

# Biometrics 2

## Face recognition



PV181 Laboratory of security and applied cryptography  
Seminar 27. 11. 2019

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# Lecture structure

## Seminar 1

1. Introduction
2. Fingerprints
3. Seminar activity
  - Fake fingerprints
4. Homework
  - Report on selected biometric system

## Seminar 2

1. Face recognition
2. Seminar activity
  - Face biometric SWOT analysis
3. Homework
  - Age estimation

# Real-life example

The Joy of Tech™



by Nitrozac & Snaggy

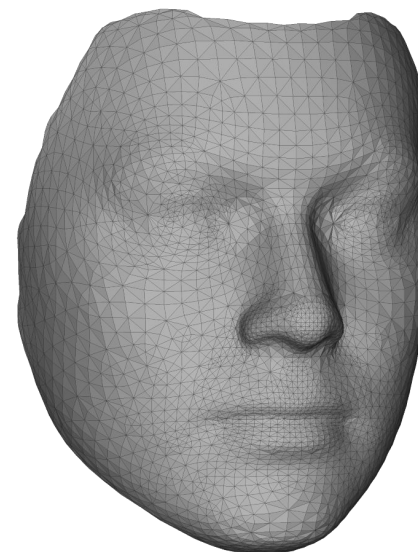
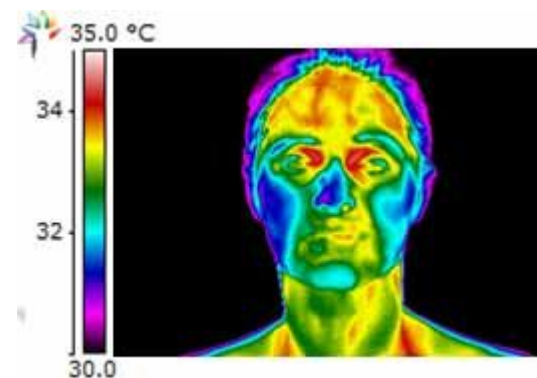


© 2009 Geek Culture

joyoftech.com

# Face recognition – Input

- Single picture
- Video sequence
- 3D image
- Facial thermograms



# Face recognition: The manual way

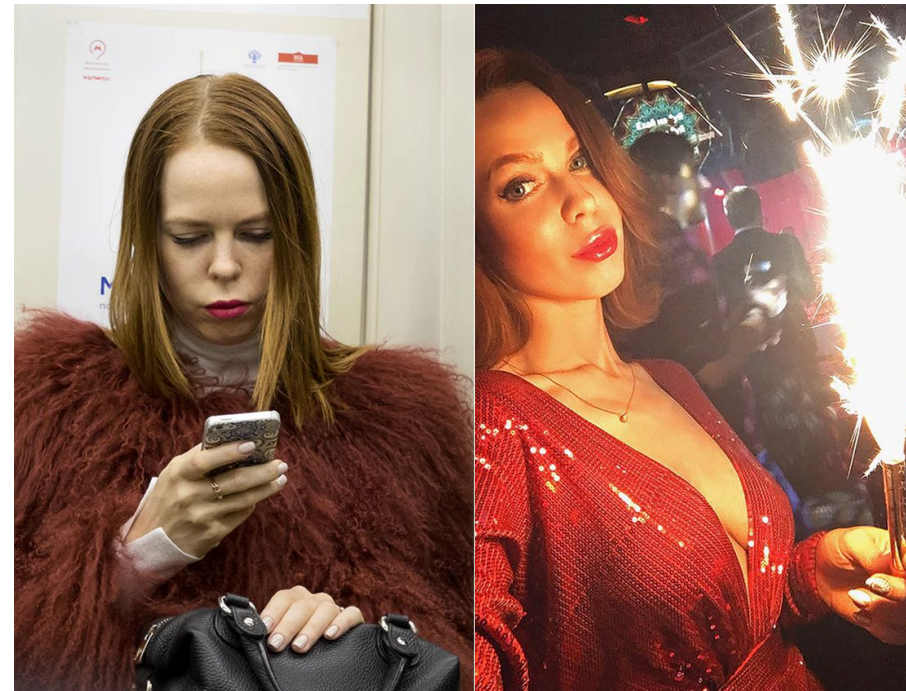
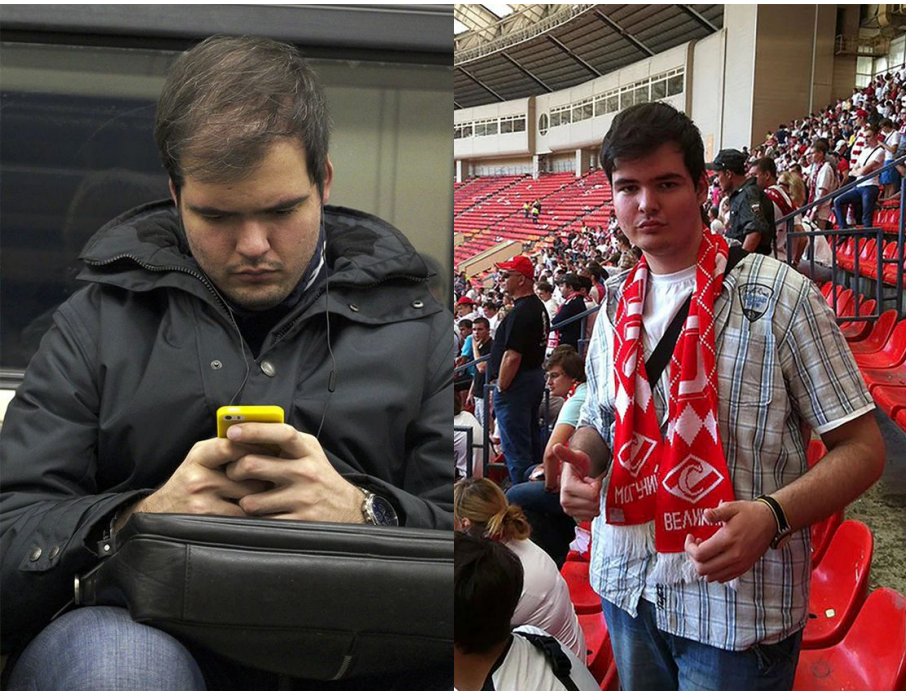


