

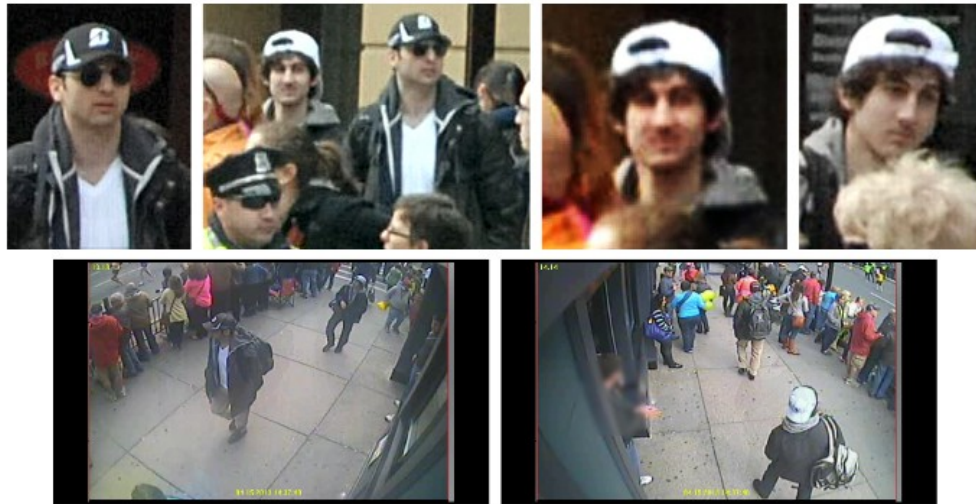


Massachusetts
Institute of
Technology

Subtasks of Unconstrained Face Recognition

Leibo*, Liao*, and Poggio

* = equal contribution



The Boston Marathon Bombings – Investigation Timeline



April 15th 2:49 p.m.
Explosions near Boston
Marathon finish line.



April 18th 5:00 p.m.
Two suspects
revealed.



April 18th 10:48 p.m.
Manhunt begins after
shooting and carjacking.



April 19th 6:45 a.m.
Suspects positively
identified.



April 19th 8:42 p.m.
Dzhokhar Tsarnaev
captured.



Opportunity for Facial Recognition

(Klontz & Jain 2013)

Unconstrained face recognition

Unconstrained face recognition



OFFICE OF THE DIRECTOR OF NATIONAL INTELLIGENCE

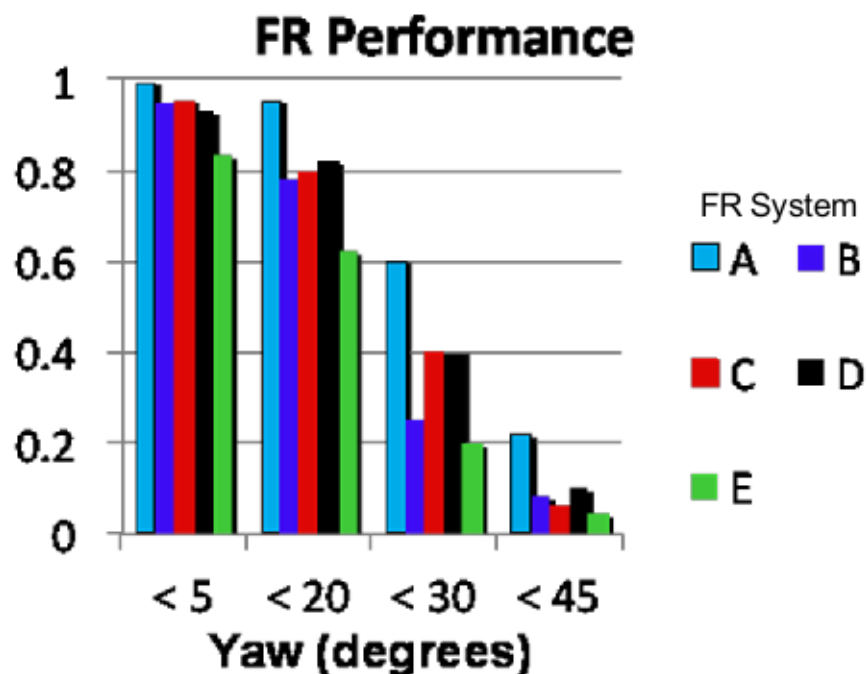
LEADING INTELLIGENCE INTEGRATION



Janus Program Concept

Advance the state-of-the-art in face recognition from using mugshots to working with operationally relevant image sources (i.e., media in the wild) using model-based recognition

