Real-World Challenges for Biometric Systems

UIDAI Bio-Challenge Webinar 2024-06-26

copy of slides available at: https://christoph-busch.de/about-talks-slides.html

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Agenda

Challenges in population scale biometric solutions in enrolment and authentication

- Biometric Sample Quality
- Presentation Attacks PAD
- Morphing Attacks MAD
- Ageing

Biometric Characteristic

Biometric activities

- 25+ years research in Biometrics
- Lecturer in Darmstadt, Gjøvik and Copenhagen
- Principal Investigator in ATHENE
- Convener of the Working Group 3 on Biometric Data Interchange Formats in ISO/IEC JTC1 SC37
- Co-Founder of the CAST Association
- Co-Founder of the European Association for Biometrics
- Chair of the TeleTrusT working group on Biometrics
- Advisor to
 - BSI (German Federal Agency for IT-Security)
 - eu-LISA (European Union Agency for the Operational Management of Large-Scale IT Systems in the Area of Freedom, Security and Justice)

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Bundesamt für Sicherheit in der Informationstechnil





European Association **Biometric Sample Quality**

Standards to test Sample Quality

Relevant international standards

- ISO/IEC 29794-1:2024 Quality Framework
 - definitions and evaluation concepts https://www.iso.org/standard/79519.html
- ISO/IEC 29794-4:2024 Finger image quality
 - based on reference implementation NFIQ2.2
 - https://github.com/usnistgov/NFIQ2
 - https://www.iso.org/standard/62791.html
- ISO/IEC 29794-5:2024 Face image quality
 - based on reference implementation OFIQ
 - https://github.com/BSI-OFIQ/OFIQ-Project
 - https://www.iso.org/standard/81005.html
- ISO/IEC 29794-6:2015 Iris image quality
 - https://www.iso.org/standard/54066.html

Quality Algorithms - Standards

Quality assessment algorithms



- Quality measures in the range of 0 to 100
- Quality scores: higher is better

Quality Algorithms - Standards

Relationship between quality and system performance



Source: ISO/IEC 29794-1

Evaluation of Predictability

Error versus reject/Discard Characteristic curve (EDC)

 Stronger decrease of the EDC curve indicates a better prediction, meaning really the poorest samples are out



[Schlett2023] T. Schlett, C. Rathgeb, J. Tapia, C. Busch: "Considerations on the Evaluation of Biometric Quality Assessment Algorithms", in IEEE Transactions on Biometrics, Behavior, and Identity Science (TBIOM), (2023)

Quality Measures for Fingerprint Images

NFIQ2.0

- The Entry Exit System implementing decision 2019/329 defines the mandatory use:
- "At the moment of enrolment, the version 2.0 (or newer version) of the Fingerprint Image Quality (NFIQ) metric shall be used for verifying that the quality of the captured fingerprint data respects the thresholds ..."



Quality Measures for Fingerprint Images

The NFIQ2 approach

• Measure quality by filtering the signal and determine the utility of a fingerprint sample.



 Providing constructive feedback only possible if cause of poor quality is known.



NFIQ2.0 constitutes the content of ISO/IEC 29794-4

http://www.christoph-busch.de/projects-nfiq2.html

Face Image Quality in the EES

The objective in the EES implementing decision 2019/329

 "The quality of the facial images, … and with the image requirements of ISO/IEC 19794-5:2011 Frontal image type

What does that mean?

Data subjects need actionable feedback

• If quality is poor, then what went wrong?

	INTERNATIONAL STANDARD	ISO/IEC 19794-5			
be		Second edition 2011-11-01			
	Information technology — Biometric dat interchange formats —				
	Part 5: Face image data				
	Technologies de l'information — Formats d'éch biométriques — Partie 5: Données d'image de la face	ange de données			





Pose







Mouth open

Inhomogenous background

Source: ISO/IEC 39794-5

Compliant image

Challenges for Biometrics

Eyes open

ISO/IEC 29794-5: Face Image Quality

ISO/IEC 29794-5 is aligned with both

- ISO/IEC 19794-5:2011
- ISO/IEC 39794-5:2019

Measures

- 7.2 Unified quality score
- 7.3 Capture-related quality elements
- 7.4. Subject-related quality elements



a) Compliant image b) Low contrast source: ISO/IEC 39794-5:2019, Annex D https://www.iso.org/standard/72156.html





Image Source: ISO/IEC 19794-5:2011

- Image Source: ISO/IEC 39794-5
- The components constitute a quality vector and quantitative ICAO compliance checklist

