

Face Image Quality Standardization

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THIS WORK SUPPORTED BY

- DHS OBIM
- NIST



A tale of two face standards





ISO/IEC 29794-5 Face Image Quality

- Has requirements on QA software applied in three use-cases
- Includes definitive imageprocessing steps for some quality checks



ISO/IEC 29794-5 Face Image Quality

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1. Capture-device related quality checks

- 6.3.2 Background uniformity
- 6.3.3 Illumination uniformity
- 6.3.4 Moments of the luminance distribution
- 6.3.5 Under-exposure
- 6.3.6 Over-exposure
- 6.3.7 Dynamic range
- 6.3.8 De-focus
- 6.3.10 Compression ratio
- 6.3.11 Unnatural colour

2. Subject-behavior related quality checks

- 6.3.9 Motion blur
- 6.4.2 Single face present
- 6.4.3 Eyes visible
- 6.4.4 Eyes open
- 6.4.5 Mouth occlusion
- 6.4.6 Mouth closed
- 6.4.8 Inter-eye distance
- 6.4.9 Horizontal position of the face
- 6.4.10 Vertical position of the face
- 6.4.11 Pose
- 6.4.13 Expression neutrality

4. October 2022 comments from:

- Christoph Busch (H. Darmstadt)
- Pierre Gacon (Idemia)
- Olaf Henniger (Fraunhofer)
- Markku Metsamaki (FI Border)
- Benjamin Tams (secunet)
- Jim Wayman (DHS)
- Andreas Wolf (BDR)

5. Progression

- 1. 2022-11 Proposed disposition of comments 2. 2023-01 Discuss comments, produce WD 6
- 3.2023-04 CD 1
- 4. 2023-07 DIS 1
- 5. 2023-12-23 DIS 2 to ISO for publication

3. Cannot be measured from an image

- 6.5.1 Shoulder presentation
- 6.5.2 Camera to subject distance
- 6.5.3 Radial distortion
- 6.5.4 Pixel aspect ratio

6. August 2022 public draft freely available here:



https://isotc.iso.org/livelink/livelink?func=ll&obj Id=22304355&objAction=Open&viewType=1

Example of quality component considered intractable

Distortion occurs in all conventional photos

How much is too much?

By analyzing JUST the image, error rates > 0 because

- some faces will naturally look distorted
- Some distorted faces will look natural



a) Good appearance

b) Good appearance

c) Good appearance

Source: ISO/IEC 39794-5:2019



d) Too strong magnification e) distortion

e) Too strong magnification distortion f) Too strong magnification distortion

Example component: Number of faces

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Source: https://www.ravelry.com/patterns/library/ anti-surveillance-jumper





SLR Camera c. 2010 6 megapixel image Standard Mugshot (SAP 50)



3000 x 2000

e-Passport 2004...2016 0.3 megapixels



640 x 480

Input layer for typical leading DNN c. 2022

112x112

32 x 24

Background face from prior slide



32 x 24 NOT TO SCALE



Up-sampled image fed into a typical CNN gives high non-mate scores

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Example component: Illumination uniformity



Proportion of pixels with greylevel i on left side

 $D = \sum_{i=0}^{M} min(p_i, q_i)$

Proportion of pixels with greylevel i on right side

 $D \rightarrow 1$ implies symmetric lighting

Quality value encoded in an interchange record QS = round(100 D)

Example component: Dynamic Range









- Compute luminance
- Compute greylevel histogram
- Compute entropy



COMPUTE QUALITY SCORE: EITHER LINEAR TO [0,100]

Quality increases linearly with entropy

QS = 12.5 H

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OR SOMETHING NON LINEAR

Quality is very low if H below ~ 5 bits QS = $\frac{100}{1+e^{-(H-5)}}$

- ISO/IEC 29794-5 is draft.
- Exact mapping here yet to be confirmed.
- Decision should be based on false non-match rates.

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Quality component: Underexposure

- 1. Dark skin is difficult to image
 - 1. Low diffuse reflection
 - 2. High specular reflection
- 2. Sufficient underexposure will lead to detection or false negative recognition error
- 3. Hot spots may give false positives
- 4. Are DNNs tolerant?



Source: NIST Special Database 32 aka "MEDS", subject S171

Google's Real Tone (in Pixel 6+ phones)

https://store.google.com/intl/en/ideas/real-tone/

- For aesthetic purposes
- Not specifically for biometrics, though it may have benefits
- Ask phone camera developers to include an API dedicated to biometric face capture leveraging e.g. high bit-depth sensors, computational photography.