Unconstrained face recognition



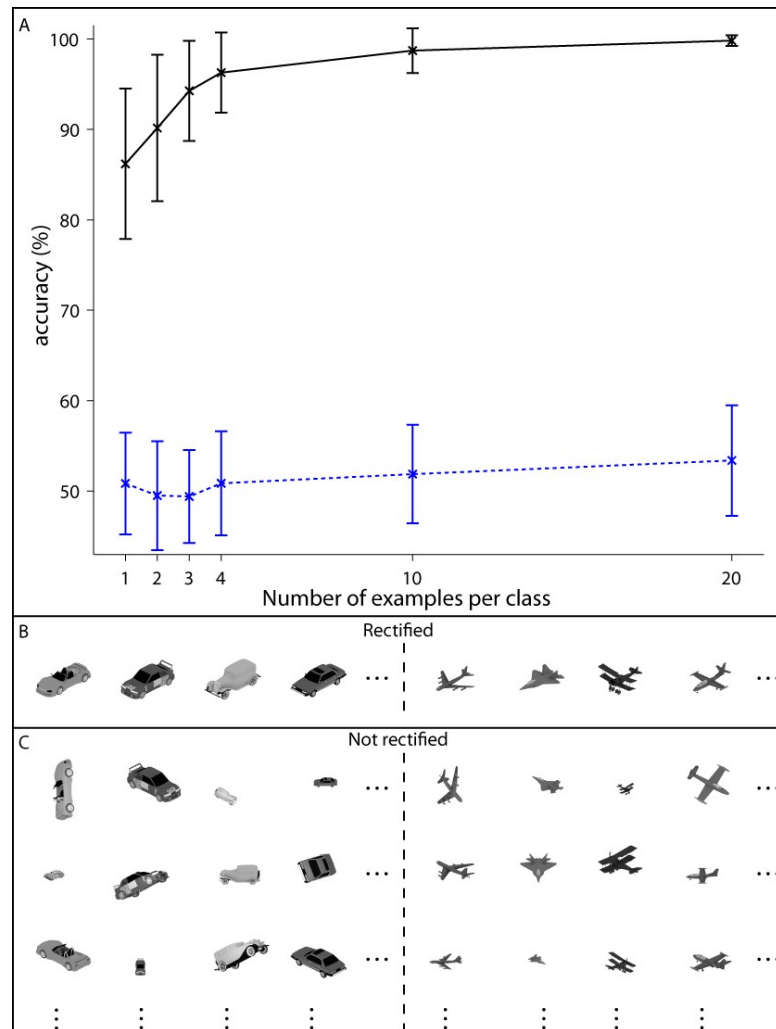
“Factors such as aging, pose, illumination, and expression (A-PIE) can not only decrease performance, they can cause its catastrophic failure”

- (IARPA JANUS announcement)

Negative impact on performance (higher better) when changing just a single factor (yaw)

