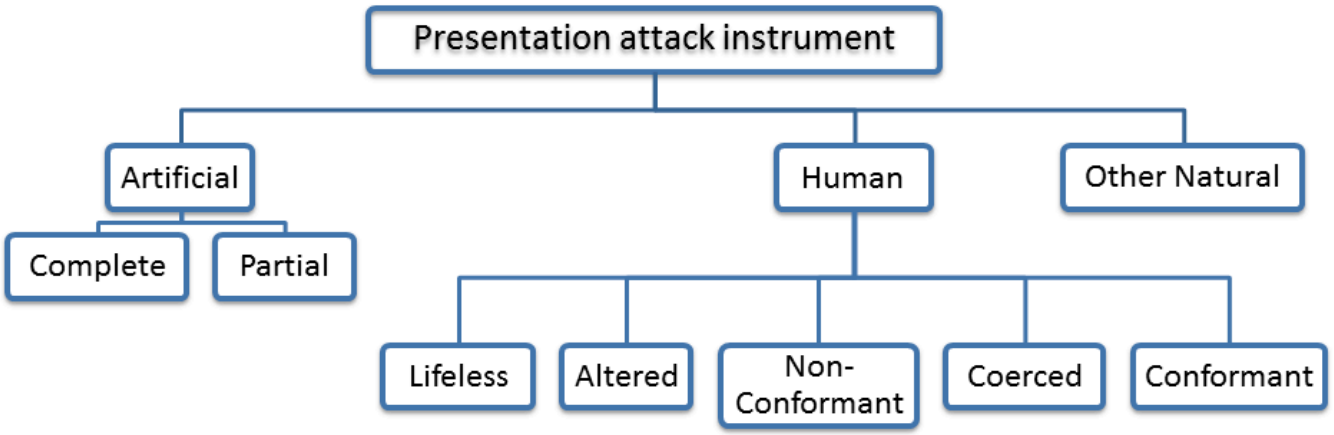
- **presentation attack instrument (PAI)**
  biometric characteristic or object used in a presentation attack
- **artefact**
  artificial object or representation presenting a copy of biometric characteristics or synthetic biometric patterns

Types of presentation attacks

(General Noun)

(Adjectives describing categories)

(Qualifying adjectives)



Source: ISO/IEC 30107-1

Presentation Attack Detection - Testing

Definition of PAD metrics in ISO/IEC 30107-3

- Testing the PAD subsystem:

- **Attack presentation classification error rate (APCER)**
  proportion of attack presentations using the same PAI species incorrectly classified as bona fide presentations in a specific scenario

- **Bona fide presentation classification error rate (BPCER)**
  proportion of bona fide presentations incorrectly classified as attack presentations in a specific scenario

Source: ISO/IEC 30107-3

Fingerphoto Presentation Attack Detection

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 skin
  - Angle of the finger to the camera

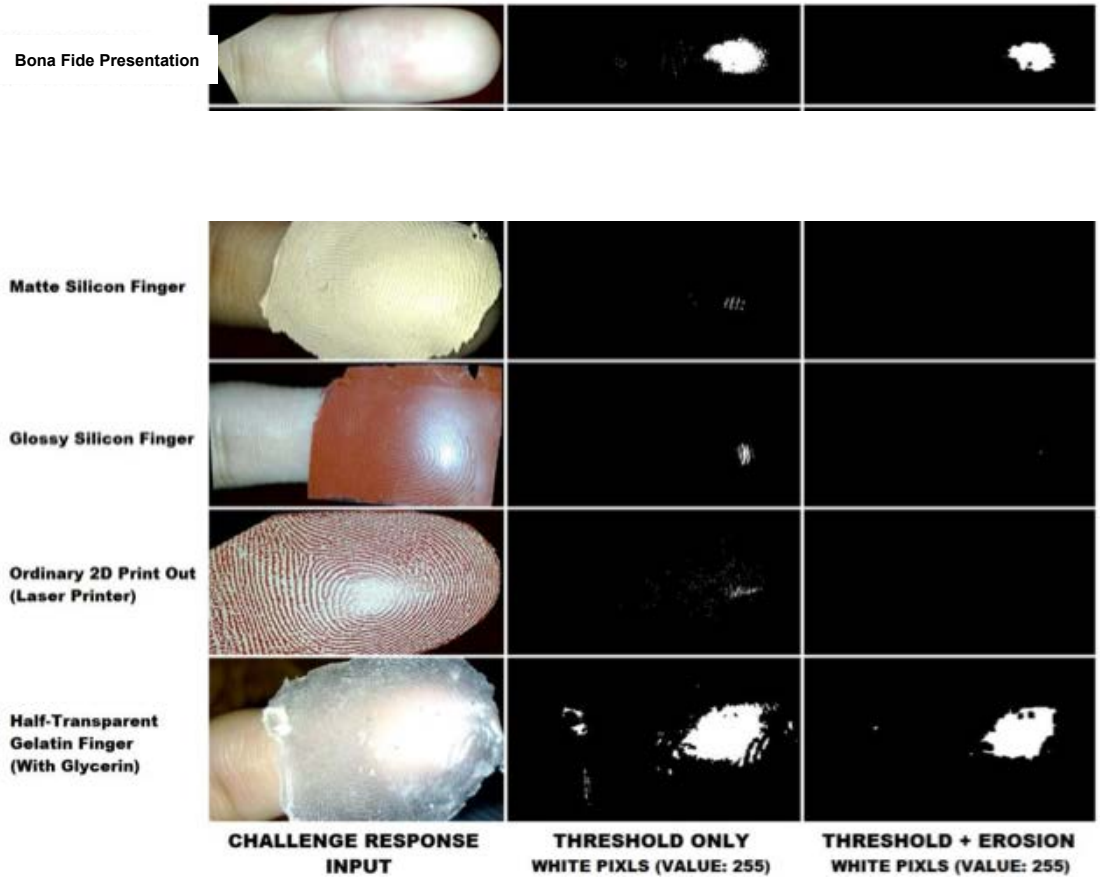
- Attack detection, as light reflection differs from artefacts to bona fide fingers

[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)

Fingerphoto Presentation Attack Detection

Finger recognition study - 2012/2013

- Results: Presentation Attack Detection (PAD)

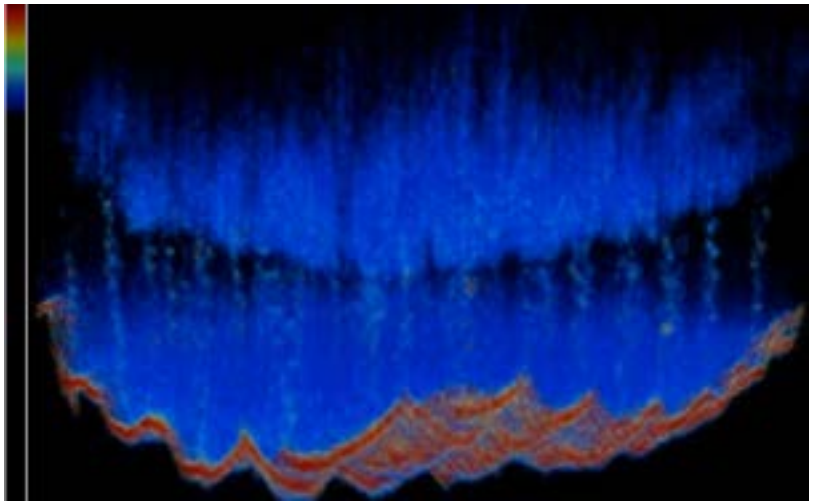


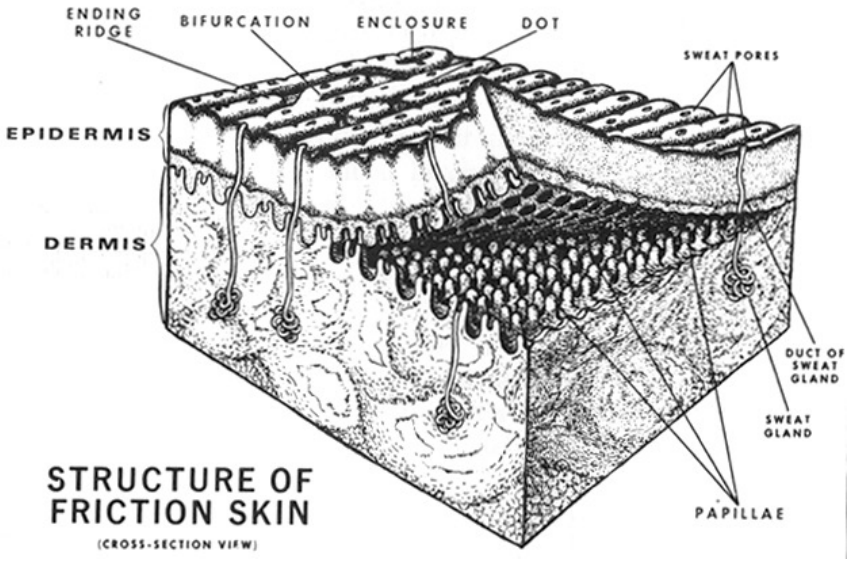
- Conclusion: Fingerphoto capture show better Presentation Attack Detection than capacitive sensors

Fingerprint Capture Device Security

Countermeasures

- Observation of the live skin properties
- Observation of the sweat glandes
- Sensor:
  - Optical Coherence Tomography (OCT)

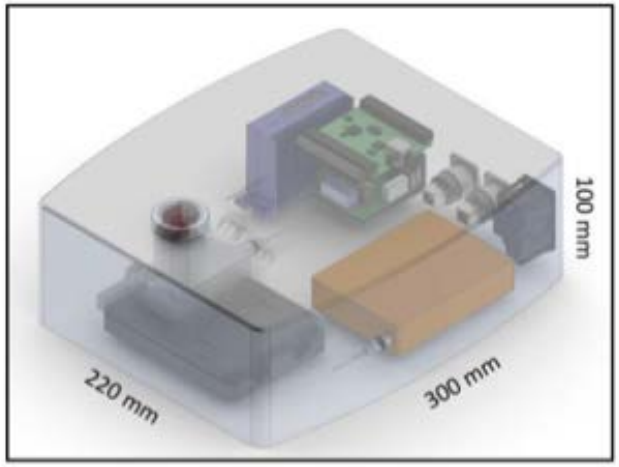




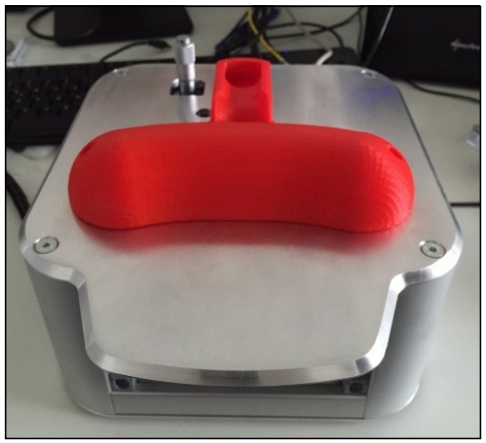
Fingerprint Capture Device Security

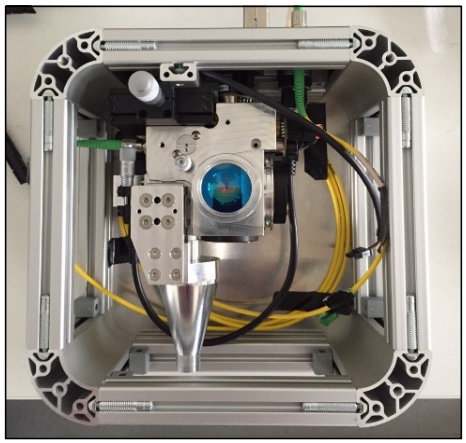
OCT

- at BSI-Germany
- Prototype for a high-end fingerprint sensor
- Requirements
  - PA robustness
  - Capture area: 20x20x6 mm
  - up to 3000 dpi
  - touchless scanning



Source: BSI



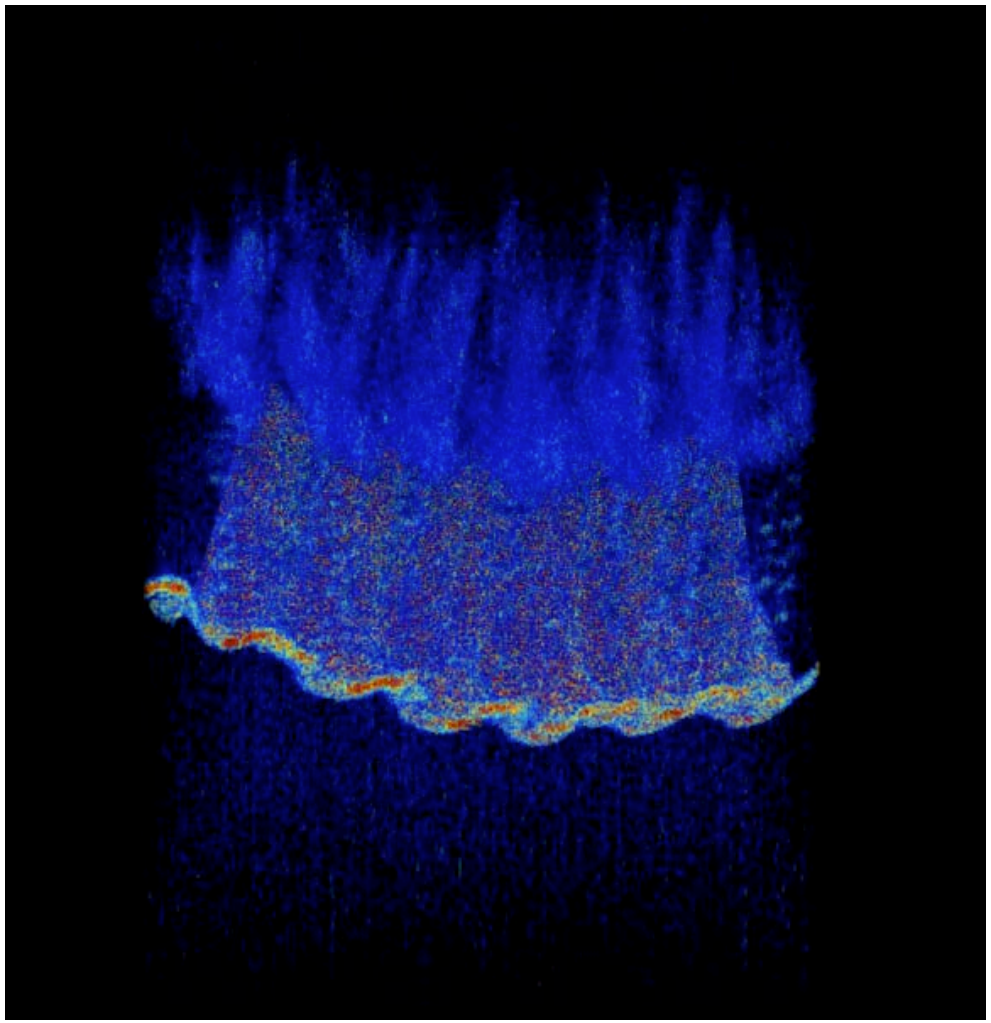


Source: BSI

Fingerprint Capture Device Security

OCT

- Visualization of sweat glands
  - good scan



Source: C. Sousedik, NTNU, 2016

Presentation Attack Detection

For finger applications:

What is an identity concealer?

Altered Fingerprint Detection in Forensics

Example for fingerprint alterations

- Left middle finger of Gus Winkler (Bank robber in the 1930s)

Image Source: H. Cummins, “Attempts to alter and obliterate finger-prints,” Journal of Criminal Law and Criminology, vol. 25, pp. 982–991, May 1935.

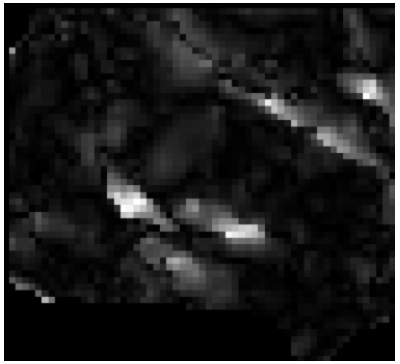
Altered Fingerprint Detection - Algorithms

- Feature: OFA and DOFTS
- Orientation Field Analysis (OFA)
  - Altered areas cause discontinuities in the OF [YoonJain2012]
- Differentials of Orientation Fields by Tensors in Scale (DOFTS)
  - Complex valued structure tensor [MikBig2014]

BonaFide fingerprint

Error map

Altered fingerprint



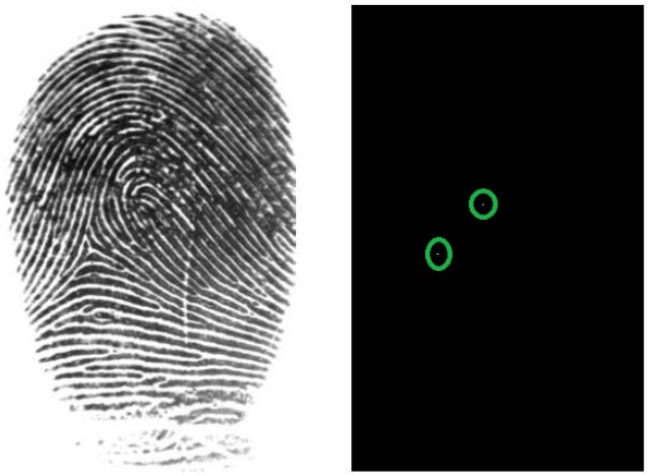
Error map

[YoonJain2012] S. Yoon, J. Feng, and A. Jain, “Altered fingerprints: Analysis and detection,” IEEE Transactions on Pattern Analysis and Machine Intelligence, vol. 34, no. 3, Mar. 2012

[MikBig2014] A. Mikaelyan and J. Bigun, “Symmetry assessment by finite expansion: application to forensic fingerprints,” in Proc. BIOSIG, Darmstadt, Germany, pp. 75–86. , (2014)

Altered Fingerprint Detection - Algorithms

- Feature: SPDA
- Singular Point Density Analysis [Ellingsg2014]
- using the Poincare´ index to detect noisy friction ridge areas



Bona fide fingerprint

altered fingerprint

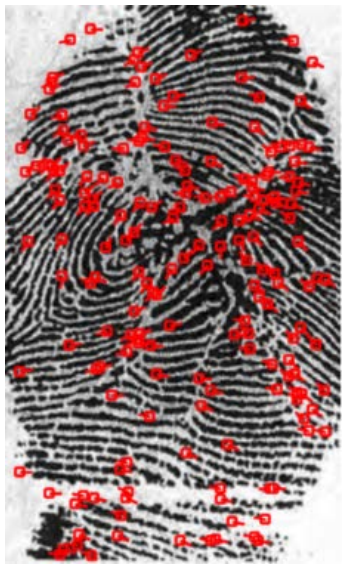
Poincare´ index response

[Ellingsg2014] J. Ellingsgaard, C. Sousedik, and C. Busch, “Detecting fingerprint alterations by orientation field and minutiae orientation analysis,” in Proc. IWBF, Valletta, Malta, (2014)

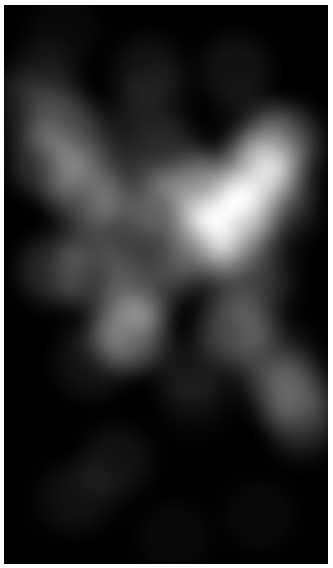
Altered Fingerprint Detection - Algorithms

- Feature: MOA
- Minutiae Orientation Analysis [Ellingsg2014]

Altered fingerprint



minutia distribution



density map

[Ellingsg2014] J. Ellingsgaard, C. Sousedik, and C. Busch, “Detecting fingerprint alterations by orientation field and minutiae orientation analysis,” in Proc. IWBF, Valletta, Malta, (2014)

Altered Fingerprint Detection - Testing

Database

- Dataset of Ellingsgaard et al. [Ellingsg2014]
  - Size: 116 altered fingerprints and 180 unaltered fingerprints
  - This data is not of sufficient size !

