

(Research in collaboration with Dr. Kevin Bowyer, University of Notre Dame) EAB/NIST FACE IMAGE QUALITY WORKSHOP (Virtual) Date: November 07, 2023

This talk is



- Not to bash the technology
- Not an anti law enforcement message
- Is a request the we implement stronger safeguards on what data gets processed by an AFR engine for low level crimes.

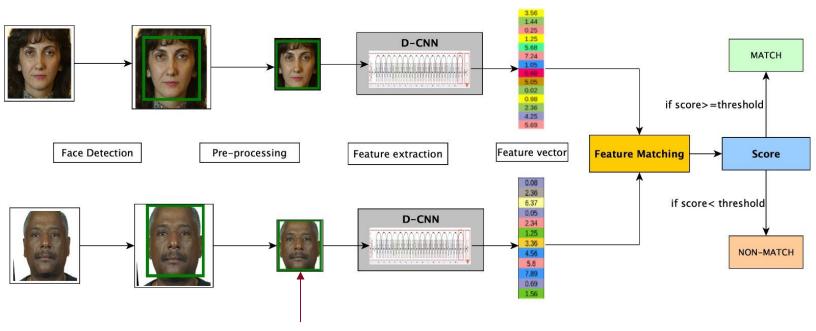
Two Common Statements on AFR (1st)



 "With good quality portrait photos, the most accurate algorithms will find matching entries, when present, in galleries containing 12 million individuals, with rank one miss rates of approaching 0.1%." [NIST FRVT 1:N Report]

DCNN FR Model/Process





Note the input resolution is usually about 112x112

Two Common Statements on AFR (2nd)



 "Facial recognition systems misidentify Black faces at a high rate. Facial recognition is less accurate in identifying people with darker skin tones—especially women. This can result in the misidentification of Black protesters or false positive matches in image databases."
 [Amnesty International]

Publicity about face recognition "bias"

The New York Eimes

ECHNOLOGY : Wrongfolly Account by an Algorithm

....

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He spent 10 days in jail after facial recognition software led to the arrest of the wrong man, lawsuit says

Updated Dec 29, 2020; Posted Dec 28, 2020



Nijeer Parks, 33, of Paterson. Photo courtesy of Daniel Sexton





MICHIGAN STATE POLICE

INVESTIGATIVE LEAD REPORT



LAW ENFORCEMENT SENSITIVE

THIS DOCUMENT IS NOT A POSITIVE IDENTIFICATION. IT IS AN INVESTIGATIVE LEAD ONLY AND IS NOT PROBABLE CAUSE TO ARREST. FURTHER INVESTIGATION IS NEEDED TO DEVELOP PROBABLE CAUSE TO ARREST.

BID DIA Identifier: BID-39641-19	Requester: CA Yager, Rathe
Date Searched: 03/11/2019	Requesting Agency: Detroit Police Department
Digital Image Examiner:	Case Number: 1810050167
	File Class/Crime Type: 3000

Probe Image	Investigative Lead





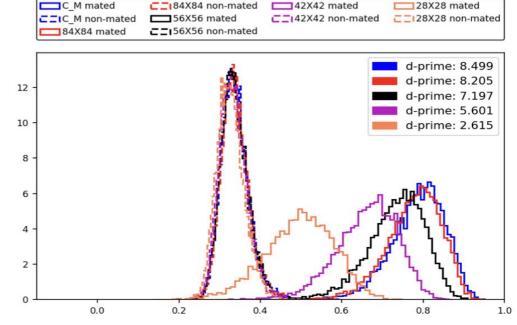






Probe Resolution Impacts Accuracy

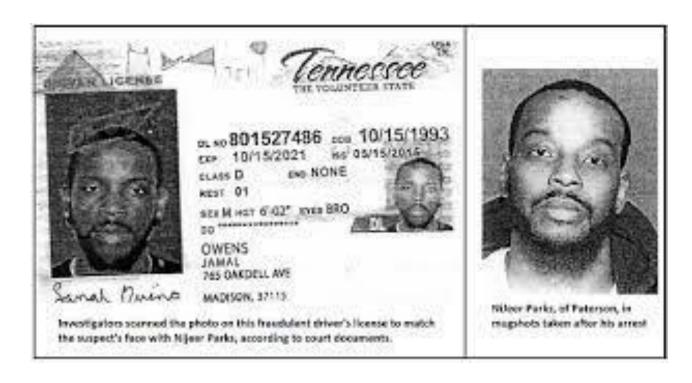
As the resolution decreases, the mated scores decrease and get closer to non-mated



Bhatta, A., Pangelinan, G., King, M. C., & Bowyer, K. W. (2023). Demographic Disparities in 1-to-Many Facial Identification. *ArXiv*. /abs/2309.04447

Probe Photo in Nijeer Parks Case

- Scan of fake driver's license dropped at scene saved as pdf doc.
- AFR Probe image was derived from PDF document.