# Face recognition: The automatic way

- Statistical
  - Eigenface, PCA, LDA, ...
- Neural networks
  - Microsoft: Face API
  - Facebook: DeepFace
  - VK: FindFace (*“best results” in MegaFace comp.*)
  - Google: FaceNet

# FindFace – example

Subway photo (left), social network photo (right)



# Challenges in face recognition

- Illumination
- Pose
- Environment
  - Noisy background
- Aging
- Feature occlusion
  - Hats, glasses, hair, ...
- Image quality
  - colour, resolution, ...





## Testing sets (databases)

- Many databases:

<http://www.face-rec.org/databases/>

- Covering:

- Aging
- Illumination
- Pose
- Expression

# OpenBR: Face recognition overview

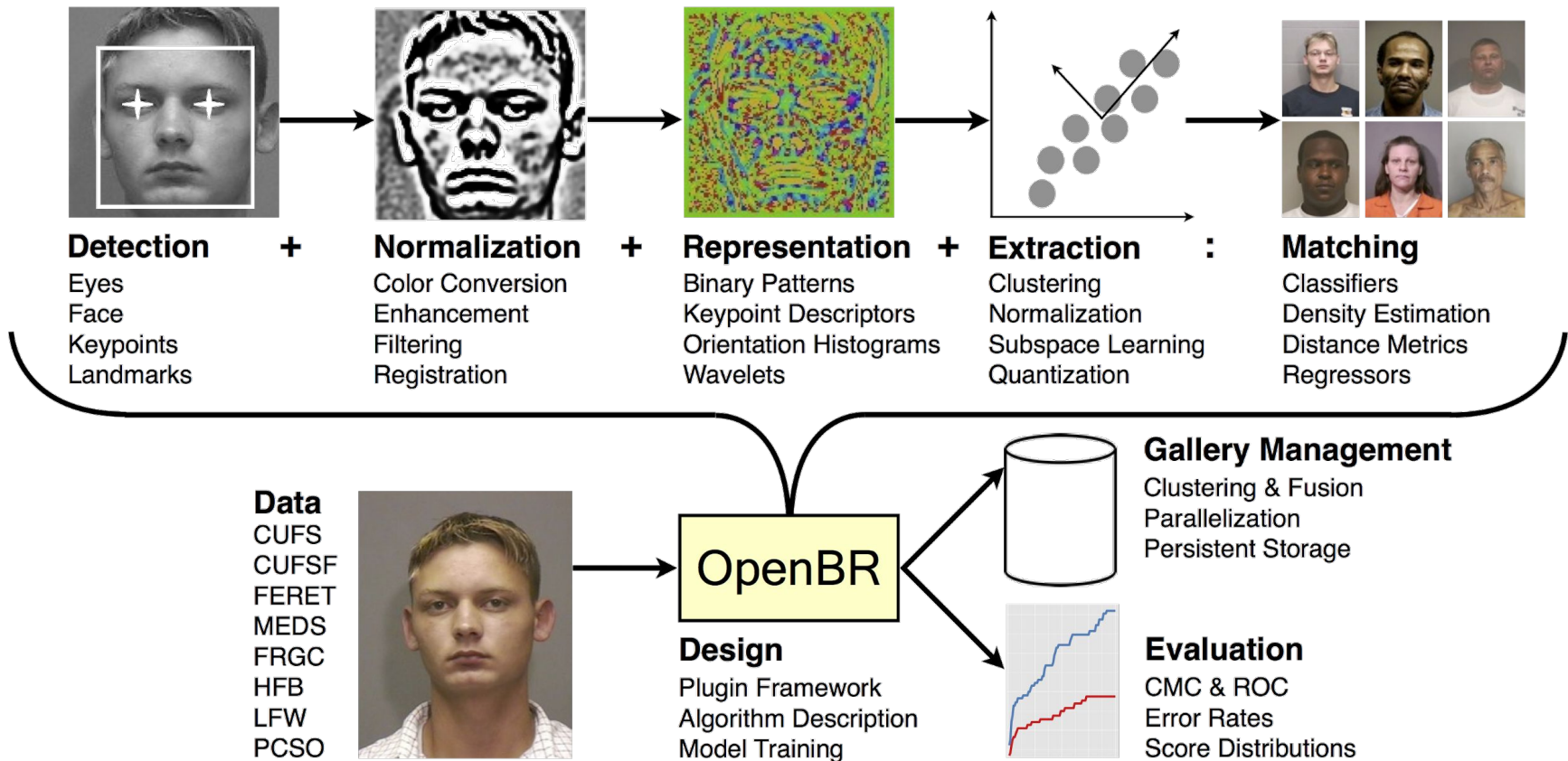
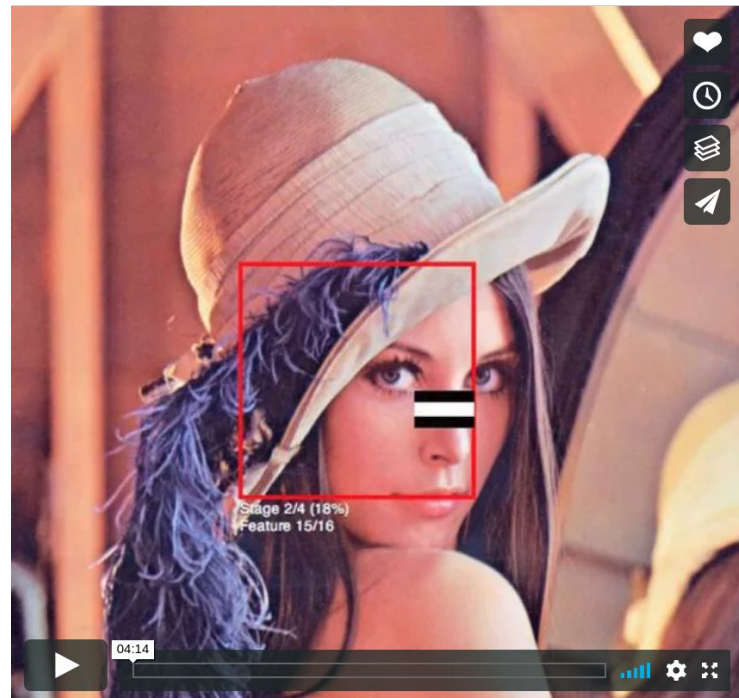


