# Demographic Effects of Race on Face Recognition

Presented by:

Dr. Michael King, Associate Professor Department of Computer Engineering and Sciences Florida Institute of Technology

*November 27<sup>th</sup>, 2018* 

Collaborative Research effort with Dr. Kevin Bowyer, University of Notre Dame

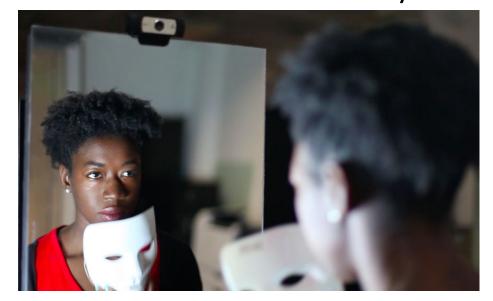
## Growing concerns about use of FR

 There have also been growing expressions of concern over "bias" in facial recognition algorithms from the Privacy and Civil Liberties community as well as several influential federal legislators.

 The Georgetown Law Center on Privacy & Technology has published reports detailing concerns relative to border security

and law enforcement and use of face recognition.

 A widely viewed TED talk by Joy Buolamwini: "How I'm fighting bias in algorithms"



## Prior Research on the topic

- Research related to face recognition accuracy varying between demographic groups goes back to at least FRVT 2002.
- A number of scholarly research publications through the years.
- However, there is relatively little work on this topic, and it more documents that effects exist than explains why they exist.

## **Objective (project ongoing)**

Investigate root causes of variations in performance accuracy of automated face recognition systems.

#### **Motivating questions:**

- 1. Verify what variations exist?
- 2. Understand why do they exist?

## **Experiment Details**

- 2 COTS algorithms + 1 open source CNN (ResNet)
- Limited publicly available data ideally suited to support this study
- Utilized the Morph Album2 Dataset collected by UNC-W
- Mugshot quality data used primarily to support academic research on facial aging

#### **Summary of Curated dataset**

	Original Total Images	No Face Visible - Dropped	Duplicate Image - Dropped	Labeling Error - Dropped	Labeling Error - Corrected	Total Images Dropped	Curated Dataset
Album 2	53,633	259	140	3	18	402	53,231
African- American	42,897	176	105	0	9	281	42,616
Caucasian	10,736	83	35	3	9	121	10,615

**Sample Images** 







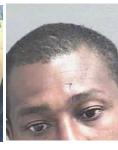


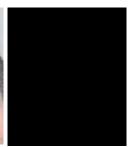


#### **MORPH Dataset: Curation Issues**

No face / partial face







Duplicate images



Original Label: Caucasian New Label: African-American









· Mis-labeled

### **RESULTS: Face Detection** Failure to Enroll (FTE) with MORPH Dataset

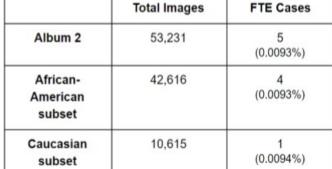
COTS A:

	Total Images	FTE Cases	
Album 2	53,231	135 (0.25%)	
African- American subset	42,616	109 (0.26%)	
Caucasian subset	10,615	26 (0.24%)	





