

Automated Face Recognition

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- Integrated Pattern Recognition and Biometrics Lab
- Currently: 8 PhD Students + 2 Post-Docs +1 UG Student
- Graduated: 24 MS Thesis Students + 8 PhD Students

Research Theme

- **Adversarial Biometrics**

- Spoofing Biometric Traits
- Digitally Altered Biometric Data
- Degraded Biometric Data

- **Ethics and Privacy**

- What Else Does Your Biometric Data Reveal?
- Privacy Preserving Biometrics

- **Biometric Fusion**

- Multiple Biometrics
- Multispectral Biometrics
- Biometrics + Demographics + Spoof Detector + Quality

Related Resources

- A. K. Jain and A. Ross, "[Bridging the Gap: From Biometrics to Forensics](#)," Philosophical Transactions of The Royal Society B, Vol. 370, Issue 1674, August 2015.
- A. K. Jain, K. Nandakumar, A. Ross, "[50 Years of Biometric Research: Accomplishments, Challenges, and Opportunities](#)," Pattern Recognition Letters, Vol. 79, pp. 80 - 105, August 2016.
- A. Ross, S. Banerjee, C. Chen, A. Chowdhury, V. Mirjalili, R. Sharma, T. Swearingen and S. Yadav, "[Some Research Problems in Biometrics: The Future Beckons](#)," Proc. of 12th IAPR International Conference on Biometrics (ICB), (Crete, Greece), June 2019.
- A. K. Jain, B. Klare, A. Ross, "[Guidelines for Best Practices in Biometrics Research](#)," Proc. of 8th IAPR International Conference on Biometrics (ICB), (Phuket, Thailand), May 2015.


Biometrics

- Automated **recognition** of individuals based on their **biological** and **behavioral** characteristics
- Traits from which **distinguishing**, **repeatable** features can be extracted

C. L. Brown

Height	1m 79.6	Head l'gth	19.8	L. Foot	27.1	Circles	Leh	Age	22	Born in	
Eng. H'ght	5-10 3/4	Head width	16.3	L. Mid. F.	11.2	Periph Z		Apparent Age			
Outs. A	1m 75.5	Cheek width	14.4	L. Lit. F.	8.7	Color of Left Eye	Leh. Mel	Native	Louisville, Ky.		
Trunk	94.9	R. Ear	6.8	L. Fore A.	46.6	Pecul		Occupation	Johnson		

Remarks Incident to Measurement



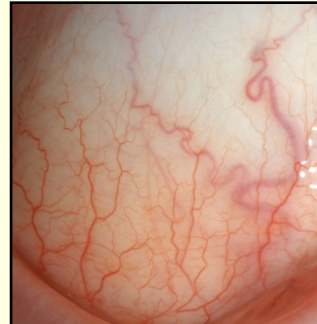
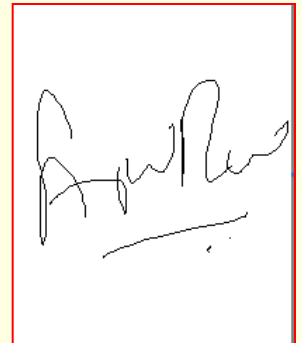
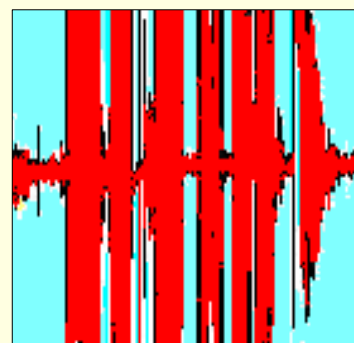
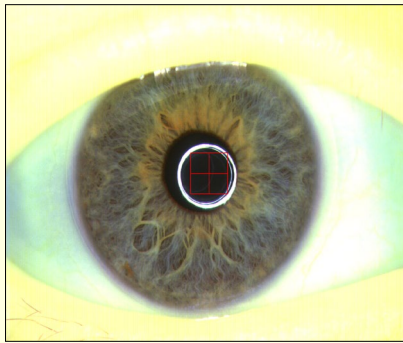
DESCRIPTIVE

Inch	Reck	Ridge	None	R. Ear		Beard	Shaved		
Height	M	Base	(Leh)	Root	Shel	Hair	Black		
Width	Br	DIMENSIONS			Teeth	Upper front	Complexion	M. Dark	
Pecul		Length	6r	Projection	6r	Build	M. Slim	Weight	165
		Pecul				Chin	M. Prom		

BUREAU OF IDENTIFICATION
Department of Police,
Tulane Ave. and Saratoga St.
New Orleans, La.

