

# Face Search at Scale

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# Information Content in a Face



**Identity:** John

**Demographics:**

Age: ~ 40; gender: male;  
ethnicity: white

**Attributes:**

Hair: Short, Brown

Moustache: Yes

Beard: Yes

Mole: Yes

Scar: Yes



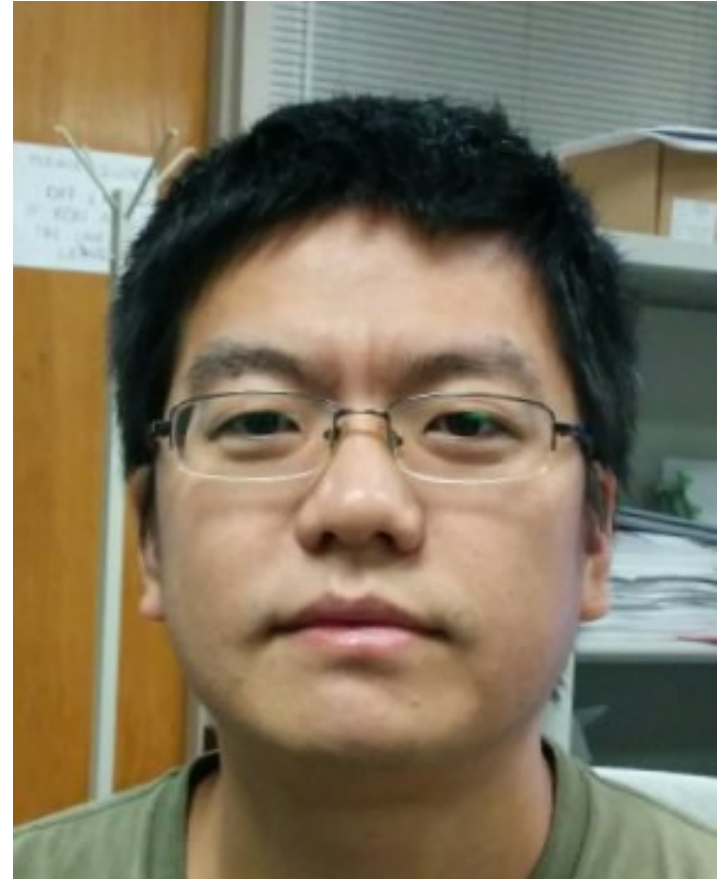
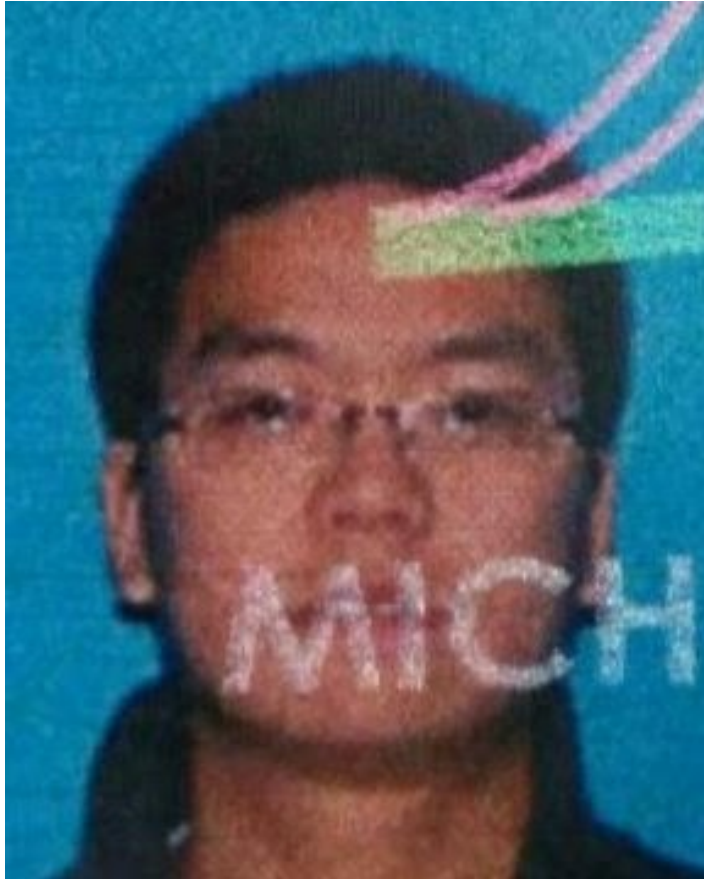
**Social cues:**

Expression, emotion,...

