

### Human-Algorithm Teaming in Face Recognition



The International Face Performance Conference 2020

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1

#### Disclaimer

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- The views presented here are those of the authors and do not represent those of the Department of Homeland Security, the U.S. Government, or their employers.
- The data used in this research was acquired under IRB protocol.
- This work was performed by a dedicated team of researchers at the Maryland Test Facility.



#### Introduction

- With recent improvements in face recognition (FR) accuracy, its adoption in DHS use-cases is growing
- DHS use-cases pose a unique context for face recognition use:
  - Face capture and matching is performed as an initial step
  - Face recognition results inform staff who are not FR experts:
    - CBP Officers, Airline Staff, or TS Officers
  - A high volume of individuals is processed each day
- If all transactions become biometric, may be largest USG use of face recognition by volume:
  - TSA screens 0.75 billion people each year (> 2 million a day, pre COVID19, [1])
  - CBP inspects over 0.39 billion people entering the US by air, land and sea each year [2]
  - Over 1.14 billion transactions combined!
- Due to the high volume, it is important to optimize system performance to reduce error, including the way staff review face recognition algorithm results:
  - What we call human-algorithm teaming.

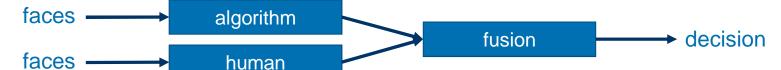


### The DHS Face Recognition Use-Case

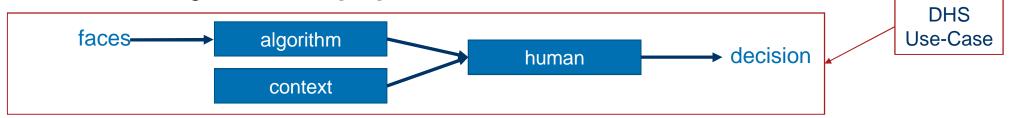
Traditionally, face recognition was performed by humans:



Parallel Human-Algorithm Process [1]



Serial Human-Algorithm Process [2, 3]





<sup>[3]</sup> White, David, et al. "Error rates in users of automatic face recognition software." PloS one 10.10 (2015): e0139827.

### **Human Face Matching Performance**

- Unlike human recognition of familiar faces, human performance on unfamiliar face matching is poor
  - 70% accuracy for passport officers with unfamiliar faces [1]
  - 50% accuracy for passport officers identifying matches on algorithm-provided candidate lists [2]
- Human identity verification performance depends on our ability to:
  - Perceive face similarity directly
    - Perceptual learning (e.g. other race effect [3])
    - Adaptation (e.g. "after-effects" [4])
    - Attention (task [5] or features [6])
  - Integrate evidence from other sources
    - Collaboration (e.g. working in pairs [7])
    - Algorithm decision aids (e.g. [8])
    - Tools and heuristics (e.g. as used by forensic examiners [9])
    - Context, including non-face information
- How humans integrate information from algorithms in face matching decisions is not well understood
- We studied how algorithm outcomes influence subsequent human judgements of face similarity



[1] White, David, et al. "Passport officers' errors in face matching." PloS one 9.8 (2014): e103510.

[2] White, David, et al. "Error rates in users of automatic face recognition software." PloS one 10.10 (2015): e0139827.

[3] Walker, Pamela M., and James W. Tanaka. "An encoding advantage for own-race versus other-race faces." Perception 32.9 (2003): 1117-1125.

[4] Leopold, David A., et al. "Prototype-referenced shape encoding revealed by high-level aftereffects." Nature neuroscience 4.1 (2001): 89-94.

[5] Wojciulik, Ewa, Nancy Kanwisher, and Jon Driver. "Covert visual attention modulates face-specific activity in the human fusiform gyrus: fMRI study." Journal of neurophysiology 79.3 (1998): 1574-1578.

[6] Fletcher, Kingsley I., Marcus A. Butavicius, and Michael D. Lee. "Attention to internal face features in unfamiliar face matching." British Journal of Psychology 99.3 (2008): 379-394. [7] Dowsett, Andrew J., and A. Mike Burton. "Unfamiliar face matching: Pairs out-perform individuals and provide a route to training." British journal of psychology 106.3 (2015): 433-445.

