

Overview State of the Art and Categorization of Methods

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Factor-specific categories

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Introduction



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“image Manipulation Attack Resolving Solutions”





- ▶ Funded with 6.988.521,25€ by the EU as part of Horizon 2020.
- ▶ Officially started 1. September 2020, planned end date is 31. August 2024.
- ▶ 24 partner organizations (industry, academia, end users).
- ▶ Image morphing/manipulation attack detection and document verification.
- ▶ **Also involves FIQA (Face Image Quality Assessment) research.**

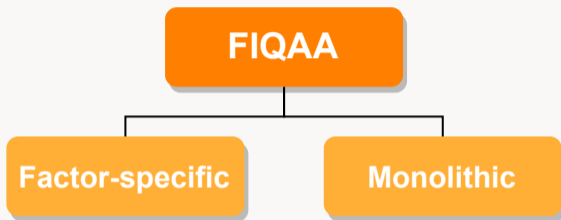
See the final slide for links with further information.

“Face Image Quality Assessment: A Literature Survey”

Torsten Schlett, Christian Rathgeb, Olaf Henniger, Javier Galbally, Julian Fierrez, Christoph Busch

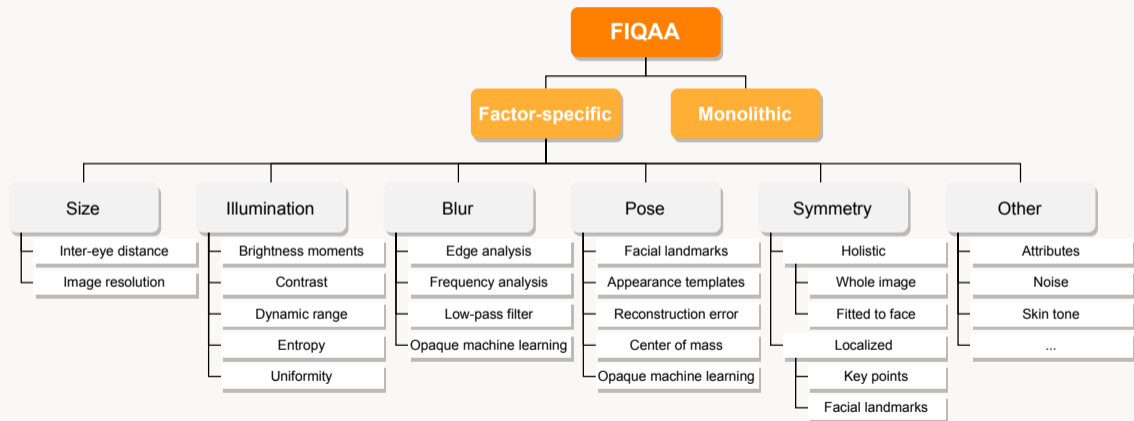
- ▶ Introductory material
- ▶ **Categorization**
→ That's what this presentation is about.
- ▶ Dataset list
- ▶ Evaluation (ERC)
- ▶ Open issues and challenges

	FIQA	IQA
Image content:	Face images 	General images 
Assesses:	Biometric utility 	Perceptual quality 
Common input:	Usually no-reference	Full/Reduced/No-reference
“Reverse usage”:	Not always general enough (modern FIQA is specialized)	Can be used as FIQA (with limited potential)



Interpretable (often actionable) factors, which could help an operator to avoid face image deficiencies in a re-capture attempt.

Opaque assessments/quality scores, which cannot be immediately interpreted with respect to some concrete separable factor.



Factor-specific categories



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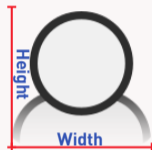


Image resolution

Implementations differ only in the way width/height are used.



Inter-Eye Distance (IED)

Relies on landmark detection.

ISO/IEC 39794-5: Distance between the eye centers.



Brightness moments

Mean / Variance / Skewness / Kurtosis.



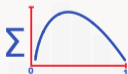
Contrast

E. g. as standard deviation or Michelson contrast.



Dynamic range

E. g. smallest value range used by $\geq 95\%$ of all pixels.



Entropy

E. g. as $-\sum h(i) \log_2(h(i))$ with normalized histogram h .



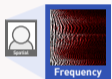
Uniformity

E. g. via landmark-local histogram comparisons.



Edge analysis

E. g. mean of the gradient image.



Frequency analysis

Involves transformation into the frequency domain.



Low-pass filter

Input compared against artificially blurred version.



Opaque machine learning

A few deep learning approaches aimed to assess blur.



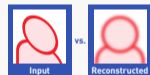
Facial landmarks

Pose angles are estimated based on facial landmark positions.



Appearance templates

Models based on training images for specific poses.



Reconstruction error

Reconstruct the input as a frontal image and check the error.



Center of mass

Selected pixel center of mass deviation from image center.



Opaque machine learning

E. g. binary frontal/non-frontal classification.



Holistic - Whole image

Comparison of fixed left/right halves.



Holistic - Fitted to face

Left/right halves are fitted to the face.



Localized - Key points

E. g. via SIFT (Scale Invariant Feature Transform) points.



Localized - Facial landmarks

E. g. symmetry of lines defined by landmark pairs.



Attributes

E. g. with/without glasses.



Noise

E. g. model trained on different synthetic noise types and levels.



Skin tone

E. g. $[-30^\circ, +30^\circ]$ hue and $[5\%, 95\%]$ saturation.



(More)

Ink mark detection, crease detection, background uniformity, . . .