Measured Feb 1 1912
By Geo. B. Jones

Biometric Traits



Biometric Applications



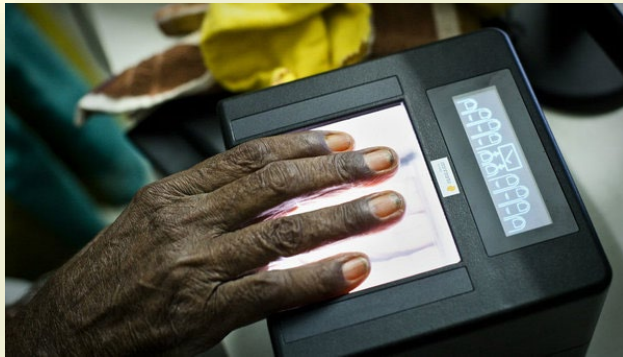
Iris: Health Care



Fingerprint: US OBIM



Face: Apple Face ID



Fingerprint: Refugee Services



**Finger Vein: Japan
ATMs**

Automated Face Recognition

- Given **two face images**, estimate two numbers:
 - the likelihood that they are of the **same person**
 - the likelihood that they are of **different people**



Components of a Biometric System

- **Sensor**: To acquire face image
- **Feature extractor**: To extract a set of discriminative features from the image
- **Matcher**: To compare two extracted feature sets
- **Database**: To store face templates of individuals

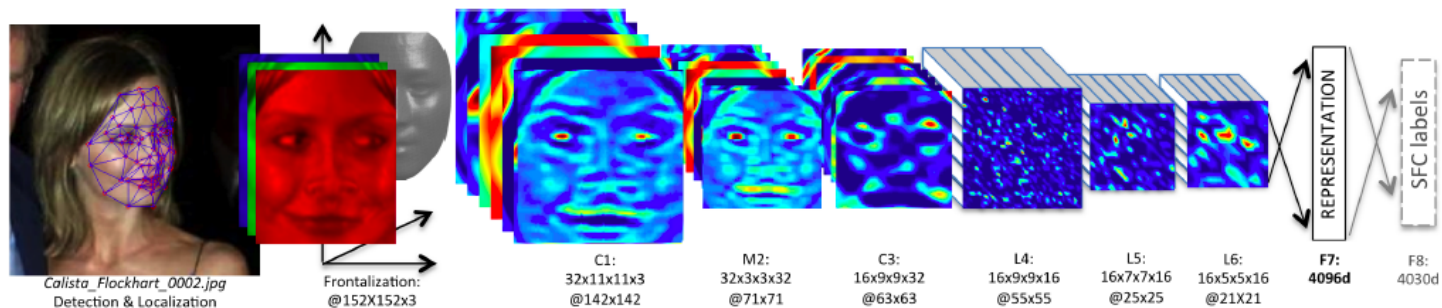
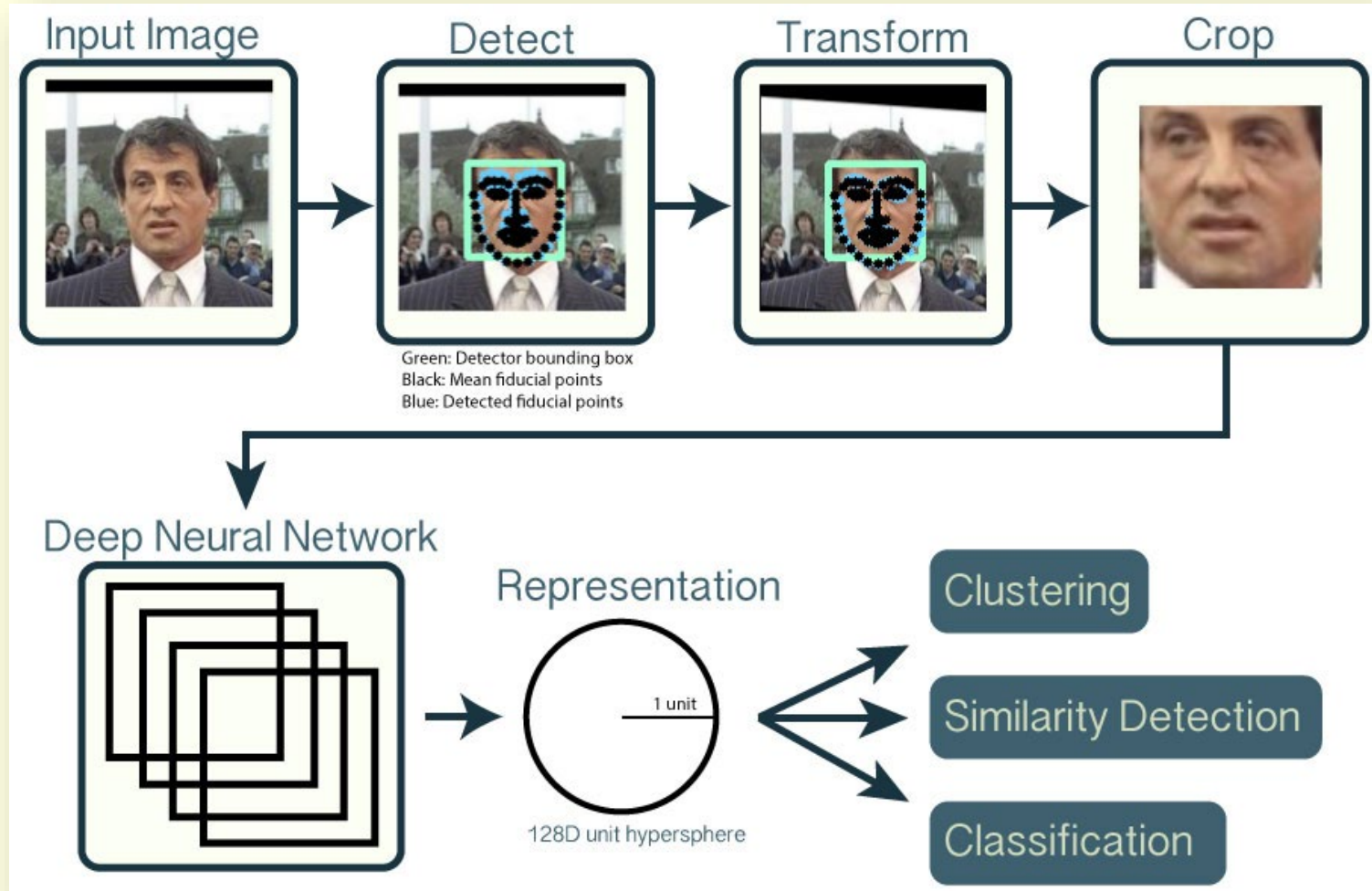


