

# Biometrics 2

## Face recognition



**PV181 Laboratory of security and applied cryptography**  
**Seminar 10, 22. 11. 2017**

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# Lecture structure

## Seminar 1

1. Introduction
2. Fingerprints
3. Hands-on:
  - Generate fingerprints
  - Fake fingerprints
4. Homework:
  - Fake fingerprints

## Seminar 2

1. Face recognition
2. Hands-on:
  - Face matching
  - Fake fingerprints validation
3. Homework:
  - Age estimation

# Real-life example

The Joy of Tech™

by Nitrozac & Snaggy

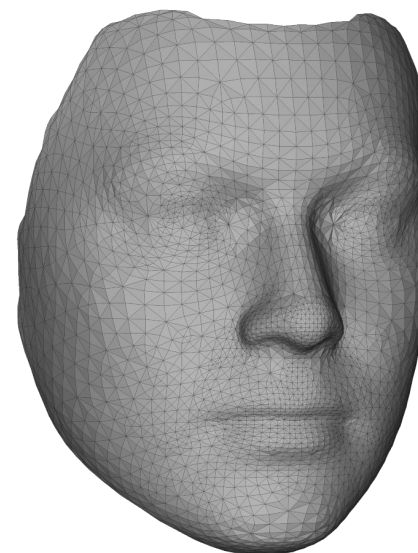
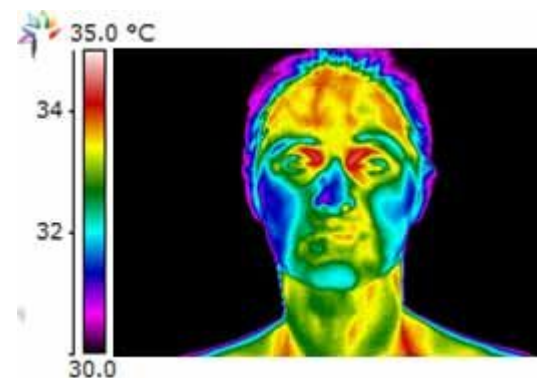