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ISO/IEC IS 29794-5: Face Image Quality

ISO/IEC DIS 29794-5 quality measures in detail

#	Face image quality measure	
1.	Quality score (unified)	
2.	Background uniformity	
3.	Illumination uniformity	
4.	Luminance mean	
5.	Luminance variance	
6.	Under-exposure prevention	
7.	Over-exposure prevention	
8.	Dynamic range	
9.	Sharpness	
10.	No compression artifacts	
11.	Natural colour	
12.	Single face present	
13.	Eyes open	
14.	Mouth closed	
15.	Eyes visible	
16.	Mouth occlusion prevention	
17.	Face occlusion prevention	
18.	Inter-eye distance	
19.	Head size	
20.	Leftward crop of face in image	
21.	Rightward crop of face in image	
22.	Downward crop of face in image	
23.	Upward crop of face in image	
24.	Pose angle yaw frontal alignment	
25.	Pose angle pitch frontal alignment	
26.	Pose angle roll frontal alignment	
27.	Expression neutrality	
		-

Capture device related

Subject related

Open Source Face Image Quality (OFIQ)

Approach

- Library with quality assessment algorithms
- Open source with liberal license (MIT)
 - enables commercial use
- Support for major OS platforms (including mobile OS)
 C/C++
- Aligned with ISO/IEC 29794-5
 - serves as reference implementation
 - providing target values for conformance tests
- Selection criteria for integrated algorithms
 - accuracy (OFIQ-evaluation or NIST FATE SIDD evaluation)
 - Iow computational complexity
 - Iberal license (MIT or alike)

Quality Measures for Facial Images



How to find the best face quality measures?

Testing



Category	ISO/IEC 29794-5 Quality Check	SIDD Quality Component	
pture	6.3.2 Background uniformity	Background uniformity	
vice-related	6.3.3 Illumination uniformity	-	
	6.3.4 Moments of the luminance distribution	-	
	6.3.5 Under-exposure	Under-exposure	
	6.3.6 Over-exposure	Over-exposure	
	6.3.7 Dynamic range	-	
	6.3.8 De-focus	Resolution	
	6.3.9 Motion blur	Motion blur	
	6.3.10 Compression ratio	Compression artifacts	
	6.3.11 Unnatural color	-	
	6.3.12 Radial distortion	-	
	6.3.13 Pixel aspect ratio	-	
	6.3.14 Camera to subject distance	-	
bject-related	6.4.2 Single face present	Face count	
	6.4.3 Eyes visible	Sunglasses + eyeglasses	
	6.4.4 Eyes open	Eyes open	
	6.4.5 Mouth occlusion	Face occlusion	
	6.4.6 Mouth closed	Mouth open	
	6.4.7 Nose occlusion	Face occlusion	
	6.4.8 Inter-eye distance	Spatial sampling rate	
	6.4.9 Horizontal position of the face	Face cropping and margin	
	6.4.10 Vertical position of the face	Face cropping and margin	
	6.4.11 Pose	Pose	
	6.4.12 Shoulder presentation	-	
	6.4.13 Expression neutrality	-	



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• FATE Quality - Unified Quality Score

https://pages.nist.gov/frvt/html/frvt_quality.html

• FATE Quality - Specific Image Defect Detection (SIDD)

SI

https://pages.nist.gov/frvt/reports/quality_sidd/frvt_quality_sidd_report.pdf



OFIQ - Unified Quality Score

General, holistic quality score

- Not limited to certain quality criteria / defects
- CNN MagFace (iResNet 50 model)
- Shows good prediction of face recognition scores
 - nigher numbers indicate better quality



OFIQ - Unified Quality Score

Excellent results in FATE SIDD (1st of 16)

- Very good prediction of low face recognition scores
- Best performing algorithm



secunet

OFIQ - Quality Components

Eyes Open and Mouth Closed

- Algorithms based on landmarks
- Maximum distance between lids / lips
- Normalized by distance T between eye's midpoint and chin



OFIQ - Quality Components

Eyes Open and Mouth Closed

- Excellent results in NIST FATE SIDD
- 1st of 6 and 1st of 5
- No ethnic bias found for Eyes Open



Face Image Quality - Expression

Quality Component: Expression Neutrality

- Expression neutrality as quality component
- Reduced biometric performance for extreme facial expressions
- Known fact: best-possible utility through neutral expressions
- Goal:

Quantify expression neutrality



[GRVB2023] M. Grimmer, C. Rathgeb, R. Veldhuis, C. Busch: "NeutrEx: A 3D Quality Component Measure on Facial Expression Neutrality", in Proceedings of International Joint Conference on Biometrics (IJCB), (2023) [GVB2024] M. Grimmer, R. Veldhuis, C. Busch: "Efficient Expression Neutrality Estimation with Application to Face Recognition Utility Prediction", in Proceedings of 12th International Workshop on Biometrics and Forensics (IWBF 2024)

Face Image Quality - Expression

Expression Neutrality Measure: NeutrEx

- Cumulative 2-Norm Distances: $D(V_i, V_A) = ||V_i V_A||_2$
- NeutrEx Measure: NeutrEx $(V_i, V_A) = 100 \cdot (1 \frac{D(V_i, V_A) D_{min}}{D_{max} D_{min}})$
- Quality measure between [0, 100]

Explainabilitv



Presentation Attacks - PAD

Fingerprint Presentation Attacks

Attack without support of an enrolled individual



- Recording of an analog 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



Source: A. Zwiesele et al. "BioIS Study - Comparative Study of Biometric Identification Systems", In: 34th Annual 2000 IEEE International Carnahan Conference on Security Technology, Ottawa, pp. 60-63, (2000)

Face Presentation Attacks

3D silicone mask

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







Face Presentation Attacks

Changing facial appearance by makeup alterations



- Makeup for impersonation
- Liveness detection is not sufficient
- Detection difficult since bona fide users may also apply makeup



[RDB2020] C. Rathgeb, P. Drozdowski, C. Busch: "Detection of Makeup Presentation Attacks based on Deep Face Representations", in Proceedings of 25th International Conference on Pattern Recognition (ICPR), (2020)

Categories of Presentation Attacks

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



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

ISO/IEC 30107-3:2023

Provides the testing methodology

	International Standard	← TC ← ISO/IEC JTC 1/SC 37 ISO/IEC 30107-3:2023		
Information technology — Biometric presentation attack detection — Part 3: Testing and reporting	ISO/IEC 30107-3:2023 Edition 2 2023-01	Information technology — Biometric presentation attack detection Part 3: Testing and reporting		
504003 NILTAN 9040 2 NILTAN	8 10 2014	Published (Edition 2, 2023)		
Read sample				

Read the sample text:

https://www.iso.org/obp/ui/en/#iso:std:iso-iec:30107:-3:ed-2:v1:en

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Definition of detection capabilities metrics

- Testing the PAD subsystem with false-negative and false-positive errors:
- 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

Definition of PAD metrics in ISO/IEC 30107-3

• DET curve reports operating points for various thresholds showing security measures versus convenience measures



New definition in the revised ISO/IEC 30107-3:2023

- Relationship between vulnerability and recognition performance
- System testing!
- Impostor attack presentation match rate (IAPMR)
- Impostor attack presentation accept rate (IAPAR) in a full-system evaluation of a verification system, proportion of impostor attack presentations using the same presentation attack instrument (PAI) species that result in accept

Source: ISO/IEC 30107-3:2023

New definition in the revised ISO/IEC 30107-3

- Relationship between vulnerability and recognition performance
- Relative imposter presentation accept rate (RIAPAR) sum of IAPAR and FRR at a fixed decision threshold

 $RIAPAR(\tau) = IAPAR(\tau) + FRR(\tau)$



Source: U. Scherhag et al.: "Biometric Systems under Morphing Attacks: Assessment of Morphing Techniques and Vulnerability Reporting", in Proceedings of the IEEE 16th International Conference of the Biometrics Special Interest Group (BIOSIG), Darmstadt, (2017)

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Challenges for Biometrics

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Morphing Attacks - MAD

Enrolment attack with morphed facial images



Subject A



Subject A+B

Subject B

Morphing attack scenario



Morphing attack scenario

Border control



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Verification against morphed facial images



Enrolment morph M

Unique Link

Principle of equality - in our society

One individual - one passport



Principle of unique link of ICAO

- One individual one passport
- ICAO 9303 part 2, 2006:

"Additional security measures: inclusion of a machine verifiable biometric feature linking the document to its legitimate holder"

image source: https://pixabay.com/de/vectors/tick-sternchen-kreuz-rot-gr%C3%BCn-40678/

Unique Link

Principle of unique link of ICAO

- One individual one passport
- We don't want this principle of unique link to be broken
 - Multiple individuals one passport

image source: https://pixabay.com/de/vectors/tick-sternchen-kreuz-rot-gr%C3%BCn-40678/

Scale of the Problem: Vulnerability of FRS

NIST IR 8430 report on FRS vulnerability [Ngan2022]

Accurate FRS are more vulnerable!