Figure 2. Outline of the DeepFace architecture. A front-end of a single convolution-pooling-convolution filtering on the rectified input, followed by three locally-connected layers and two fully-connected layers. Colors illustrate outputs for each layer. The net includes more than 120 million parameters, where more than 95% come from the local and fully connected layers.

Deep Neural Networks



© <https://cmusatyalab.github.io/openface/>

Verification vs Identification

Verification



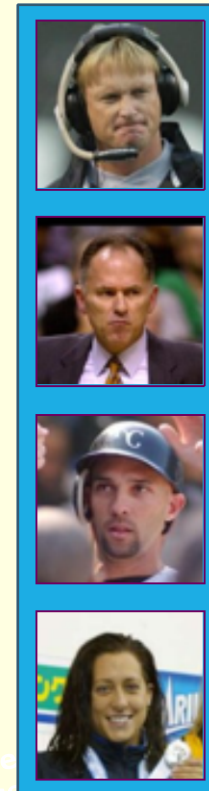
(Known Image of
Claimed Identity)



MATCH / NON-MATCH

Identification

Gallery



**RANKED
LIST OF
MATCHES**

Intra-user variations



© Nostra

FNMR: False Non-Match Rate (False Negative)

Inter-user similarity



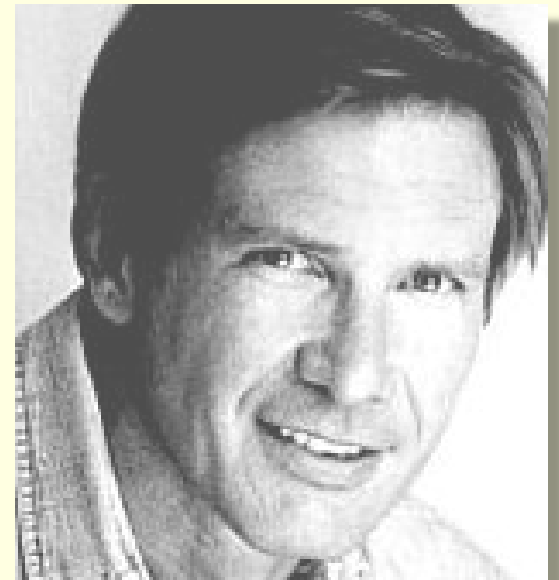
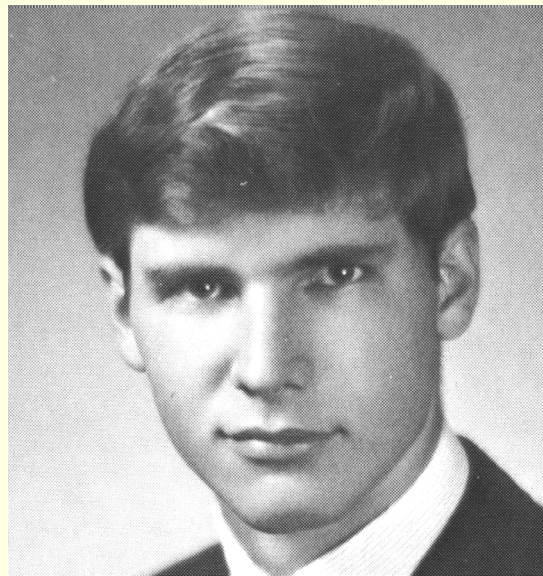
TWIN BROTHERS
© Martin Schoeller



MOTHER DAUGHTER
© PleasantonWeekly.Com

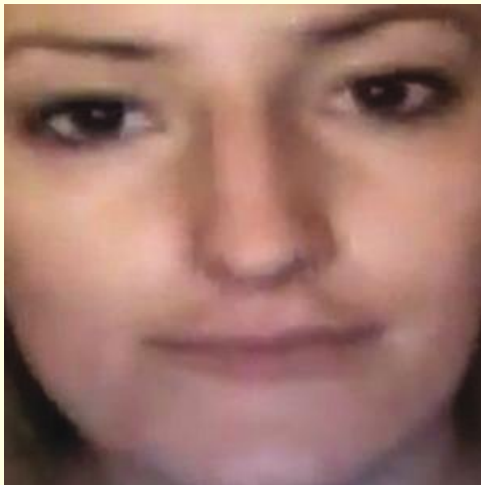
FMR: False Match Rate (False Positive)

Impact of Ageing



Impact of Cosmetics

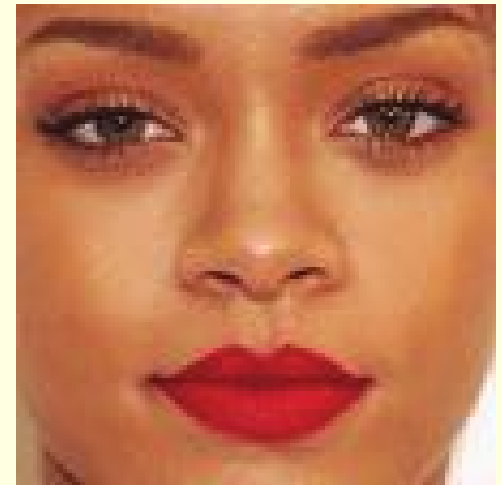
- **Cosmetics:** To spoof another person's face image



Before-makeup



After-makeup



Target identity

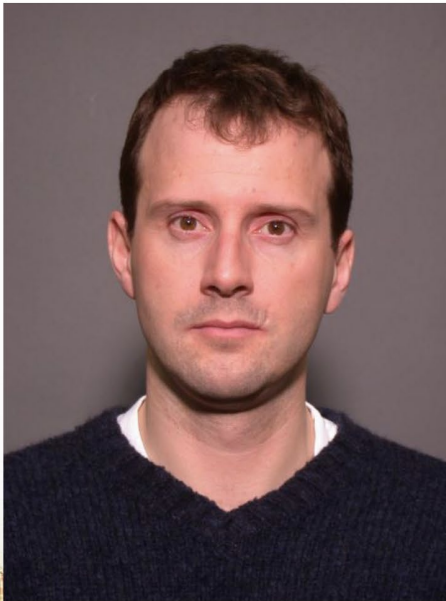
Rank 734 → Rank 1
[13,334 gallery images]