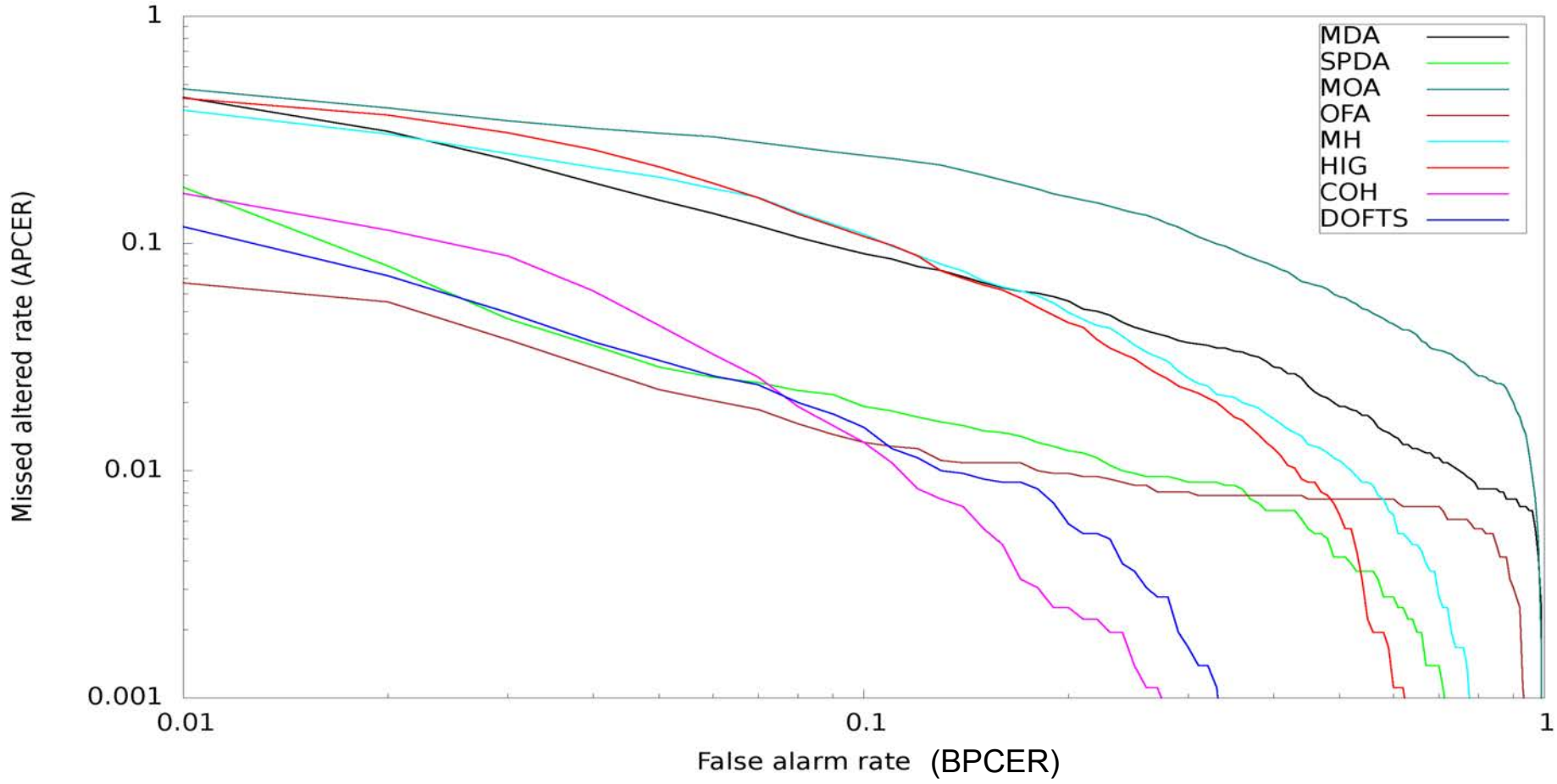
- Sources:
  - subset of GUC-100 (NTNU)
  - subset of Samischenko (Book)
  - subset of Brno
    (collection of fingerprints with dermatological diseases)
  - subset of NIST Special Database 14

[Ellingsg2014] J. Ellingsgaard, C. Sousedik, and C. Busch, “Detecting fingerprint alterations by orientation field and minutiae orientation analysis,” in Proc. IWBF, Valletta, Malta, (2014)

[Ellingsg2017] J. Ellingsgaard, C. Busch: "Altered Fingerprint Detection", in Handbook of Biometrics for Forensic Science, Springer, February, (2017)

Altered Fingerprint Detection - Testing

Results [Gottsch2015]



MDA = Minutia Distribution Analysis, SPDA = Singular Point Density Analysis, MOA = Minutia Orientation Analysis, OFA = Orientation Field Analysis, MH = Minutiae Histograms, HIG = Histograms of Invariant Gradients, COH = coherence, DOFTS = Differentials of Orientation Fields by Tensors in Scale,

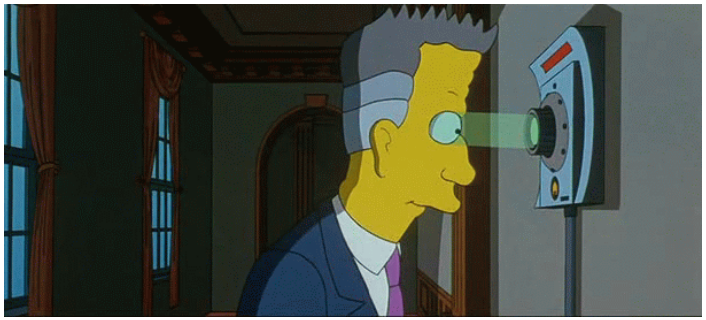
[Gottsch2015] C. Gottschlich, A. Mikaelyan, M. Olsen, J. Bigun, C. Busch: "Improving Fingerprint Alteration Detection", in Proceedings 9th International Symposium on Image and Signal Processing and Analysis,(2015)

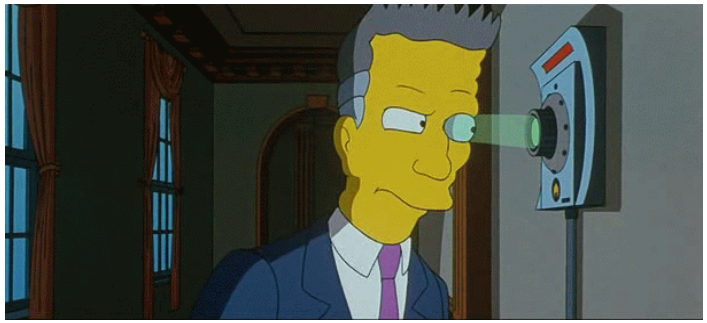
What about other modalities?
Presentation Attacks with Eye Artefacts

Eye Recognition Security

Presentation attacks

- in the Movie “The Simpsons” (2007)





PAD for Eye Recognition Security

Eye recognition study - 2015

- Presentation Attack Detection (PAD) videos on iPhone 5 S and Nokia 1020

- Method based on Eulerian Video Magnification (EVM)
  - Normalized Cumulative Phase Information

PAD for Eye Recognition Security

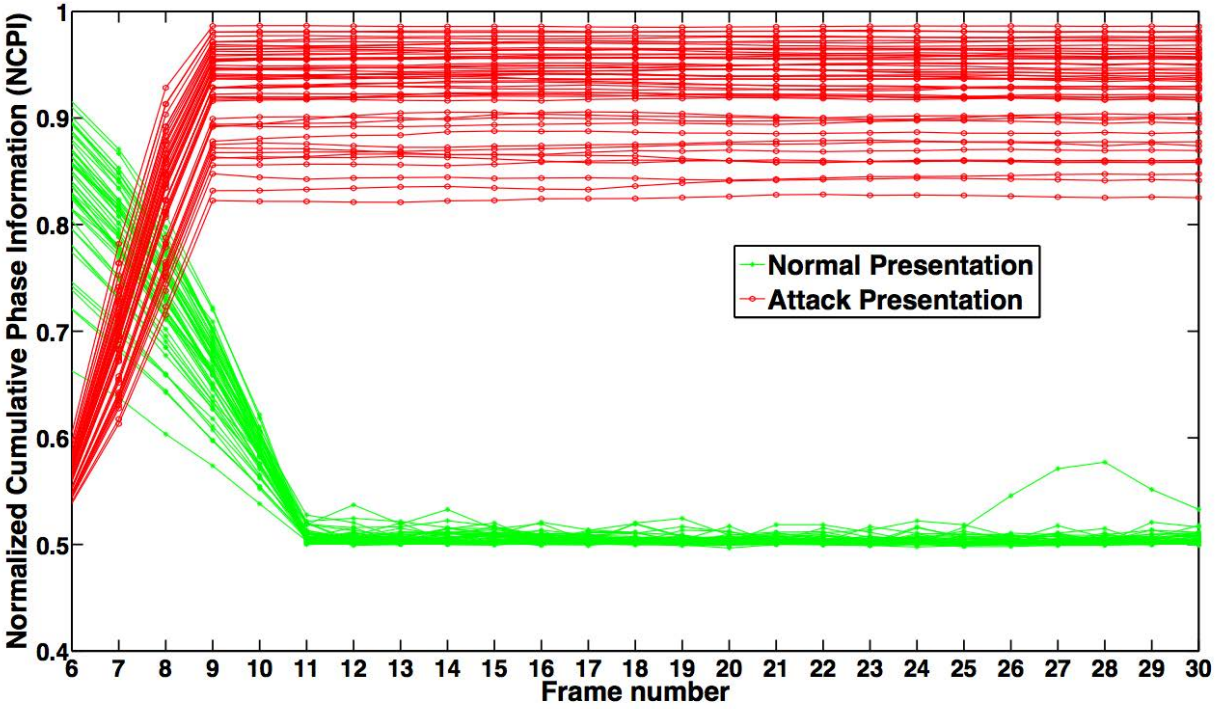
Method based on Eulerian Video Magnification (EVM)

[RRB2015] K. Raja, R. Raghavendra, C. Busch: "Video Presentation Attack Detection in Visible Spectrum Iris Recognition Using Magnified Phase Information",
in IEEE Transactions on Information Forensics and Security (TIFS), June, (2015)

PAD for Eye Recognition Security

Eye recognition study - 2015

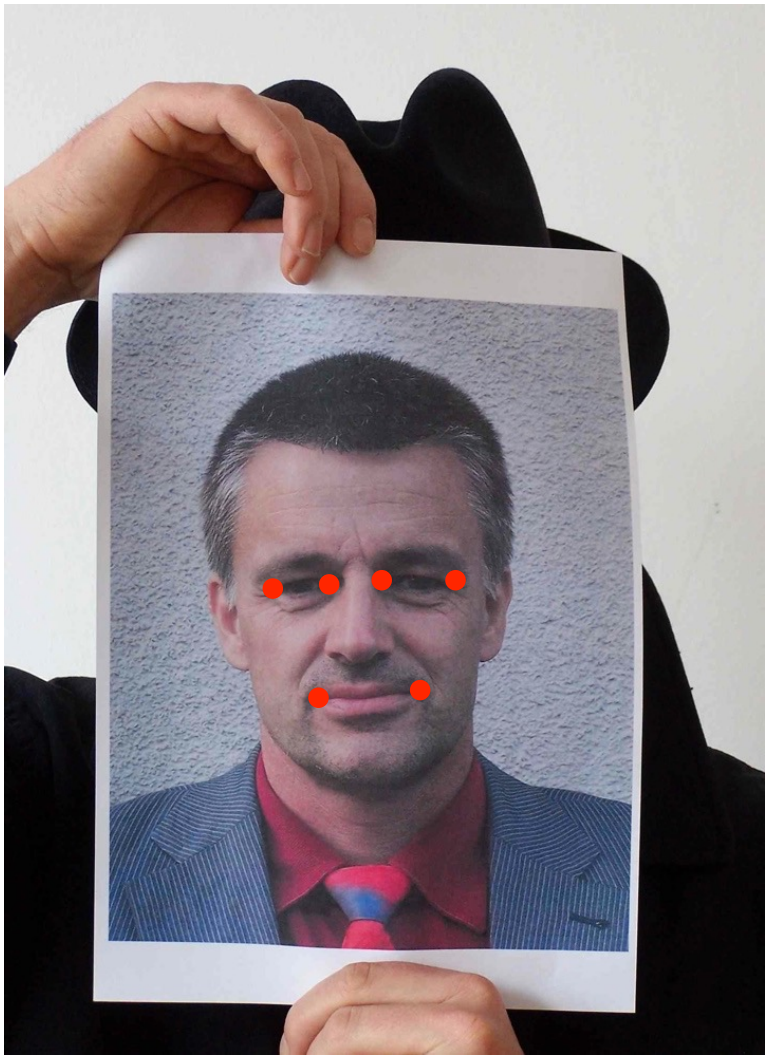
- Method based on Eulerian Video Magnification (EVM)
  - Normalized Cumulative Phase Information
- Zero Error Rates:
  - APCER = 0 %
  - BPCER = 0 %



[RRB2015] K. Raja, R. Raghavendra, C. Busch: "Video Presentation Attack Detection in Visible Spectrum Iris Recognition Using Magnified Phase Information", in IEEE Transactions on Information Forensics and Security (TIFS), (2015)

Widely used at borders is Face Recognition!
Presentation Attacks with Face Artefacts

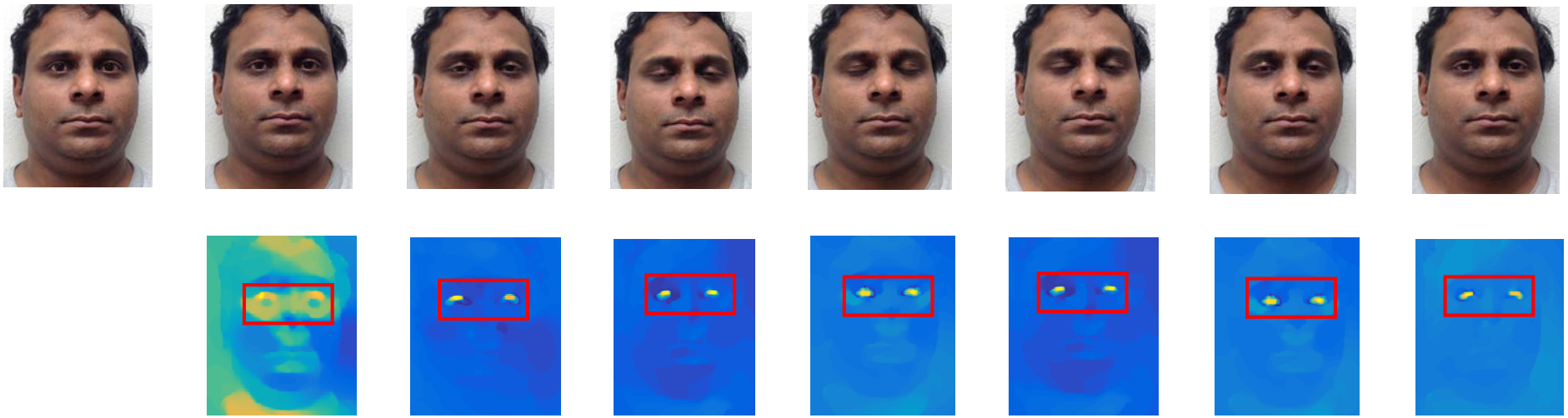
Face Presentation Attacks



Face Presentation Attack Detection

Hardware based

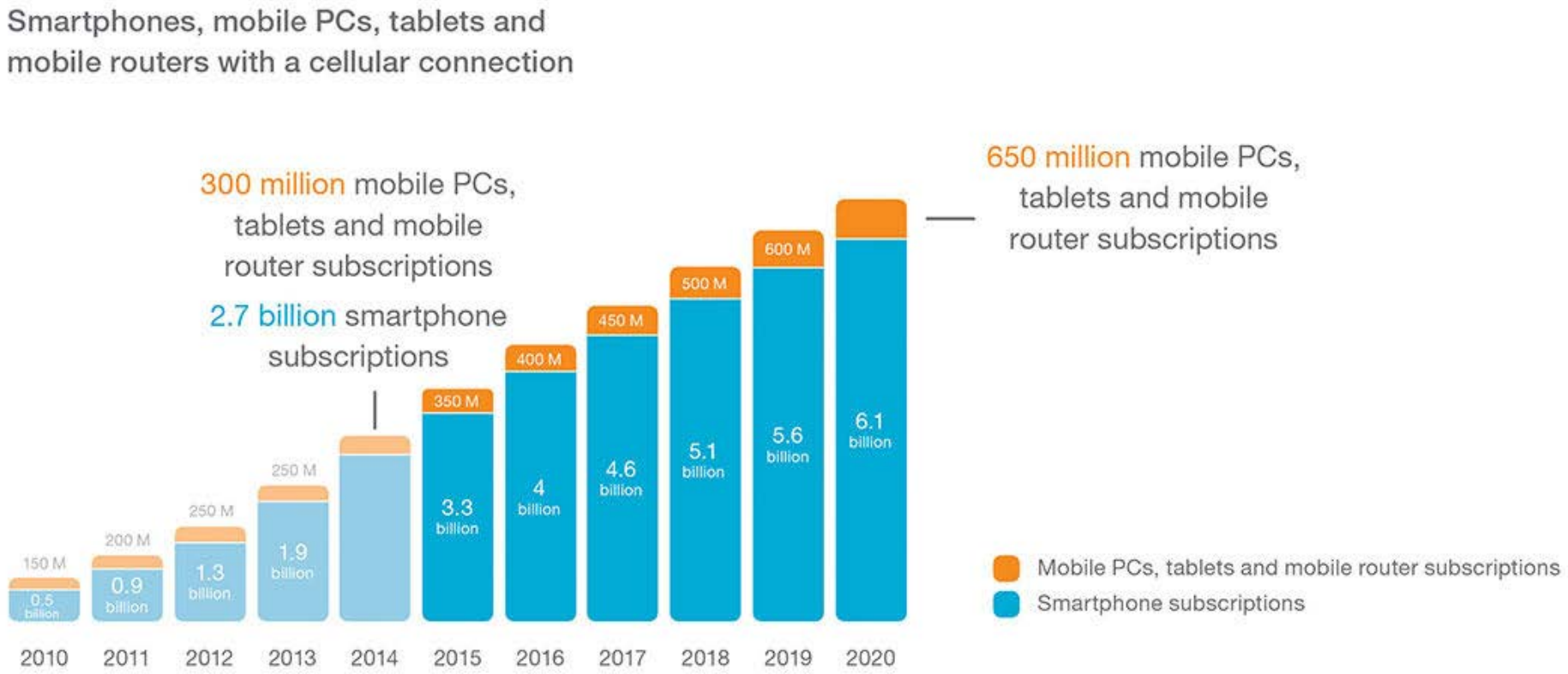
- Challenge Response
  - challenge the subject instructions and then compare the response to reference model for a bona fide behaviour
    - Instructions to the user to change head pose.
    - Reads user's lips after playing audio tracks of words or numbers.
- Blink detection