# Outline

- Face Recognition
- Applications
- Challenges
- State of the Art
- Summary

# Face Verification



**Same Person?**

# Face Search

*Probe*

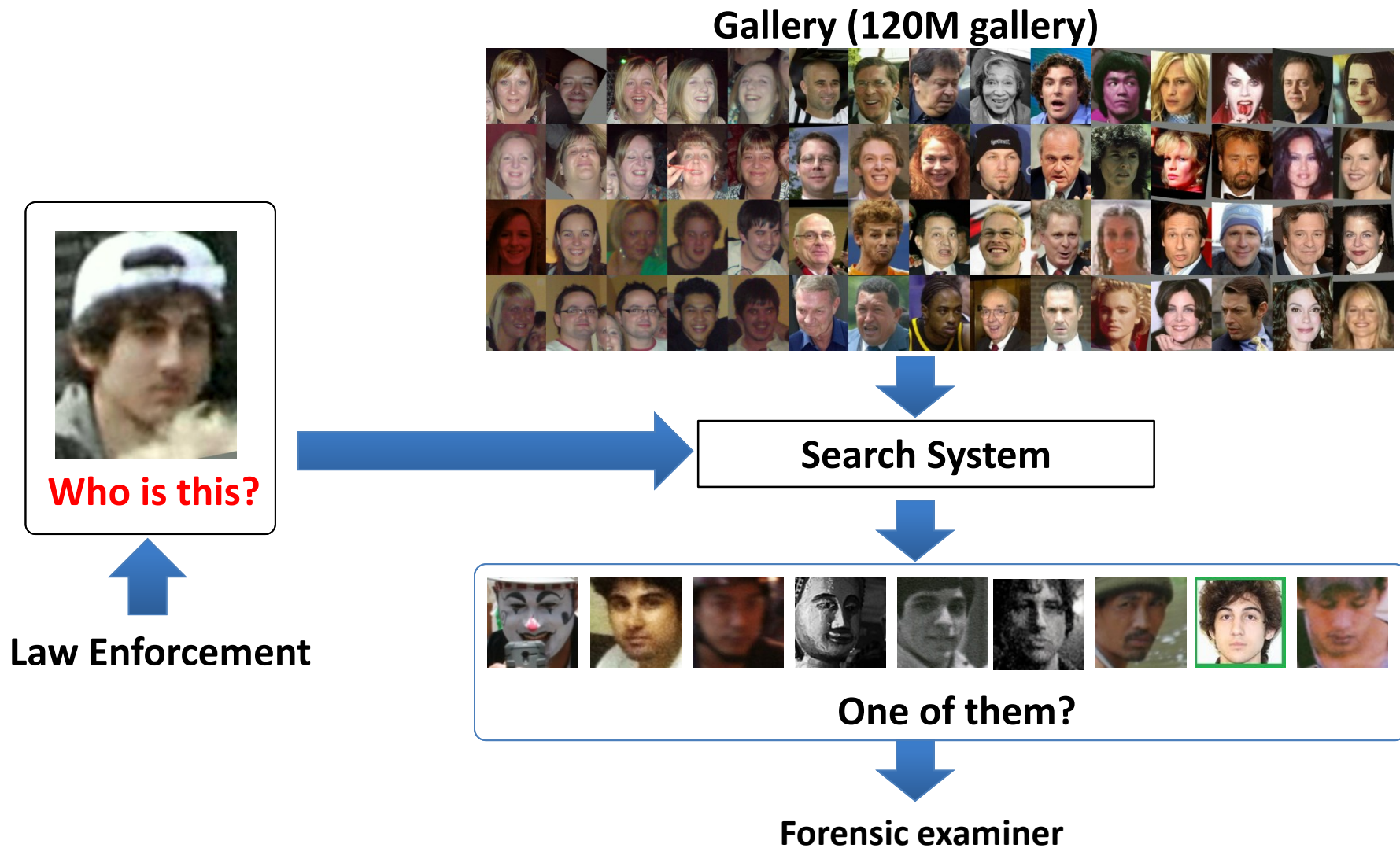
*Gallery*



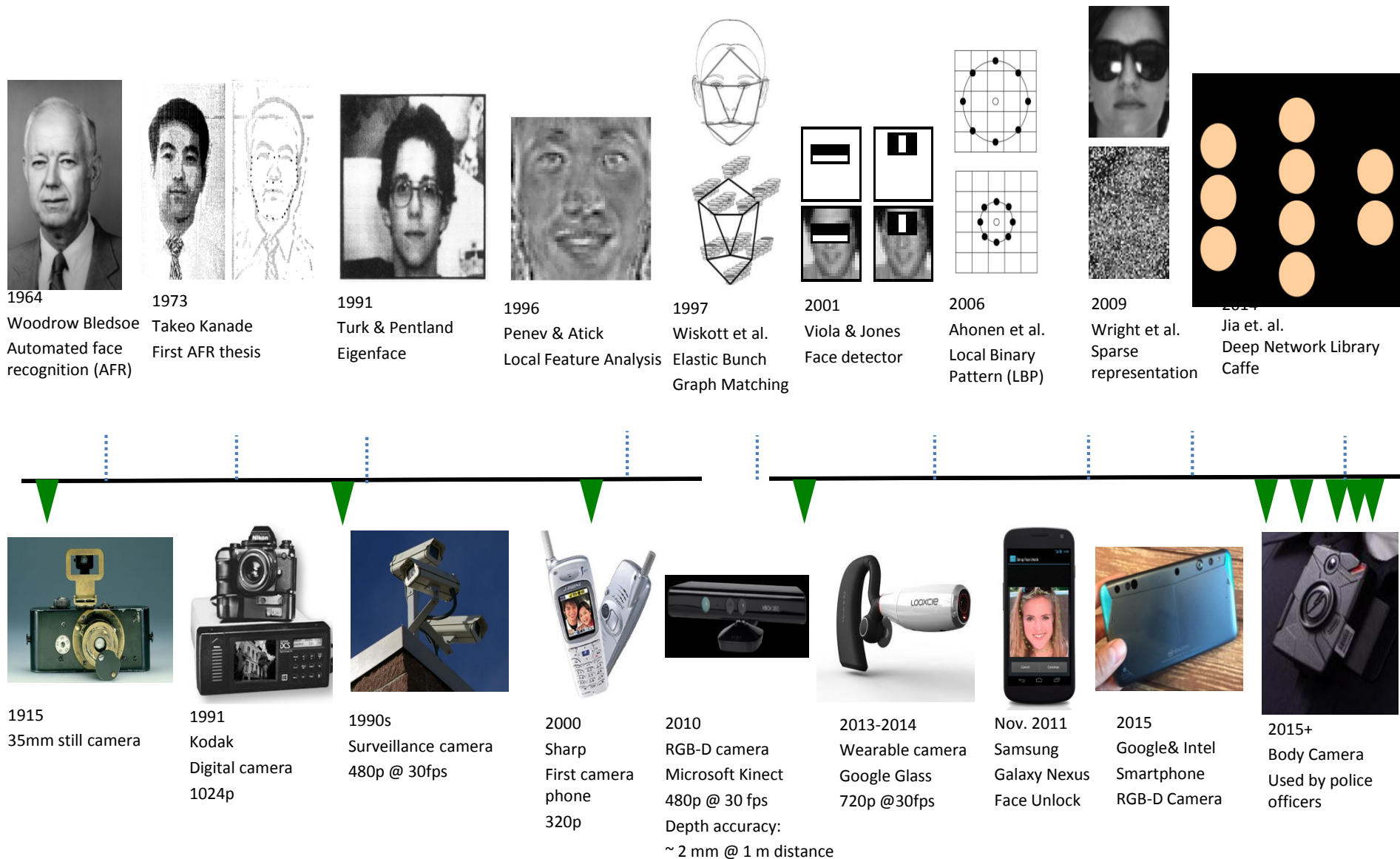
**Closed-set v. Open-set Search**



# Forensic Face Search



# Face Recognition Milestones



# Growing Interest in Face Recognition

- **Technology Drivers**

- Security (covert acquisition, IR, thermal,..)
- Prevalence of surveillance cameras
- Mobile phones
- Social media

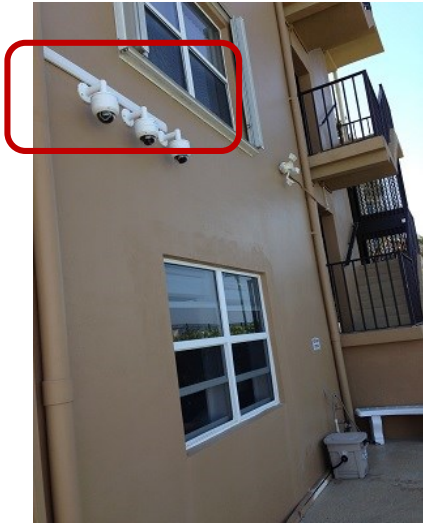
- **Technology Enablers**

- Processors (2M comparisons/sec/core)
- Deep networks
- Large training sets
- Benchmark datasets of increasing complexity
- Legacy database s

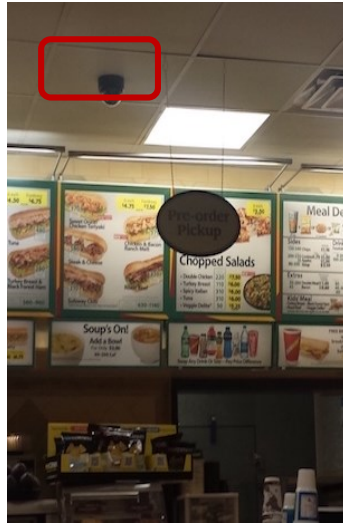


# Applications

# Surveillance Cameras Everywhere!



6:00 AM, Home



6:15 AM, Fast Food



6:35 AM, ATM



6:45 AM, Gas Station



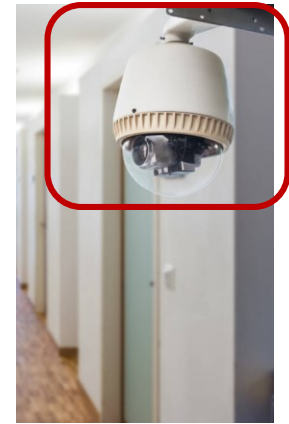
7:00 AM, Parking Lot



7:10 AM, Airport



7:30 AM, Security



3:00 PM, Hotel

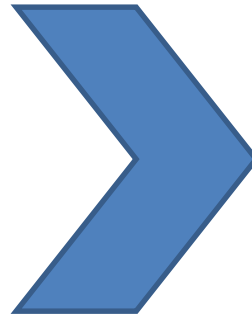
**~200 million surveillance cameras; billion of hours of videos/week!**

# Face Recognition in Video

Widespread looting and rioting



Face recognition lead to many arrests



Extensive CCTV Network



Many suspects could not be identified

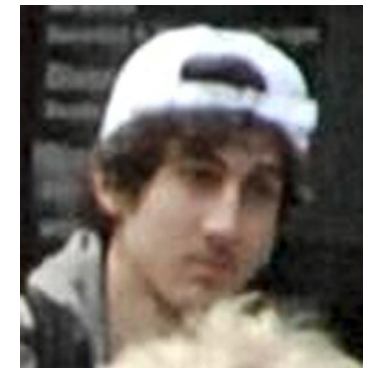


**2011 London riots**

# Boston Marathon Bombing

(April 2013)

Tamerlan Tsarnaev



Dzhokhar Tsarnaev





# International Border Crossing



SmartGate, Australia & NZ

**ePassports from eligible countries**



HK-Schenzen border

**Fusion of face & fingerprint**



# Passenger Verification

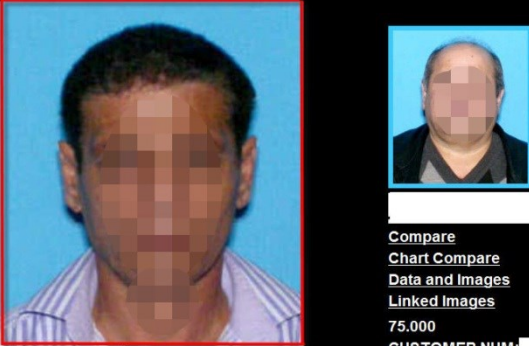


**Matching face image to photo on ID card**

# De-duplication: Driver License Database

Face-based scrubbing of  
**13.5M records (~30M photos)**  
in Michigan DMV database;  
photos of different subjects in  
the same record!

Check the highlighted Images.

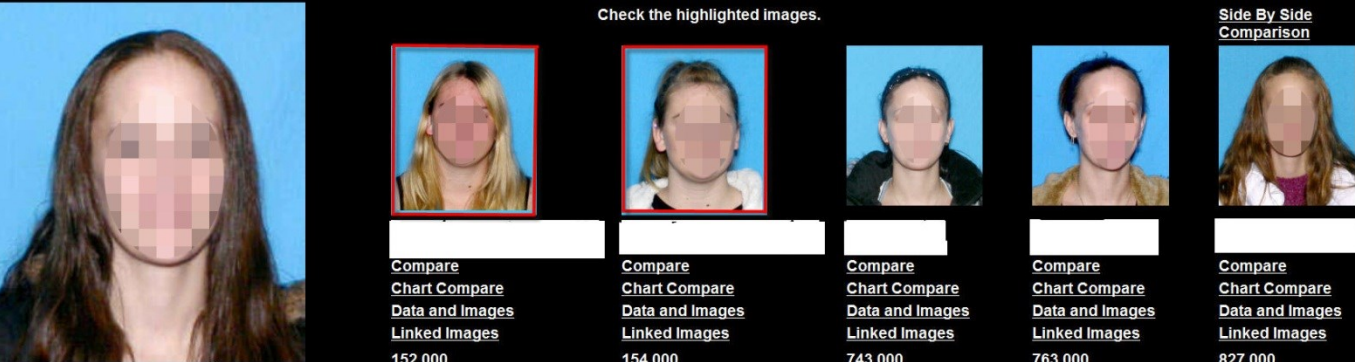


Compare  
[Chart Compare](#)  
[Data and Images](#)  
[Linked Images](#)  
75.000  
CUSTOMER NUM: [REDACTED]  
NAME: [REDACTED]  
DOB: [REDACTED]  
ISSUE DATE: 03/01/2013

Identifier: [REDACTED]  
Description: [REDACTED]  
W: [REDACTED] 64  
Date/Time: 9/28/2013 7:18:06 AM  
Status: [REDACTED]  
Comment: [REDACTED]  
CUSTOMER NUM: [REDACTED]  
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DOB: [REDACTED]  
ISSUE DATE: 08/30/2013

Check the highlighted Images.