Chen et al, "Spoofing Faces Using Makeup: An Investigative Study", ISBA 2017

FRVT: Verification Scenarios



(a) Visa



(b) Mugshot



(c) Wild



(d) Border

Images from:
NIST 2019 Report

Ongoing Face Recognition Vendor Test (FRVT) Part 1: Verification
Patrick Grother, Mei Ngan, Kayee Hanaoka

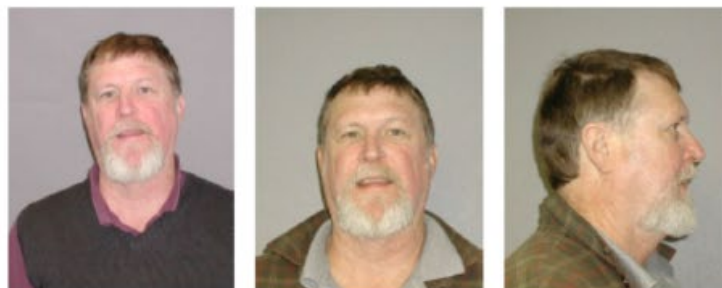
FRVT: Identification Scenarios

Unconstrained Faces



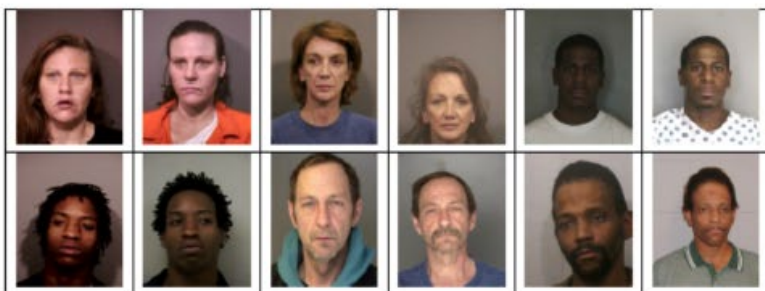
Surveillance and Access Systems

Profile Faces



Surveillance and Transactional Systems

Mugshot Images



Law Enforcement and Passport Type of Applications

Webcam Images



Immigration and Transactional Systems

Images from:
NIST 2019 Report

Ongoing Face Recognition Vendor Test (FRVT) Part 2: Identification
Patrick Grother, Mei Ngan, Kayee Hanaoka

Search Scenario

Error Rates on a 12M Face Image Search Database

Algorithm	Error Rates <i>FNIR @ FPIR = 0.001</i>	Template Size <i>Bytes</i>	Memory Requirements <i>GB</i>	Search Speed* <i>milliseconds</i>
NEC	0.058	1712	20.5	697
Paravision	0.106	4096	49.2	1417
RankOne	0.116	165	2.0	393
Innovatrics	0.142	1076	12.9	414
Microsoft	0.154	1024	12.3	2312
Idemia	0.166	528	6.3	880
Cognitec	0.184	2052	24.6	2088
Neurotechnology	0.214	2048	24.6	1604
Toshiba	0.214	1548	18.6	7250
Cogent	0.224	1043	12.5	3131
Aware	0.264	3100	37.2	924