Face Recognition in unsupervised environments

Smartphone Deployment

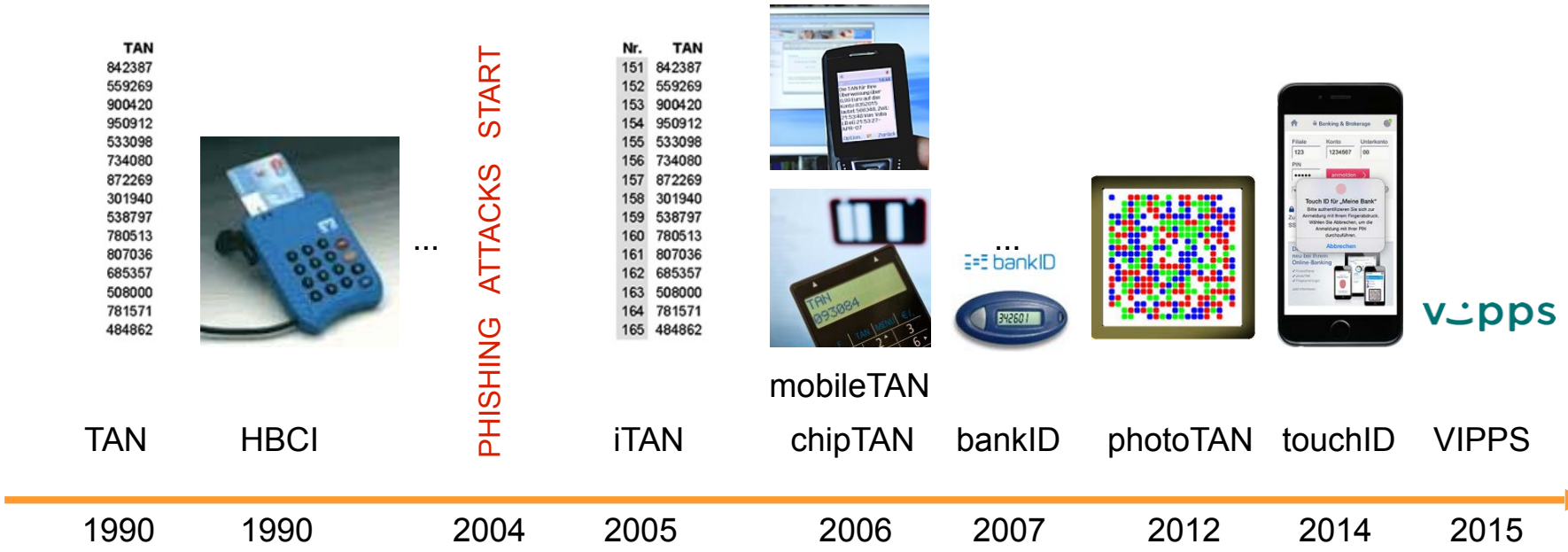
The Smartphone as personal device



Source: https://thenextweb.com/insider/2014/11/18/2020-90-worlds-population-aged-6-will-mobile-phone-report/

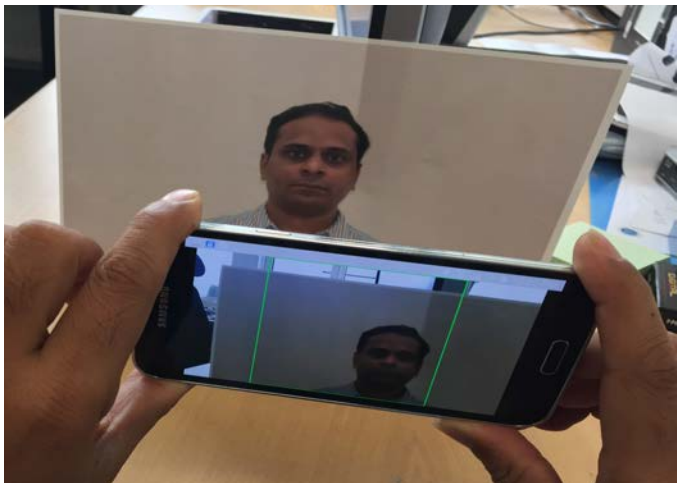
Access Control in the Banking Environment

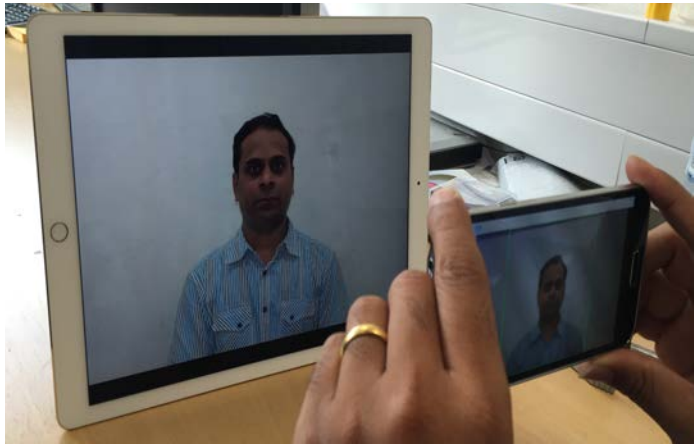
A European perspective



Inspired by: BdB (2015)

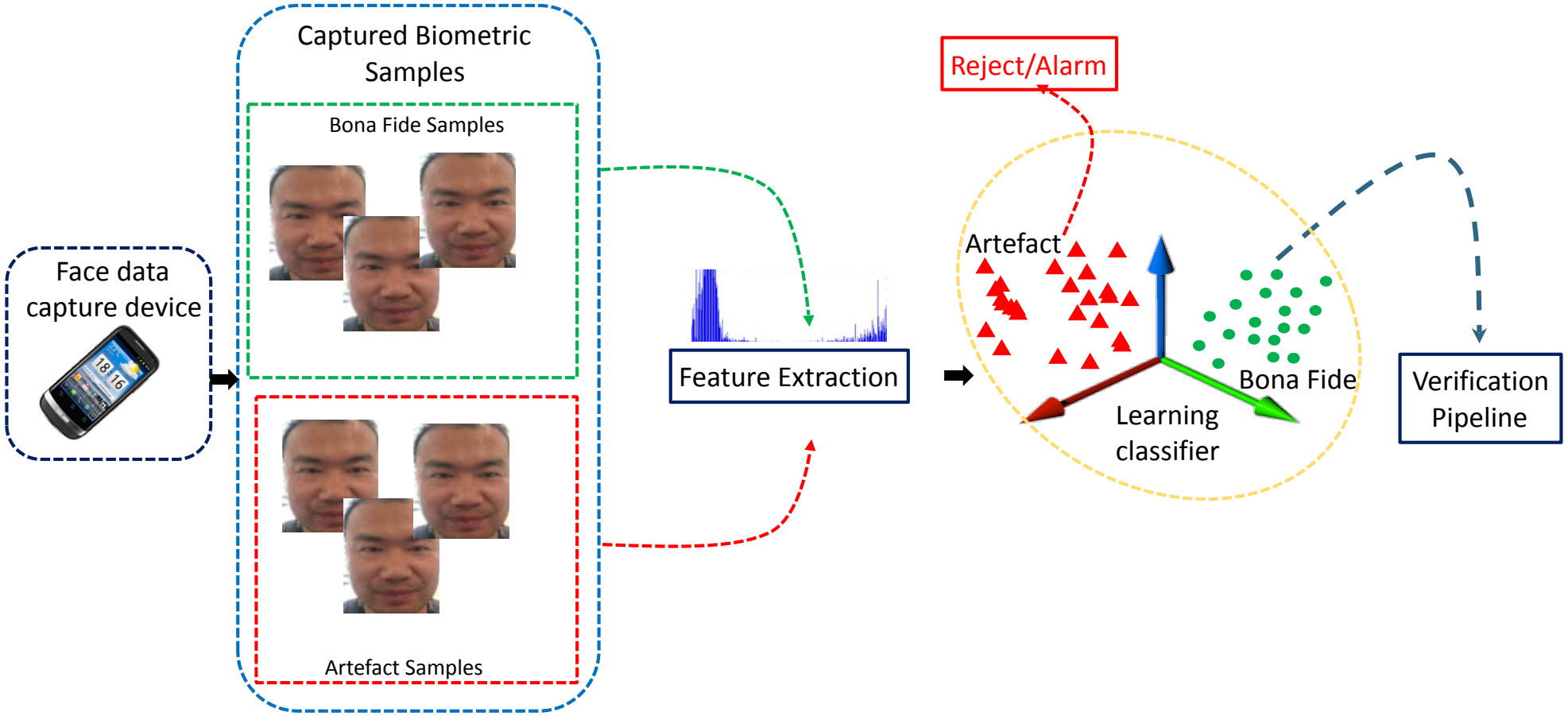
Smartphone - Presentation Attacks





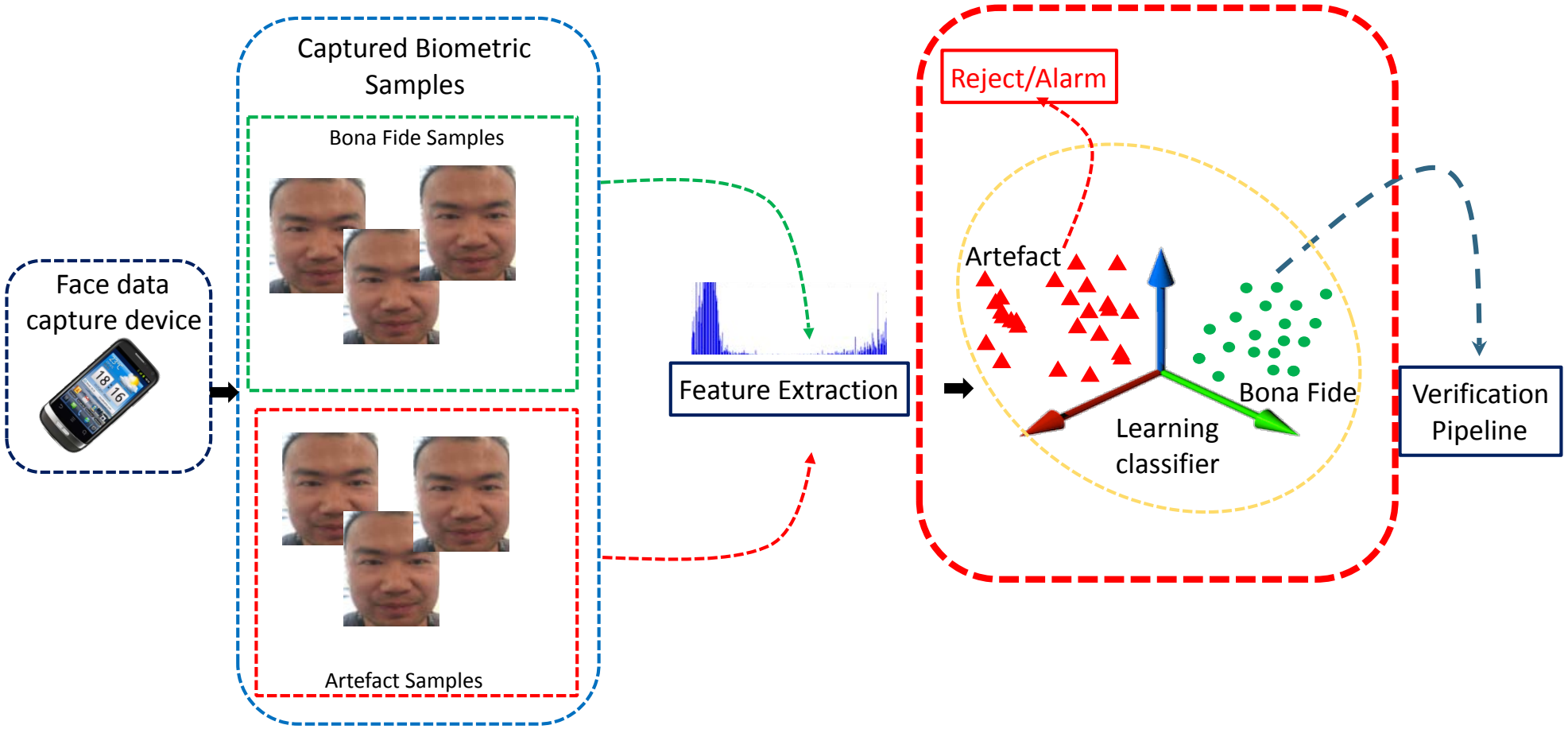
Smartphone - Presentation Attack Detection

- Augmenting the processing pipeline



Smartphone - Face PAD

- Augmenting the processing pipeline



Smartphone - Face PAD

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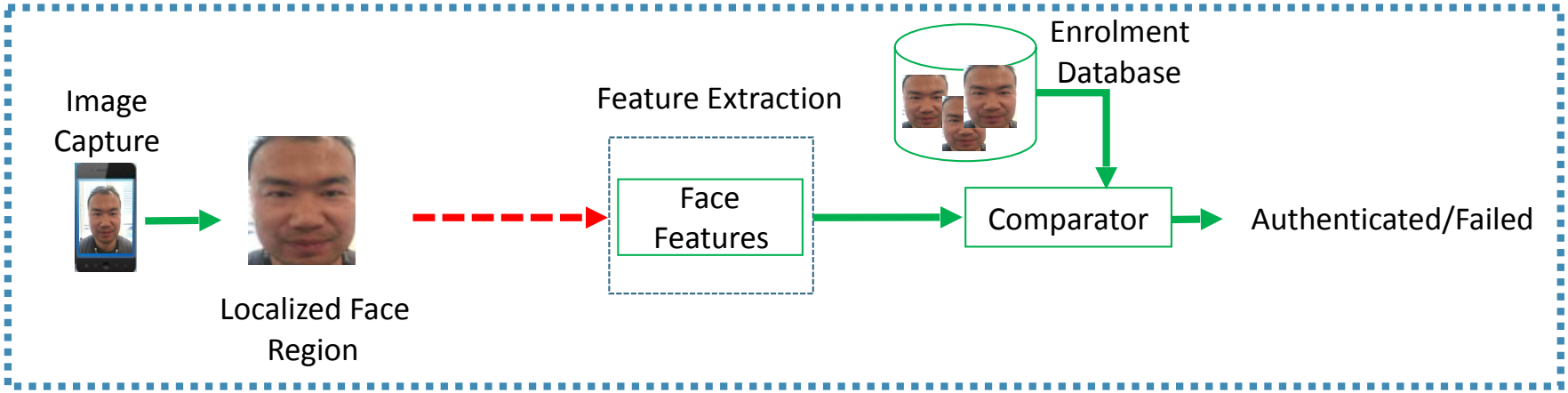

[Wasnik2016] P. Wasnik, K. Raja, R. Raghavendra, and C. Busch. “Presentation attack detection in face biometric systems using raw sensor data from smartphones”. In Proc. 12th International Conference on Signal-Image Technology & Internet-Based Systems (SITIS), (2016)

Smartphone - Face PAD

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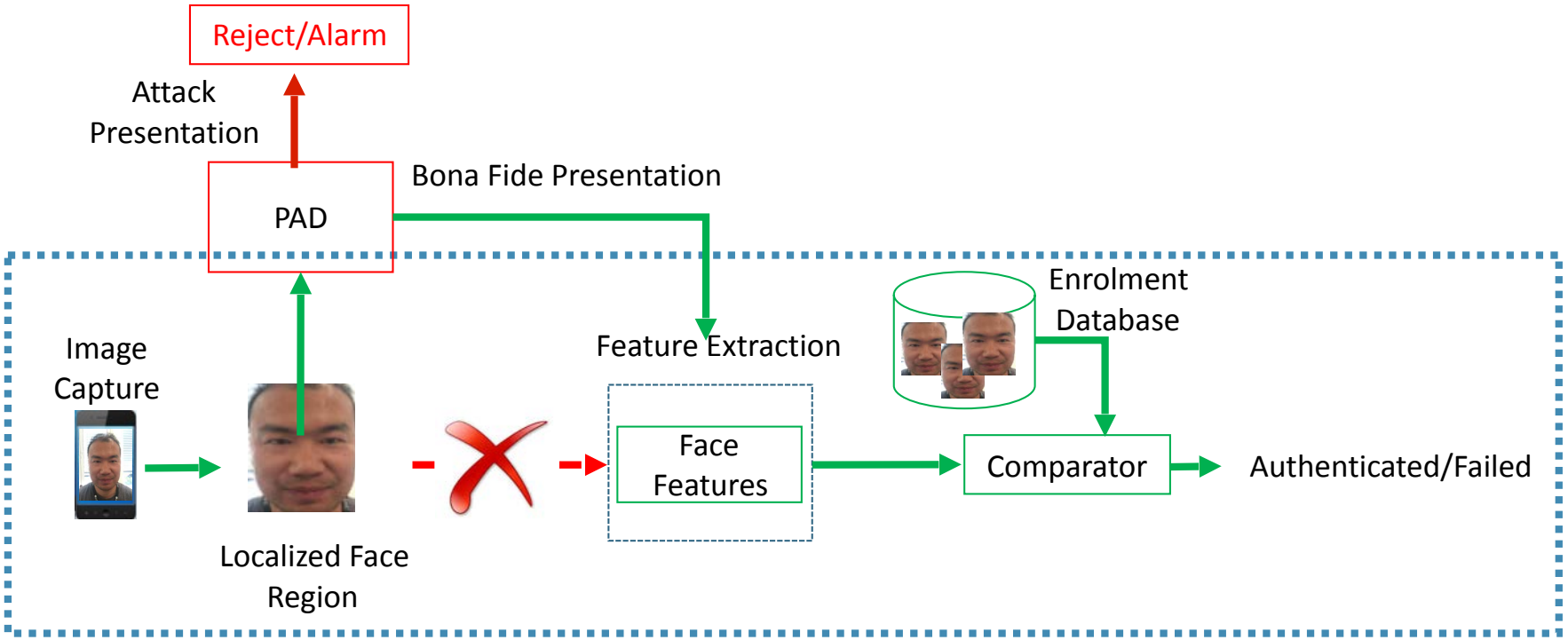

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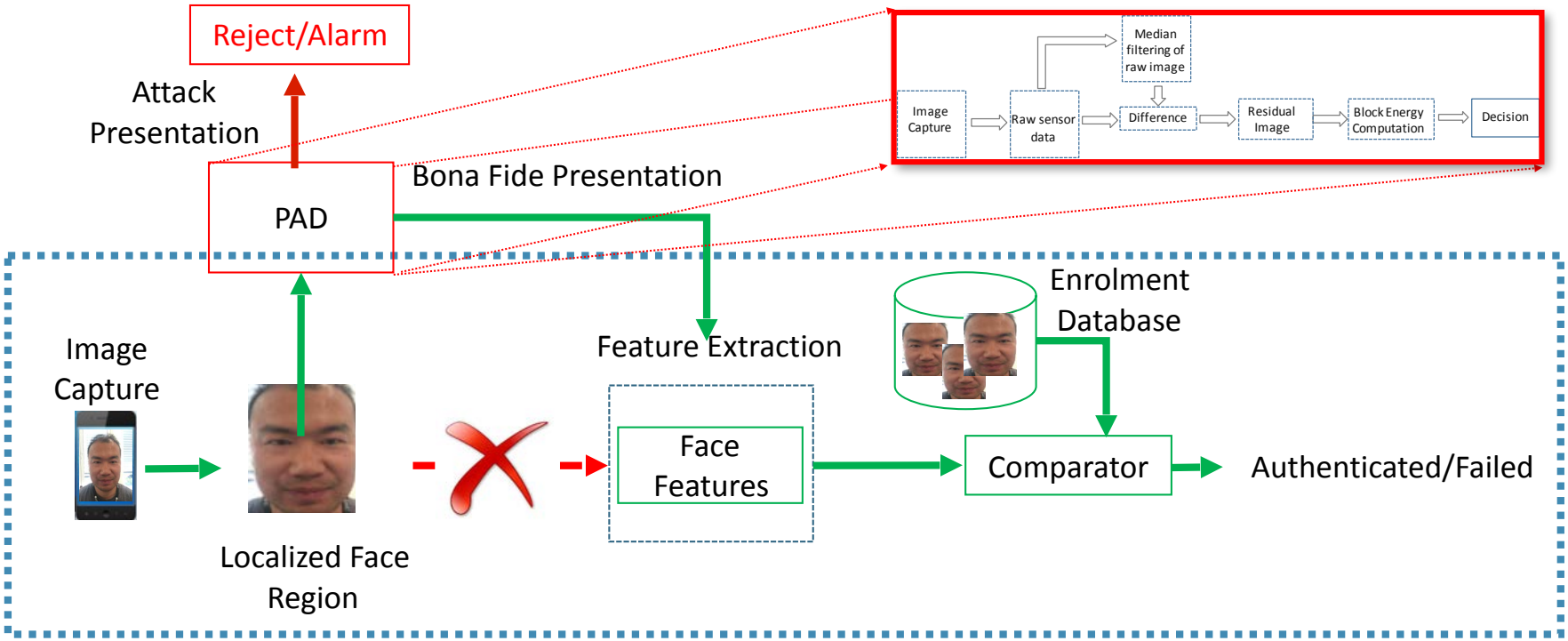
Smartphone - Face PAD

