

How good is my Al System? Measures and Evaluation for Face Recognition

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Automatic Face Analysis Motivation

Active research topic

- Age estimation
 - age groups
 - real age
- Accessories
 - glasses
 - hat
 - beard
 - masks
 - ...

Google Cloud

- Gender classification
 - male
 - female
- Identification
 - recognition
 - verification
 - clustering

DEEP.va

Azure

• ...

Meta aws





VALOSSA

Animal identification using face recognition

SAISBECO

- "A Semi-Automated Audiovisual Species and Individual Identification System for Behavioral Research and Conservation"
- Collaboration between Fraunhofer and Max-Planck Institutes
- Identification and facial analysis of great apes
- Estimation and monitoring of population sizes

SIMSALACAT

- Cat flap based on face recognition
- Including detection of prey in mouth
- Implemented on edge device (Jetson Nano)

BearID

- Face recognition, age estimation, and gender classification of grizzly bears
- Collaboration with Applied Conservation Science Lab, Canada



SAISBECO, BearID, SimsalaCat

Face Recognition for animals



SPIRIT

- EU funded project
- Collaboration with law enforcement agencies

Goal:

Development of novel tools and approaches for resolving digital identities

Result:

Platform that allows to create social graphs to perform social and criminal network analysis

https://www.spirit-tools.com/





- FaceMatch Climate Change in German TV Program
 - How salient are climate change & biodiversity in the German TV programme?
 - (a) Which representatives from politics, science, economy, law and culture are visible within the climate discourse?
 - (b) How is the **share of men and women** in the climate discourse?
 - Recording of 20 German TV stations for two months (01.09.2022 – 31.10.2022), 19h/day (05:30 am – 0:30 am)
 - 37, 476 recorded programs
 - 65 TB video data
 - 23,181.7 h video
 - **4,655,706** detected face-tracks
 - 806,390 distinct persons
 - **146** reference persons









- FaceMatch Unsupervised Face Clustering
- Grouping, recognition and search of persons in private image collections
- Extracted information stored in local database
- Data remains on host system: 100% compliance with GDPR regulations
- Demo available for Windows 10 and as Docker Image for Linux







- IDCheck
- Robust and privacy-compliant verification
- Extraction of personal data from any identity document
- Extraction of personal data from ticket or certificate
- Extraction of facial data via live camera on site
- Confirmation of an "identity match" when facial and personal data correspond
- Implemented on edge device (Jetson Nano)

https://www.idmt.fraunhofer.de/idcheck









- Nvidia Jetson Nano
 - Quad-core 64-bit ARM CPU
 - 128-core NVIDIA GPU
 - 4GB or 2GB LPDDR4 memory





© https://developer.nvidia.com/embedded/jetson-nano-developer-kit





















Evaluation of Face Analysis Algorithms

Performance Metrics



















Evaluation of Face Recognition Algorithms Large Scale Training Datasets



Source: M. Wang, W.Deng, "Deep Face Recognition: A Survey", ArXiv preprint, arXiv:1804.06655v9



Evaluation of Face Recognition Evaluation Tasks and Performance Metrics

- Gallery Set:
 - Set of images of known subjects initially enrolled in the system
- Probe Set:
 - Set of images of <u>known (or unknown)</u> subjects different from the gallery set
- Face Verification:
 - Computes <u>one-to-one</u> similiarity between the gallery and the probe image to determine if the two images are from the same subject
 - Relevant to access control systems and re-identification
- Face Identification:
 - Computes <u>one-to-many</u> similarity to determine the specific identity
 - Closed-Set Identification:
 - Person in probe image is known to the system
 - Open-Set Identification:
 - Person in probe image is known or unknown to the system



Evaluation of Face Recognition Algorithms

Verification vs. Open-set vs. Closed-Set



Evaluation of Face Recognition Evaluation Tasks and Performance Metrics (Verification)

- Face Verification:
- Computes <u>one-to-one</u> similarity between the gallery and the probe image to determine if the two images are from the same object
- Relevant to access control systems and re-identification
- Receiver Operating Characteristic (ROC) curve
 - True Acceptance Rate (TAR) vs. False Acceptance Rate (FAR) at different thresholds
 - <u>TAR:</u> fraction of genuine individuals that exceed acceptance threshold
 - <u>FAR:</u> fraction of impostor individuals that exceed acceptance threshold
- Comparison of systems
 - Area Under Curve (AUC)
 - Equal Error Rate (EER)
 - PaSC: TAR@10⁻²FAR
 - IJB-A: TAR@10⁻³FAR
 - Megaface: TAR@10⁻⁶FAR
 - MS-celeb-1M: TAR@10⁻⁹FAR