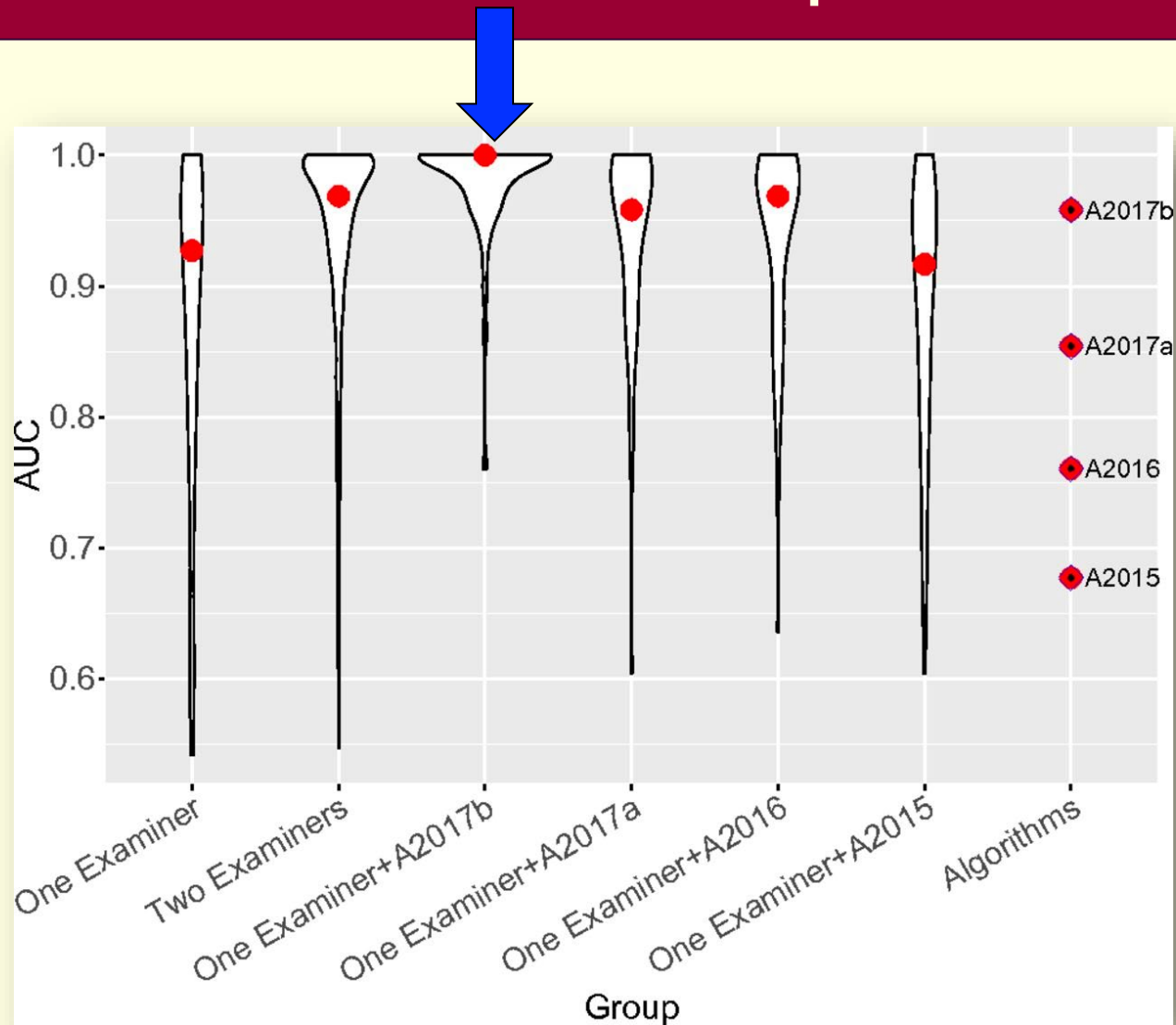
* Search time includes template generation and search speed

Humans versus Computers

- “We present data comparing **state-of-the-art** face recognition technology with the **best human** face identifiers”
- “The best machine performed **in the range** of the best humans: professional facial examiners”
- “However, optimal face identification was achieved only when humans and machines worked in **collaboration**”

Phillips et al., “Face recognition accuracy of forensic examiners, superrecognizers, and face recognition algorithms”, PNAS 2018

Humans + Computers



Phillips et al., "Face recognition accuracy of forensic examiners, superrecognizers, and face recognition algorithms", PNAS 2018

NIST Evaluation

- “Between 2014 and 2018, facial recognition software got **20 times better** at searching a database to find a matching photograph, according to the National Institute of Standards and Technology’s (NIST) evaluation of 127 software algorithms from 39 different developers—the bulk of the industry”

Please also see

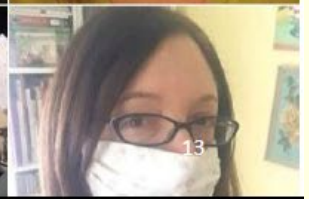
Grother *et al.*, “Ongoing Face Recognition Vendor Test (FRVT) Part 2: Identification,” NISTIR 8238, 2018

Novel Challenge



Masks

Briefly remove your mask for identity verification



Summary

- Performance of face recognition has **improved considerably** over the past decade
- Face recognition systems must be **used in conjunction** with human examiners/reviewers as well as other pieces of evidence
- Factors impacting performance:
 - **Quality** of probe and gallery images
 - Face recognition **algorithm** that was used
 - Size and composition of the **database**

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