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joyoftech.com

## Face recognition – Input

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



# Face recognition

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

# Open source frameworks

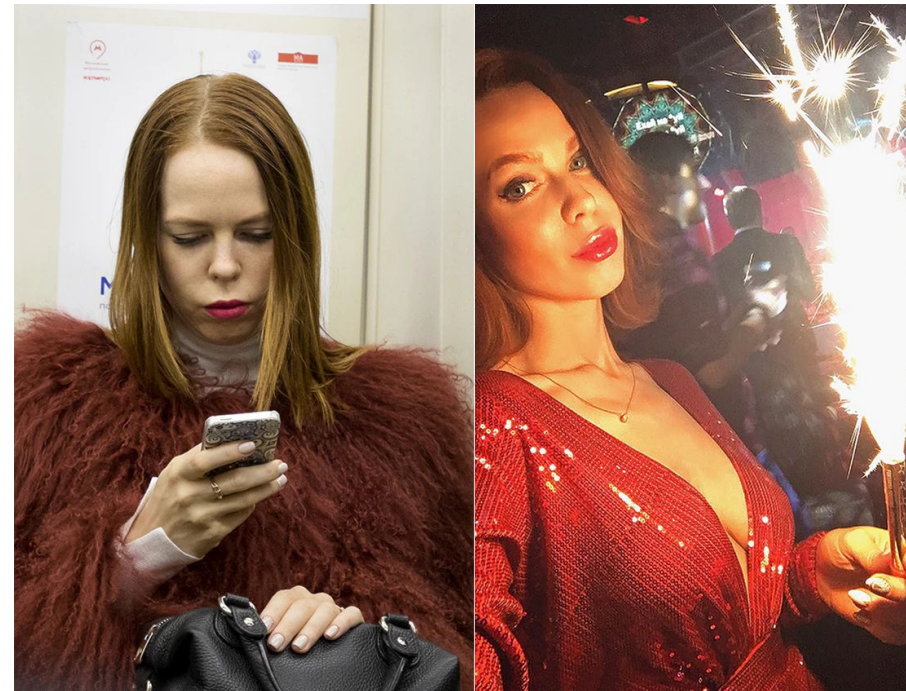
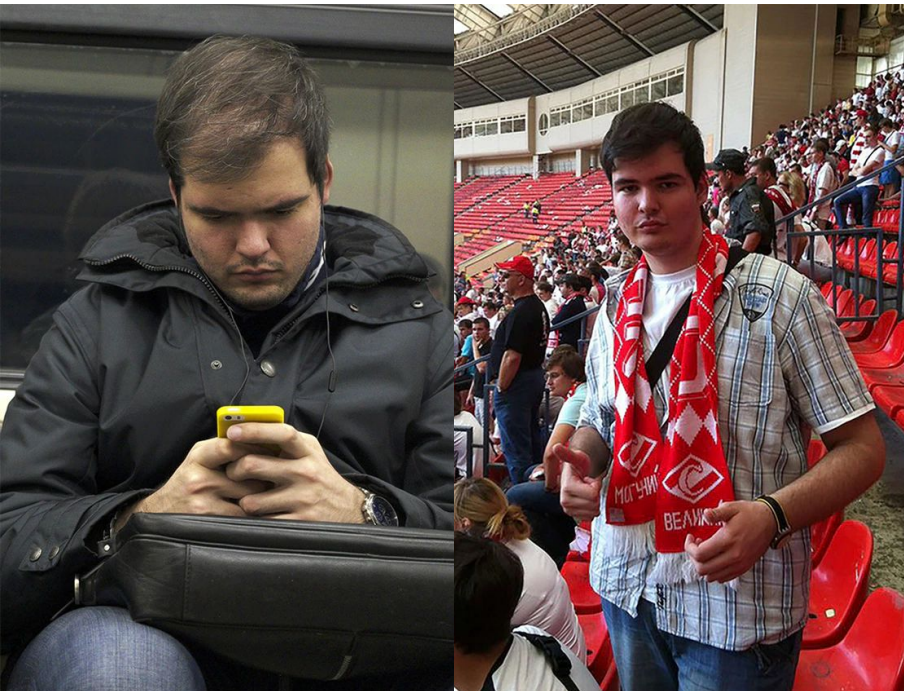
| Project    | Modern | Active | Deployable |
|------------|--------|--------|------------|
| CSU [17]   | Yes    | No     | No         |
| OpenCV [4] | No     | Yes    | Yes        |
| OpenBR     | Yes    | Yes    | Yes        |

Table 1: Existing open source face recognition software. A project is considered *modern* if it incorporates peer-reviewed methods published in the last five years, *active* if it has source code changes made within the last six months, and *deployable* if it exposes a public API.

J. Klontz, B. Klare, S. Klum, A. Jain, M. Burge. "Open Source Biometric Recognition", *Proceedings of the IEEE Conference on Biometrics: Theory, Applications and Systems (BTAS)*, 2013.