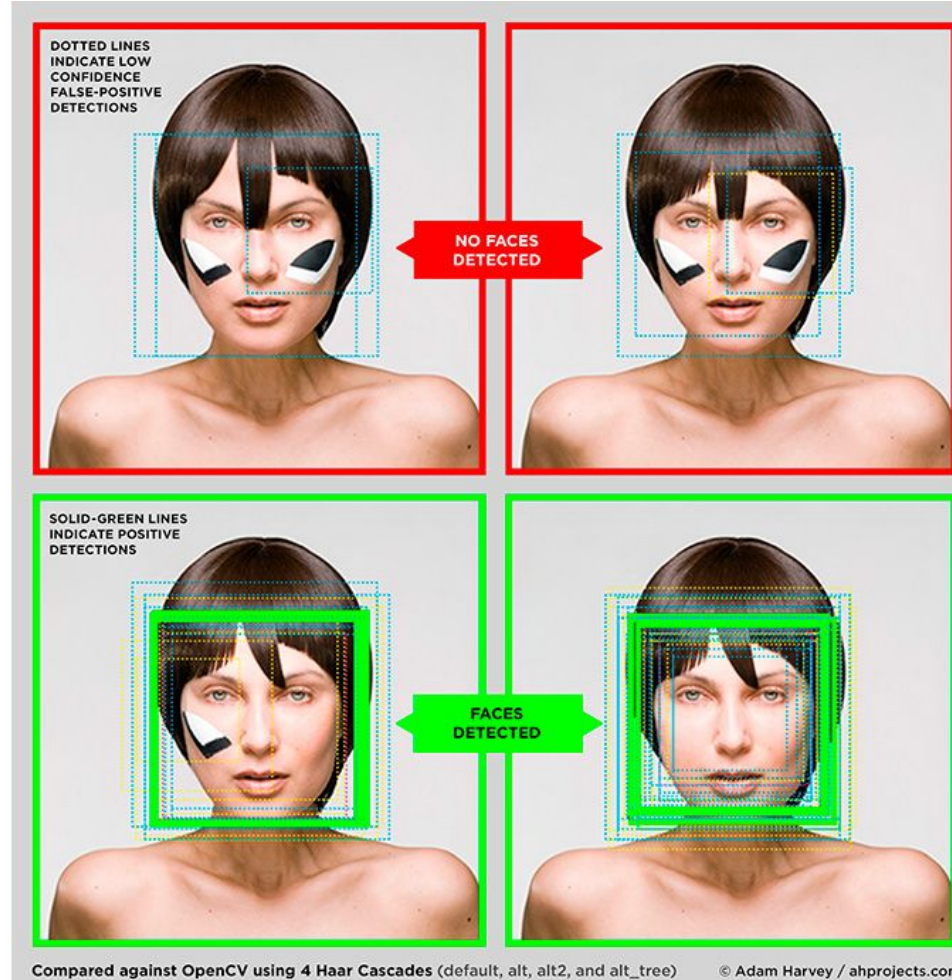
Photo © 2016 openbiometrics.org

# OpenBR face recognition – visualization

- Haar-cascade Detection
- Machine learning based approach where a cascade function is trained from a lot of positive and negative images.
- See video:  
*OpenCV Face Detection: Visualized*  
<https://vimeo.com/12774628>