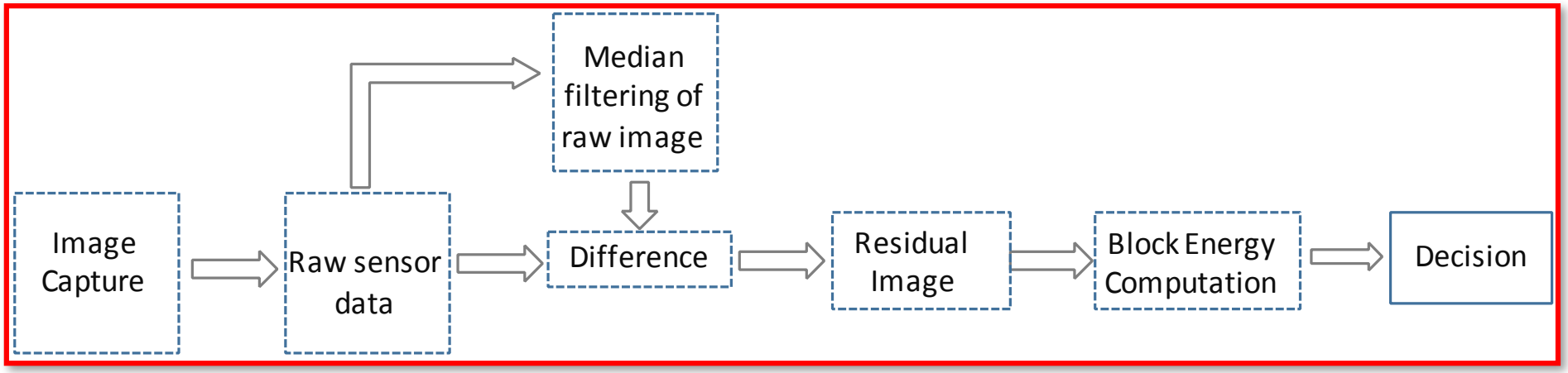
- Augmenting the processing pipeline



[Wasnik2016] P. Wasnik, K. Raja, R. Raghavendra, and C. Busch. “Presentation attack detection in face biometric systems using raw sensor data from smartphones”. In Proc. 12th International Conference on Signal-Image Technology & Internet-Based Systems (SITIS), (2016)

Smartphone - Face PAD

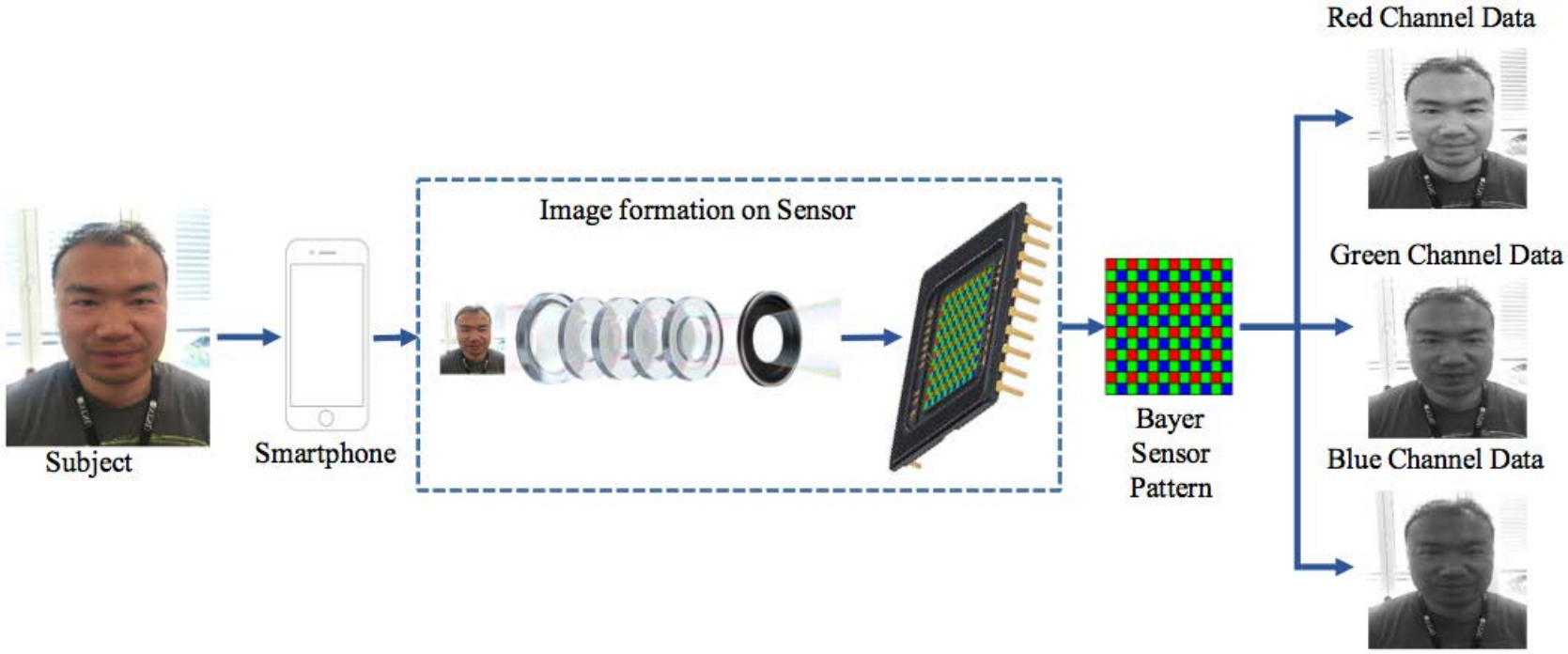
- The Presentation Attack Detection subsystem



[Wasnik2016] P. Wasnik, K. Raja, R. Raghavendra, and C. Busch. “Presentation attack detection in face biometric systems using raw sensor data from smartphones”. In Proc. 12th International Conference on Signal-Image Technology & Internet-Based Systems (SITIS), (2016)

Smartphone - Face PAD

- The biometric sample



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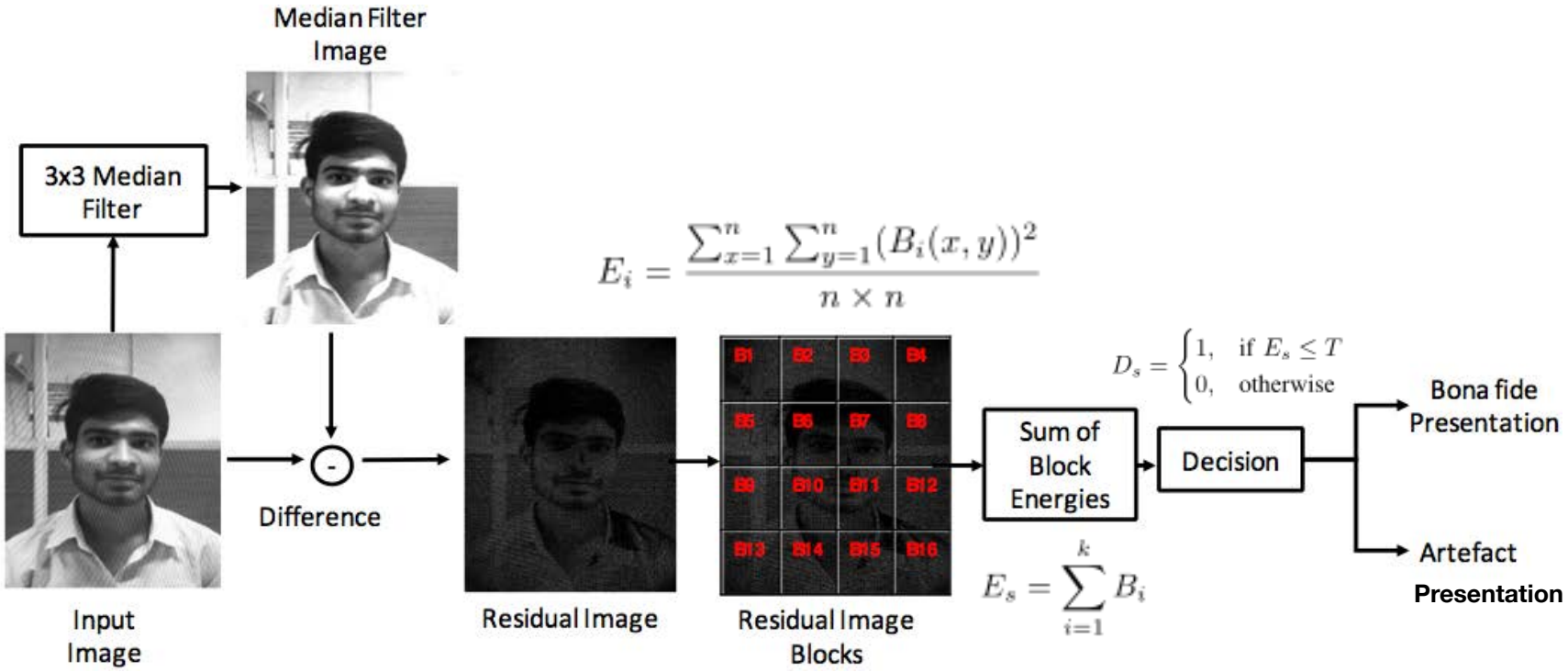
Smartphone - Face PAD

- Channel based processing



[Wasnik2016] P. Wasnik, K. Raja, R. Raghavendra, and C. Busch. “Presentation attack detection in face biometric systems using raw sensor data from smartphones”. In Proc. 12th International Conference on Signal-Image Technology & Internet-Based Systems (SITIS), (2016)

Smartphone - Face PAD

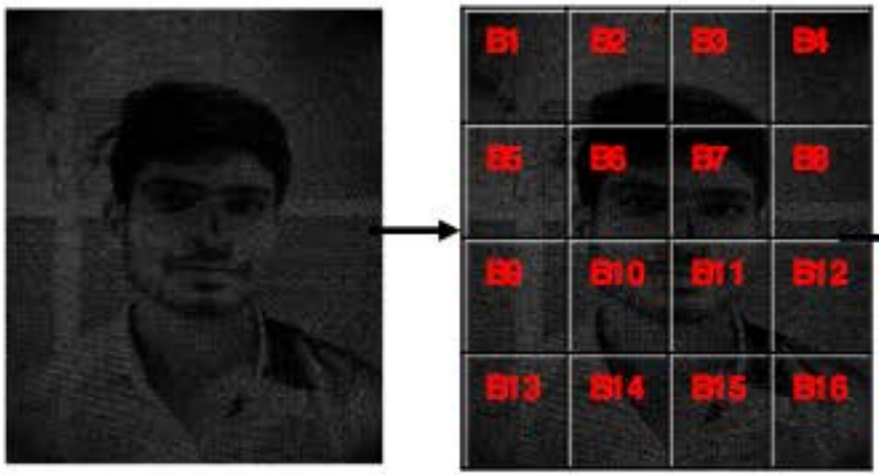
- Residual image computation

$$E_i = \frac{\sum_{x=1}^{n} \sum_{y=1}^{n} (B_i(x,y))^2}{n \times n}$$

$$E_s = \sum_{i=1}^{k} B_i$$

$$D_s = \begin{cases} 1, & \text{if } E_s \leq T \\ 0, & \text{otherwise} \end{cases}$$

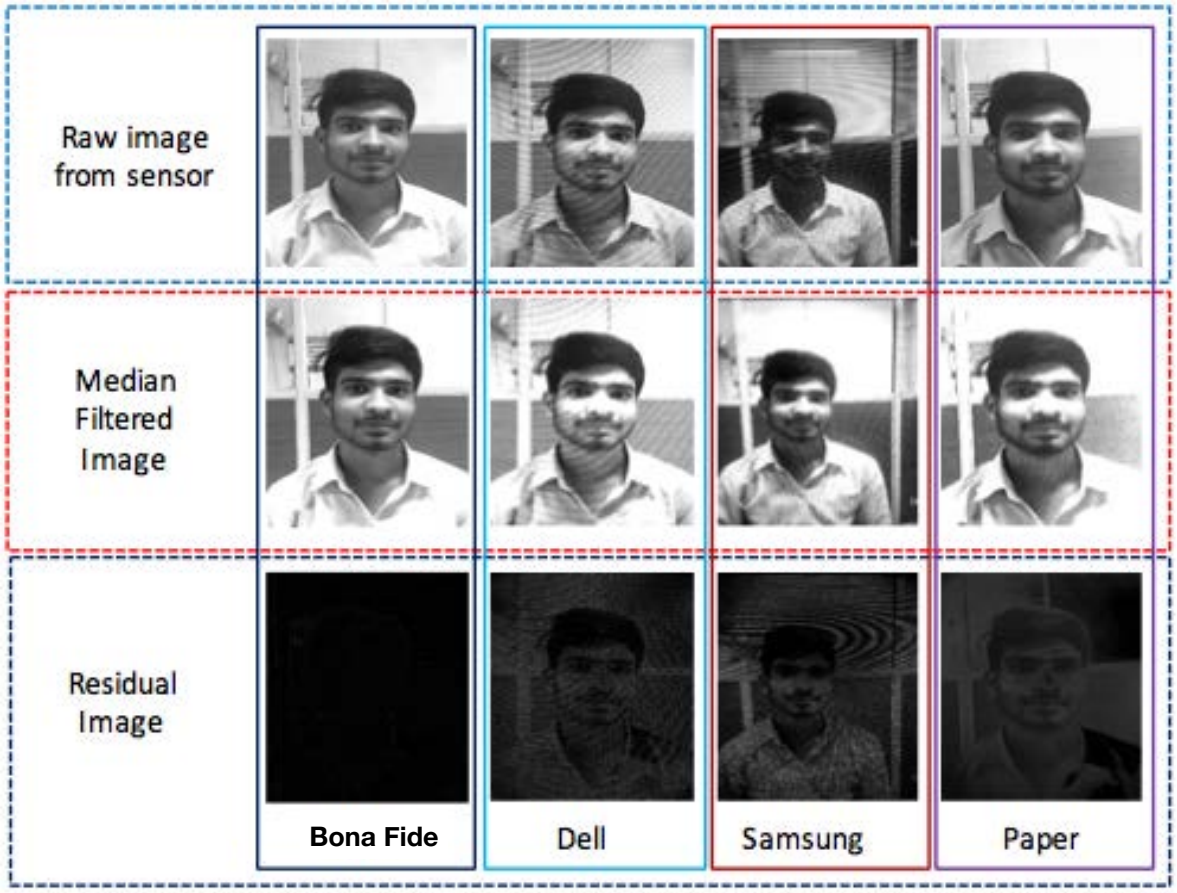
$$D = \begin{cases} 1, & \text{if } majority\{D_r, D_g, D_b\} = 1 \\ 0, & \text{otherwise} \end{cases}$$



[Wasnik2016] P. Wasnik, K. Raja, R. Raghavendra, and C. Busch. “Presentation attack detection in face biometric systems using raw sensor data from smartphones”. In Proc. 12th International Conference on Signal-Image Technology & Internet-Based Systems (SITIS), (2016)

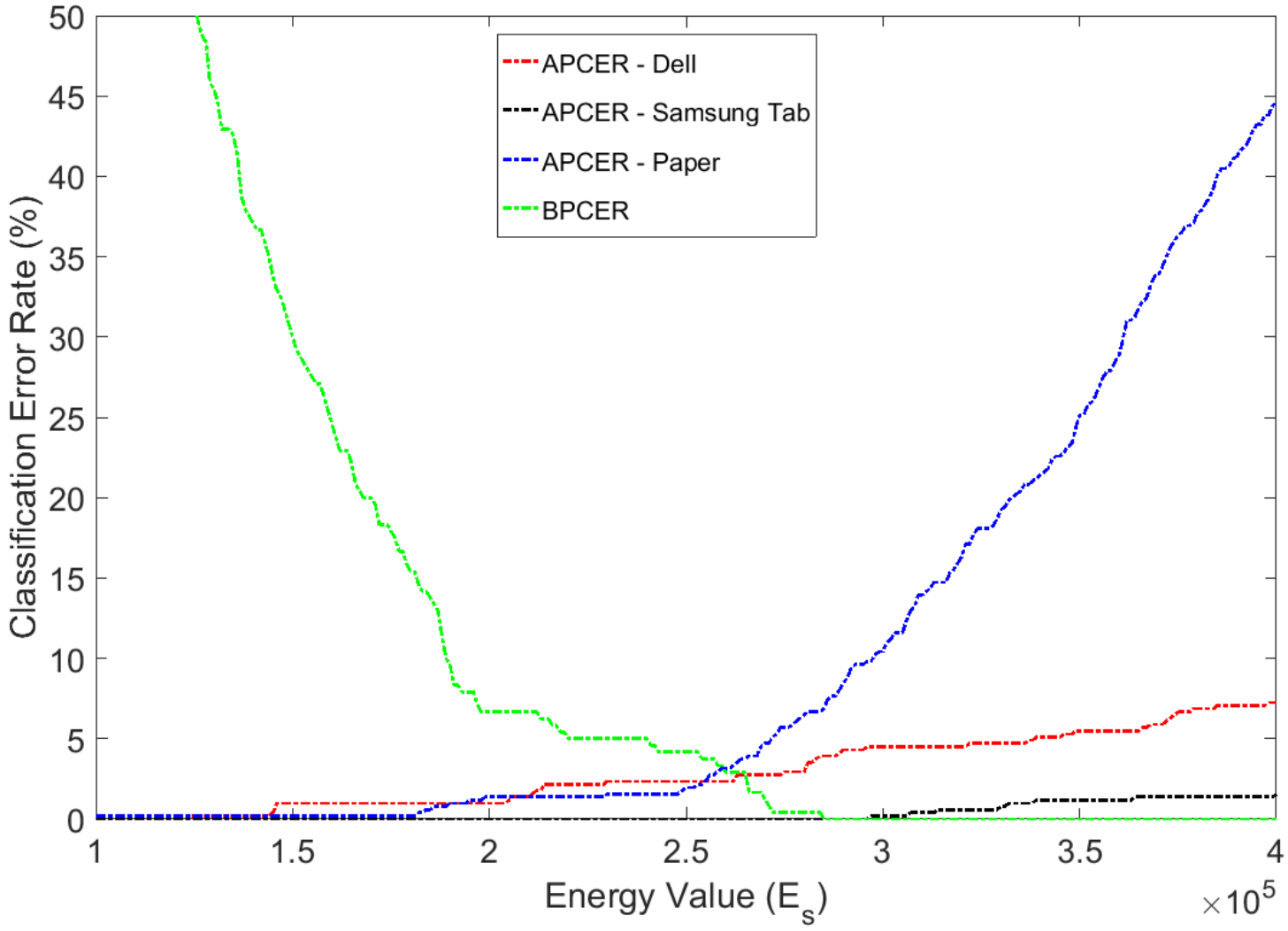
Smartphone - Face PAD

- Residual image computation



Smartphone PAD – Results

Classification Error Rates



Smartphone PAD – Results Majority Voting

Classification Error Rates

- Error rates for different thresholds of with majority voting on all three channels