Side By Side Comparison



Compare  
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[Data and Images](#)  
[Linked Images](#)  
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NAME: [REDACTED]  
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ISSUE DATE: 02/12/2013

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[Data and Images](#)  
[Linked Images](#)  
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Courtesy: Pete Langenfeld, Michigan State Police



# De-duplication

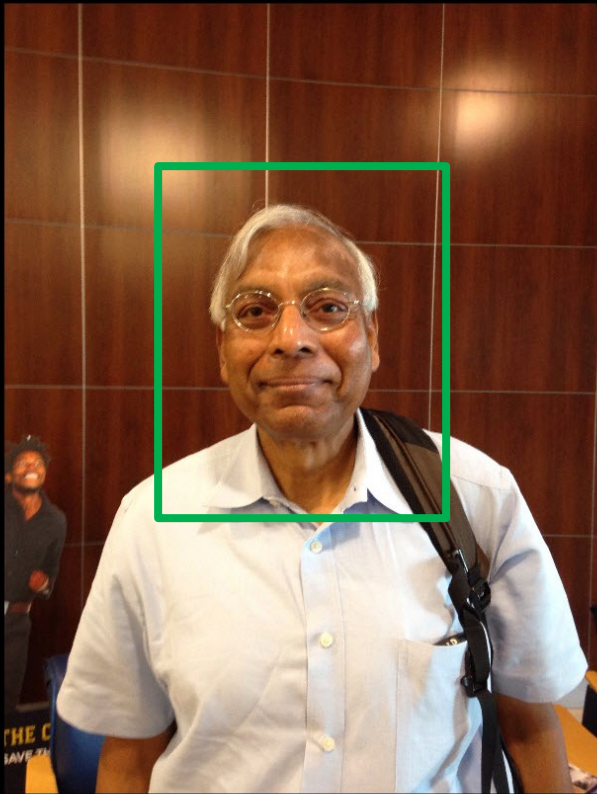
< Previous

2 of 10

Next >

Return

Close



SearchImage

Driver's License or  
Personal ID Number:  
[redacted] x

DOB:  
[redacted]

Last Name:  
[redacted]

First Name:  
[redacted]

Middle Name:  
[redacted]

Street Address:  
[redacted]

City:  
[redacted]

State:  
[redacted]

Zip:  
[redacted]

Sex:  
[redacted]

Height:  
[redacted]

Eye Color:  
[redacted]

Issue Date:  
[redacted]

Photo Date:  
[redacted]

Expiration Date:  
[redacted]

Printed Date:  
[redacted]


License Type:  
[redacted]

Country:  
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IDENTITY IN QUESTION:  
[redacted]

Comment:  
[redacted]

Driver License  
Information



<

Front View

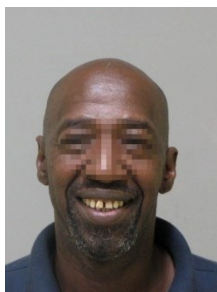
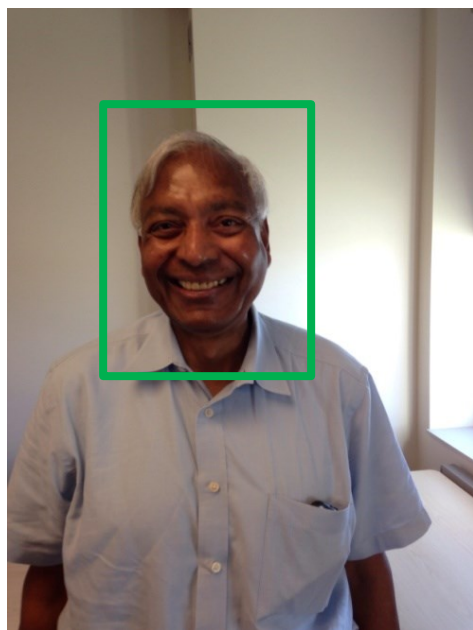
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1 of 2

**Gallery: 34 million (30M DMV photos, 4M mugshots) 2009 driver license**

# Smile Makes a difference!

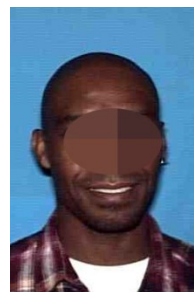
Top-10 retrievals



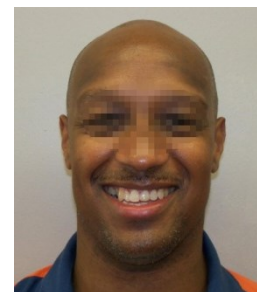
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2



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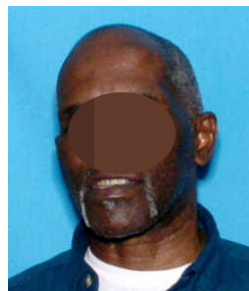
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5



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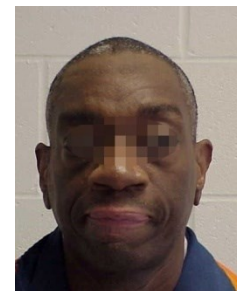
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8