# CV Dazzle: Anti face-detection



# CV Dazzle: Anti face-detection

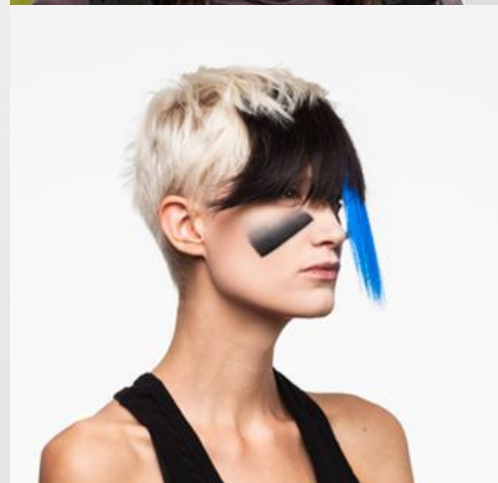
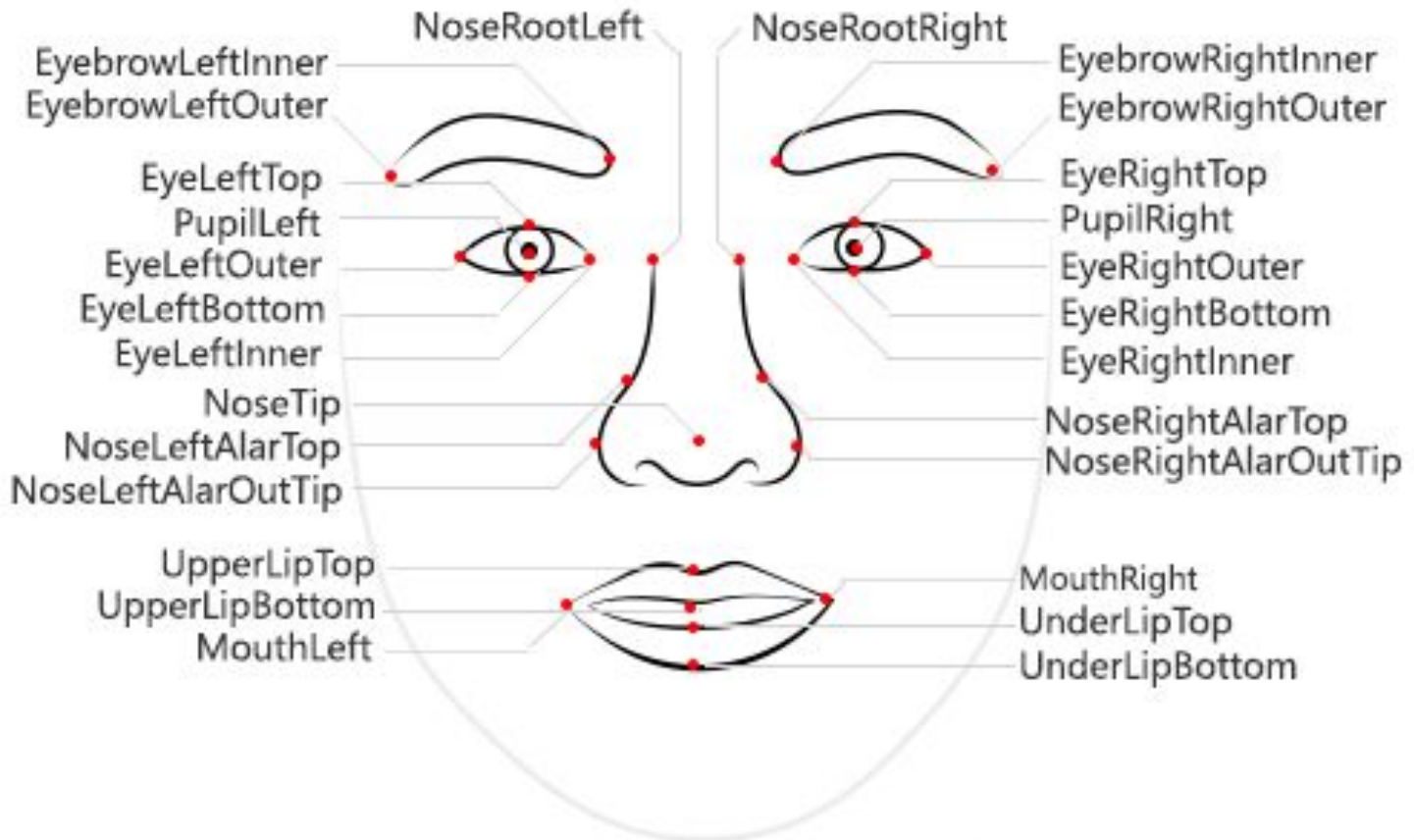


Photo © 2010-2016 Adam Harvey, CV Dazzle

# Microsoft: Face API



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# Automatic passport control



## Biometric passports

- “Smart card”, contain NFC chip
- Two security levels:
  - BAC: Reading your photo+personal information  
(Try Android app Passport reader)
  - EAC: Reading your biometrics
    - Fingerprint, Face and Iris support



# Face impersonation

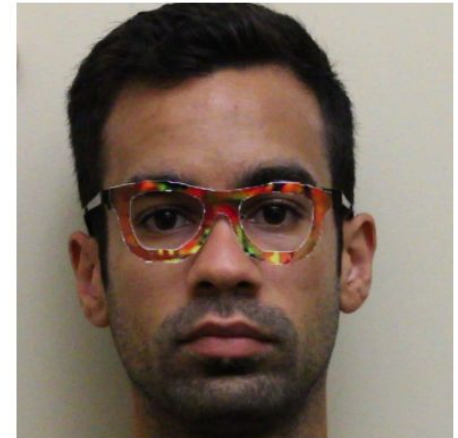


Photo © 2016 Carnegie Mellon University, *Accessorize to a Crime: Real and Stealthy Attacks on State-of-the-Art Face Recognition*