[Ngan2022] NIST IR 8430: "FRVT MORPH: Utility of 1:N Face Recognition Algorithms for Morph Detection", 2022 https://pages.nist.gov/frvt/reports/morph/frvt_morph_4A_NISTIR_8430.pdf

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Morphing Attack - Testing

ISO/IEC 20059

Will provide the testing methodology

	← TC ← ISO/IEC JTC 1/SC 37 ISO/IEC CD 20059.2
OMMITTEE	Methodologies to evaluate the resistance of biometric recognition systems to morphing attacks
DRAFT	Under development A draft is being reviewed by the committee.
-	

Read the committee draft text:

https://www.iso.org/committee/313770.html?t=JmZzqEdifOdEbu4MS_njlaN-2xjfvCVPgJ4nLW72ITIBGGwuoL_2beUixRNm4Nk&view=documents#section-isodocuments-top:~:text=37_N7648_Consultation

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Morphing Attack - Testing

ISO/IEC 20059

• Defines the morphing attack potential (MAP) "measure of the capability of a morphing attack to deceive one or more BRSs using multiple recognition attempts"

generality

						\longrightarrow
			# BRSs (<i>c</i>)			
				2	3	4
robustness	(r)	1	85%	73%	60%	48%
	ots (2	80%	68%	55%	43%
		3	75%	63%	50%	38%
	atte	4	70%	58%	45%	33%
	, #	5	65%	53%	40%	28%

[FFMB2022] M. Ferrara, A. Franco, D. Maltoni, C. Busch: "Morphing Attack Potential", in Proceedings of 10th International Workshop on Biometrics and Forensics (IWBF 2022), Salzburg, AT, April 20-21, (2022)

Morphing Attack Detection Scenarios

Real world scenarios

- Single image morphing attack detection (S-MAD)
 - One single facial image is analysed (e.g. in the passport application office)



- Differential morphing attack detection (D-MAD)
 - A pair of images is analysed and one is a trusted Bona Fide image
 - Biometric verification (e.g. at the border)



[SRB2018a] U. Scherhag, C. Rathgeb, C. Busch: "Towards Detection of Morphed Face Images in electronic Travel Documents", in Proceedings of the 13th IAPR International Workshop on Document Analysis Systems (DAS 2018), April 24-27, (2018)

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Face Pre-processing and Feature Extraction

Morphing Attack Detection (S-MAD) with texture analysis

Image descriptors as hand-crafted features



[SRB2018b] U. Scherhag, C. Rathgeb, C. Busch: "Detection of Morphed Faces from Single Images: a Multi-Algorithm Fusion Approach", in Proceedings if of the 2nd International Conference on Biometric Engineering and Applications (ICBEA 2018), Amsterdam, The Netherlands, May 16-18, (2018)

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Differential Morphing Attack Detection

D-MAD with deep latent vectors

Deep Face representations of Deep CNNs



- Deep representations extracted by the neural network (on the lowest layer)
- Feature space with small dimension: 512 (for ArcFace)
- SVM with radial basis function

[SRMB2020] U. Scherhag, C. Rathgeb, J. Merkle, C. Busch: "Deep Face Representations for Differential Morphing Attack Detection", in IEEE Transactions on Information Forensics and Security (TIFS), (2020)

NIST-FATE-MORPH

NIST IR 8292 report presented June, 2024

FATE-MORPH since 2019

https://pages.nist.gov/frvt/html/frvt_morph.html

- Results for MAD algorithms from eleven research labs:
 - University of Bologna (UBO)
 - Norwegian University of Science and Technology (NTNU)
 - Hochschule Darmstadt (HDA)
 - West Virginia University (WVU)
 - Universidade de Coimbra (VIS)
 - Kempelen Institute of Intelligent Technologies
 - Fraunhofer (HHI)
 - Idemia (IDM)
 - secunet (SEC)
 - Neurotechnology (NET)
 - Vision Box (VIS)

Challenges for Biometrics



NISTIR 8292 DRAFT SUPPLEMENT

Face Analysis Technology Evaluation (FATE)