Image equity: Making image tools more fair for everyone

As part of Google's Product Inclusion efforts, our teams are building more equitable camera and imaging products Oct 19, 2021 for people of color.



3 min read

Florian Koenigsberger Google Image Equity Lead Share



Pictures are a big part of how we see each other and the world around us, and historically racial bias in camera technology has overlooked and excluded people of color. That same bias can carry through in our modern imaging tools if they aren't tested with a diverse

- https://blog.google/products/pixel/image-equity-real-tone-pixel-6-photos/
- https://blog.google/inside-google/company-announcements/super-bowl-ad-2022/ 10

Example component: Dynamic Range









- Compute luminance
- Compute greylevel histogram
- Compute entropy



EITHER

Quality increases linearly with entropy QS = 12.5 H

OR

Н

Quality is very low if H below ~ 5 bits QS = $\frac{100}{1+e^{-(H-5)}}$

THREE USE CASES

| Mandatory | UC1: Creation of reference samples Seek rules like accept if QS > 75 say |
|-----------|--|
| Optional | UC2: Probe for instantaneous recognition |
| Optional | UC3: System enrolment, current or later creation of a reference, delayed recognition. |

Mandatory or Optional

| # | Image quality aspect | Sub-clause | Collection of reference | Collection of probe for | System e | enrolment, current or later creation of a |
|-----|-------------------------------------|----------------------|--------------------------|---------------------------|----------------|---|
| | | | samples for ID documents | instantaneous recognition | reference | e, delayed recognition |
| 1. | Unified quality score | 6.2 | Μ | Μ | Μ | |
| 2. | Background uniformity | 6.3.2 | Μ | 0 | 0 | |
| 3. | Illumination uniformity | 6.3.3 | Μ | 0 | 0 | |
| 4. | Illumination mean | 6.3.4.2 | Μ | 0 | 0 | |
| 5. | Illumination variance | 6.3.4.3 | Μ | 0 | 0 | |
| 6. | Illumination skewness | 6.3.4.4 | Μ | 0 | 0 | |
| 7. | Illumination kurtosis | 6.3.4.5 | Μ | 0 | 0 | |
| 8. | Under-exposure | 6.3.5 | Μ | 0 | 0 | |
| 9. | Over-exposure | 6.3.6 | Μ | 0 | 0 | |
| 10. | Dynamic range | 6.3.7 | Μ | 0 | 0 | |
| 11. | De-focus | 6.3.8 | Μ | 0 | 0 | Snapshot from 2022 |
| 12. | Image sharpness | 6.3.9 | Μ | 0 | 0 | draft of ISO/IEC |
| 13. | Motion blur | 6.3.10 | Μ | 0 | 0 | 20704 5 |
| 14. | Edge density | 6.3.11 | Μ | 0 | 0 | 23734-3 |
| 15. | Compression ratio | 6.3.12 | 0 | 0 | 0 | |
| 16. | Unnatural colour and colour balance | 6.3.13 | Μ | 0 | 0 | Requirements on what |
| New | Polarization filter | | M | <mark>o</mark> | <mark>0</mark> | software library must |
| New | Radial distortion | | <mark>o</mark> | <mark>o</mark> | <mark>0</mark> | software library must |
| New | Pixel aspect ratio | | <mark>0</mark> | <mark>0</mark> | <mark>0</mark> | support? |
| 17. | Camera lens focal length | 6.3.14 | 0 | 0 | 0 | |
| 18. | Camera subject distance | 6.3.15 | 0 | 0 | 0 | Boquiromonts on what |
| new | Single subject | | M | M | M | - Requirements on what |
| new | Shoulders squared on | | <mark>o</mark> | <mark>o</mark> | <mark>0</mark> | must be done in each |
| 19. | Eyes visible | 6.4.2 | Μ | 0 | M | use-case? |
| 20. | Mouth occlusion | 6.4.3 | Μ | 0 | М | |
| 21. | Nose occlusion | 6.4.4 | Μ | 0 | Μ | |
| 22. | Inter-eye distance | 6.4.5 | Μ | Μ | Μ | |
| 23. | Horizontal position of the face | 6.4.6 | Μ | Μ | Μ | |
| 24. | Vertical position of the face | 6.4.7 | М | М | Μ | |
| new | No head coverings | New | M | 0 | M | |
| new | Pose angle yaw frontal alignment | <mark>6.4.8.1</mark> | M | <mark>o</mark> | M | |
| new | Pose angle pitch frontal alignment | <mark>6.4.8.2</mark> | M | <mark>o</mark> | M | |
| new | Pose angle roll frontal alignment | <mark>6.4.8.3</mark> | M | <mark>o</mark> | M | |
| 26. | Expression neutrality | 6.4.9 | Μ | 0 | 0 | |
| 27. | Mouth closed | 6.4.10 | Μ | 0 | Μ | |
| 28. | Eyes open | 6.4.11 | Μ | 0 | 0 | |