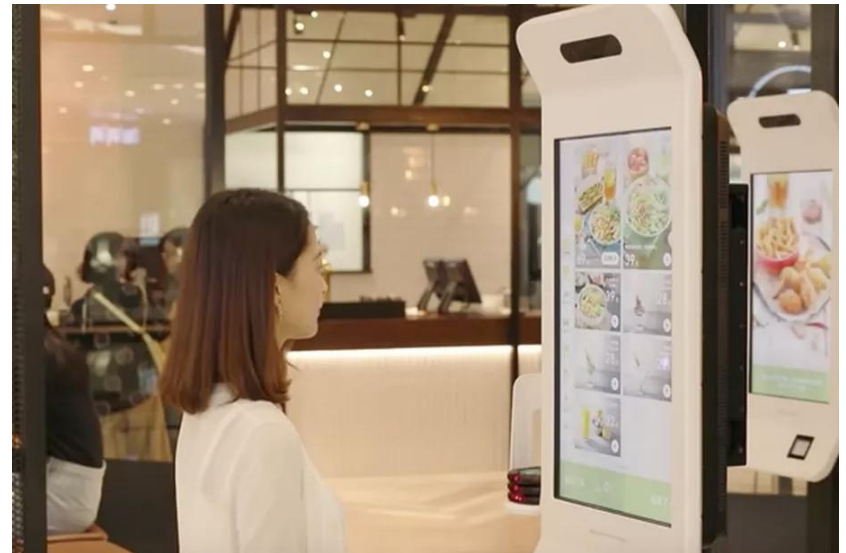
## Face impersonation

- Fooling deep-neural-networks-based face recognition systems (e.g. Face++)
  - Over 90% success rate
  - The principle is more general
- *"physically realizable and inconspicuous"*

*Sharif, Mahmood, et al. "Accessorize to a crime: Real and stealthy attacks on state-of-the-art face recognition." Proceedings of the 2016 ACM SIGSAC Conference on Computer and Communications Security. ACM, 2016.*

## KFC AliPay

- Introduced 2015
- Only one KFC in China



- See AliPay promo video at <https://www.theverge.com/2017/9/4/16251304/kfc-china-alipay-ant-financial-smile-to-pay>

# Apple FaceID hacked

- Liveness detection feature
- In 2019 by researchers
- Hacked by usage of pair of modified glasses
- A victim has to sleep :-)

Source: <https://threatpost.com/researchers-bypass-apple-faceid-using-biometrics-achilles-heel/147109/>



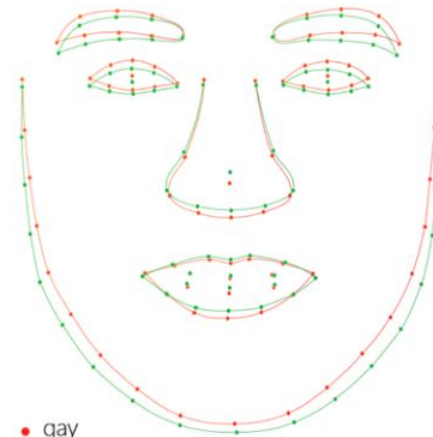
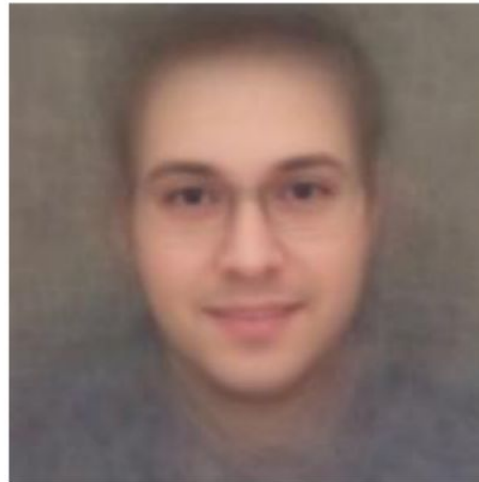
# Detecting sexual orientation from faces

Composite heterosexual faces

Composite gay faces

Average facial landmarks

Male



Female

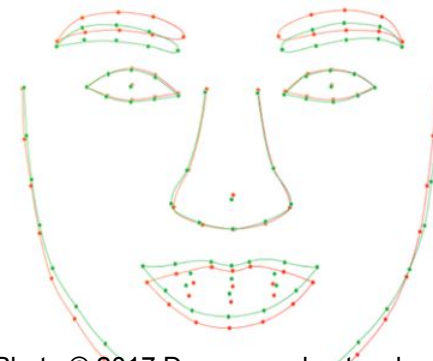
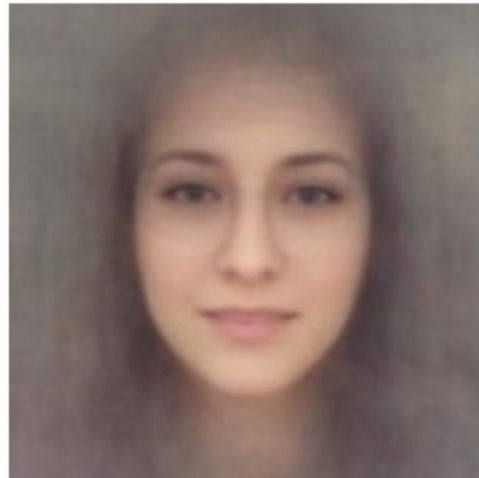


Photo © 2017 Deep neural networks are more accurate than humans at detecting sexual orientation from facial images. Journal of Personality and Social Psychology

## Detecting sexual orientation from faces

- Classifying sexual orientation (straight vs. gay) on men/women photos
  - Human success: 61% / 54%
  - Neural networks: 81% / 71%
  - Neural networks (5 images): 91% / 83%
- May be a privacy issue!