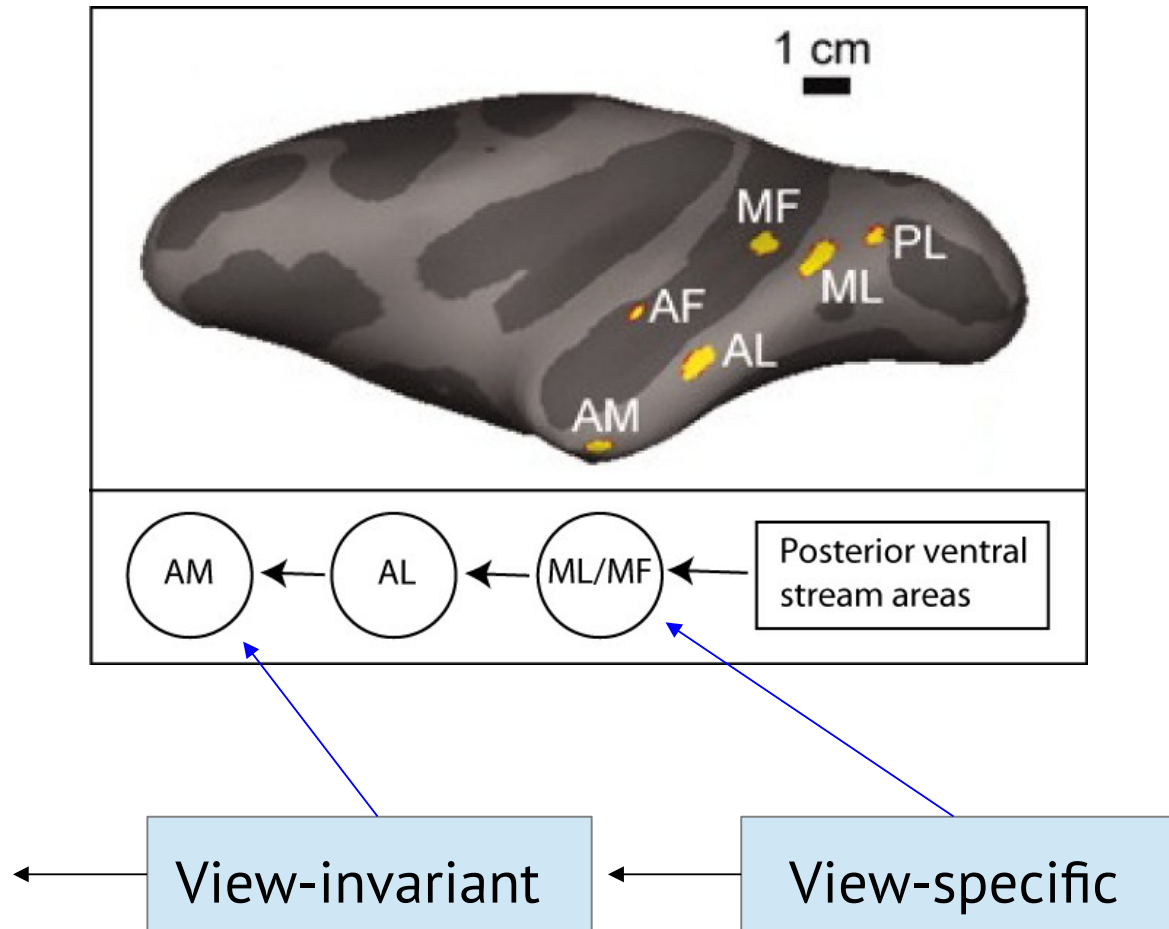
– NIST Multiple Biometric Evaluation 2010

Transformations



(Anselmi et al. 2013)

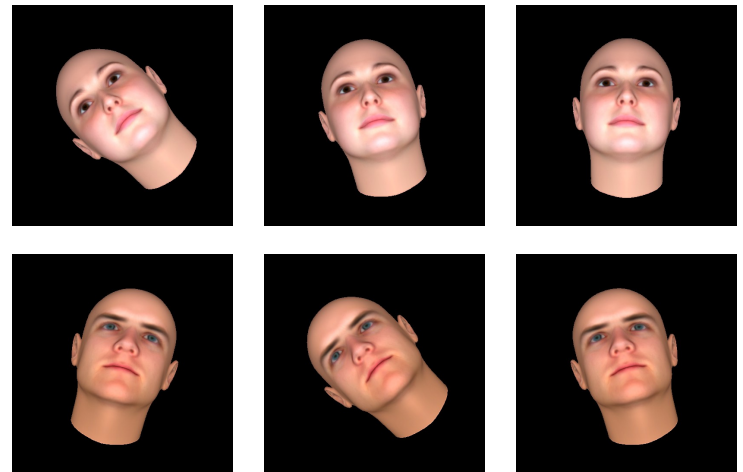
Transformations



(Freiwald & Tsao 2010)