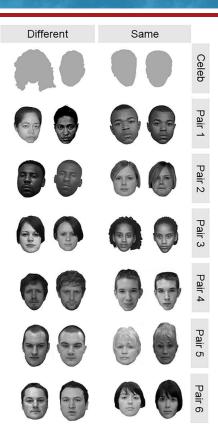
[7] Dowsett, Antidew C., and A. wince button. Ornamina face matering. Fars our-perform introducts a foreign foreign foreign face of the fa

[9] Phillips, P. Jonathon, et al. "Face recognition accuracy of forensic examiners, superrecognizers, and face recognition algorithms." Proceedings of the National Academy of Sciences 115.24 (2018): 6171-6176.

### **Face Matching Task**

- Experiments carried out at MdTF during the course of technology testing
  - Including 2019 Biometric Technology Rally
  - https://mdtf.org/Rally2019
- 343 volunteers performed the task
  - Diverse age, race, and gender
- Face matching task modified from the Glasgow Face Matching Test [1]
  - Added diverse face stimuli from the MEDS dataset to better match volunteer demographics [2]
  - Added familiar celebrity faces to detect appropriate task performance
- There was no time limit, but most volunteers finished in less than 15minutes





### **Face Matching Task**

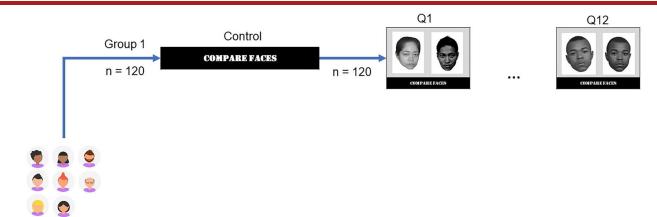


- -3 I am absolutely certain these are different people
- -2 I am mostly certain these are different people
- -1 I am somewhat certain this is the different person
- 0 I am not sure
- 1 I am somewhat certain these are same people
- 2 I am mostly certain this is the same person
- 3 I am absolutely certain this is the same person



# Face Matching Task with Prior Identity Information

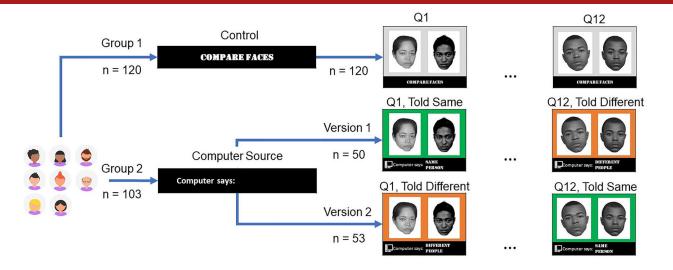
- Volunteers were assigned to 1 of 3 groups
- Control:
  - no prior identity information





# Face Matching Task with Prior Identity Information

- Volunteers were assigned to 1 of 3 groups
- Control:
  - no prior identity information
- Computer Source:
  - told that a computer previously reviewed the faces and made a same / different decision

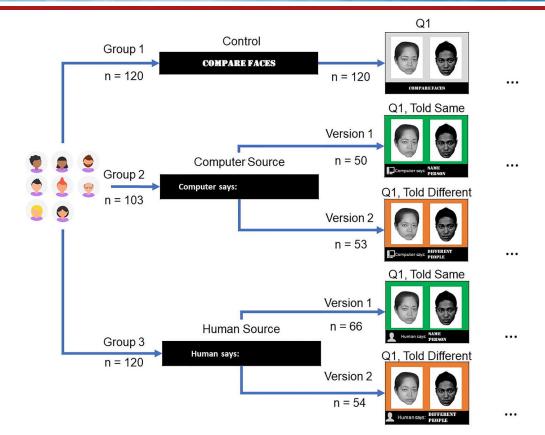




# Face Matching Task with Prior Identity Information

- Volunteers were assigned to 1 of 3 groups
- Control:
  - no prior identity information
- Computer Source:
  - told that a computer previously reviewed the faces and made a same / different decision
- Human Source:
  - Told that a human previously reviewed the faces and made a same / different decision





Q12

Q12. Told Different

Q12, Told Same

Q12, Told Different

Q12, Told Same

## What is the Impact of Prior Identity Information?

Does the source of prior identity information (human vs. computer) affect face similarity judgements?