*Wang, Y., & Kosinski, M. (in press). Deep neural networks are more accurate than humans at detecting sexual orientation from facial images. Journal of Personality and Social Psychology, 2017.*

# Mugshots



BUDDSDJ\_10



CAUGHMANMD\_3



CLYMANN\_1



DELAROSAJ\_2



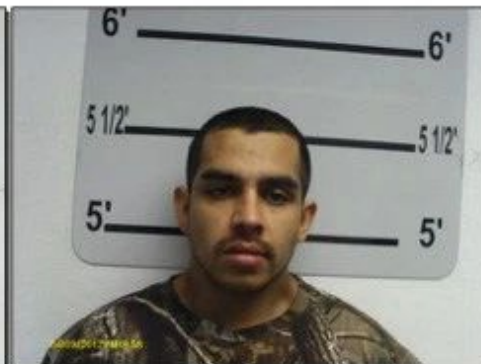
CHEWEYSR\_22



CLARKJ\_6



DELOACHAM\_1



GILLEYNK\_1

## Face recognition ban

- San Francisco
- “Threat to civil liberties”
- Ban for government agencies (city police and sheriff)
- Federal agencies not affected
- Reason: privacy issue
  - Less accurate at people of colour
- For the supplier: step back
- [www.banfacialrecognition.com](http://www.banfacialrecognition.com)

Gregory Barber, *San Francisco Bans Agency Use of Facial-Recognition Tech*. 2019, Wired.



## Code of Ethics (ACM)

1. Society and human well-being
2. No harm for participants & risk analysis
3. Honesty (transparency)
4. No plagiarism
5. Respect privacy
6. Confidentiality
7. High quality & standards (competence)
8. Professional review
9. Inform society

Advancing Computing as a Science & Profession, *ACM Code of Ethics and Professional Conduct*. Online [2019]: [acm.org/code-of-ethics](https://www.acm.org/code-of-ethics)