Role for Quality Assessment in Presentation Attack Detection

Presentation Attack: Impersonation

by maximizing similarity score

ReplayCosmetics

A: Instruments

Masks

Presentation Attack: Evasion

by minimizing similarity score

B: Instruments

- Replay
- Masks
- Cosmetics
- Image manipulation



C: Behaviours

- Adverse Pitch | Yaw
- Pronounced expression
- Occlusion
- Motion Blur

Common Image Quality Problems

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D: Imaging problems

- Blur
- Illumination
- Exposure
- ••

Incorrect Arrests in Michigan and New Jersey



PROBE

INCORRECT PERSON ROBERT WILLIAMS



https://www.cbsnews.com/news/facial-recognition-60-minutes-2021-05-16/



Future roles for quality?

- 1. Issue caution to an investigator when signal is "low / unavailable" (Image A) or "noisy" (Image B)
- 2. Issue caution to someone putting a poor image (Image C) in an authoritative database

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The FRVT Program

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| | 1:1 | 2017-02Verification Authorization Non-repudiation | | |
|---|--|---|--|--|
| | 1:N | 2018-01Investigation Duplicate detection Identification | | |
| U.S. Government Department of Commerce | Morph Detection | ■2018-09 ■Single-image Differential two-image | | |
| 2000 – 2022 Public | Quality Summarization | 2019-02Produce a number predicting mate matching outcomes | | |
| Free Open worldwide | Quality Defect Detection | 2022-08Find specific image quality problems (blur, pose, sunglasses etc) | | |
| Collaboration with USG agencies | Presentation Attack Detection | 2023-01Is the image or video showing subversive behavior? | | |
| | Twins Low False Match Demonstration | ■2022-09 ■High resolution images | | |

FRVT Quality Tracks



| TRACK A Quality Summarization | SCALAR: DECISION | Q = 98BOX 0. QUALITY BENGN: Y, AcceptOne "visa – bo— No longer use— No longer use— Extend to use r | CHMARK rder" dataset wild new "kiosk" dataset |
|---|---|---|---|
| TRACK B Specific Image Defect Detection | BOX 1.QUALITY BENCHMARK-Concept presented at the Nov Q Workshop 2021-11-API + Concept Published 2021- 07-07 for commentFinal specifications 2022-09-Algorithms to NIST 2022-09-Align with ISO/IEC 29794-5 | BOX 2. IMAGING VARIABLES THAT INFLUENCE ACCURACY Illumination adequacy + uniformity Exposure Focus, blur Resolution / Sp. Sampling Rate | BOX 3.SUBJECT VARIABLES THATINFLUENCE ACCURACY-Head orientation (R, P, Y)-Expression neutrality-Sunglasses, face masks-Motion blur-No, or additional, faces |
| | | | |

Noise

Over-

exposure

Under-

exposure

Hot Spots

Two People

No People

Cropped

Mis-focus

Non-frontal

Summary



ISO/IEC 29794-5

- » Defines a set of quality measurements
 - Compute a raw component value
 - Some computations are explicit
 - Some computations are abstract e.g. "estimate pose" without saying how
 - Compute an interpretable quality score [0,100]
 - Actionable feedback "focus score too low"
- Mandates subsets of those computations depending on use-case
- » Conceives of three use-cases

Quality evaluation

- » FRVT Specific Image Defect Detection
 - Do 29794-5 quality assessment algorithms work well?
 - Do 29794-5 relate to face recognition accuracy?

THANKS!

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Beatle John Lennon between the release of the Red Album and the Blue Album, ~5 years.

| Year | Developer | Algorithm | Score | FMR | Outcome |
|------|---------------|-----------|---------|---------|--------------|
| 2022 | Cogent Thales | 007 | 3322 | < 5e-07 | Strong match |
| 2021 | Idemia | 008 | 7438.78 | < 5e-07 | Strong match |
| 2022 | Paravision | 010 | 0.38308 | < 5e-07 | Strong match |
| 2014 | Cogent Thales | A20A | 2521 | 0.48 | Failed match |
| 2014 | NEC | E20A | 0.562 | 0.002 | Failed match |