# FindFace – example

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



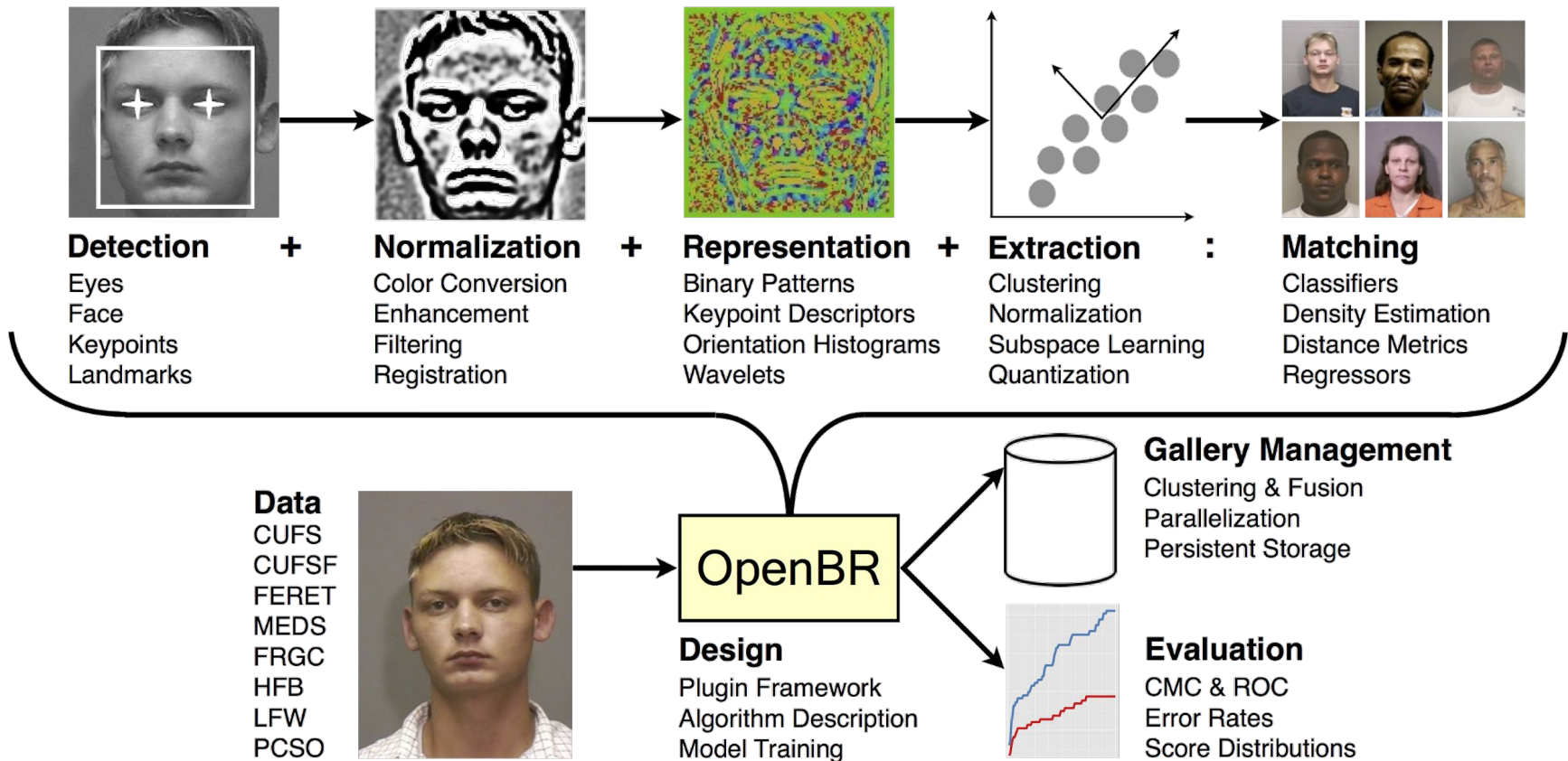
# Challenges in face recognition

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





# OpenBR: Face recognition overview

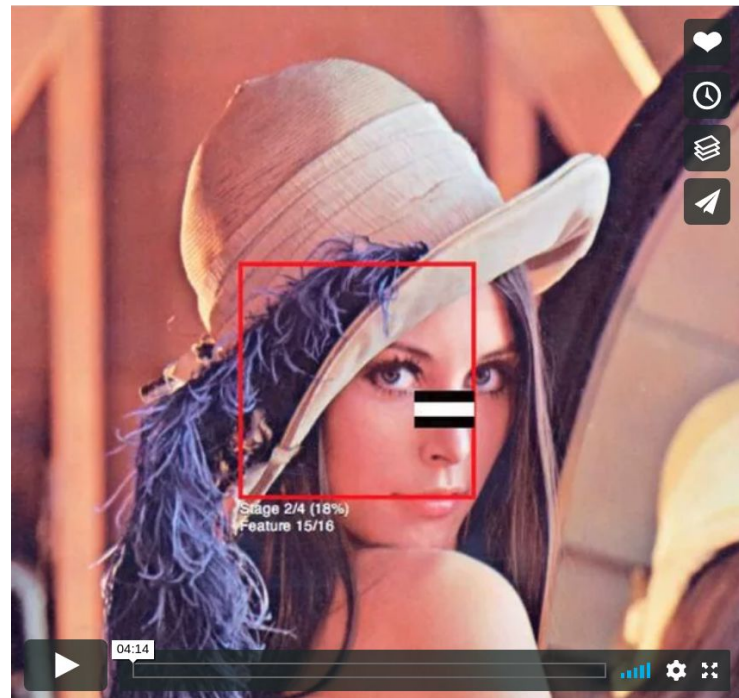


## Step 1 – Face detection

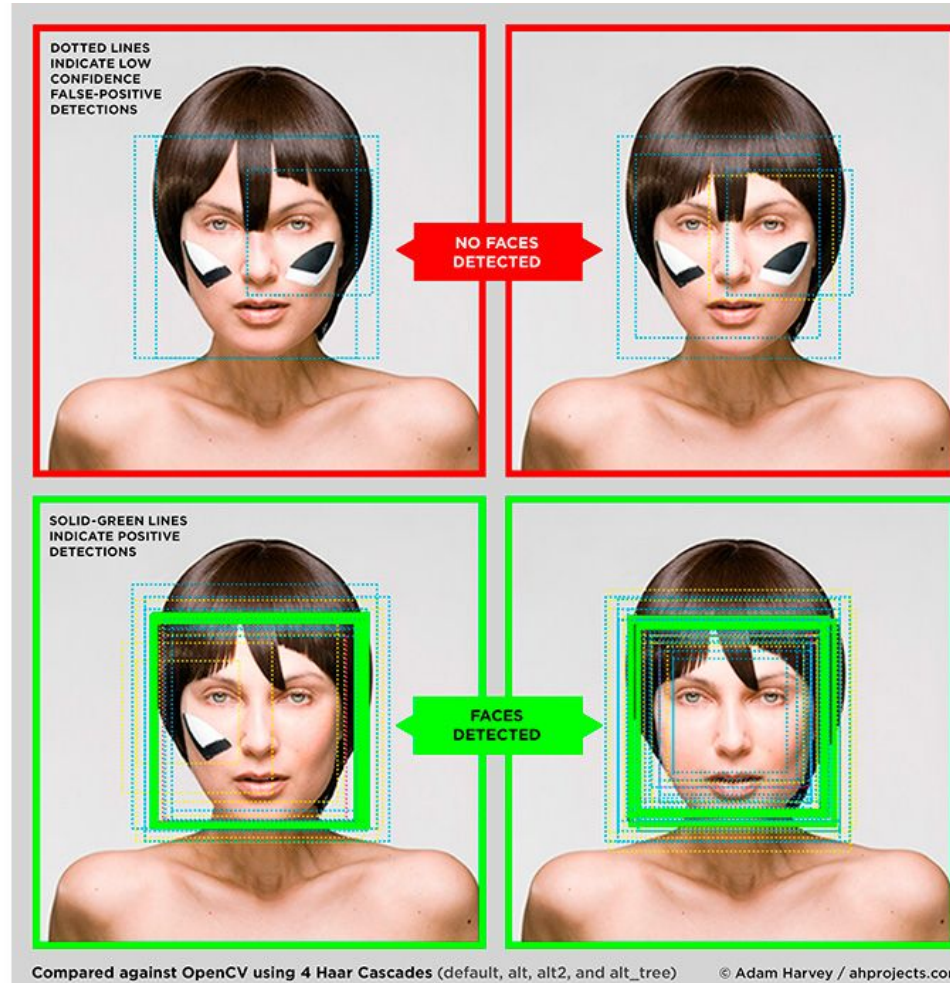
- Knowledge-based methods.
  - Ruled-based methods that encode our knowledge of human faces.
- Template matching methods.
  - These algorithms compare input images with stored patterns of faces or features.
- Appearance-based methods.
  - A template matching method whose pattern database is learnt from a set of training images.