# Fun with biometrics

- InterSoB task
  - <https://how-old.net/>
  - Try to appear as old as possible

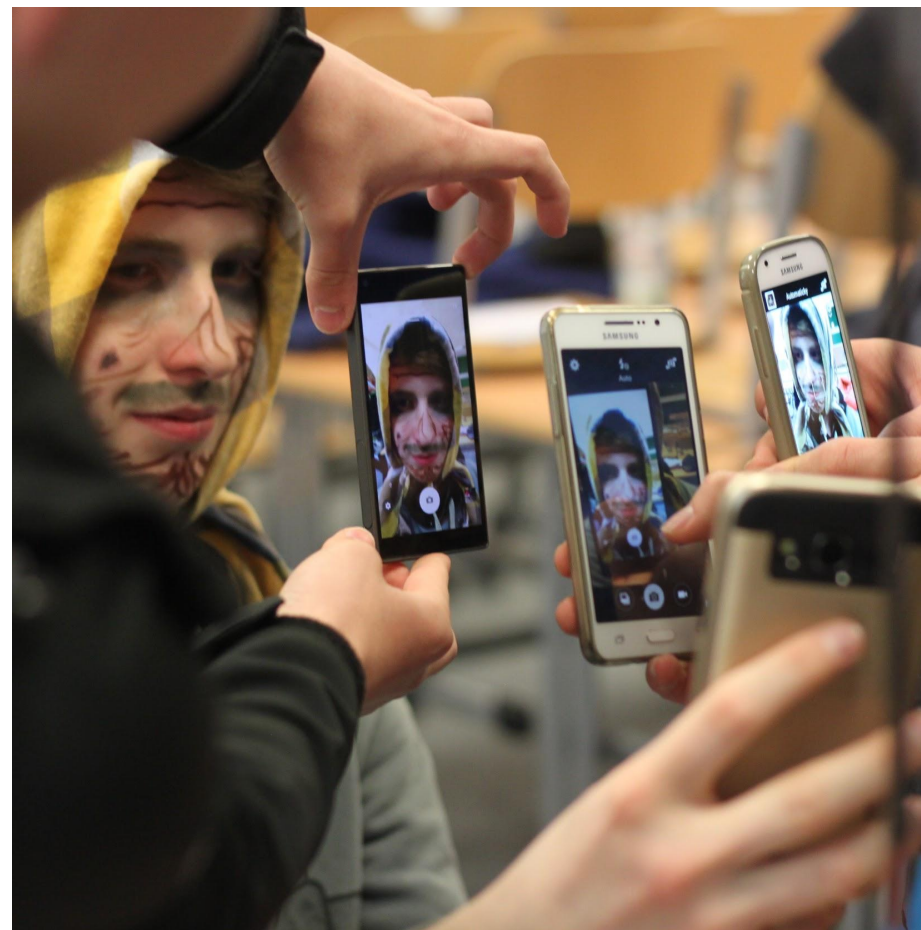


Photo © 2016 Dominika Krejčí, InterSoB

## Detour: SWOT analysis

- A.k.a. “SWOT matrix”
- From 1960s
- Strategic planning technique related to business competition or project planning
- Widely applicable



# SWOT example: Passwords

## Strengths

- Well understood
- Legacy
- Intuitive usage
- Possibility of high entropy

## Opportunities

- FIDO 2.0 system
- Integration of SMS/OPT and Push-to-Approve

## Weaknesses

- Often low entropy
- Infinite ways to implement
- Policy differences
- Sticky note syndrome
- Threats related to storage

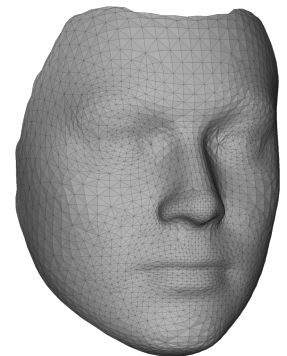
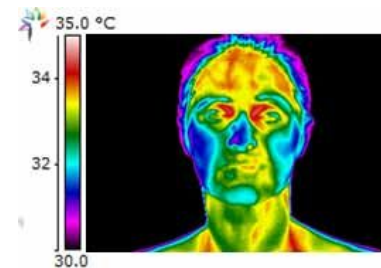
## Threats

- Bad attack understanding
- Long tail of replacement
- Usability issues
- The dark web

Example inspired by the RSAC 2018 talk *Passwords and fingerprints and faces – Oh my! Comparing old and new authentication* by Jackson Shaw

## Seminar task

- Do a SWOT analysis for a use case on face recognition biometrics, work in groups of three
- Use cases:
  - a. Face authentication on border crossing (passports)
  - b. “Pay by a smile” for Internet card payments
  - c. 3D face authentication for accessing bank vaults
  - d. Thermal face scans securing nuclear power plant



# Homework

Exploring automatic age estimation

## Homework: Overview

- Investigate what influences age estimation
  - In <https://how-old.net/> (neural-networks based)
  - Adjust our pictures again
- Submit to IS MUNI a **single ZIP file** with