Threshold	Paper			Dell			Samsung		
	BPCER (%)	APCER (%)	ACER (%)	BPCER (%)	APCER (%)	ACER (%)	BPCER (%)	APCER (%)	ACER (%)
200000	3.33	0.32	1.83	3.33	3.23	3.28	3.33	0.00	1.67
210000	3.33	0.32	1.83	3.33	3.23	3.28	3.33	0.00	1.67
220000	3.33	0.32	1.83	3.33	3.23	3.28	3.33	0.00	1.67
230000	2.67	0.65	1.66	2.67	4.19	3.43	2.67	0.00	1.33
240000	2.67	0.65	1.66	2.67	4.19	3.43	2.67	0.00	1.33
250000	2.00	1.29	1.65	2.00	5.48	3.74	2.00	0.00	1.00
260000	2.00	2.27	2.13	2.00	5.48	3.74	2.00	0.00	1.00
270000	2.00	3.24	2.62	2.00	5.48	3.74	2.00	0.00	1.00
280000	2.00	4.21	3.10	2.00	6.13	4.06	2.00	0.00	1.00
290000	1.33	8.41	4.87	1.33	6.77	4.05	1.33	0.00	0.67
300000	1.33	9.71	5.52	1.33	6.77	4.05	1.33	0.00	0.67

[Wasnik2016] P. Wasnik, K. Raja, R. Raghavendra, and C. Busch. “Presentation attack detection in face biometric systems using raw sensor data from smartphones”. In Proc. 12th International Conference on Signal-Image Technology & Internet-Based Systems (SITIS), (2016)

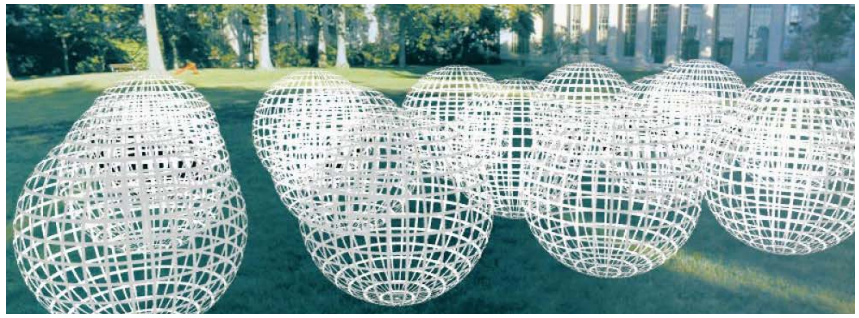
PAD – based on Depth Information

Light-field camera recently proposed for PAD

- panoptic or directional camera

Why light-field camera?

- Multiple focus/depth images in one shot.
- No need to adjust the lens to set focus.
- Portable and hand-held, low cost.

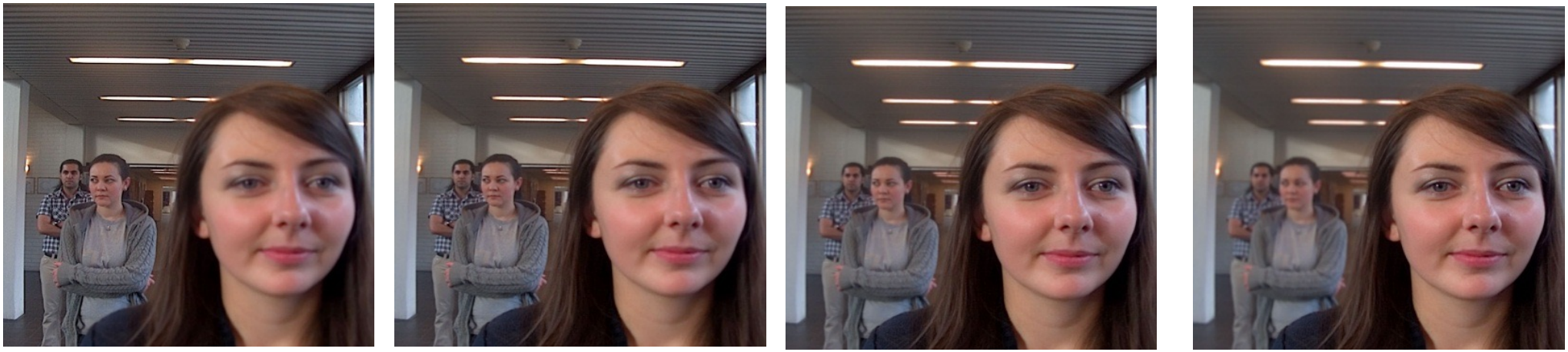


$$P(\theta, \phi, \lambda, t, Vx, Vy, Vz)$$

[Raghu2015] R. Raghavendra, K.B. Raja, and C. Busch: "Presentation Attack Detection for Face Recognition using Light Field Camera", in IEEE Transactions on Image Processing, vol. 24, no. 3, pp. 1060–1075, (2015)

PAD – based on Depth Information

Example of light-field imaging (LYTRO)



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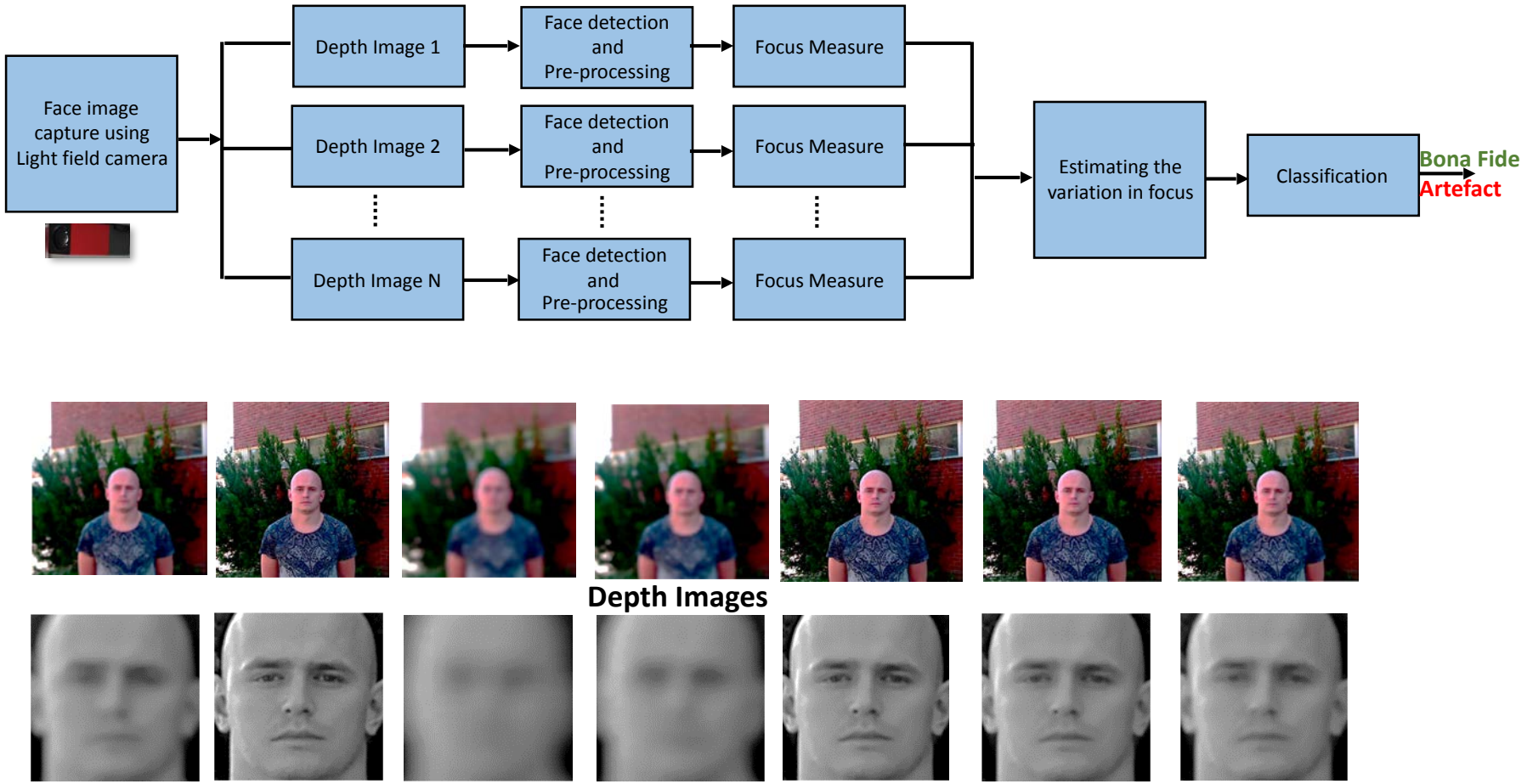
PAD – based on Depth Information

Presentation Attack Detection



[Raghu2015] R. Raghavendra, K.B. Raja, and C. Busch: "Presentation Attack Detection for Face Recognition using Light Field Camera", in IEEE Transactions on Image Processing, vol. 24, no. 3, pp. 1060–1075, (2015)

PAD – based on Variable Focus

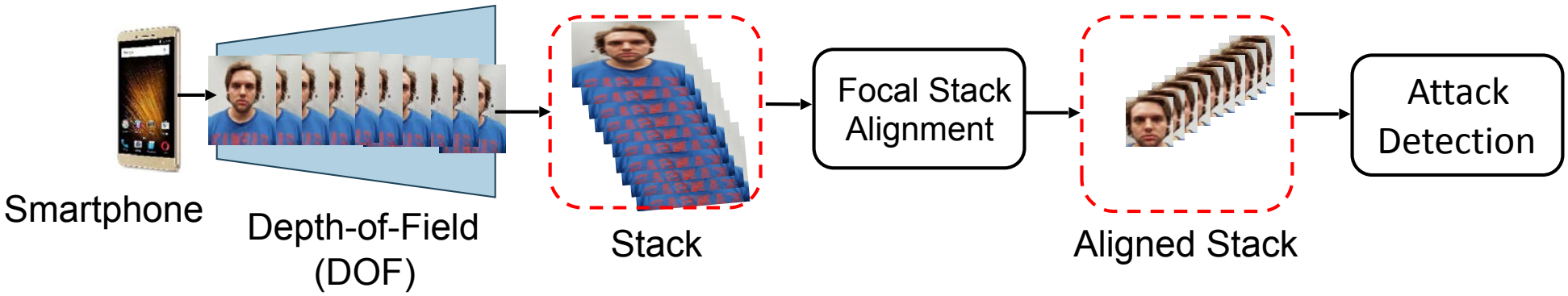
A recently proposed PA detection - for Smartphones

- based on variable camera focus

[Raja2017] K. Raja, P. Wasnik, R. Raghavendra, C. Busch: "Robust Face Presentation Attack Detection On Smartphones: An Approach Based on Variable Focus", in Proceedings of International Joint Conference on Biometrics (IJCB 2017), Denver, Colorado, October 1-4, (2017)

PAD – based on Variable Focus

- Stack alignment



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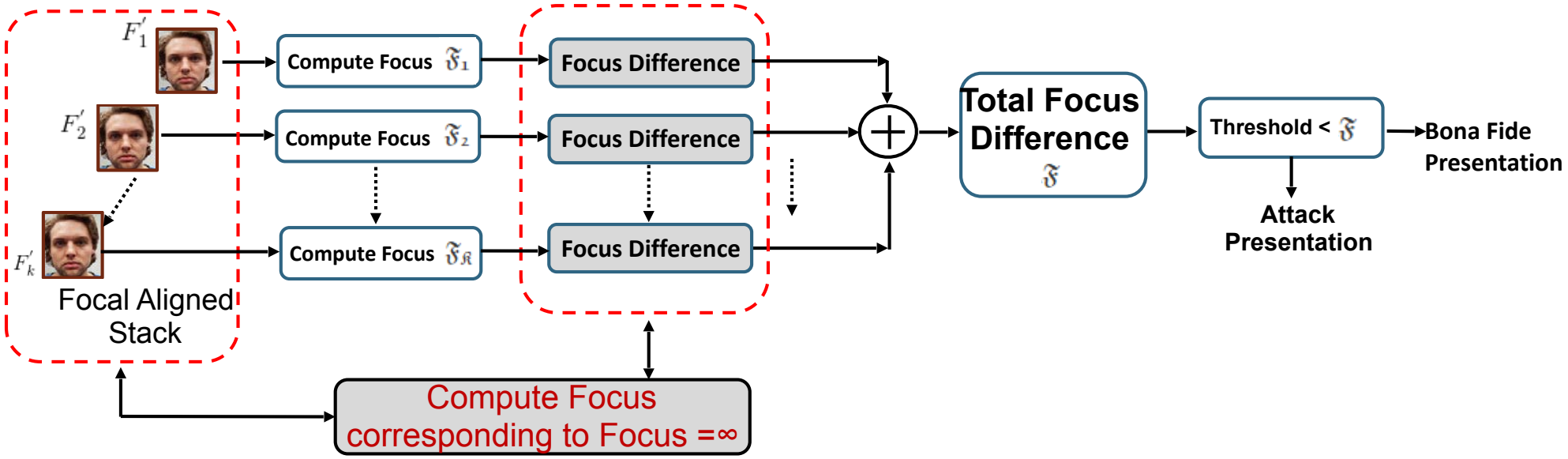
PAD – based on Variable Focus

- The proposed approach



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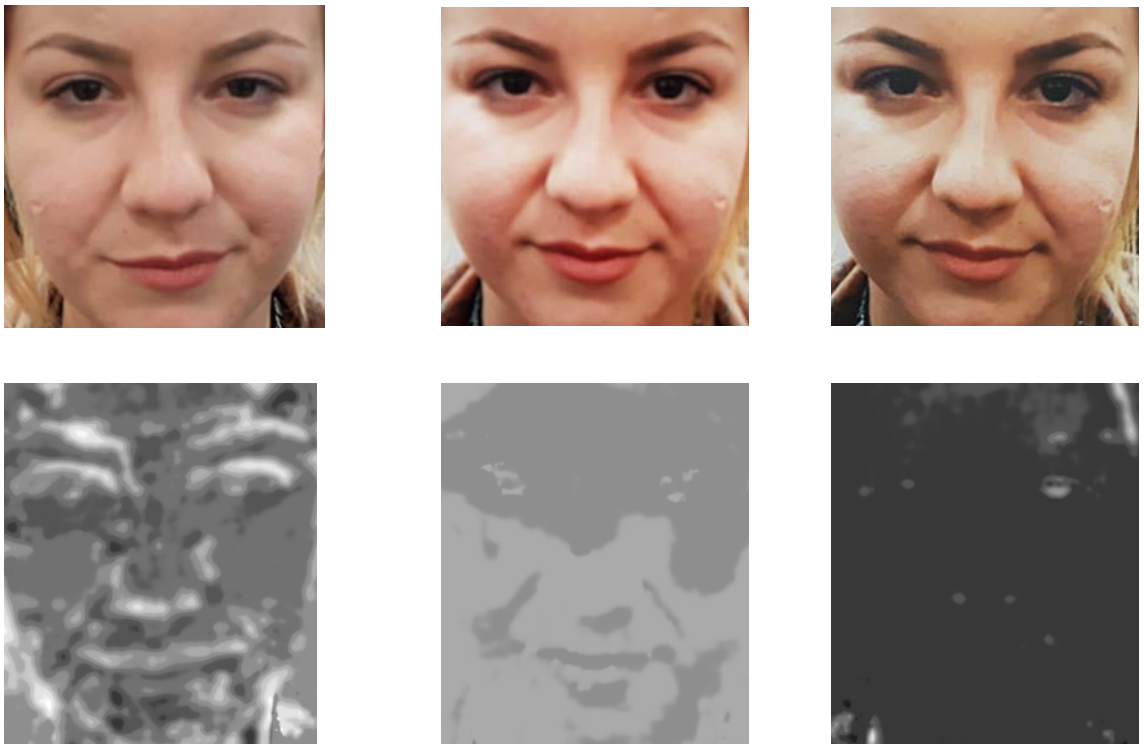
PAD – based on Variable Focus

- Focus difference



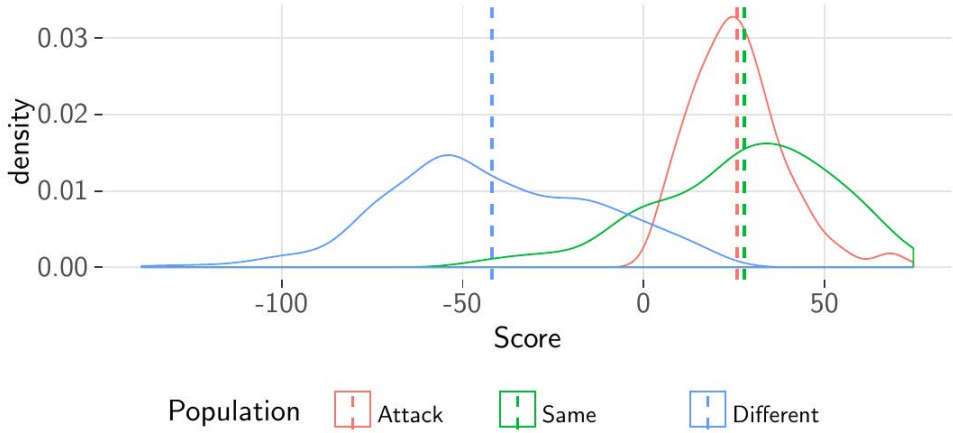
Bona fide | Attack - Screen | Attack - Print

Presentation Attack Detection - Testing

Definition of full system vulnerability metric w.r.t attacks

- **Impostor attack presentation match rate (IAPMR)**
  <in a full-system evaluation of a verification system> the proportion of impostor attack presentation using the same PAI species in which the target reference is matched

Source: ISO/IEC 30107-3



- **Concealer attack presentation non-match rate (CAPNMR)**
  in a full-system evaluation of a verification system, the proportion of concealer attack presentation using the same PAI species in which the target reference is not matched.