# 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

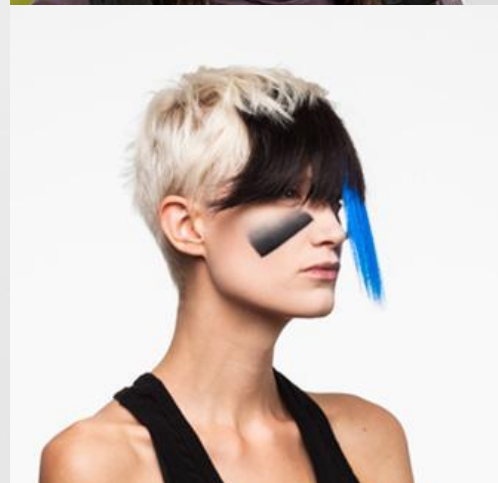


Photo © 2010-2016 Adam Harvey, CV Dazzle

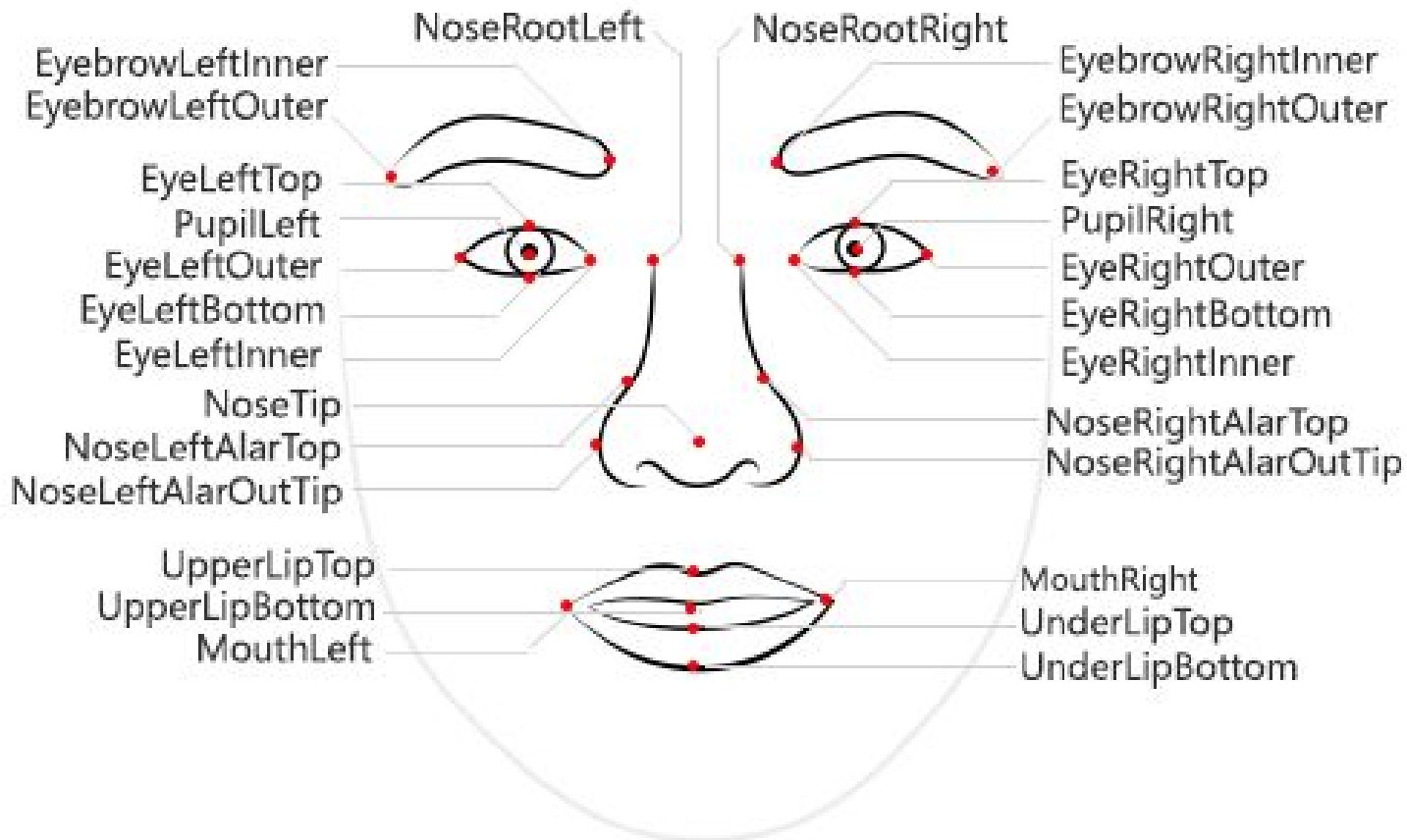
## Step 2 – Normalization and Representation

- Picture preprocessing
- OpenBR approach (Eigenface):
  - Detects eyes in detected faces
  - Normalize the face with respect to rotation and scale using the eye locations
  - Converts the image to floating point format
  - Embeds the image in a PCA subspace trained on face images

## Step 3 – Extraction

- Extracting relevant information from image
- Face color? Position of eyes, mouth, nose?  
Between eyes ratio? Width-length ratio?
- Information must be valuable to the later step of identifying the subject
- “Reducing dimension”

# Microsoft: Face API



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## Step 4 – Matching

- Template matching
  - Patterns are represented by samples, models, pixels, curves, textures. The recognition function is usually a correlation or distance measure.
- Statistical approach
  - Patterns are represented as features. The recognition function is a discriminant function.
- Neural networks
  - The representation may vary. There is a network function in some point.