9



10

**Gallery: 34 million (30M DMV photos, 4M mugshots)**

# Mobile Phones

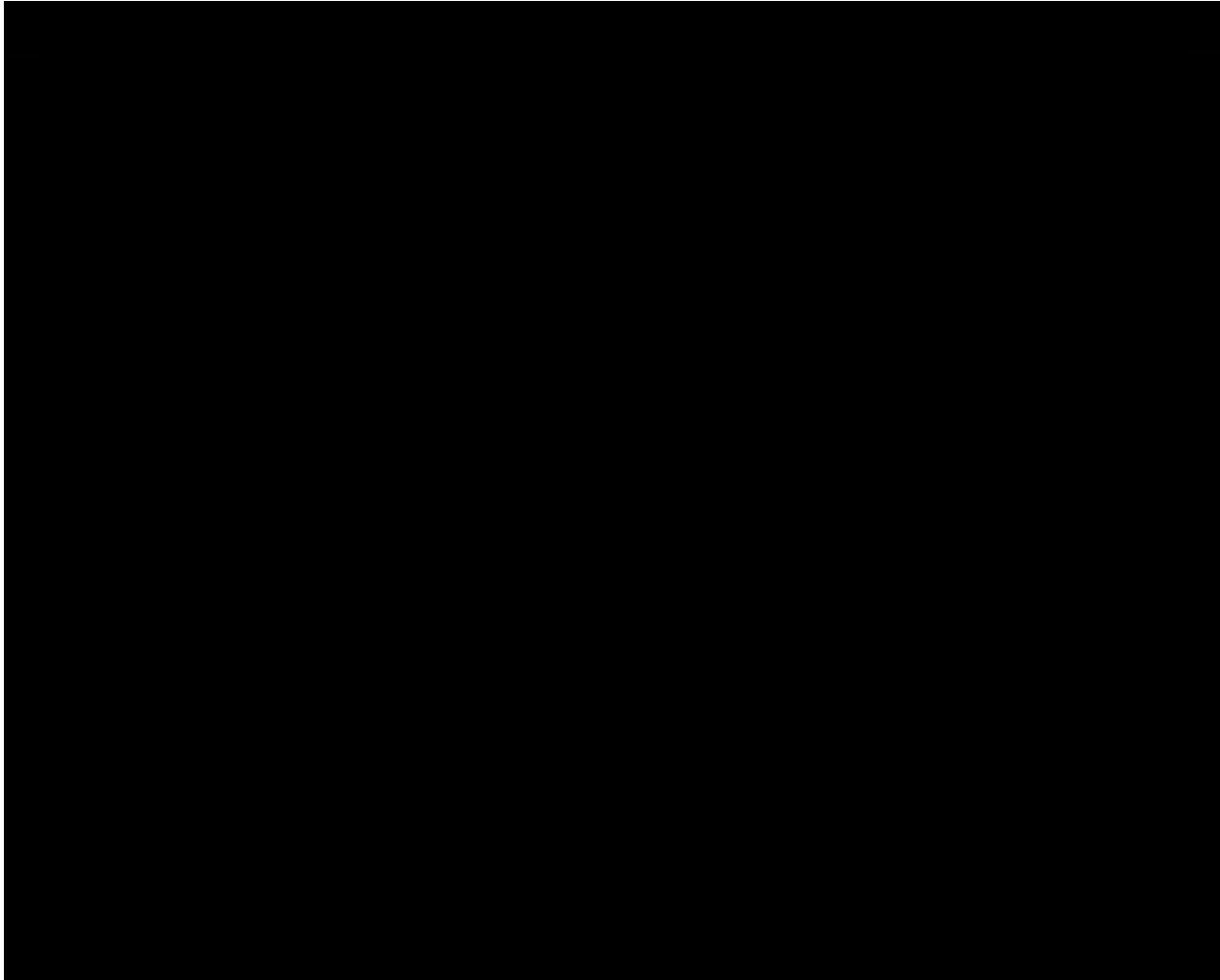


Joseph Van Os / Getty Images

**More cell phone accounts than world's population; \$1 Trillion in mobile payments**



# Mobile Face Unlock



Uploaded: Dec 6, 2011 YouTube

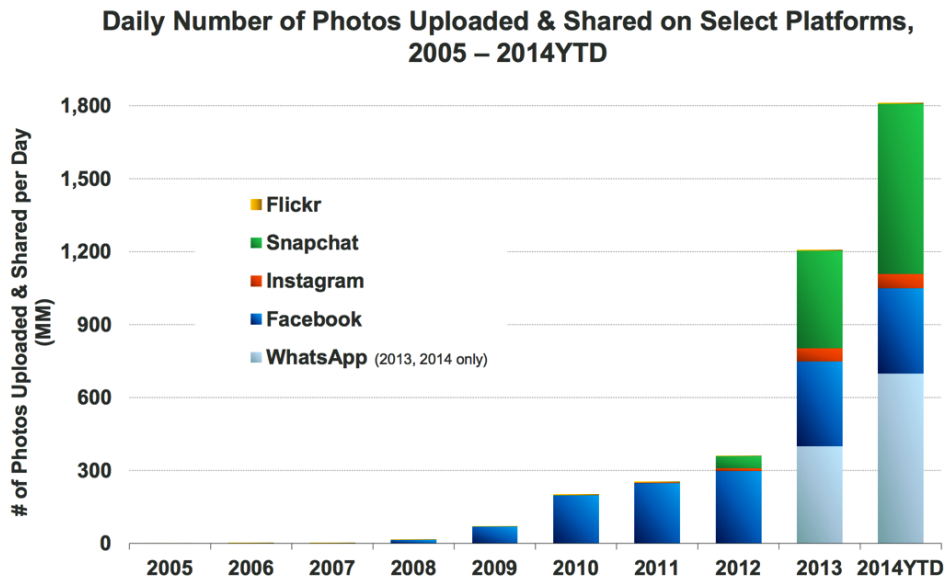
# Photo Tagging







# Social Media



@KPCB

Source: KPCB estimates based on publicly disclosed company data, 2014 YTD data per latest as of 5/14.

62

- *~trillion image* shares per year, and increasing
- Challenges: accuracy and efficiency



# Constrained Face Recognition

- Cooperative subjects:  
Small intra-subject  
variations (**FERET**)
- Operational face data  
(**mugshots, visa images**)
  - Limited user cooperation
  - Effect of aging



FERET Images



**PCSO** Mugshot Images



# Unconstrained Face Recognition

- **LFW**

- Images of celebrities and public figures
- Faces detect by Viola-Jones detector



LFW Images

- **IJB-A**

- Semi-automatic data collection
- Manually selected identities & annotation



IJB-A Images

# State of the Art: Verification

FRGC v2.0 (2006)



MBGC (2010)



LFW (2007)



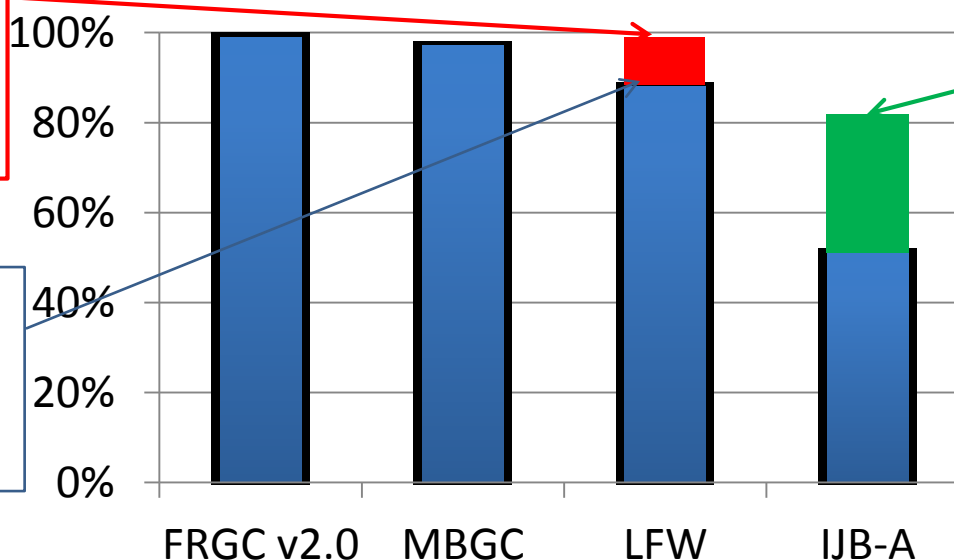
IJB-A (2015)



**LFW Standard Protocol**  
99.77% (Accuracy)  
3,000 genuine & 3,000  
imposter pairs;  
10-fold CV

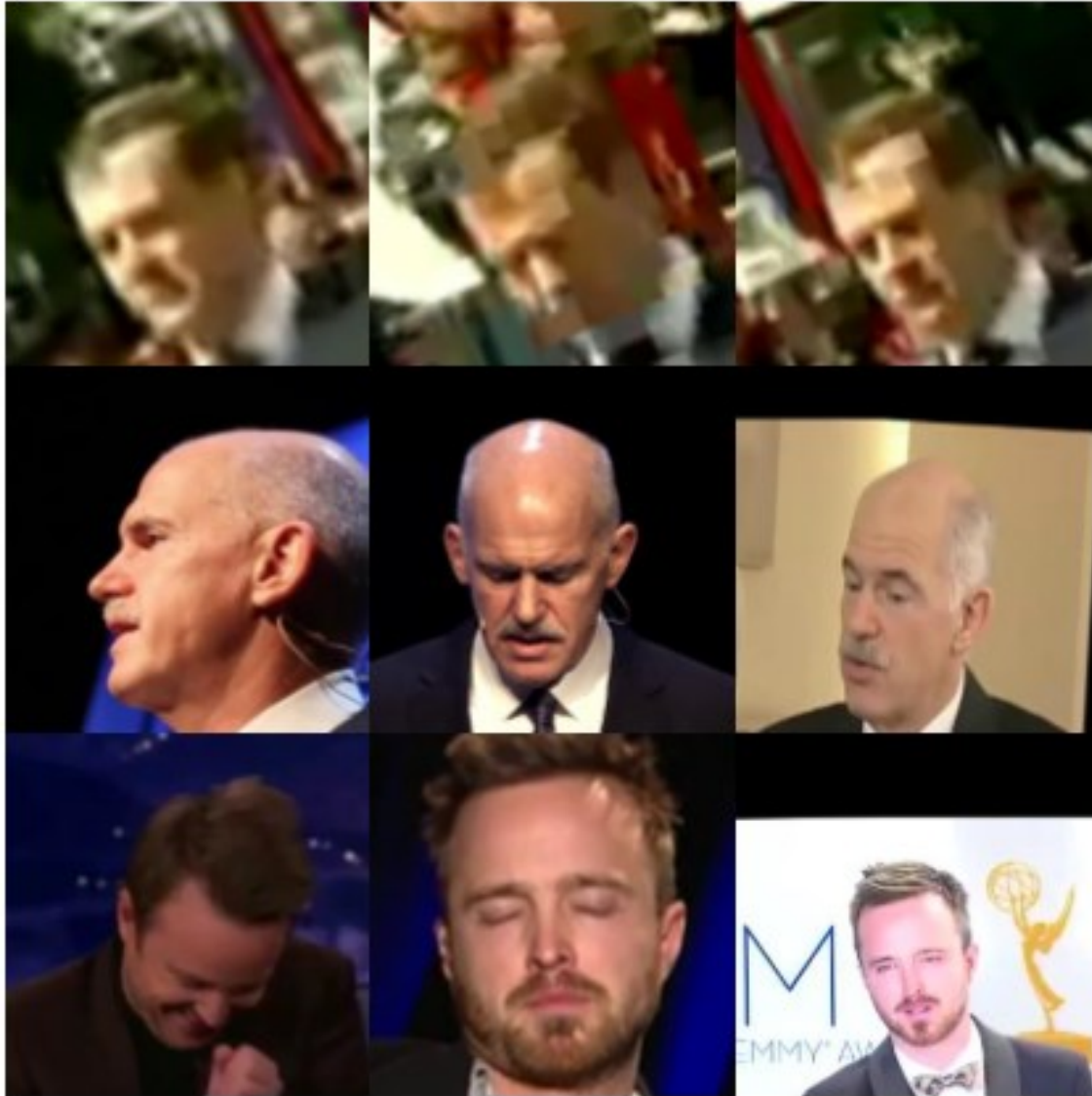
**LFW BLUFR Protocol**  
88% TAR @ 0.1% FAR  
156,915 genuine, ~46M  
imposter pairs;  
10-fold CV