Part 4: MORPH - Performance of Automated Face Morph Detection

> Mei Ngan Patrick Grother Kayee Hanaoka Jason Kuo Information Access Division Information Technology Laboratory

This publication is available free of charge from: https://www.nist.gov/programs-projects/face-recognition-vendor-test-frvt-ongoing

NIST-FRVT-MORPH

NIST IR 8292 report presented June, 2024

- Performance of Automated Face Morph Detection https://pages.nist.gov/frvt/reports/morph/frvt_morph_report.pdf
- Results for high quality morphs versus print and scanned
 - note the low number of print and scanned images

Dataset: Manua





Dataset: Print and Scanned APCER @ BPCER=0.01 and Algorithm

Challenges for Biometrics

1.000 unibo 001 (differential)

Human Experts in MAD

Border guards, case handlers, document examiners, ID experts

- S-MAD: 410 participants, 180 trials
- D-MAD: 469 participants, 400 trials (4 x 100 tasks)





[GOD2022] S. Godage, F. Løvåsdal, S. Venkatesh, K. Raja, R. Raghavendra, C. Busch: "Analyzing Human Observer Ability in Morphing Attack Detection - Where Do We Stand?", https://arxiv.org/abs/2202.12426

Human Experts in MAD

Overall accuracy





	D-MAD		S-MAD	
	Number of	Average	Number of	Average
Line of work	participants	Accuracy	participants	Accuracy
Border Guard	30	64.66	26	55.17
Case handler- Passport, visas, ID, etc	150	63.45	137	56.65
Document examiner- 1st line	38	60.79	30	57.63
Document examiner- 2st line	40	68.64	34	62.56
Document examiner- 3rd line	30	65.74	25	61.51
Face comparison expert (Manual examination)	44	72.56	39	64.63
ID Expert	53	63.09	50	57.21
Other	84	64.66	69	55.17
Student	103	56.91	-	-
Total participants	572		410	
Experts	469		410	

[GOD2022] S. Godage, F. Løvåsdal, S. Venkatesh, K. Raja, R. Raghavendra, C. Busch: "Analyzing Human Observer Ability in Morphing Attack Detection - Where Do We Stand?", https://arxiv.org/abs/2202.12426

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Ageing

Impact of Ageing

Aged reference data can pose a challenge

- For face recognition
 - changing skin properties and facial morphology
 - droopy or baggy eyelid area



2001

2006

2007

2010

2015

Not for fingerprint recognition

[Kess2021] R. Kessler, O. Henniger, C. Busch: "Fingerprints, forever young?", in Proceedings of 25th International Conference on Pattern Recognition (ICPR), Milan, IT, January 10-15, (2021)

Not for iris recognition

[Grother2013] P. Grother, J. Matey, E. Tabassi, G. Quinn: "IREX VI - Temporal Stability of Iris Recognition Accuracy", in NISTIR 7948, (2013)

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Generative Adversarial Networks

InterFaceGAN - semantic face editing

• Mated sample generation (e.g. with aging)



[Grimm2021] M. Grimmer, R. Raghavendra, C. Busch: "Deep Face Age Progression: A Survey", in IEEE Access, (2021)

[JBGB2023] E. Jensen, M. Bjerre, M. Grimmer, C. Busch: "Lifespan Face Age Progression using 3D-Aware Generative Adversarial Networks", in Proceedings IJCB, (2023)

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Testing Age Impact

Comparison score (cs) degradation for aged reference images

- Open source Face Recognition Systems (FRS)
 - ArcFace, VGGFace2, MagFace
 - Reference Face Image (RFI) versus Probe Face Image (PFI)



Testing Age Impact

Reduced comparison score (cs) degradation for synthetically adjusted (aged) reference images ?

- Open source Face Recognition Systems (FRS)
 - Adjusted Reference Face Image (ARFI) versus Probe Face Image (PFI)



Conclusions

Address the challenges

- Control the quality of reference data
- Live enrolment only
 - no unsupervised capture process, unless strong PAD is in place
 - no import of face images from passports unless strong MAD is in place
- Renew reference data
 - unless age-resistant FRS are in place
- Interact with academia and research

Contact

For more information

- on face image quality: https://christoph-busch.de/projects-ofiq.html
- on finger image quality: https://christoph-busch.de/projects-nfiq2.html
- on morphing attack detection: https://christoph-busch.de/projects-mad.html
- on biometric standards: https://christoph-busch.de/standards-sc37wg3.html



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Challenges for Biometrics

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