Album 2 53,231 African-42,616 American



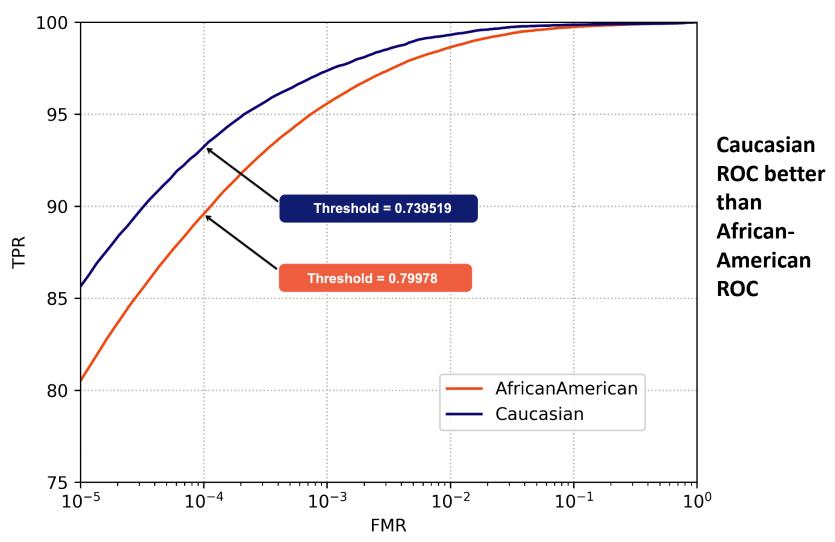




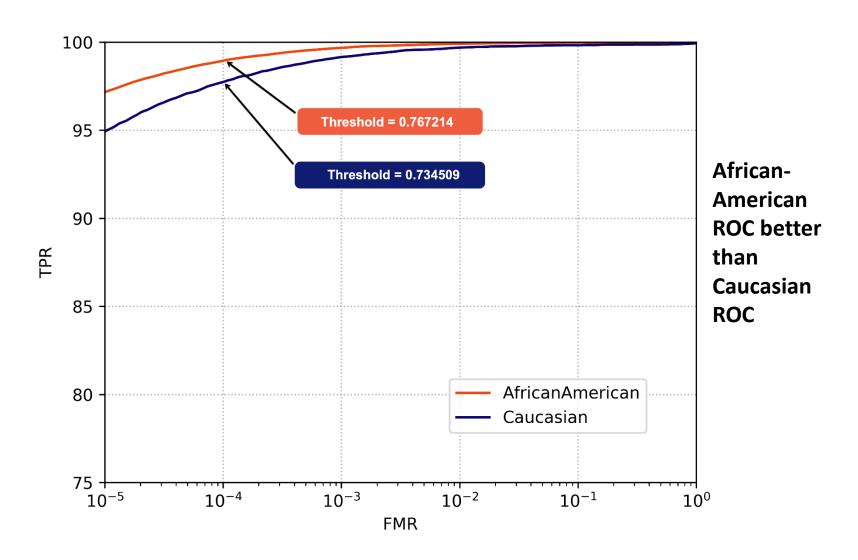
FTE rates are very similar

COTS B:

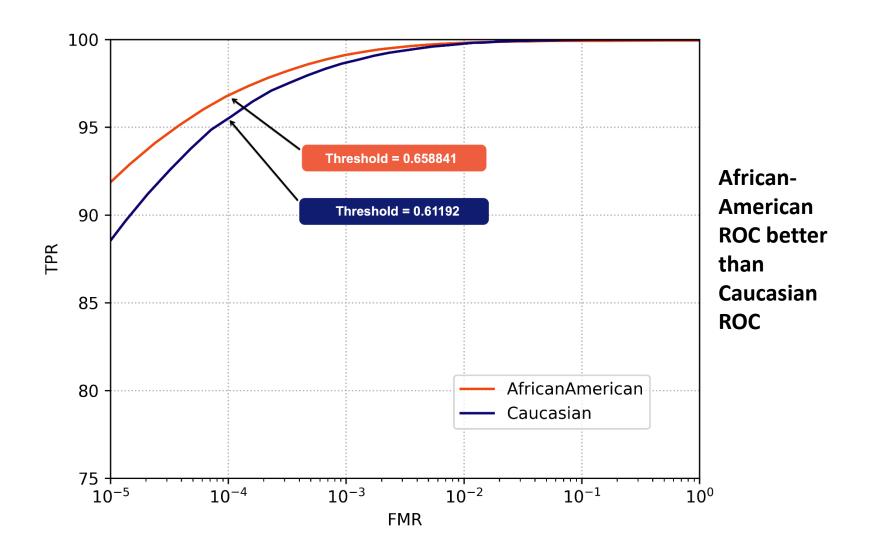
#### **COTS-A ROC Curves (Race)**



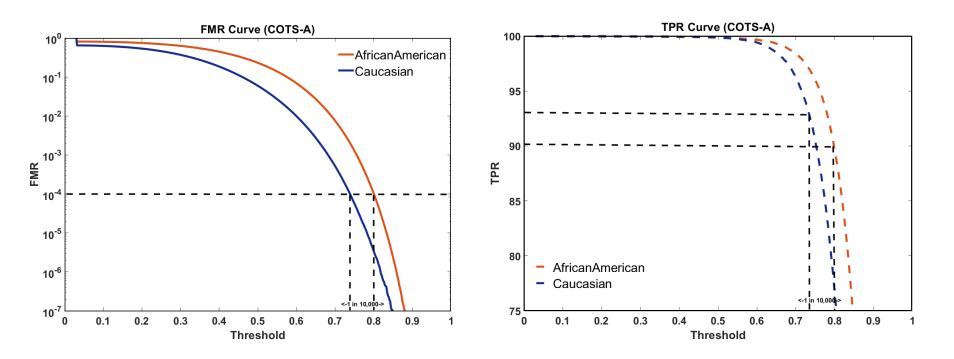
#### **COTS-B ROC Curves (Race)**



#### **ResNet ROC Curves (Race)**

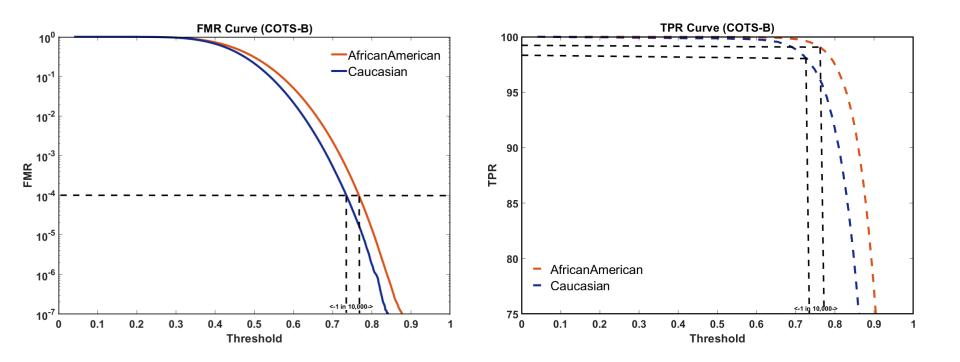


#### **COTS-A FMR and TPR Curves (Race)**



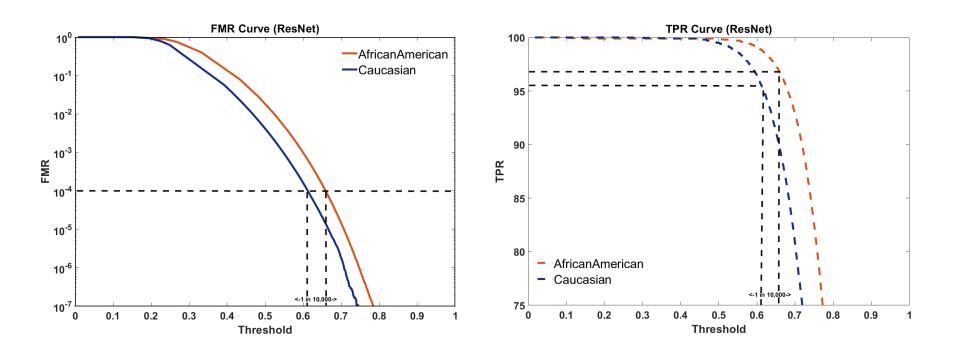
- For a desired FMR, the threshold setting would need to be different for each demographic group.
- On Morph, at a fixed decision threshold value, we find that the African-American scores generally result in a higher FMR and TPR than the Caucasian scores.

#### **COTS-B FMR and TPR Curves (Race)**



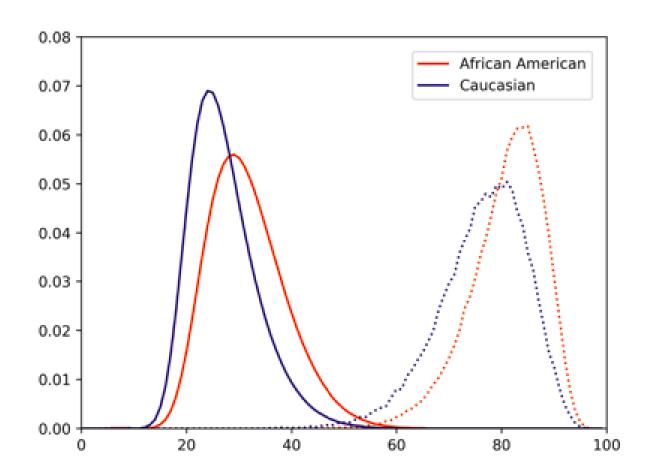
- For a desired FMR, the threshold setting would need to be different for each demographic group.
- On Morph, at a fixed decision threshold value, we find that the African-American scores generally result in a higher FMR and TPR than the Caucasian scores.

#### ResNet FMR and TPR Curves (Race)



- For a desired FMR, the threshold setting would need to be different for each demographic group.
- On Morph, at a fixed decision threshold value, we find that the African-American scores generally result in a higher FMR and TPR than the Caucasian scores.

#### **ResNet Authentic & Imposter Distributions (Race)**



African-American Impostor and Authentic both shifted to higher similarity scores.