TAR at 0.1% FAR



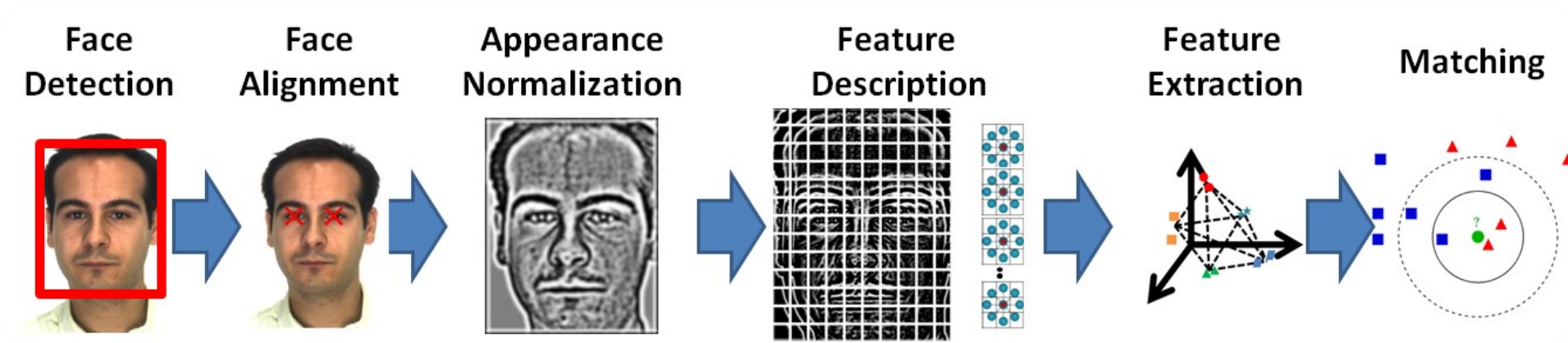
**IJB-A**  
80% TAR @ 1% FAR  
10-fold CV

# State of the Art: Verification



**NIST IJB-B database: TAR @0.01% FAR = 70%**

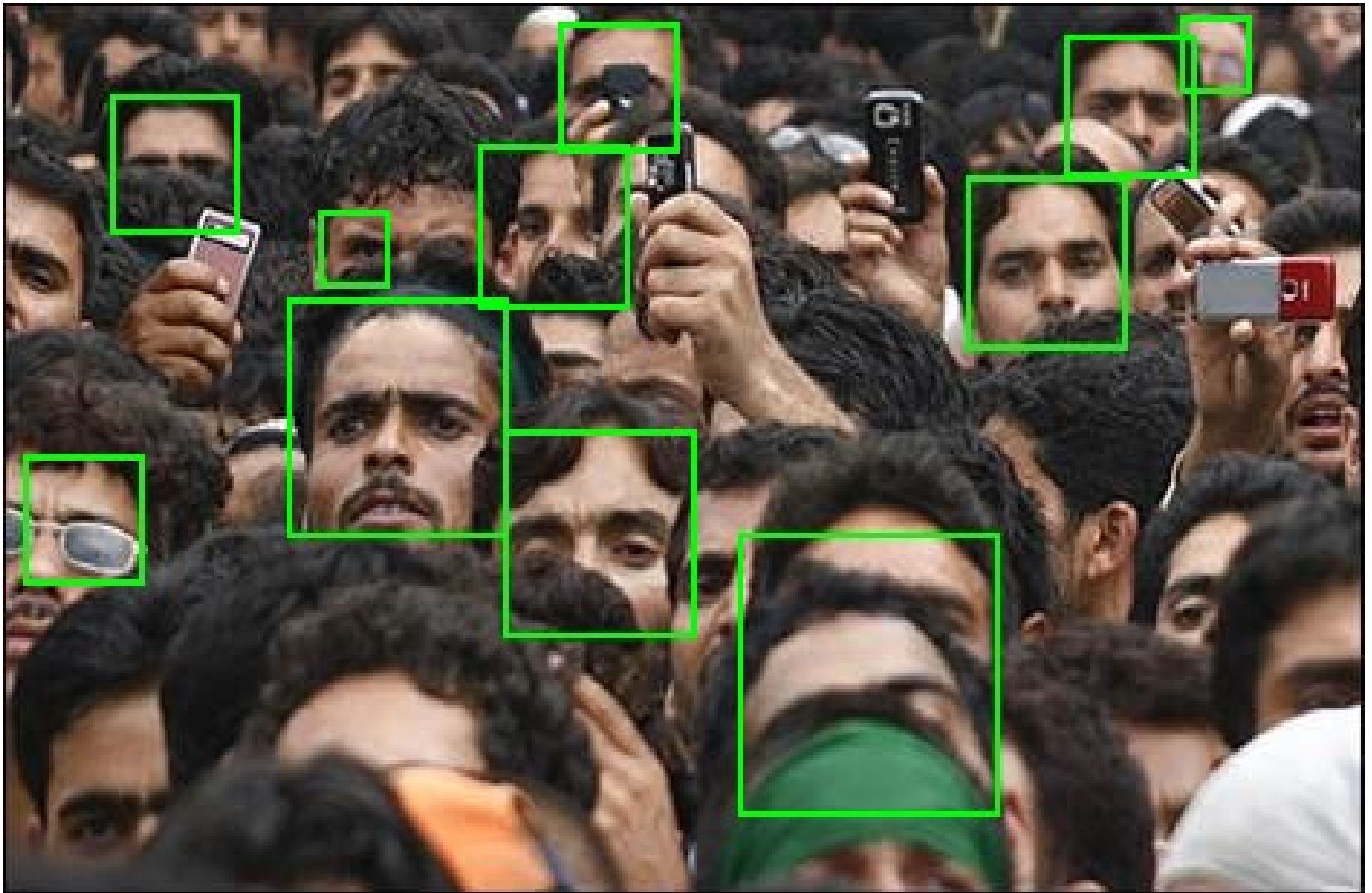
# Automated Face Recognition



- Most face recognition algorithms follow this pipeline



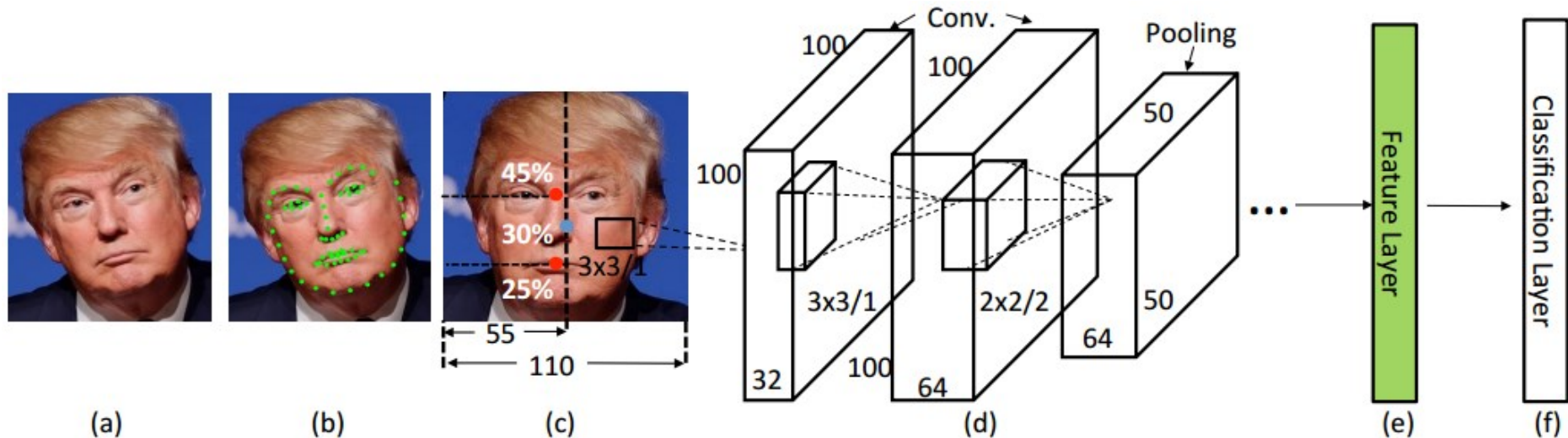
# Face Detection



Number of mobile phone users worldwide in 2016 is estimated to be about 4.8 billion



# Learning Face Representation



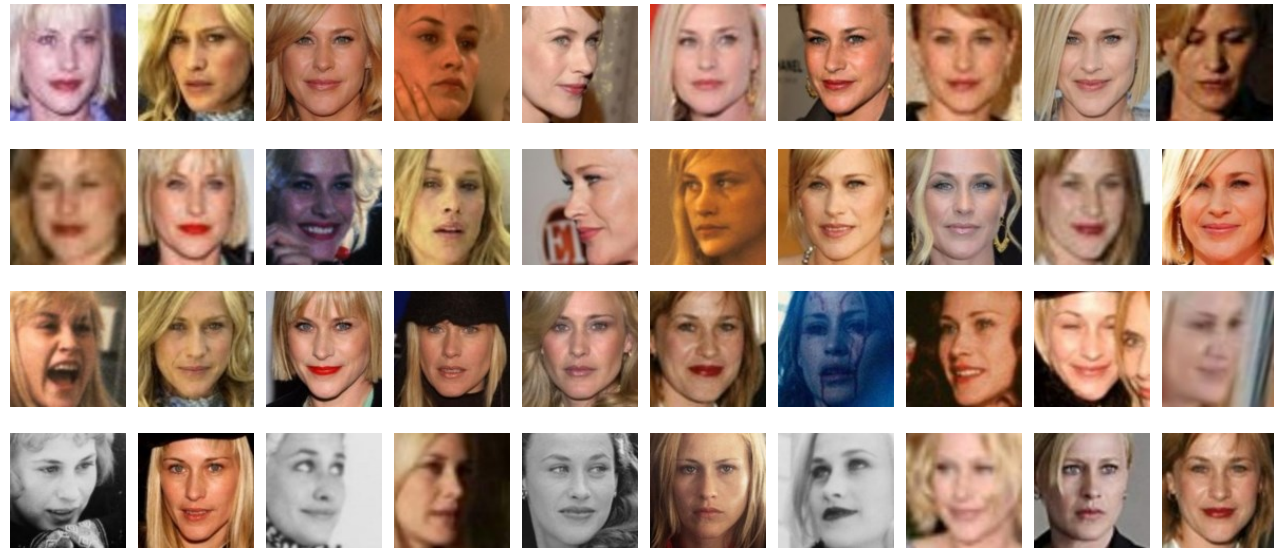
(a) Input RGB image, (b) detected keypoints, (c) normalized face image, (d) a convolutional neural network, (e) 320-dimensional feature vector and (f) softmax classification layer for training only