Does the provided prior identity information (same vs. different) affect face similarity judgements?

Human vs. Computer

| Image: Computer says: SAME | Computer says: PERSON | PERSON |

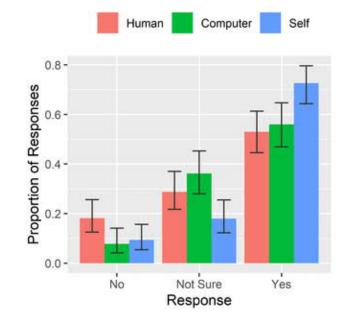
#### Same vs. Different





### Trust in Prior Identity Information Source

- Trust is an important factor in determining reliance on decision aids [1]
- We asked volunteers in each group to indicate their trust in the prior identity source:
  - Control: Do you trust yourself to identify a person?
  - Human: Do you trust a human to identify a person?
  - Computer: Do you trust a computer to identify a person?
- Responses indicated that:
  - Volunteers trust their own abilities to identify
  - Volunteers distrusted other people more than volunteers distrusted algorithms





### No Effect of Prior Identity Information Source

Introducing prior identity information did not affect overall task performance

Source	N	Accuracy	FPR	TPR
Control	120	0.75	0.19	0.70
Human	120	0.74	0.20	0.69
Computer	103	0.73	0.22	0.69

- Performance results were comparable to standard GFMT norms:
  - GFMT (short version) accuracy average is 0.81 [1]

Human vs. Computer







# Prior Identity Information Biases Human Responses

 Prior identity information decisions (same vs. different) did not change accuracy, but modulated False Positives and True Positives

Source	N	Accuracy	FPR	TPR
Control	120	0.75	0.19	0.70
Same	223	0.73	0.25	0.72
Different	223	0.75	0.17	0.66

- False positives and true positives:
  - Increased together if the prior identity decision was "same"
  - Decreased together if the prior identity was "different"
- But this is just at one threshold, what do things look like across thresholds?



Same vs. Different



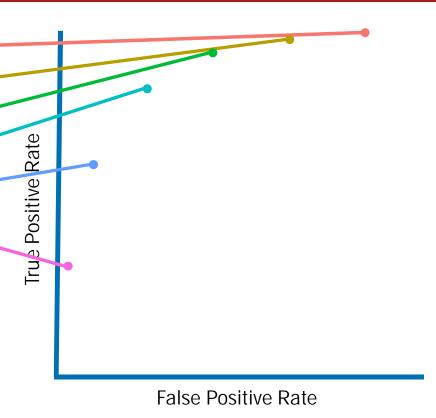
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False Positive Rate

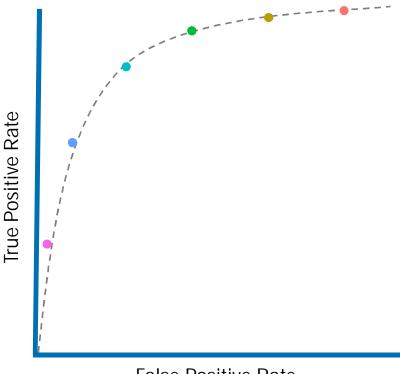


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- I am absolutely certain these are different people -3
- I am mostly certain these are different people -2
- I am somewhat certain this is the different person -1
- I am not sure 0
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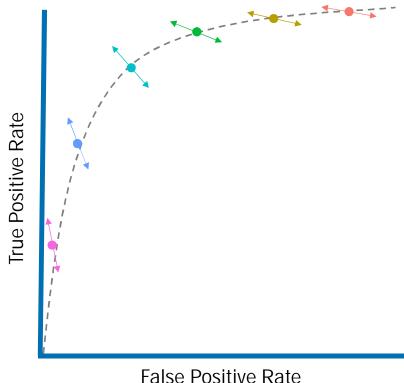