  - Report (PDF),  
see next slide
  - Used adjusted images
- Deadline:  
4. 12. 2018 8:00



# Homework: Overview

## **Step 1: State the hypotheses.**

E.g., Wrinkles around the tails of eyes increase the estimated age.

## **Step 2: Set the criteria for a decision.**

Set baseline (no wrinkles) and repeat measurement for different wrinkles around tails of eyes.

## **Step 3: Compute the test statistic (if you know how).**

In our simplistic case, take a look on measurements.

This is not necessary, if you don't understand statistics well.

## **Step 4: Interpret the results.**

The hypothesis should not be regarded as true based on these data.



# Homework: Good methodology



## Measurements:

Martin 1 - 27

Martin 2 - 27

Martin 3 - 27

Martin 4 - 27

Martin 5 - 27

# Homework: Report

- Write a summarizing report
  - Your hypotheses and how you tested them
  - Test at least 5 distinct features
- Concentrate on:
  - Having a formulated hypotheses for each feature (e.g. smoother skin decreases estimated age)
  - Having several images supporting/falsifying your idea
- Avoid:
  - Many changes in the face at once
  - Radical changes (deleting half the face)
  - Overgeneralization

## Homework: Methodology, scoring

- Up to 10 points awarded
  - Scoring rubric available in the Information system
  - The rubric can help you understand what is important in the task!
- Have a look at old homework submissions with good methodology in the Study Materials.
  - Special thanks to Jan Kvapil and Rao Arvind for providing them.

# Homework: Bad methodology (but at least funny)