# Network Training

- ConvNet is trained with CASIA-Webface
  - 494,414 images of 10,575 subjects (**training bias?**)
- Preprocessing: face and landmark detection
  - Align face images using the centers of eyes & mouth



#subjects = 10,575



Total # images with landmarks = 435,689

# Face v. Non-Face

- In 120M faces, we estimate  $\sim 2.3\%$  non-faces
- Some “non-faces” are of statues, toys, etc., but some are completely wrong

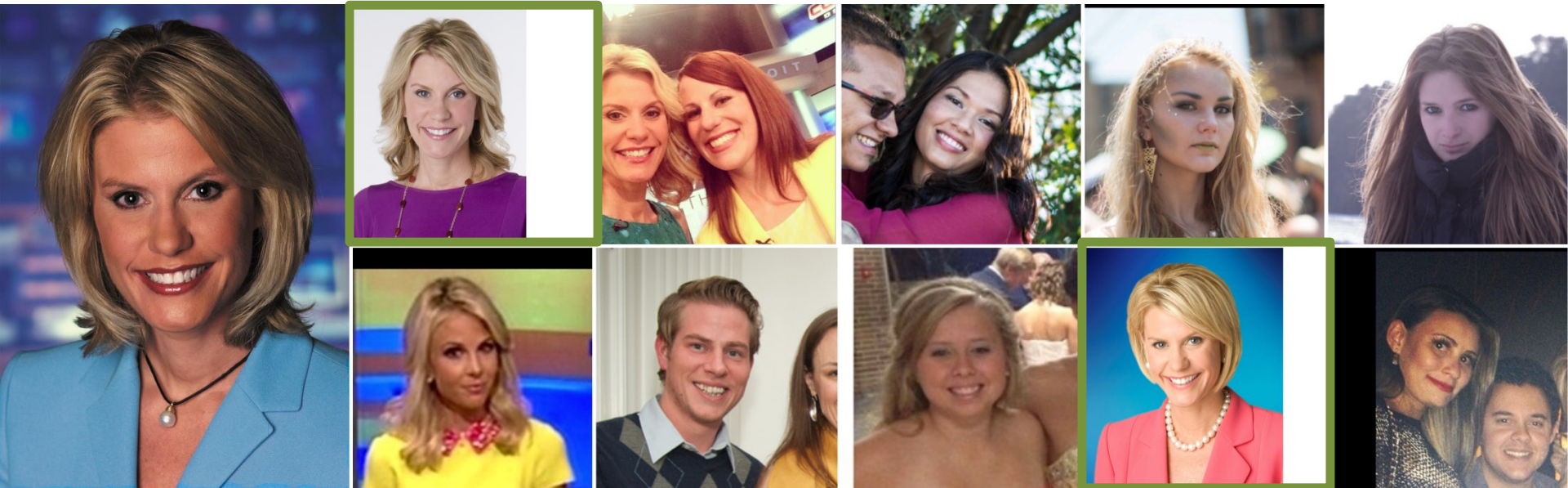


False-Positive Face Detections



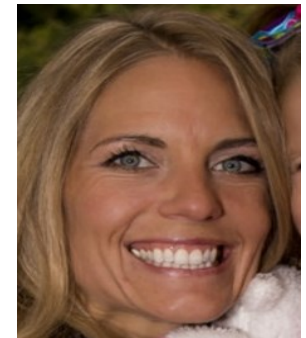
# Search Example

- 3 Images of TV Anchor Tammy Leitner, added to 120M background set
- Top-10 retrieval results for one query:

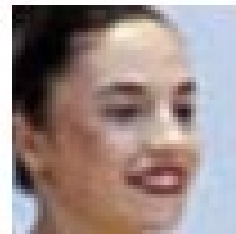
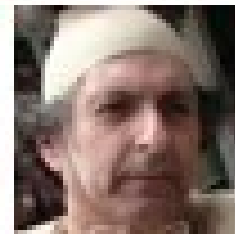
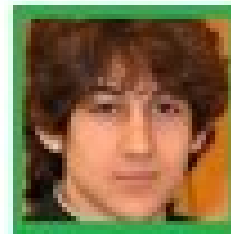
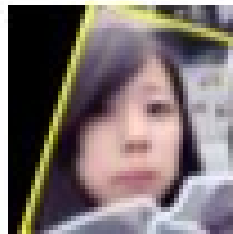
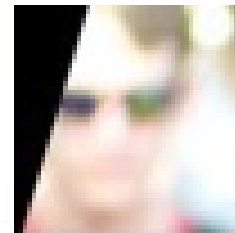
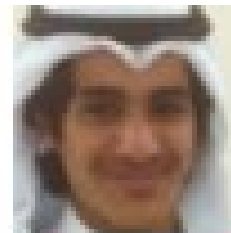
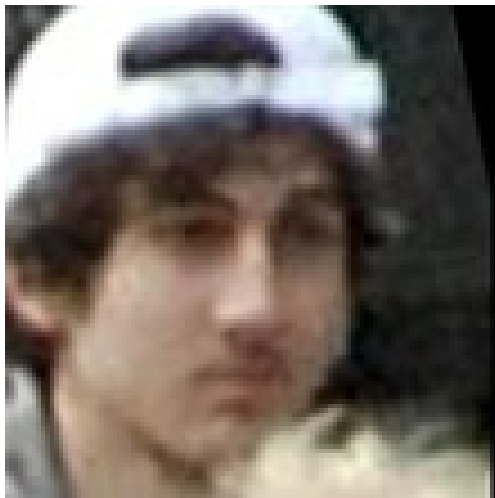




# Search for Sherry Jones (120M Gallery)



# Search for Dzhokhar Tsarnaev (120M Gallery)



# Network Architecture & Training

|      | Training Set                   | Network                               | VR@FAR=0.1% |
|------|--------------------------------|---------------------------------------|-------------|
| (1)  | CASIA-Webface                  | From Scratch [50]                     | 84.41%      |
| (2)  | CASIA-Webface                  | From Scratch [50], fusion of 9 models | 88.00%      |
| (3)  | CASIA-Webface                  | 18-layer ResNet                       | 82.06%      |
| (4)  | VGG-CASIA                      | 50-layer ResNet                       | 88.67%      |
| (5)  | VGG-CASIA                      | 50-layer ResNet, 10-crop              | 89.74%      |
| (6)  | CASIA-Webface                  | 50-Layer Pre-ResNet                   | 88.36%      |
| (7)  | CASIA-Webface                  | 50-Layer Pre-ResNet, 10-crop          | 89.64%      |
| (8)  | VGG-Face                       | 50-Layer Resnet                       | 81.40%      |
| (9)  | VGG-Deduplicated               | 50-Layer Pre-ResNet                   | 86.98%      |
| (10) | VGG-Deduplicated+CASIA-Webface | 50-Layer Pre-ResNet                   | 91.04%      |
| (11) | VGG-Deduplicated+CASIA-Webface | 50-Layer Pre-ResNet, 10-crop          | 92.22%      |
| (12) | VGG-Deduplicated+CASIA-Webface | 101-Layer Pre-ResNet                  | 91.18%      |
| (13) | VGG-Deduplicated+CASIA-Webface | 101-Layer Pre-ResNet, 10-crop         | 92.10%      |