Source: ISO/IEC 30107-3

PAD – based on Variable Focus

Evaluation

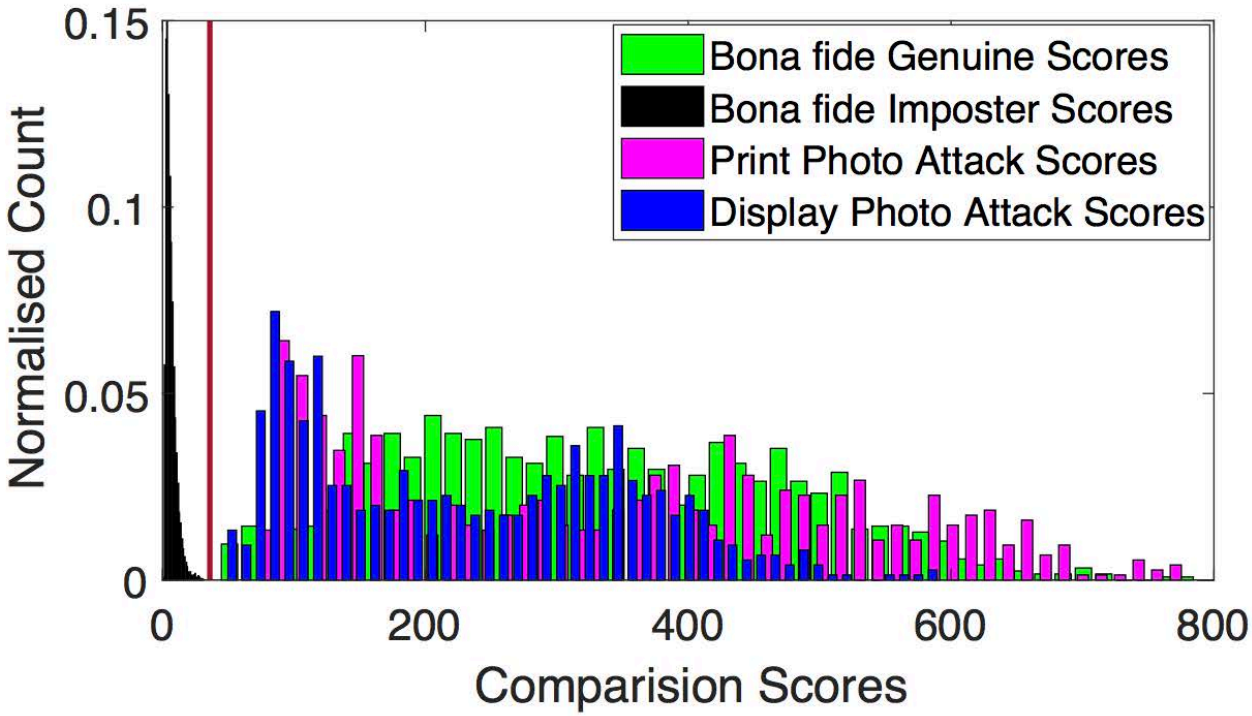
- Vulnerability analysis of commercial face recognition system
- Impostor Attack Presentation Match Rate (IAPMR)

Method	PAI	IAPMR @ FAR =	
		0.1%	0.01%
Neurotech	Display Monitor	100	100
	Laptop	100	100
	iPad-Pro	100	100
	iPhone 6S	100	100
	Printed-Photo	100	100

PAD – based on Variable Focus

Evaluation

- Vulnerability analysis of commercial face recognition system
- Comparison score distribution



PAD – based on Variable Focus

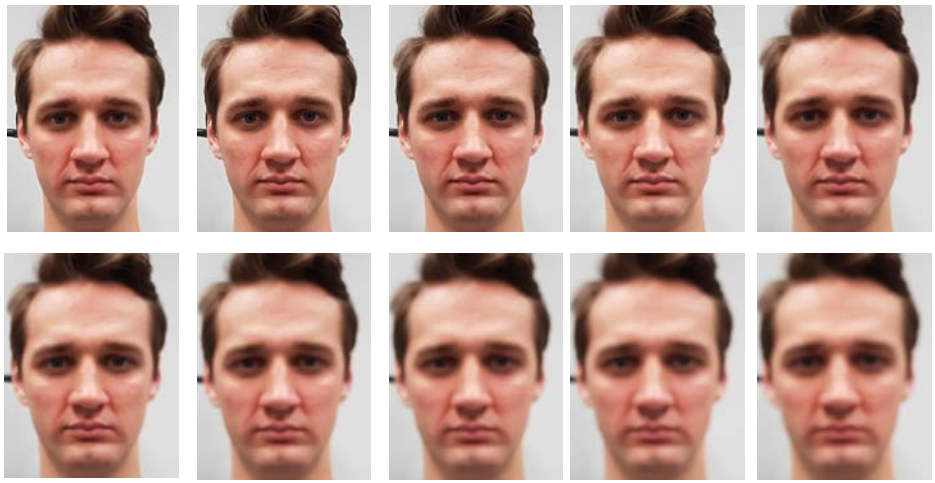
Evaluation

- Qualitative detection performance
- Proposed approach for various displays Presentation Attack Instruments (PAI)

PAI	EER (%)	BPCER @ APCER =	
		5 %	10 %
Display Monitor	4.00	2.67	1.33
Print Photo	1.33	0.00	0.00
Laptop Screen	1.33	0.00	0.00
iPad-Pro	1.33	0.00	0.00
iPhone 6S	0.00	0.00	0.00

PAD – based on Variable Focus

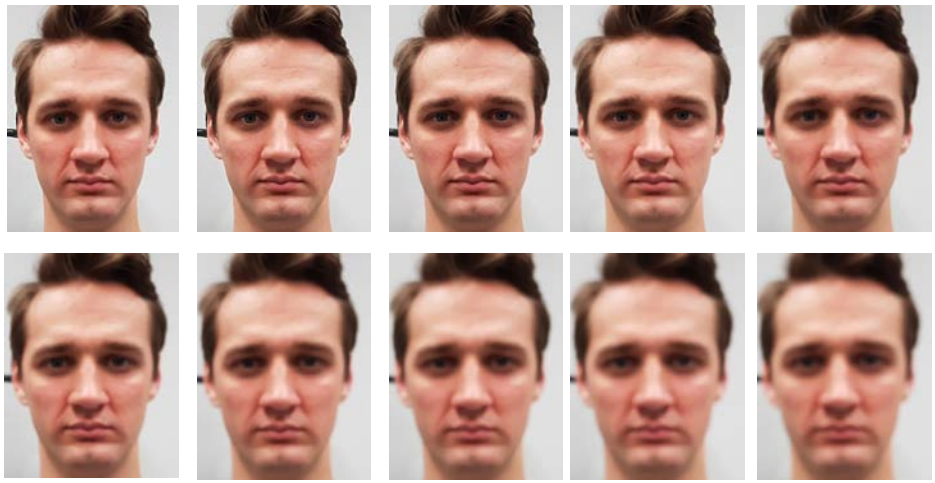
Reconstructing a 3D model



[Raja2017] K. Raja, P. Wasnik, R. Raghavendra, C. Busch: "Robust Face Presentation Attack Detection On Smartphones: An Approach Based on Variable Focus", in Proceedings of International Joint Conference on Biometrics (IJCB 2017), Denver, Colorado, October 1-4, (2017)

PAD – based on Variable Focus

Reconstructing a 3D model



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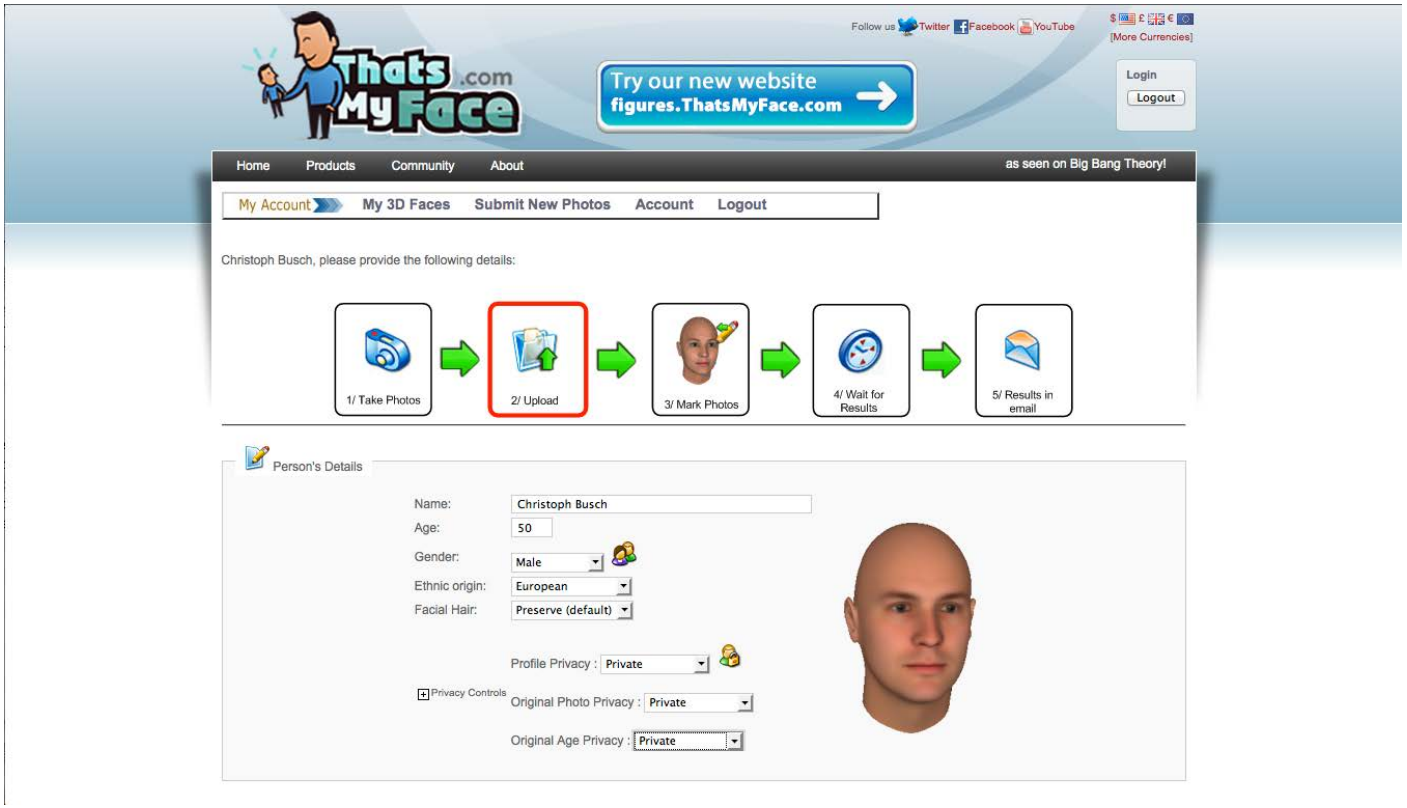
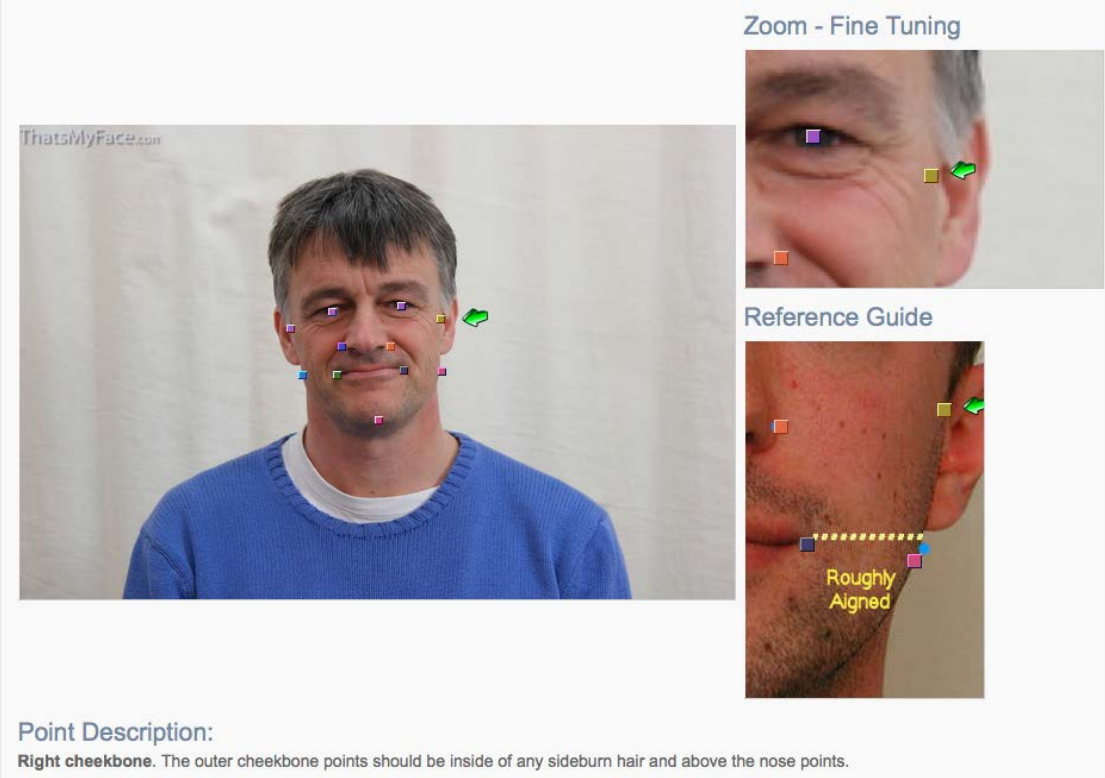
3D Face Mask Production

Attack again without support of an enroled individual

- Frontal and profile photos are uploaded
- 3D face dataset rendered and produced



3D Face Mask Production



3D-reconstruction

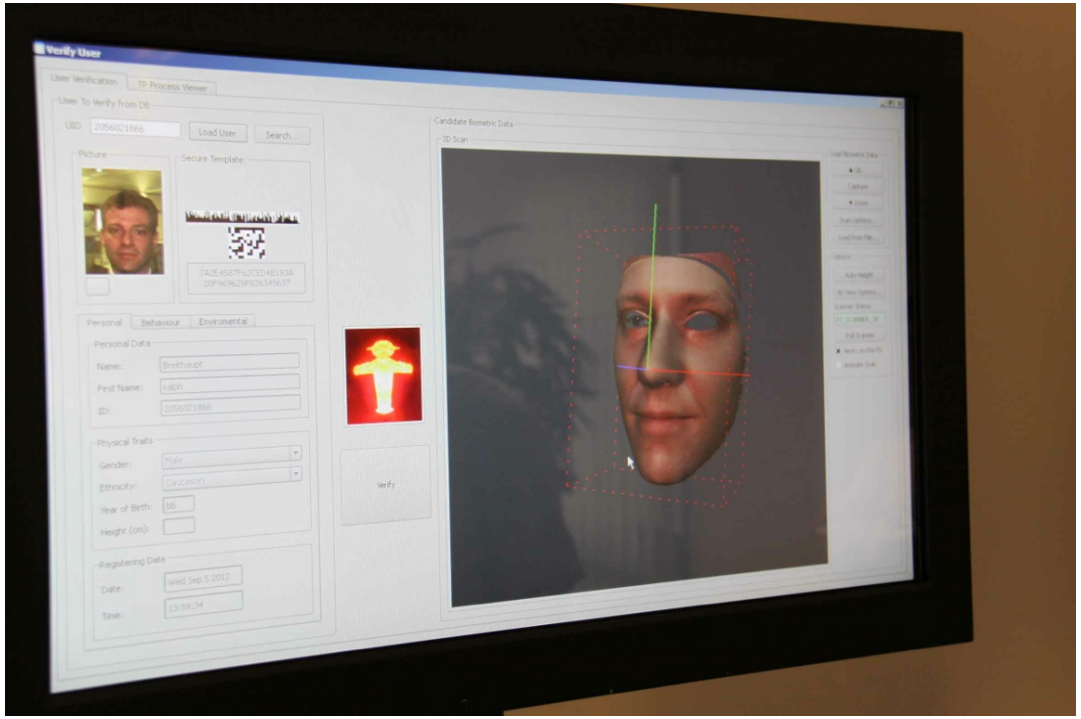
mask production preview (“beautified”):

3D Face Mask Production

Attack again without support of an enroled individual

- A static mask is produced and shipped

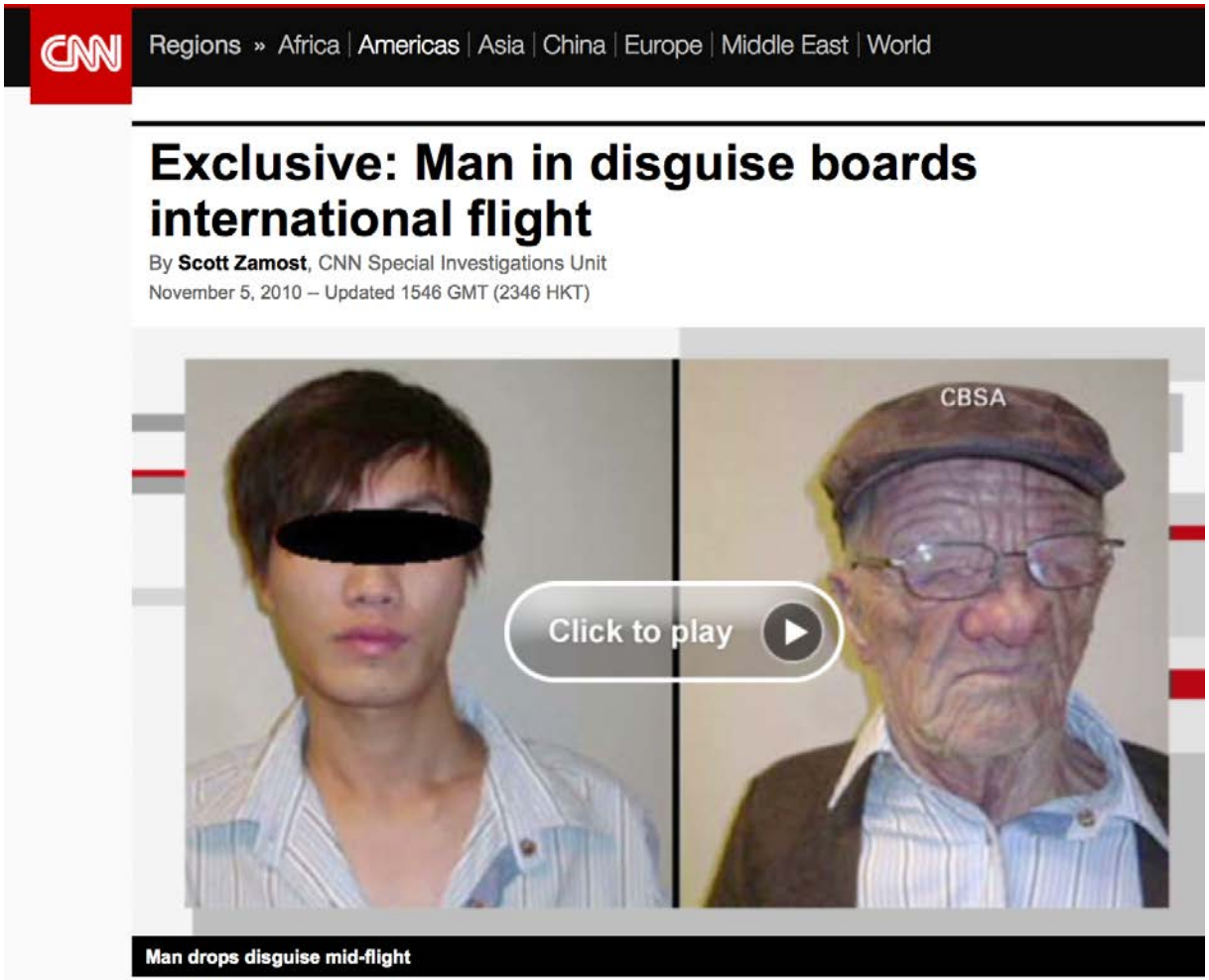
Face Capture Device Security



Impostor Presentation Attack

3D silicon mask

- Young asian traveller under identity of an elderly man
  http://edition.cnn.com/2010/WORLD/americas/11/04/canada.disguised.passenger



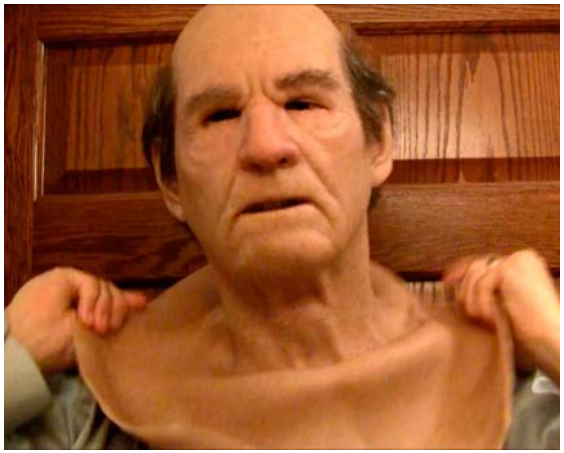
Impostor Presentation Attack

3D silicon mask

- Targeted attack with 3D silicon custom mask
- Cost more than 3000 USD

Image Source: Sebastien Marcel (Idiap)