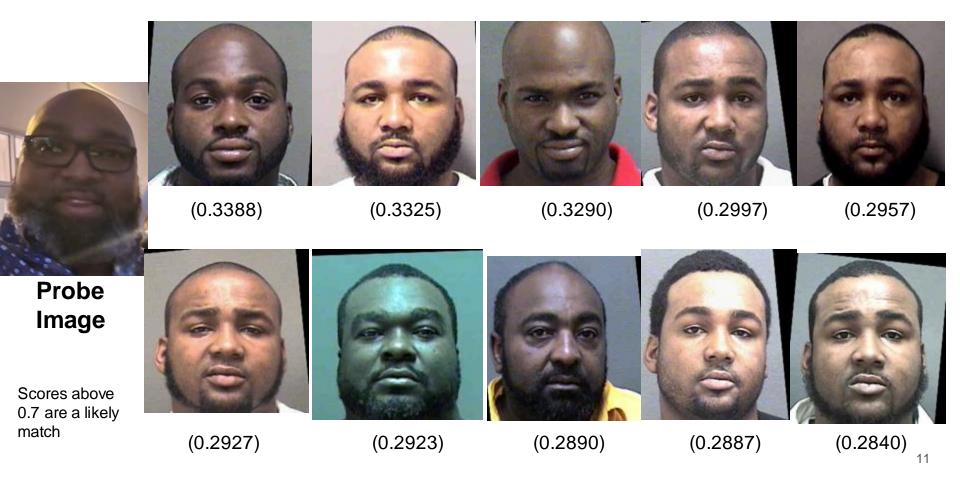




Probe Images

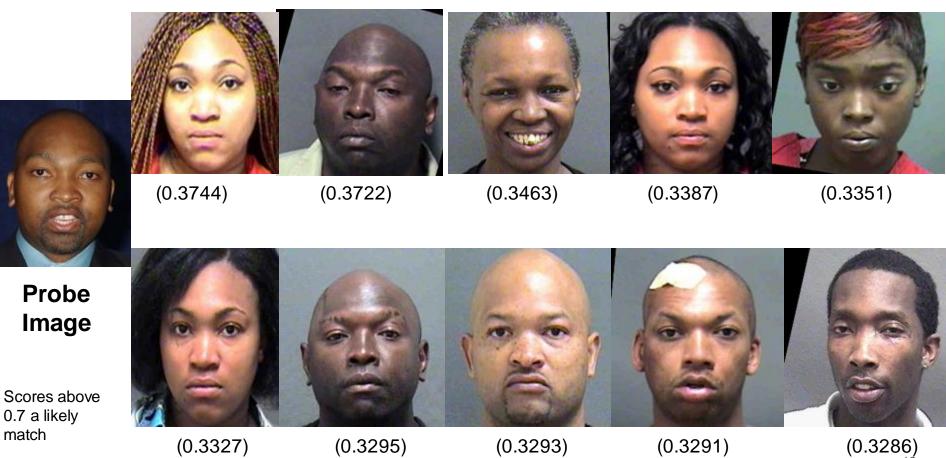


My Face Compared to Images in Morph (Top 10). I am not in the dataset)



My Face Compared to Images in Morph (Top 10). I am not in the dataset)

0.7 a likely match



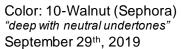
Key Messages



- Without a doubt, automated face recognition technology is highly accurate
- Demographic differentials in accuracy are also an issue (more of an issue for some algorithms than others.)
- For the investigative lead use-case, the error rate using top performing algorithms should be nearly the same for each demographic. (with caveats: such as size/composition of the gallery, number of searches performed, etc.)
- Rejecting poor quality probe images in low-level criminal searches may be necessary to prevent wrongful arrests
- Continue to correct the narrative: "leads" versus "hits" or "matches"

What's My Skin Tone?









Lancôme, Chicago, II; October 26th, 201914



The CHROMA-FIT Dataset:

Characterizing Human Ranges of MelAnin For Increased Tone-awareness

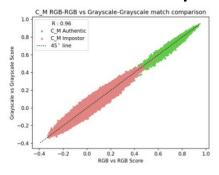


Florida Tech Identity Lab Team in Collaboration with the University of Houston, University of Miami, and New Mexico State University

Our Deep CNN Face Matchers Have Developed Achromatopsia



scatter plots of similarity scores from color-trained matcher

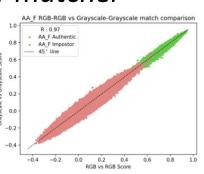


C_F RGB-RGB vs Grayscale-Grayscale match comparison

R: 0.97
C_F Authentic
C_F Impostro
C_F Impostro
C_F Impostro
C_F RGB-RGB vs Grayscale match comparison

R: 0.97
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R: 0.97
C_F RGB vs Grayscale match c



(a) Caucasian Male

(b) Caucasian Female

(c) African-American Male

(d) African-American Female

similarity score on X axis









similarity score on Y axis

https://arxiv.org/pdf/2309.05180.pdf



Thank you for your attention.

Questions?

Email: Michael C. King (michaelking@fit.edu); or Kevin Bowyer (kwb@nd.edu)