LFW Under BLUFR Protocol

# IARPA Janus Program

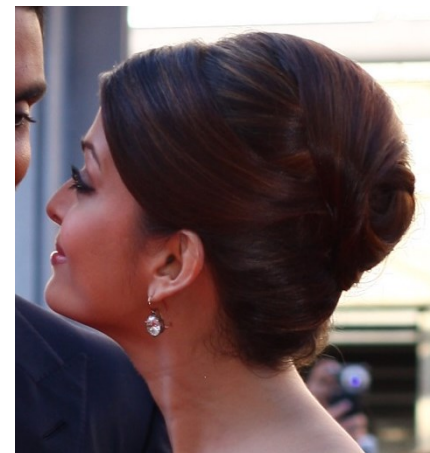
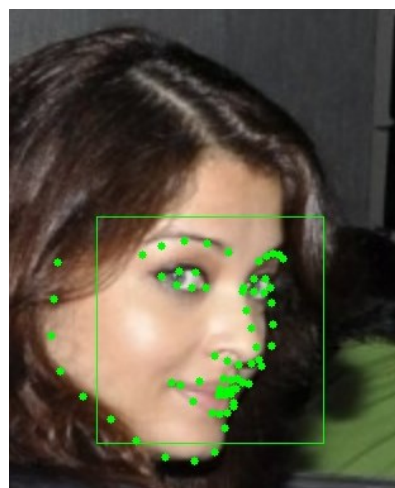
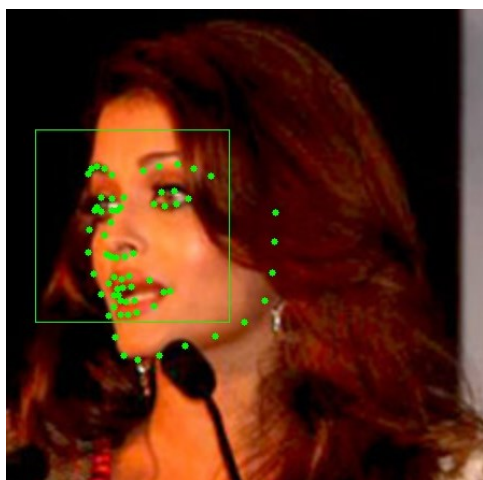
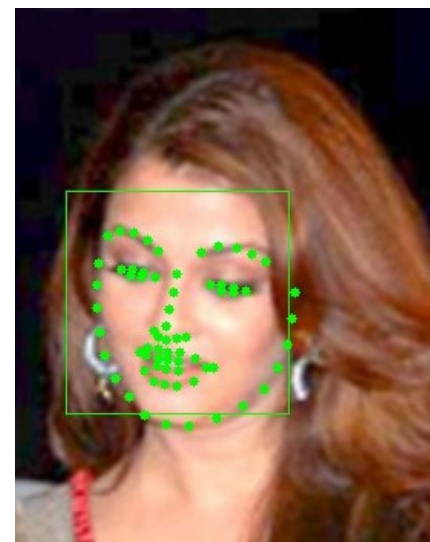
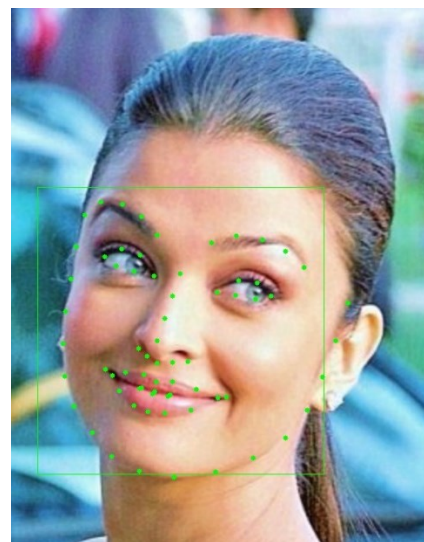
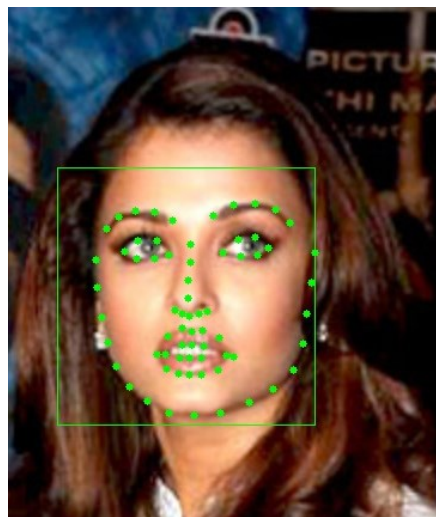
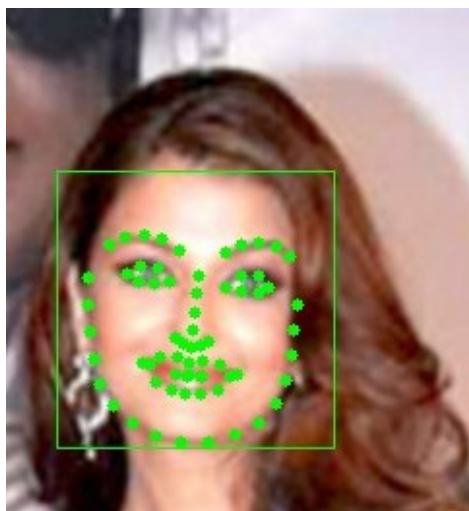
IARPA's Janus program aims to dramatically improve the current performance of face recognition tools by fusing the rich spatial, temporal, and contextual information available from the multiple views captured by today's "media in the wild".

<https://www.iarpa.gov/index.php/research-programs/janus>



# Some Challenges

# Pose, Illumination, Expression



Images of one subject in NIST IJB-A data, overlaid with V-J detector & dlib landmarks

# Facial Aging and Doppelgangers

Jan 1995



Gallery seed

Jul 1998



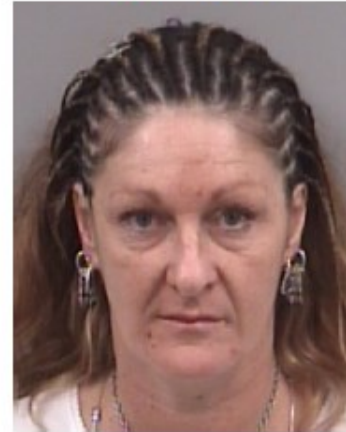
Score=0.99

Nov 1999



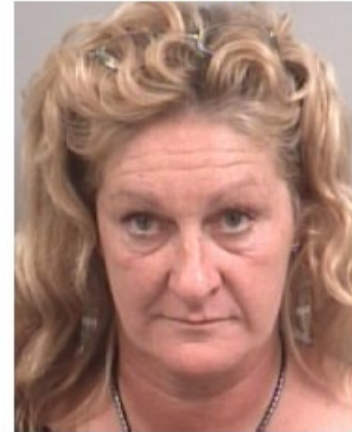
Score=0.62

Nov 2003



Score=0.41

Feb 2005



Score=0.26



<http://www.theguardian.com/theguardian/2010/dec/05/barack-obama-doppelganger-ilham-anas>

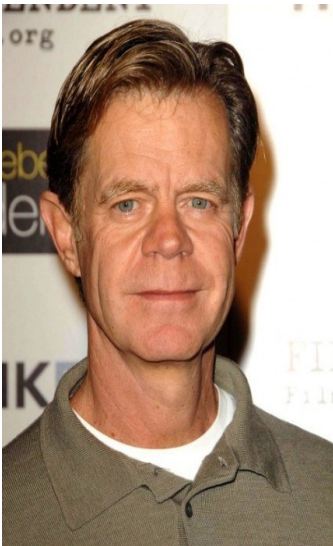
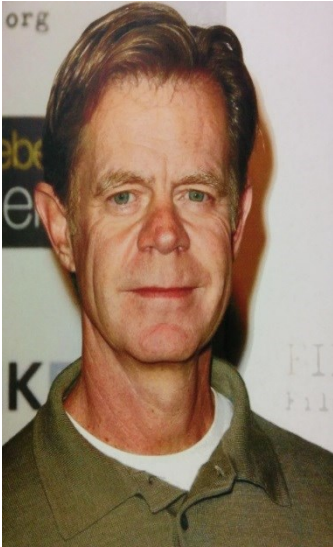
# Scars, Marks & Tattoos



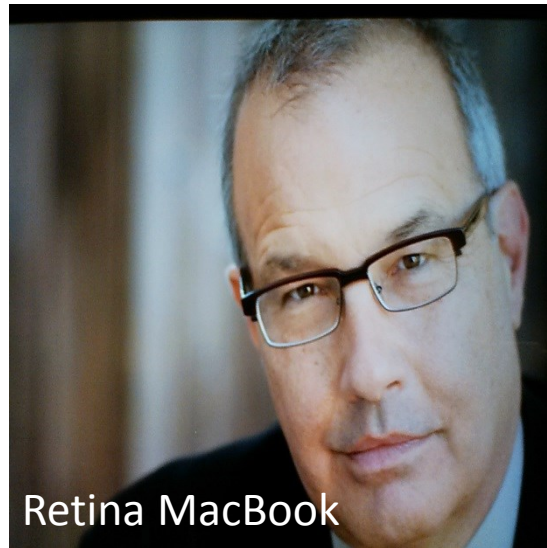
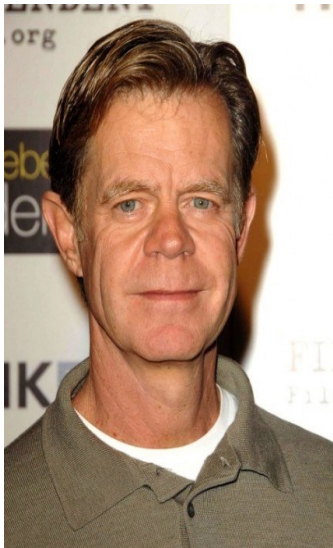
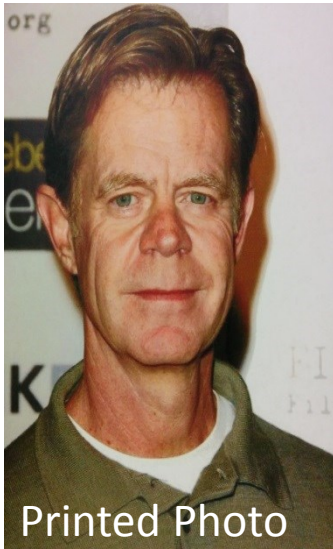
Detroit police linked at least six armed robberies at an ATM after matching a tipster's description of the suspect's distinctive tattoos