Aspect categories



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Aspects in addition to the factor-specific/monolithic categorization:

- ▶ **Capture-/Subject-related**
 - ▶ Per factor.
 - ▶ More on the following slides.
- ▶ **Video-frame context**
 - ▶ Still FIQA, only the surrounding context differs.
- ▶ **Deep learning**
 - ▶ Either used or not used.
- ▶ **Fusion**
 - ▶ More on the following slides.
- ▶ **Data**
 - ▶ More on the following slides.



- ▶ **Capture-related:** FIQA is influenced by circumstances external to the capture subject, such as the used sensor (e.g. camera focus, resolution) or the illumination setup.
- ▶ **Subject-related:** FIQA is influenced by the subject, e.g. pose, expression, or movement.

Per factor-specific category:

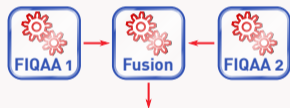
- ▶ **Size:** Subject-r. (subject to camera distance) & Capture-r. (camera resolution).
- ▶ **Illumination:** Capture-r. (lighting), but also Subject-r. (e.g. lighter/darker facial hair).
- ▶ **Pose:** Predominantly Subject-related.
- ▶ **Blur:** Subject-r. (movement) & Capture-r. (focus).
- ▶ **Symmetry:** Considers illumination & (often) pose.

(**Monolithic:** By definition a mixture.)



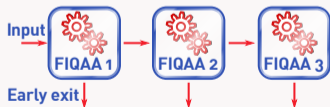
Explicit

E. g. weighted sum with manually set weights.



Trained

E. g. using neural networks or random forests.



Cascade

Can exit early, which can reduce computational workload.



Hand-crafted

No training required. E. g. image resolution FIQAAs.

Utility-agnostic training

Trained, but not on utility data. E. g. landmark-based pose.

Human utility ground truth

Trained on utility assessments from humans.

FR-based ground truth

Trained on FR-derived scores.

FR-based inference

Utilize FR models during inference.

FR-integration

FR and FIQA trained in a single model.

State of the Art

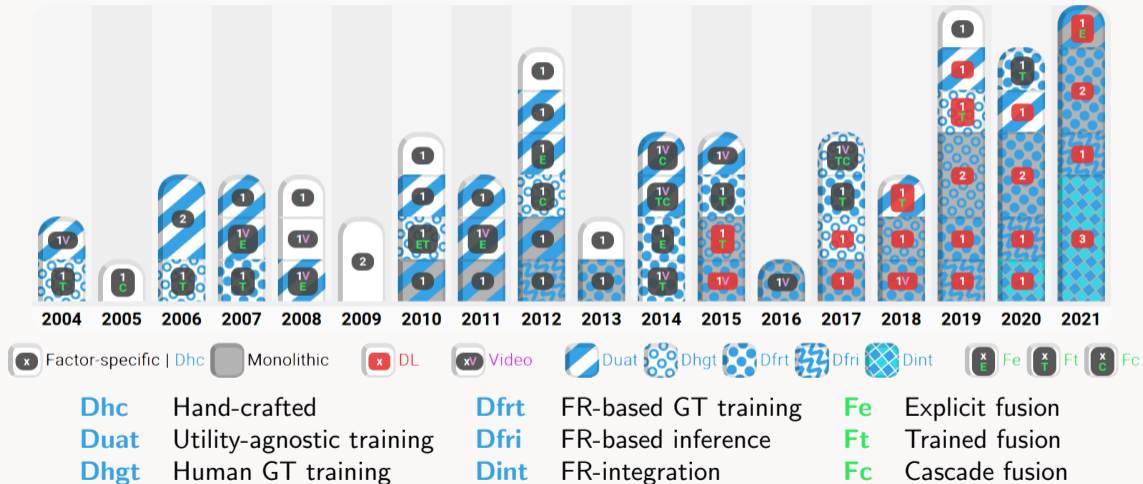


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Timeline



Approaches with dedicated presentations in the workshop:

▶ **FaceQnet v0 & v1:** FR-based training

v0: *"FaceQnet: Quality Assessment for Face Recognition based on Deep Learning"* (2019)

v1: *"Biometric Quality: Review and Application to Face Recognition with FaceQnet"* (2020)

▶ **PCNet:** FR-based training

"Inducing Predictive Uncertainty Estimation for Face Recognition" (2020)

▶ **SDD-FIQA:** FR-based training

"SDD-FIQA: Unsupervised Face Image Quality Assessment with Similarity Distribution Distance" (2021)

▶ **MagFace:** FR-integration

"MagFace: A Universal Representation for Face Recognition and Quality Assessment" (2021)

Some other modern approaches:

▶ **SER-FIQ:** FR-inference

"SER-FIQ: Unsupervised Estimation of Face Image Quality Based on Stochastic Embedding Robustness" (2020)

▶ **LightQNet:** FR-based training

"LightQNet: Lightweight Deep Face Quality Assessment for Risk-Controlled Face Recognition" (2021)

Thank you!

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“Face Image Quality Assessment: A Literature Survey” preprint:

<https://arxiv.org/abs/2009.01103>

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image manipulation attack
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