Impostor Presentation Attack

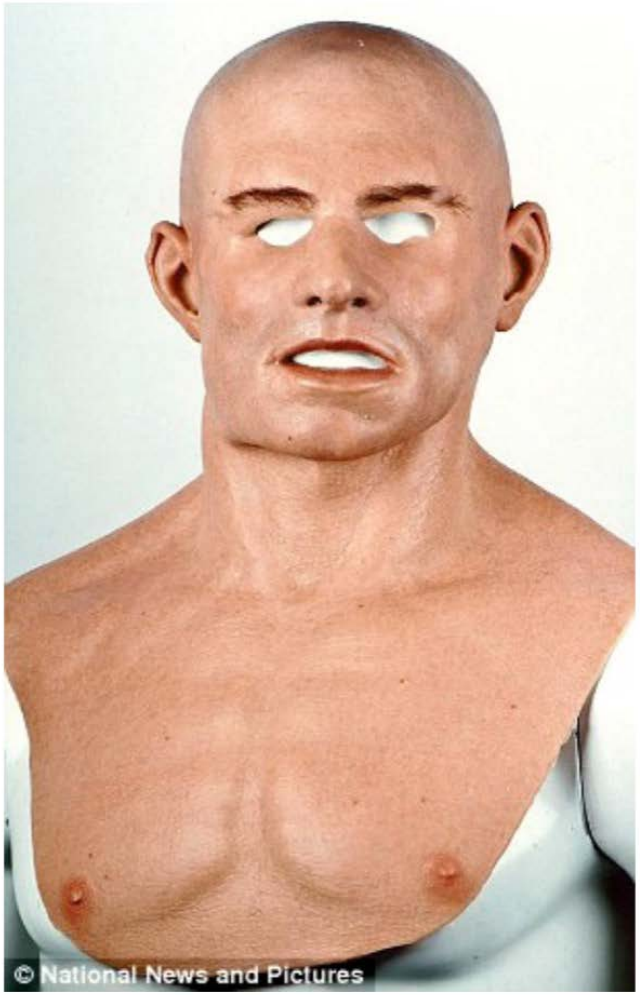


Source: BSI

Face Capture Device Security

Face disguise for organized crime (June 2012)

- http://www.dailymail.co.uk/news/article-2153346/Black-armed-robber-disguised-white-man-using-latex-mask.html



The man in the latex mask: BLACK serial armed robber disguised himself as a WHITE man to rob betting shops

- **Henley Stephenson wore the disguise during a 12-year campaign of hold-ups at betting shops and other stores across London**
- **He was part of a three-man gang jailed for a total of 28 years**
- **CCTV footage showed him firing a semi-automatic pistol into the ceiling during a raid on a betting shop**
- **The mask was bought from the same London shop which supplied masks used in the £40m Graff Diamonds heist**

By ROB PREECE and REBECCA CAMBER FOR THE DAILY MAIL

PUBLISHED: 17:22 GMT, 1 June 2012 | **UPDATED:** 16:21 GMT, 2 June 2012

Most masked robbers opt for a balaclava to hide their identity.

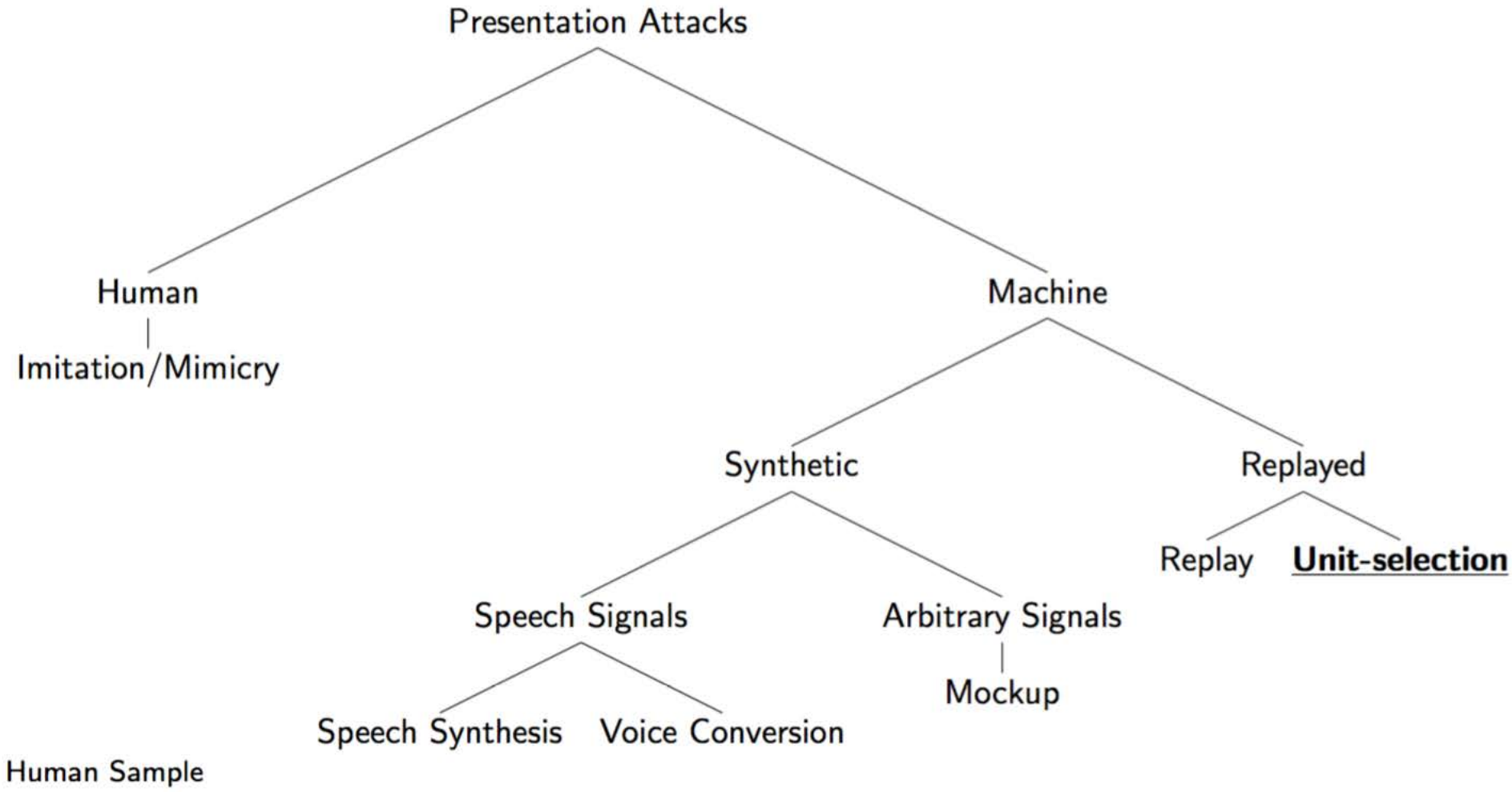
Not this one. Henley Stephenson, 41, eluded police for more than ten years thanks to an extraordinarily lifelike latex mask, which turned him into a white skinhead.

Officers discovered that their man was in fact black when they finally caught up with Stephenson after a string of armed raids dating back to 1999.

Capture Device - Presentation Attacks with Voice Artefacts

Voice Recognition with PAD

Presentation attacks on speaker recognition systems



Voice Recognition with PAD

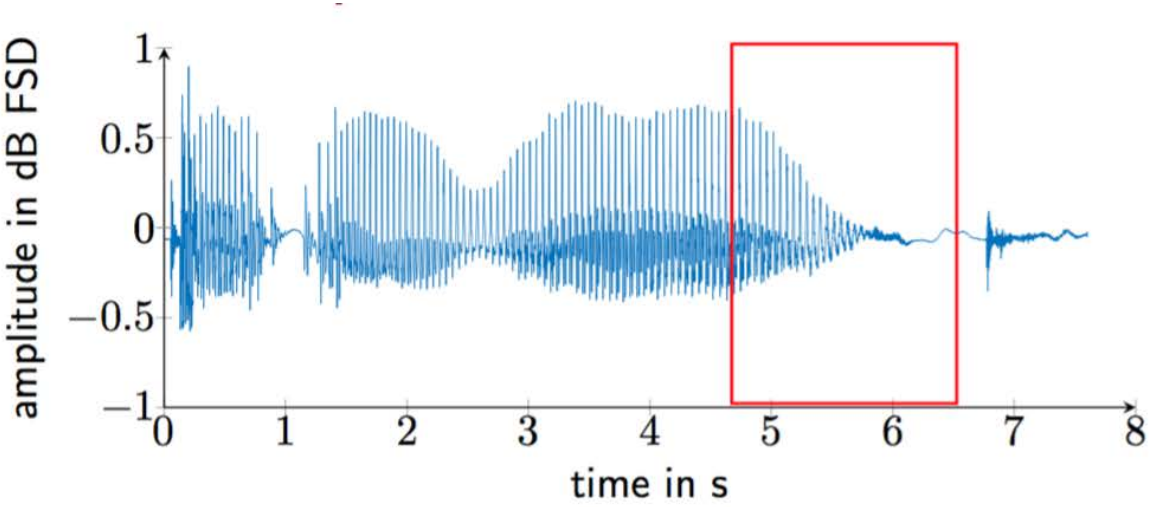
Unit selection attacks

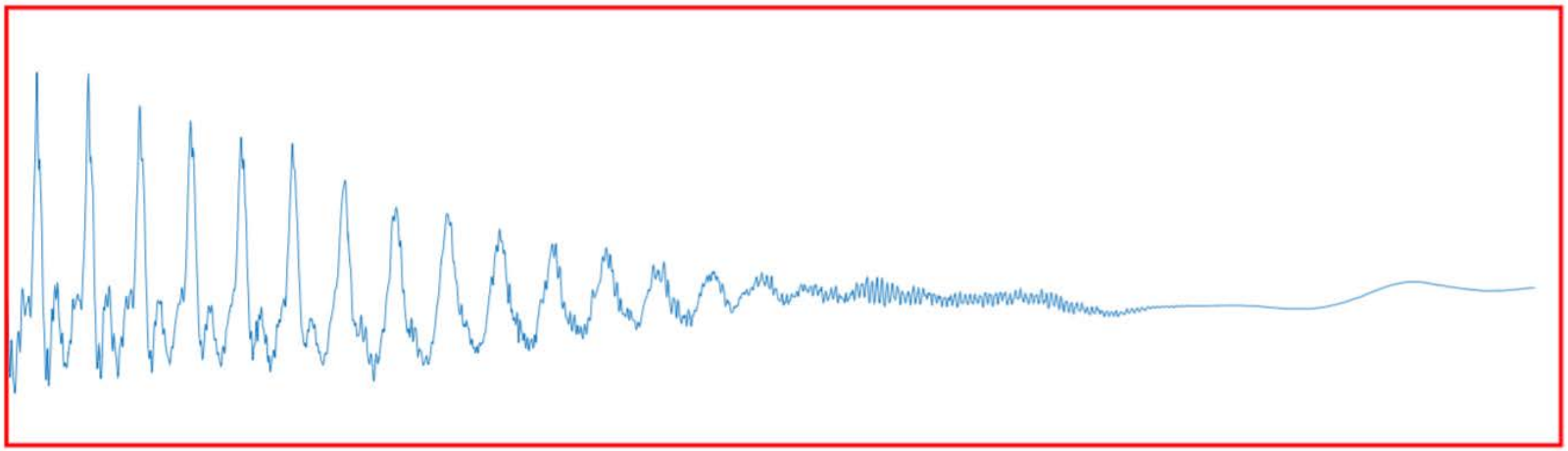
- Can be interpreted as short-time-replay
- Process:
  - 1. Speech samples are recorded
  - 2. Samples are segmented into units
  - 3. Units are concatenated
- Replay is a special case of unit-selection

Voice Recognition with PAD

Presentation attack detection

- Transition in human speech

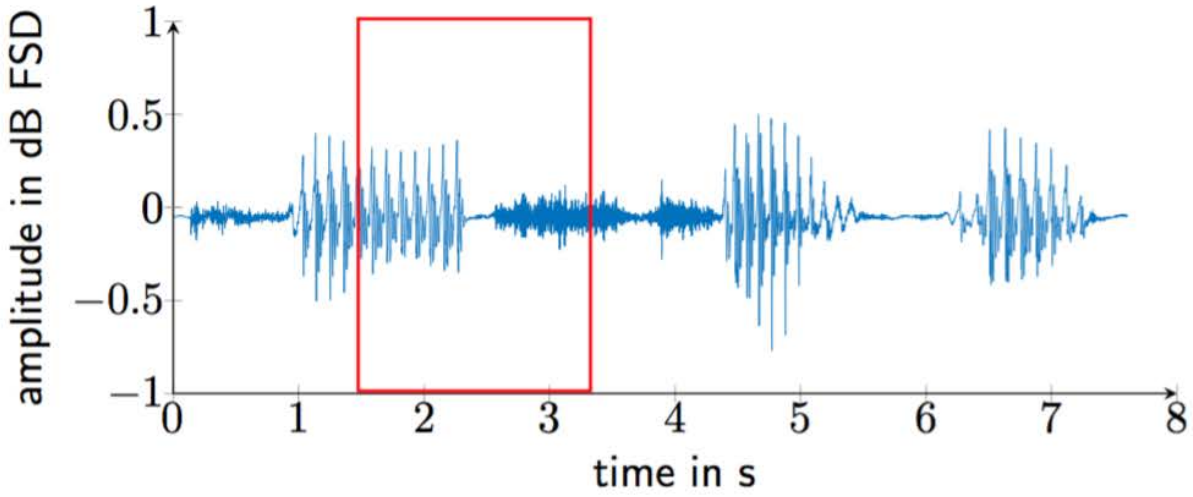


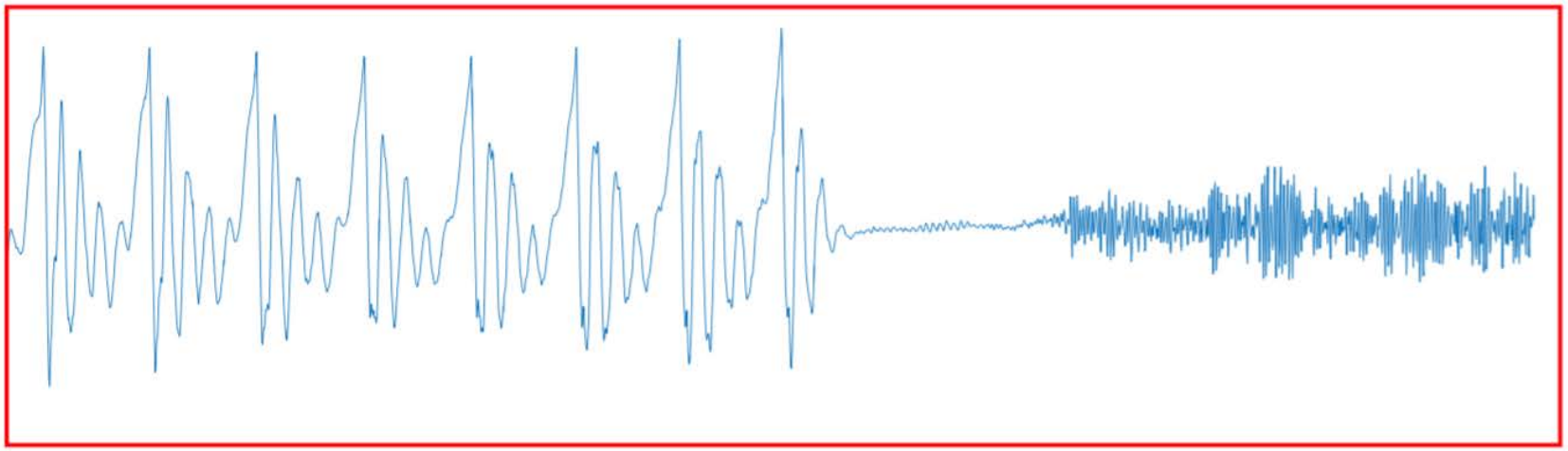


Voice Recognition with PAD

Presentation attack detection

- Transition in unit-selection attack samples





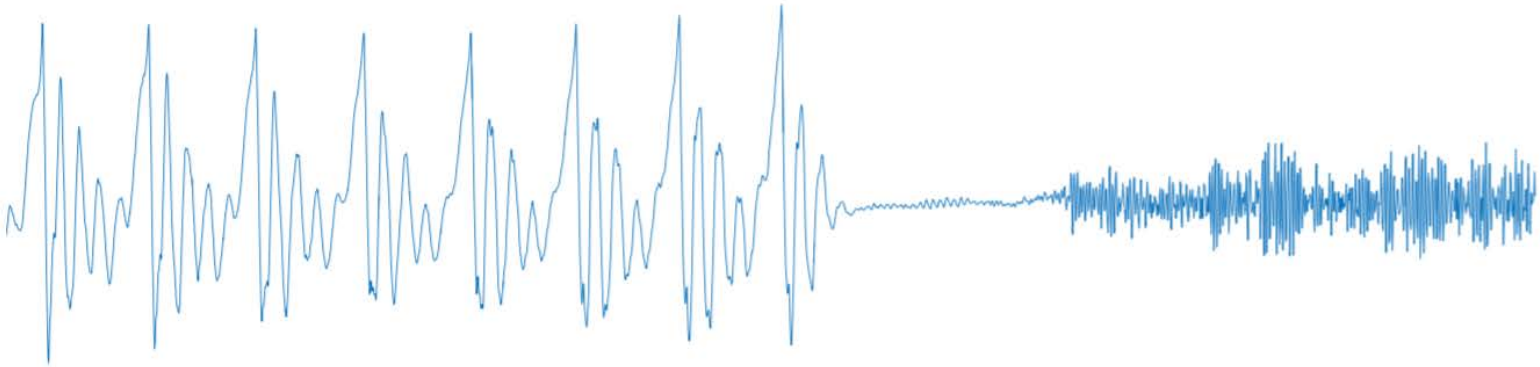
Voice Recognition with PAD

Presentation attack detection

- Comparison human speech and unit-selection samples

human:

unit-selection:

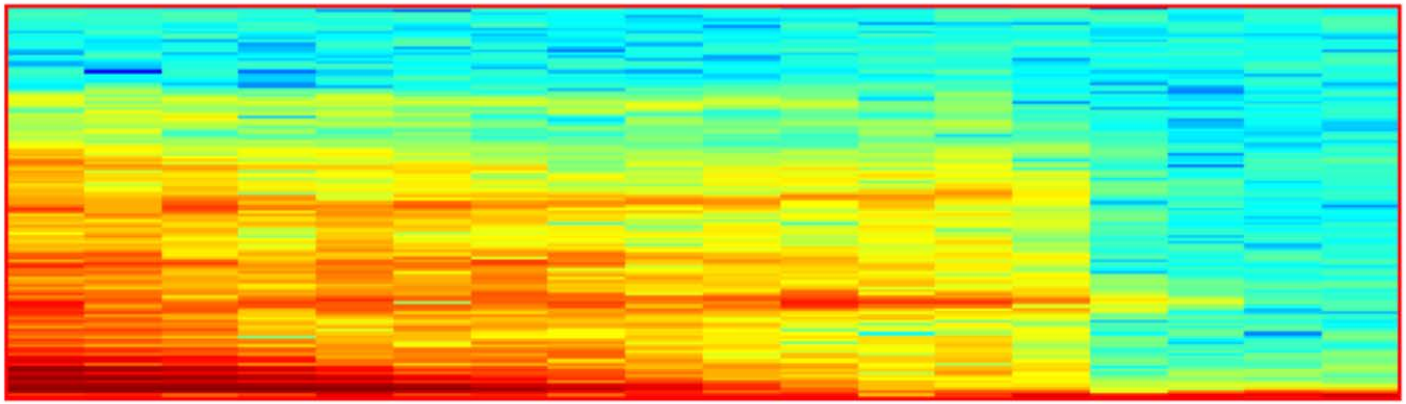


Voice Recognition with PAD

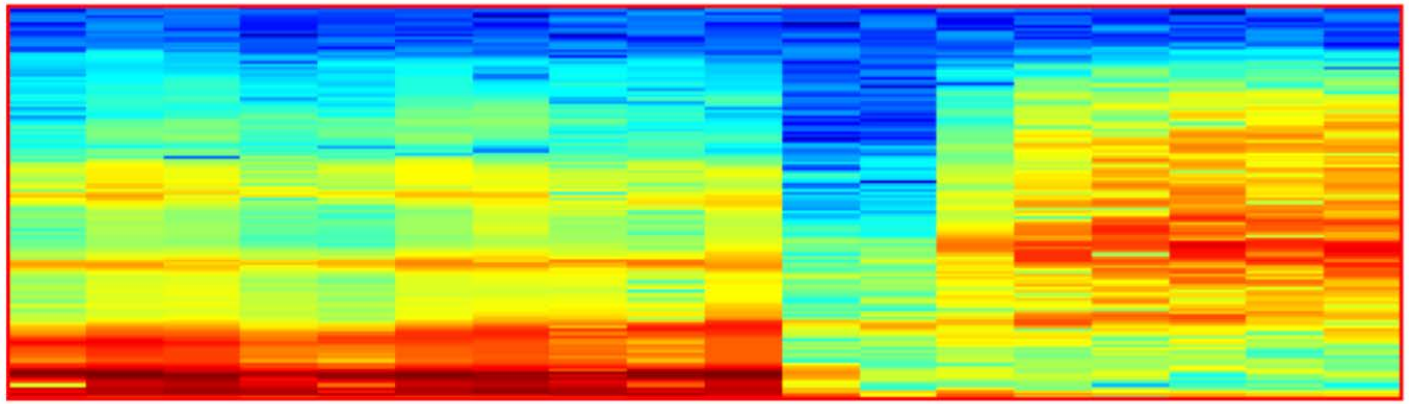
Presentation attack detection

- Spectrogram comparison human and unit-selection samples

human:



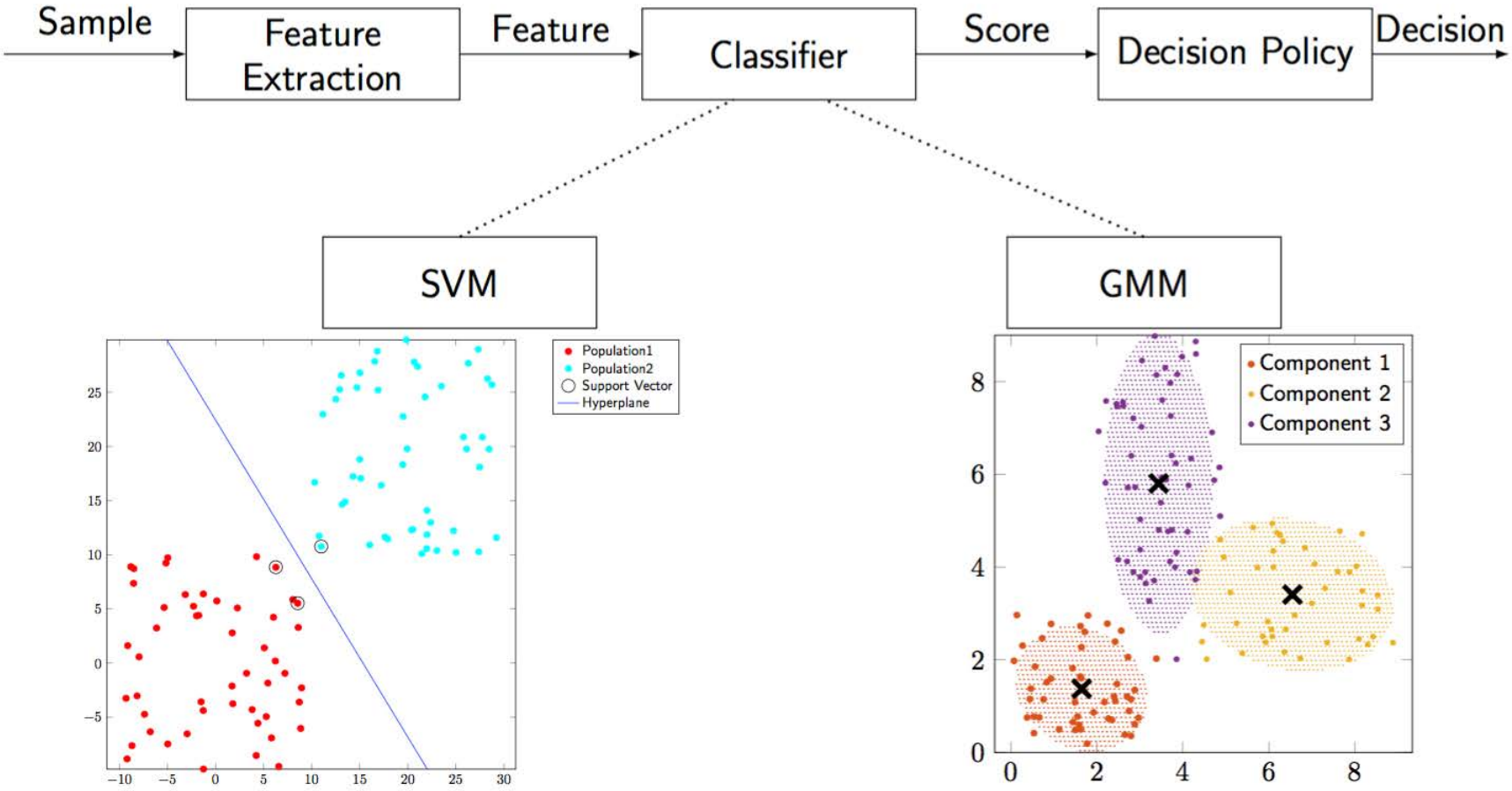
unit-selection:



Voice Recognition with PAD

Presentation attack detection

- Processing



Voice Recognition with PAD

Presentation attack detection - Feature vectors

- Fourier-Based feature:
  - Fourier transformation provides fix-dimension
  - Dimensions of feature-vector can be influenced by resolution of FFT
  - Magnitude of the Fourier transformed is calculated
- Wavelet-Based feature:
  - DWT preceding to FFT (filter function)
  - Utilization of wavelet iteration 3-5 and 5
  - FFT for fix-dimension reduction

Voice Recognition with PAD

Presentation attack detection - Results

- Evaluation set (TU Darmstadt)
- ASVspoof

Frontend	Backend	EER Eval-set	EER ASVspoof
DWT-3-5 & FFT	SVM	7.1%	11.7%
DWT-3-5 & FFT	GMM	15.0%	24.6%
FFT	SVM	8.5%	22.6%
FFT	GMM	9.5%	27.7%
DWT-5 & FFT	SVM	27.0%	11.7%
DWT-5 & FFT	GMM	40.1%	45.7%

We are close to the end of this talk!

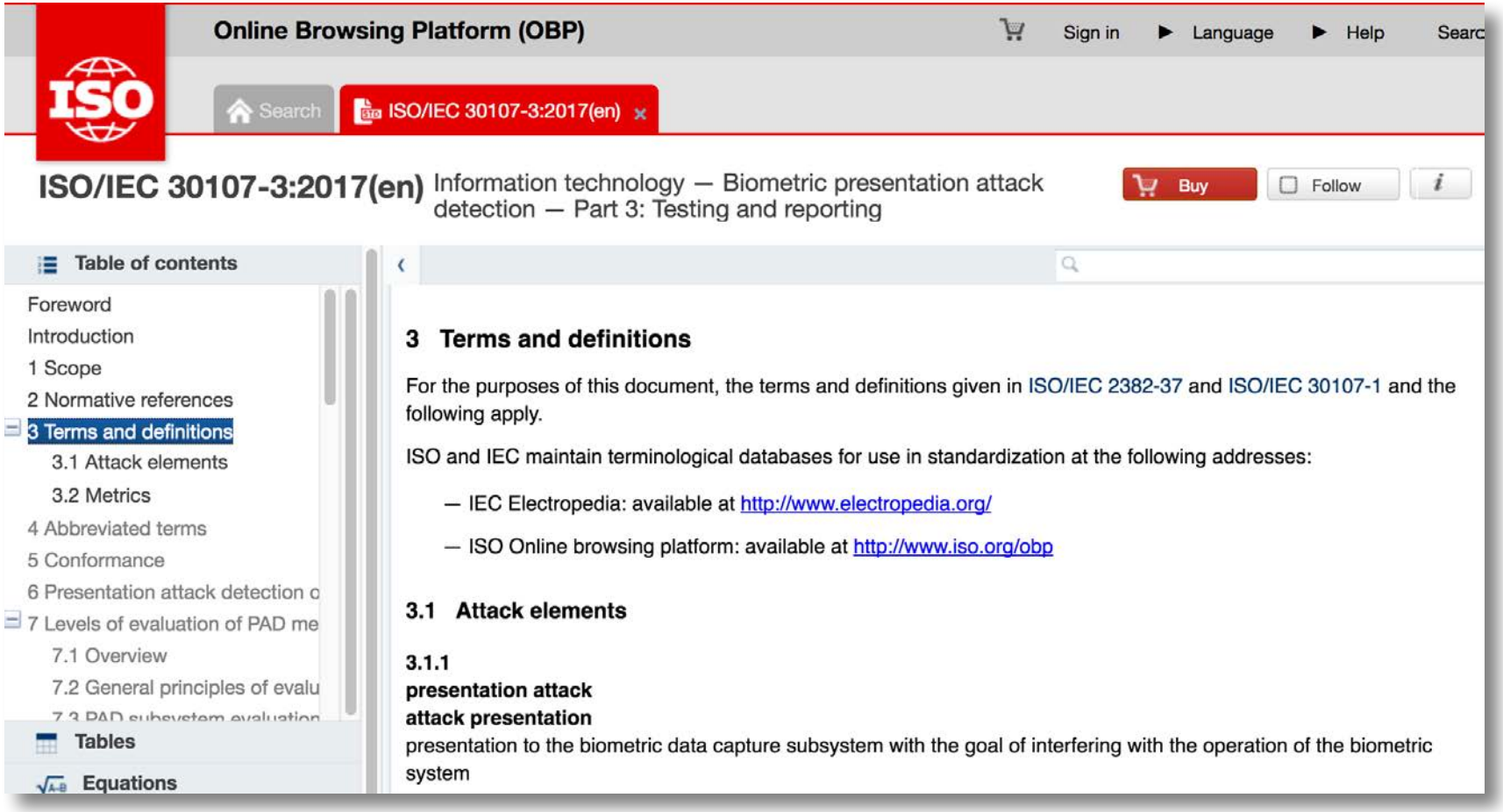
Now - the bonus material in this talk:

More on
Standardized Metrics

Presentation Attack Detection - Testing

ISO/IEC 30107-3

- available in the ISO/IEC Portal
  https://www.iso.org/obp/ui/#iso:std:iso-iec:30107:-3:ed-1:v1:en



Presentation Attack Detection - Testing

Definition of detection capabilities metrics

- Testing the PAD subsystem with security measure:

- **Attack presentation classification error rate (APCER)** *proportion of attack presentations using the same PAI species incorrectly classified as bona fide presentations in a specific scenario*

$$APCER_{PAIS} = 1 - \left(\frac{1}{N_{PAIS}}\right) \sum_{i=1}^{N_{PAIS}} Res_i$$

Source: ISO/IEC 30107-3

- N_{PAIS} *is the number of attack presentations for the given PAI species*
- Res_i *takes value 1 if the* i^{th} *presentation is classified as an attack presentation, and value 0 if classified as a bona fide presentation*

Presentation Attack Detection - Testing

Definition of detection capabilities metrics

- Testing the PAD subsystem with security measure:

- **Attack presentation classification error rate (APCER)** *the highest APCER (i.e. that of the most successful PAI species) should be reported as follows:*

$$APCER_{AP} = \max_{PAIS \in \mathcal{A}_{AP}} (APCER_{PAIS})$$

Source: ISO/IEC 30107-3

where A_{AP} is a subset of PAI species with attack potential at or below AP.

Presentation Attack Detection - Testing

Definition of detection capabilities metrics

- Testing the PAD subsystem with convenience measure:

- **Bona fide presentation classification error rate (BPCER)**
  BPCER shall be calculated as follows:

$$BPCER = \frac{\sum_{i=1}^{N_{BF}} RES_i}{N_{BF}}$$

Source: ISO/IEC 30107-3

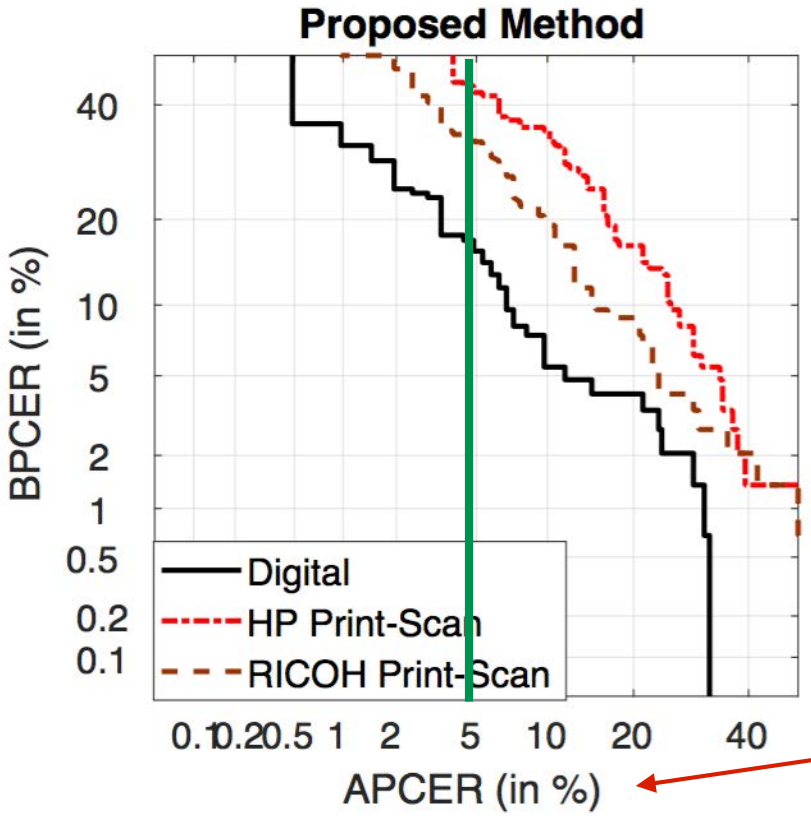
- N_{BF} *is the number of bona fide presentations*
- Res_i *takes value 1 if the* it^h *presentation is classified as an attack presentation, and value 0 if classified as a bona fide presentation*

Presentation Attack Detection - Testing

Definition of detection capabilities metrics

- DET curve analyzing operating points for various security measures and convenience measures
- Example:

convenience measure



security measure (strength of function)

Source: IR. Raghavendra, K. Raja, S. Venkatesh, C. Busch: "Transferable Deep-CNN features for detecting digital and print-scanned morphed face images", in Proceedings of 30th International Conference on Computer Vision and Pattern Recognition Workshop (CVPRW 2017), Honolulu, Hawaii, July 21-26, (2017)

Presentation Attack Detection - Testing

Definition of detection capabilities metrics

- Testing a specific security level:

  PAD mechanism may be reported in a single figure

- *BPCER at a fixed APCER:*

  One may report BPCER when $APCER_{AP}$ is 5% as BPCER20

Source: ISO/IEC 30107-3

References

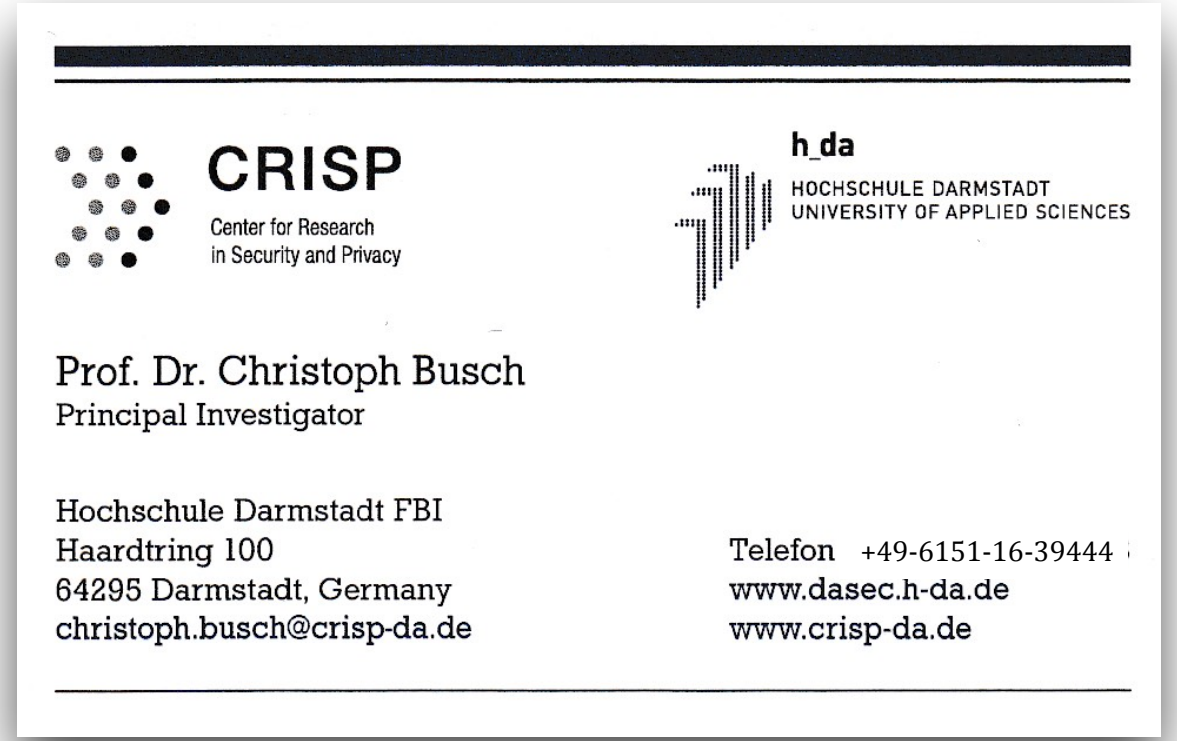
Standards

- ISO/IEC Standards
  http://www.iso.org/iso/iso_catalogue/catalogue_tc/catalogue_tc_browse.htm?commid=313770&published=on
- ISO/IEC 30107-1, “Biometric presentation attack detection - Part 1: Framework”, 2016
  http://standards.iso.org/ittf/PubliclyAvailableStandards/c053227_ISO_IEC_30107-1_2016.zip
- ISO/IEC 30107-3, “Biometric presentation attack detection - Part 3: Framework”, 2017
  http://www.iso.org/iso/home/store/catalogue_tc/catalogue_detail.htm?csnumber=67381
- ISO/IEC 2nd WD 19989-1, “Criteria and methodology for security evaluation of biometric systems - Part 1: Framework”
  https://www.iso.org/standard/72402.html
- ISO/IEC 2nd WD 19989-3, “Criteria and methodology for security evaluation of biometric systems - Part 3: Presentation attack detection
  https://www.iso.org/standard/73721.html

Contact

If you have a student interested in an internship

- then please contact:



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