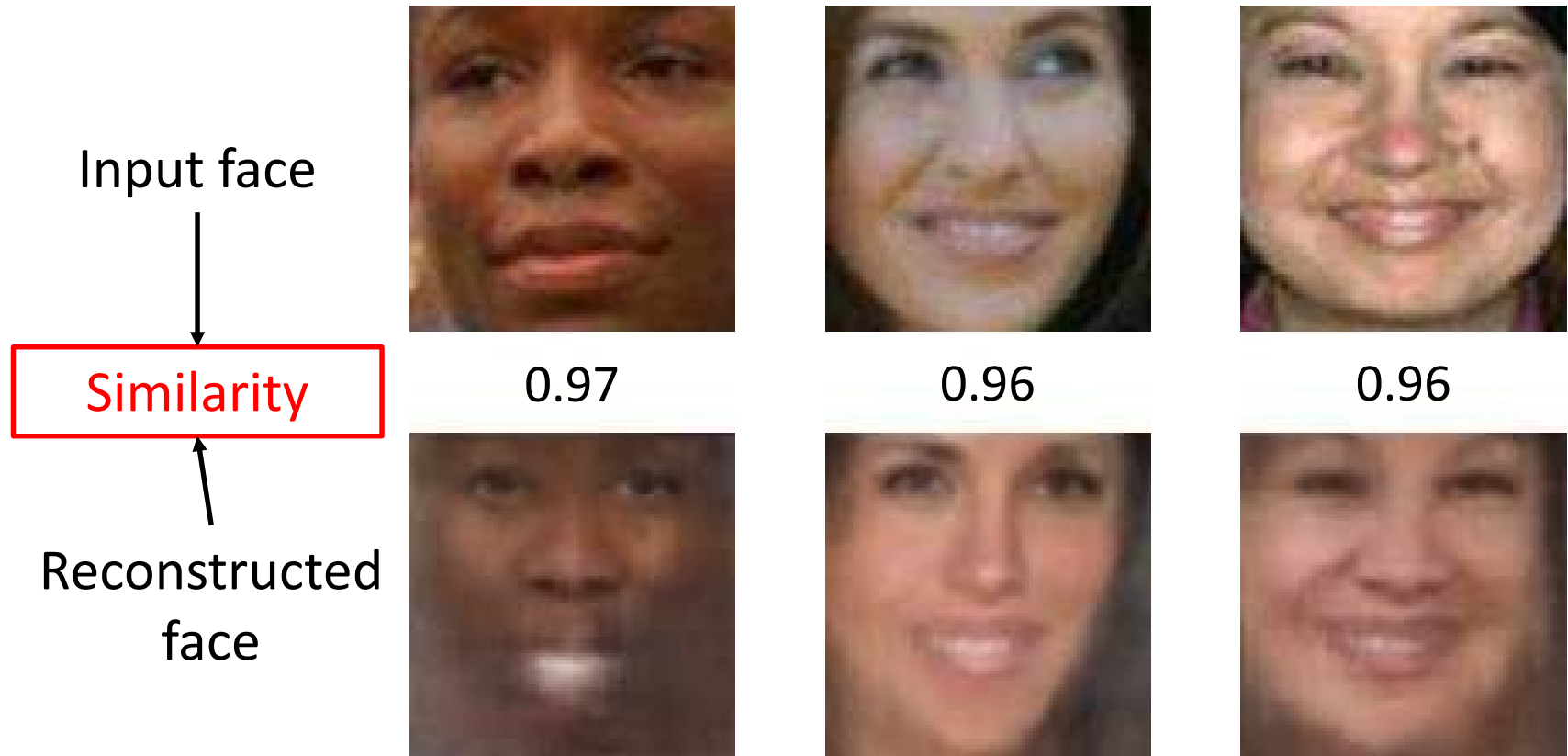
# Which Ones Are Real?



# Which Ones Are Real?



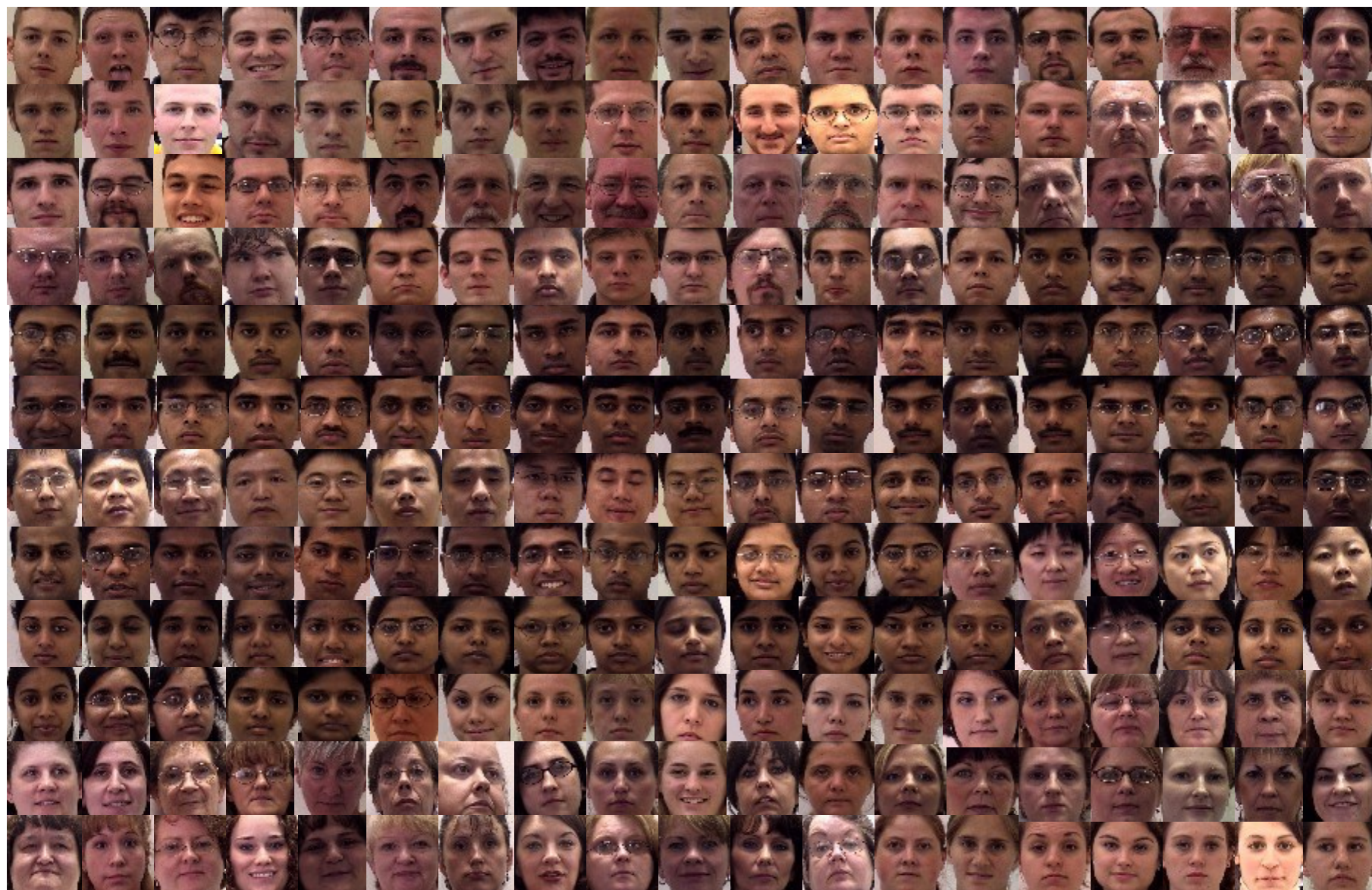
# Face Image Recovery from Templates



**Threshold@ FAR = 0.1% is 0.78**



# Capacity of Face Recognition?





# Summary

- Face recognition is now a major topic of research; growing no. of FR systems deployed
- State-of-the-art: High accuracy for constrained & cooperative subjects; low accuracy for unconstrained face recognition of non-cooperative subjects
- Need recognition systems robust to pose, illumination & expression, aging, and low resolution video
- User concerns: Data security & privacy

# References

- L. Best-Rowden and A. K. Jain, "Automatic Face Image Quality Prediction", arXiv preprint arXiv:1706.09887, 2017
- G. Mai, K. Cao, P.C. Yuen and A.K. Jain, "Face Image Reconstruction from Deep Templates", arXiv preprint arXiv:1703.00832, 2017
- L. Best-Rowden and A.K. Jain, "Longitudinal Study of Automatic Face Recognition", *IEEE Trans. Pattern Analysis & Machine Intelligence*, 2017 DOI:10.1109/TPAMI.2017.2652466
- C. Otto, D. Wang and A. K. Jain, "Clustering Millions of Faces by Identity", *IEEE Trans. Pattern Analysis & Machine Intelligence*, 2017 <https://arxiv.org/abs/1604.00989>
- D. Wang, C. Otto and A. K. Jain, "Face Search at Scale", *IEEE Transactions on Pattern Analysis and Machine Intelligence*, DOI 10.1109/TPAMI.2016.2582166, June 2016