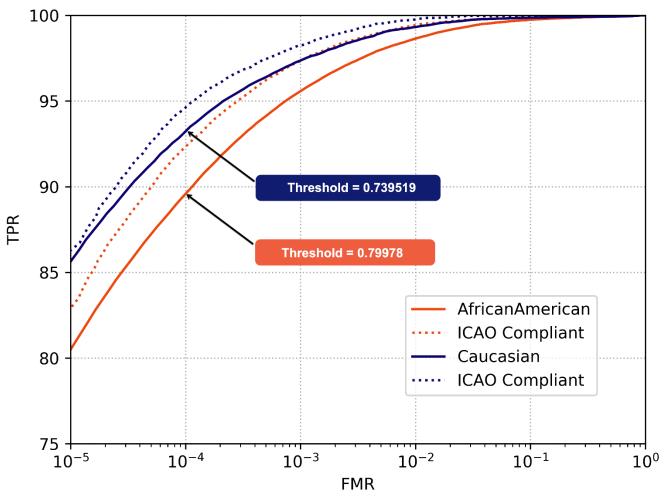
#### **Image Quality and ICAO Compliance**

- Used IFace SDK to check for ICAO compliance
- 48% of the African-American and 57% of Caucasian images are rated as ICAO compliant.



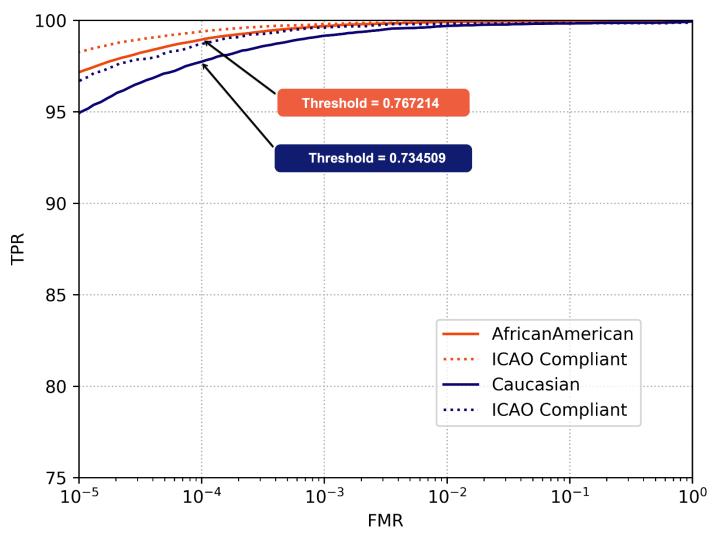
Example Images In Curated MORPH Dataset Rated Not ICAO-Compliant By IFace SDK (extended)

#### **COTS-A ROC Curves (ICAO)**



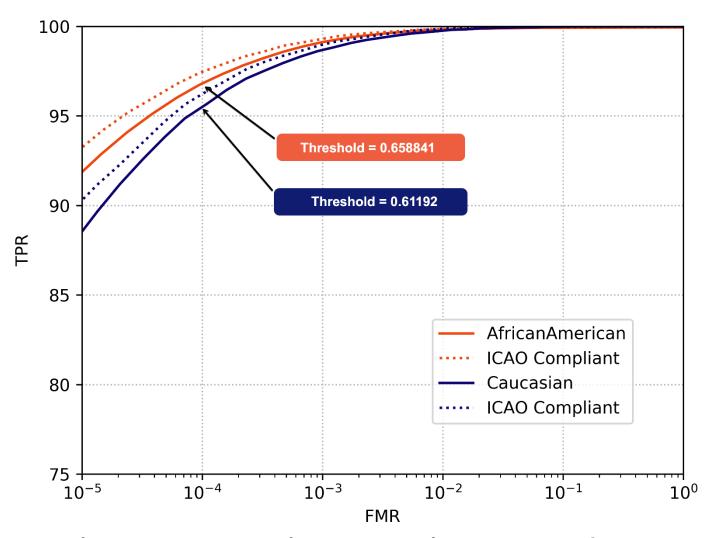
A simple ICAO compliance check improves performance and decreases gap between curves.

#### **COTS-B ROC Curves (ICAO)**



A simple ICAO compliance check improves performance

#### **ResNet ROC Curves (ICAO)**



A simple ICAO compliance check improves performance

#### **Takeaways:**

#### On the MORPH dataset,

- COTS A results in a better ROC curve for Caucasians than for African-Americans.
- COTS B and ResNet result in a better ROC curve for African-Americans than for Caucasians.
- ROC curves hide that comparison between groups is not achieved at the same decision threshold. A given TPR and FMR line drawn across the plot intersects the curves at different decision threshold values.
- A simple quality gate improves accuracy.