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0.4

False Acceptance Rate FAR

0.6

0.2

0.0



0.8

1.0

Evaluation Tasks and Performance Metrics (Verification)

- **Face Verification:**
- Computes one-to-one similarity between the gallery and the probe image to determine if the two images are from the same object
- Relevant to access control systems and re-identification
- **Detection (Decision) Error Trade-Off Curve (DET)**
 - False Rejection Rate (FRR) vs. False Acceptance Rate (FAR) at different thresholds
 - FRR : fraction of genuine individuals classified as unknown .
 - FAR: fraction of impostor individuals that exceed acceptance threshold
 - Might be easier to visually compare different algorithms
 - The user can deduct directly from the DET-curve plot at which rate FRR will improve when willing to accept an increase in FAR (or vice-versa)



99%

95%

80%

50%

1%



Evaluation Tasks and Performance Metrics (Identification – Closed Set)

- Face Identification: Closed-Set Identification:
 - Computes <u>one-to-many</u> similarity to determine the specific identity
 - Person in probe image is known to the system
 - Cumulative Match Characteristic (CMC)
 - Reports the percentage of probes identified within a given rank
 - Cumulative Identification Rate vs. Rank
 - Most frequent metric: Rank-1 Accuracy





Evaluation Tasks and Performance Metrics (Identification – Open Set)

- Face Identification: Open-Set Identification:
- Computes <u>one-to-many</u> similarity to determine the specific identity
- Person in probe image is known or unknown to the system
- Receiver Operating Characteristic (ROC)
 - Correct Acceptance and Identification Rate (CDIR) vs.
 False Acceptance Rate (FAR) at different thresholds
 - <u>CDIR</u>: fraction of genuine individuals that exceed acceptance thresho and are correctly identified
 - <u>FAR:</u> fraction of impostor individuals that exceed acceptance threshold
- Comparison of Systems
 - Area under Curve
 - Equal Error Rate





Evaluation Tasks and Performance Metrics (Identification – Open Set)

- Open-Set Identification:
- Computes <u>one-to-many</u> similarity to determine the specific identity
- Person in probe image is known or unknown to the system
- Receiver Operating Characteristic (ROC)
 - Correct Acceptance and Identification Rate (CDIR) vs.
 False Acceptance Rate (FAR) at different thresholds
 - <u>CDIR</u>: fraction of genuine individuals that exceed acceptance thresh_{0.4} and are correctly identified
 - <u>FAR:</u> fraction of impostor individuals that exceed acceptance
 - Threshold

Comparison of Systems

- Area under Curve
- Equal Error Rate
- Red: Error from False Classification
- Blue: Error from False Rejection





Cross-factor face recognition

- Cross-Pose Face Recognition
- Severe decrease in accuracy from full-frontal to profile face verification
- Face frontalization in image and feature space
- Pose Invariant Models (PIM) to learn pose invariant features
- Cross-Age Face Recognition
- Extremely challenging due to changes in facial appearance over time
- Synthesize desired image with target age using GANs
- Methods that decompose ageing and identity components and thus extract age invariant representations
- Cross-Makeup Face Recognition
 - Increases intra-class variations
 - Synthesize non-makeup versions from makeup images
- Masked Face Recognition
- Face Recognition for masked faces
- Algorithms trained on eye region only





Evaluation of Face Recognition Biases in face recognition software

- Gender bias
 - Male
 - Female
 - Diverse
- Age bias
 - Children
 - Juvenile
 - Adult
 - Elderly people
- Ethnicity bias
 - Caucasian
 - Asian
 - Indian
 - African
 - ...



Bias in face recognition software

Gender, Age, Ethnicity

© Netflix: "Coded Bias"



How good is my Al System? Summary

- > 10 years of experience in the field of face analysis
 - Face recognition for media applications
 - Automatic annotation of images and videos
 - Unsupervised clustering of similar looking faces
 - TV program study
 - Face recognition on edge devices
 - IDCheck
 - Maskcognizer
 - Special use cases
 - Face recognition for animals
 - Evaluation of face recognition algorithms as a service
- Evaluation
 - The whole face recognition pipeline matters!
 - What use-cases are you interested in?
 - Use the correct evaluation metric for your use case!
 - Interpret the results correctly, and look at the errors in particular!
 - What biases and cross-matching scenarios can occur?







Contact

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Thank you for your attention!