Subtasks of unconstrained face recognition

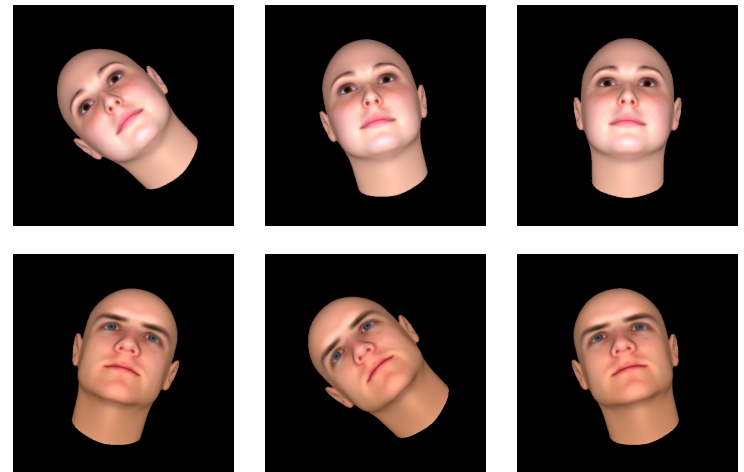
A collection of synthetic datasets constituting a partial decomposition of the unconstrained problem into subtasks



Subtasks of unconstrained face recognition

A collection of synthetic datasets constituting a partial decomposition of the unconstrained problem into subtasks

- 400 faces
- All rendered under specific transformation conditions for each “subtask”
- Same-different tasks (like Labeled Faces in the Wild).

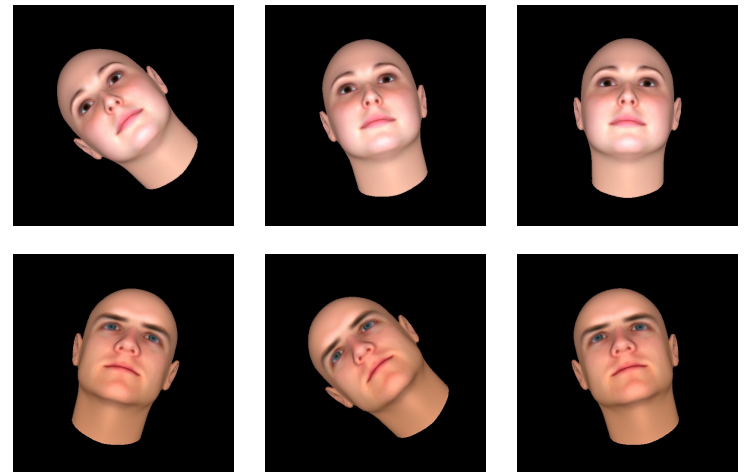


Subtasks of unconstrained face recognition

A collection of synthetic datasets constituting a partial decomposition of the unconstrained problem into subtasks