**False Positive Rate** 



#### Sensitivity

measures how well volunteers distinguish "same" and "different" face pairs





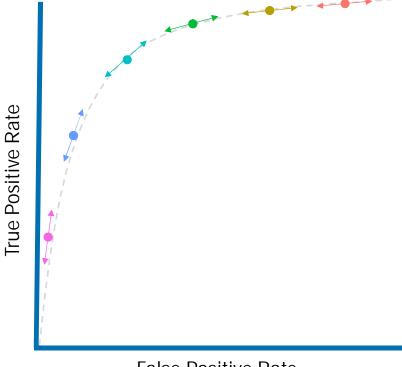


#### Sensitivity

 measures how well volunteers distinguish "same" and "different" face pairs

#### Criterion

 measures whether volunteers are biased toward higher or lower similarity ratings



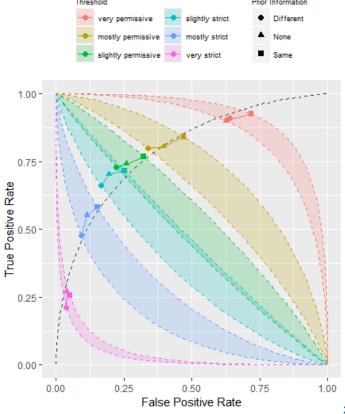




# Prior Identity Information Cognitively Biases Human Responses

- At each threshold (color) prior information moved responses along the ROC curve
- This is consistent with a shift in the Criterion and no change in Sensitivity
- The overlap in some shaded regions means prior identity information could shift responses by a whole step on the confidence scale:
  - I am not sure → I am somewhat sure
- The effect of the prior identity decision was present, but modest, humans trusted their own perceptual abilities





### Reducing Available Face Information

- What happens when perceptual abilities are degraded?
  - You may have noticed that it has recently become more difficult to recognize even familiar faces
  - We are all wearing face masks in public!
- We repeated the experiment on a new sample of 153 volunteers, but using face pairs where one of the faces was occluded with a digital face mask
- Face matching performance was reduced in the presence of digital masks:

Source	Mask Usage	N	Accuracy
Control	No Mask	52	0.83
Computer	No Mask	51	0.80
Computer	Mask	50	0.71

Same vs. Different

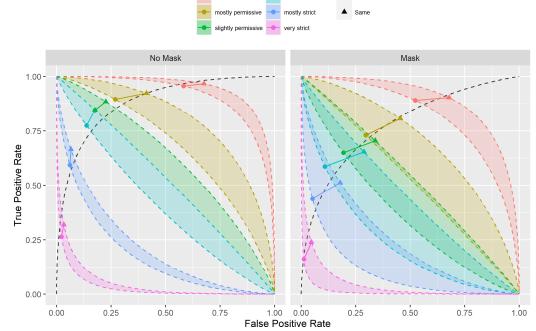






### Reducing Available Face Information

- With unmasked pairs
  - Prior identity information moved responses primarily along the ROC curve
  - Slight overlap in some shaded regions
  - Effect of prior identity information was present but modest
- With masked face pairs
  - Prior identity information effects were magnified
  - Now almost all shaded regions overlap
- When sensory input is degraded, humans will rely more on prior identity decisions

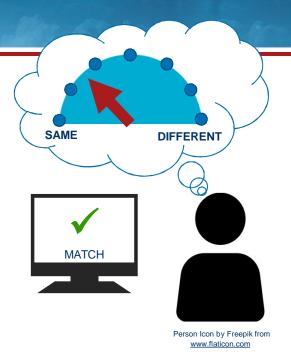




#### Conclusions

- Human face similarity judgements are systematically biased by prior identity decisions
  - However, humans trust in and rely on their own perception when making decisions
- The influence of prior identity decisions grows markedly when the matching task becomes harder
  - In the presence of face masks, humans altered their responses based on prior identity decisions to a greater extent
- These interactions should be taken into account when considering the performance of human-algorithm teams
- For the DHS use-case, human-algorithm team performance may not faces be easily predicted from studies investigating humans and algorithms in isolation







#### **Questions?**

- This work was performed by a team of researchers at the Maryland Test Facility.
  - Full paper available: <a href="https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0237855">https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0237855</a>
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