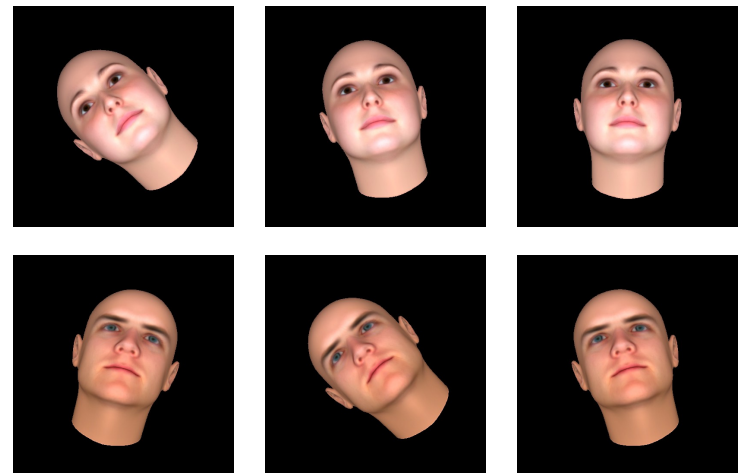
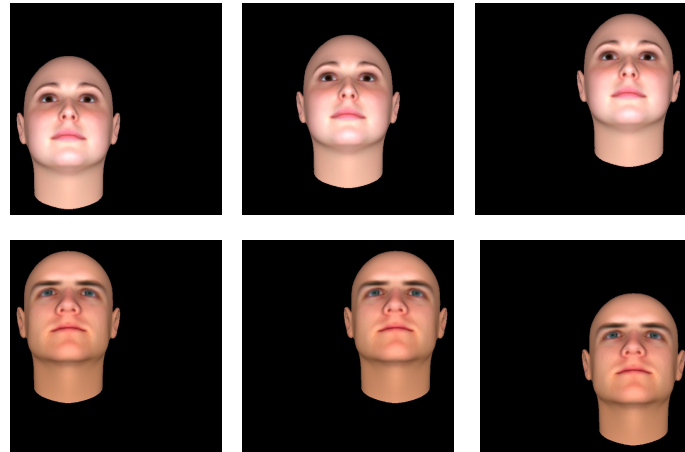
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Unit tests for unconstrained face recognition



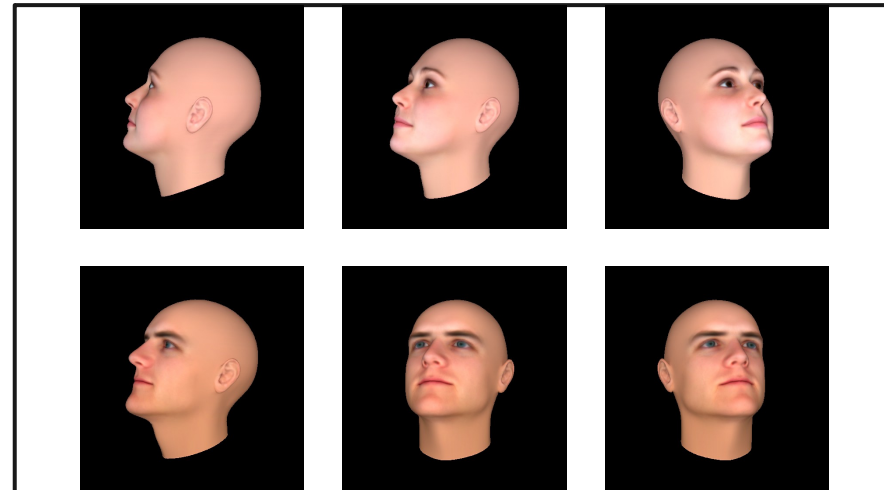
Affine subtasks

- Translation
- Scaling
- In-plane rotation

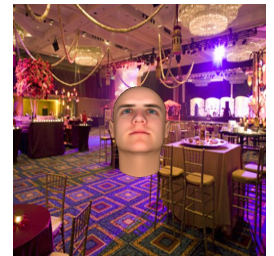
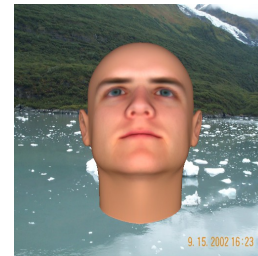
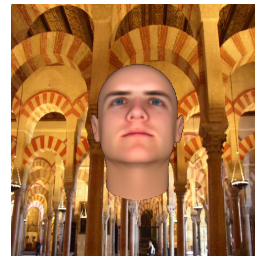
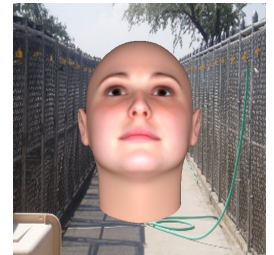
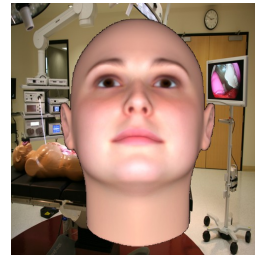
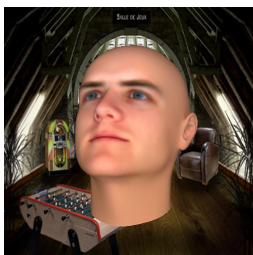
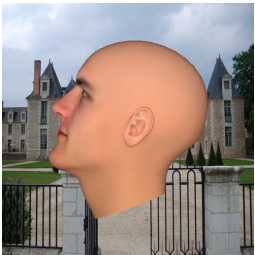
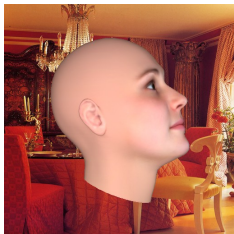
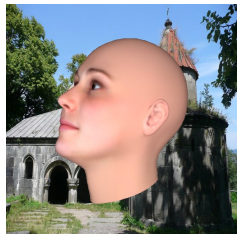


Non-affine subtasks

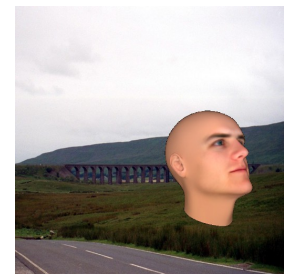
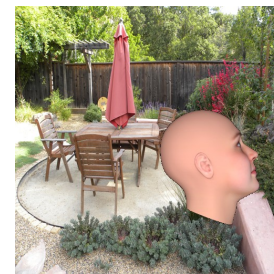
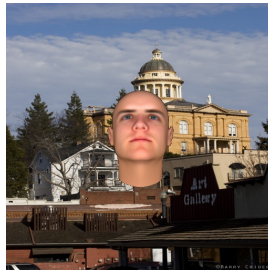
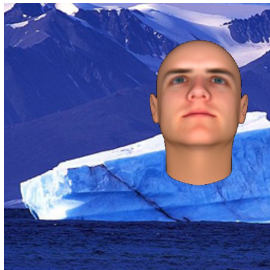
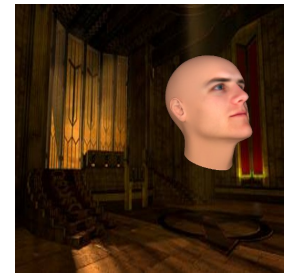
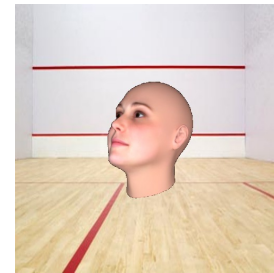
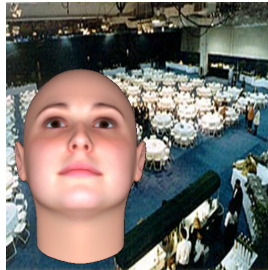
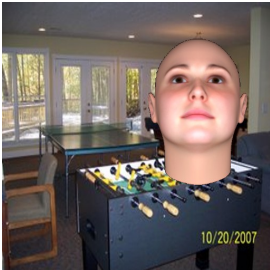
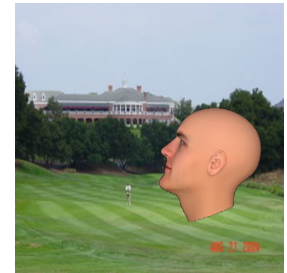
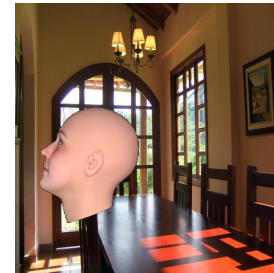
- Yaw rotation
- Pitch rotation
- Illumination



Clutter subtasks



Interaction subtasks



Alternative subtasks



Alternative subtasks



**Labeled faces
in the wild**

- Huang et al. 2008
- Leibo et al. 2014

Alternative subtasks



**Labeled faces
in the wild**

- *Aligned*

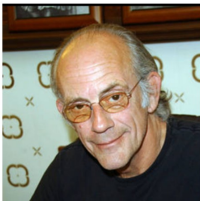
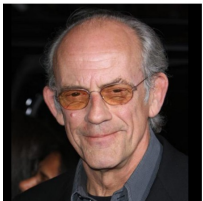
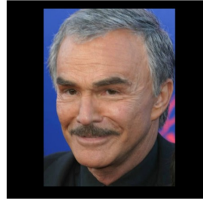
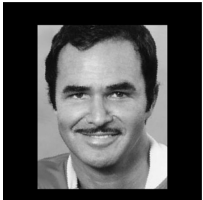
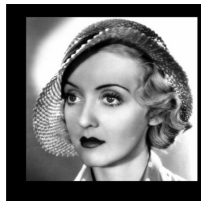
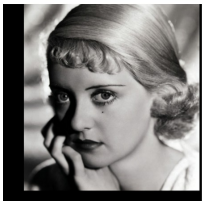


95+ of 123 papers on LFW indexed
by Google Scholar actually used
LFW-a, closely cropped

- Huang et al. 2008
- Leibo et al. 2014
- Taigman et al. Etc.

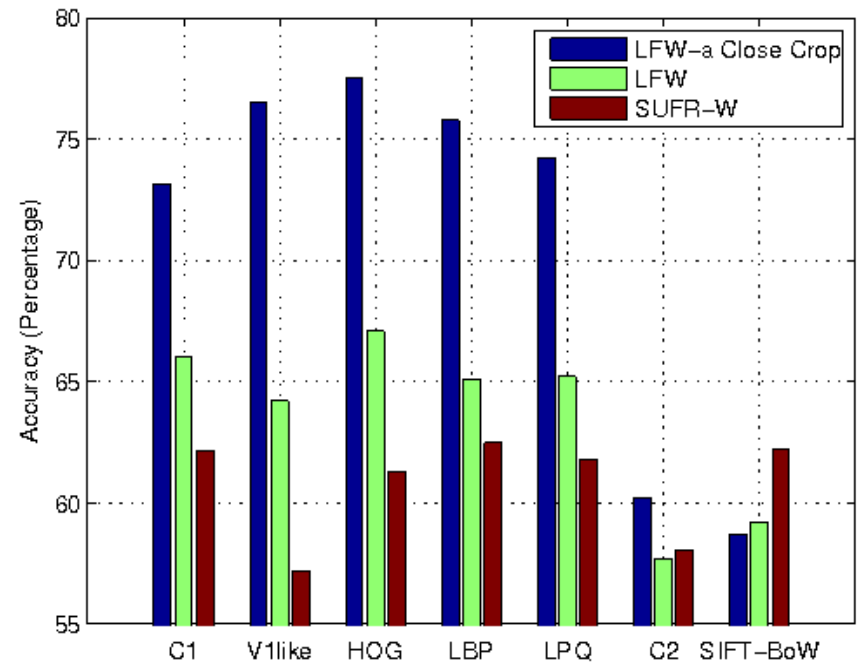
SUFR-in the Wild (SUFR-W)

- A new benchmark we proposed.
- Comparable to LFW.
- 13,661 images to LFW's 13,233



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Conclusion

- We advocate an algorithm development approach combining both synthetic and natural, unconstrained data.
- Even if you will ultimately be working within a DAR pipeline, recognition systems that can handle transformation invariance is useful for recovering from errors of alignment. No single point of failure.
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All datasets available from CBMM.MIT.EDU

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