## Step 5 – Output

- Confidence:
  - Euclidian distance as match measure
  - Interval 0 (=bad match) to 1 (=perfect match)
  - Cca  $>0.6$  to detect similarity
- Similarity value for comparing two templates
  - The higher value the more likely the same
  - Computed as  $-\log(\text{distance}+1)$  where distance is the sum of the Euclidean distances between two face images
  - Smaller distances (Euclidean) indicate higher similarity

# 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

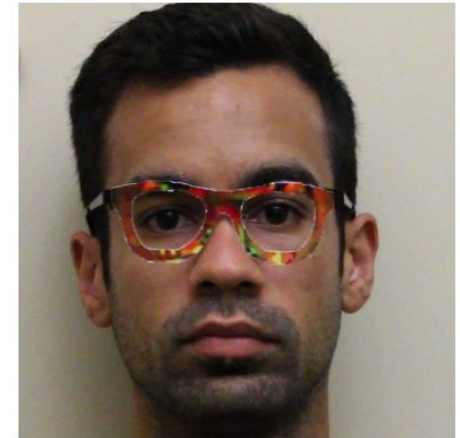


Photo © 2016 Carnegie Mellon University, *Accessorize to a Crime: Real and Stealthy Attacks on State-of-the-Art Face Recognition*

## 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.*

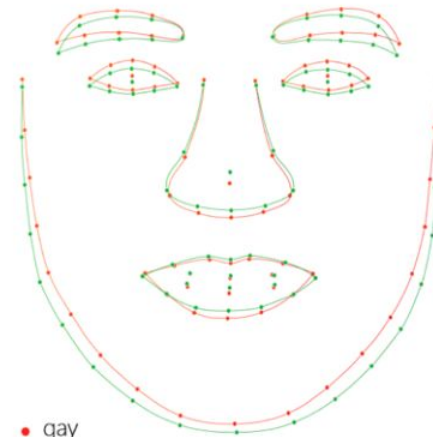
# Detecting sexual orientation from faces

Composite heterosexual faces

Composite gay faces

Average facial landmarks

Male



Female

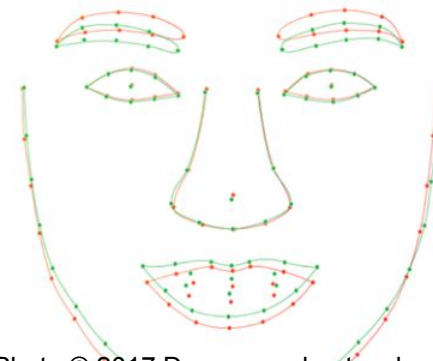
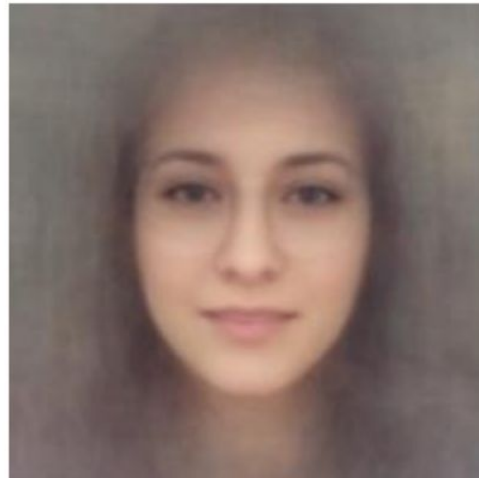


Photo © 2017 Deep neural networks are more accurate than humans at detecting sexual orientation from facial images. Journal of Personality and Social Psychology

## 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.*



## Testing sets (databases)

- Many databases:

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

- Covering:

- Aging
- Illumination
- Pose
- Expression

## Fun with biometrics

- Attractiveness measurement
  - <https://www.howhot.io/>
- InterSoB task
  - <https://how-old.net/>
  - Try to appear as old as possible

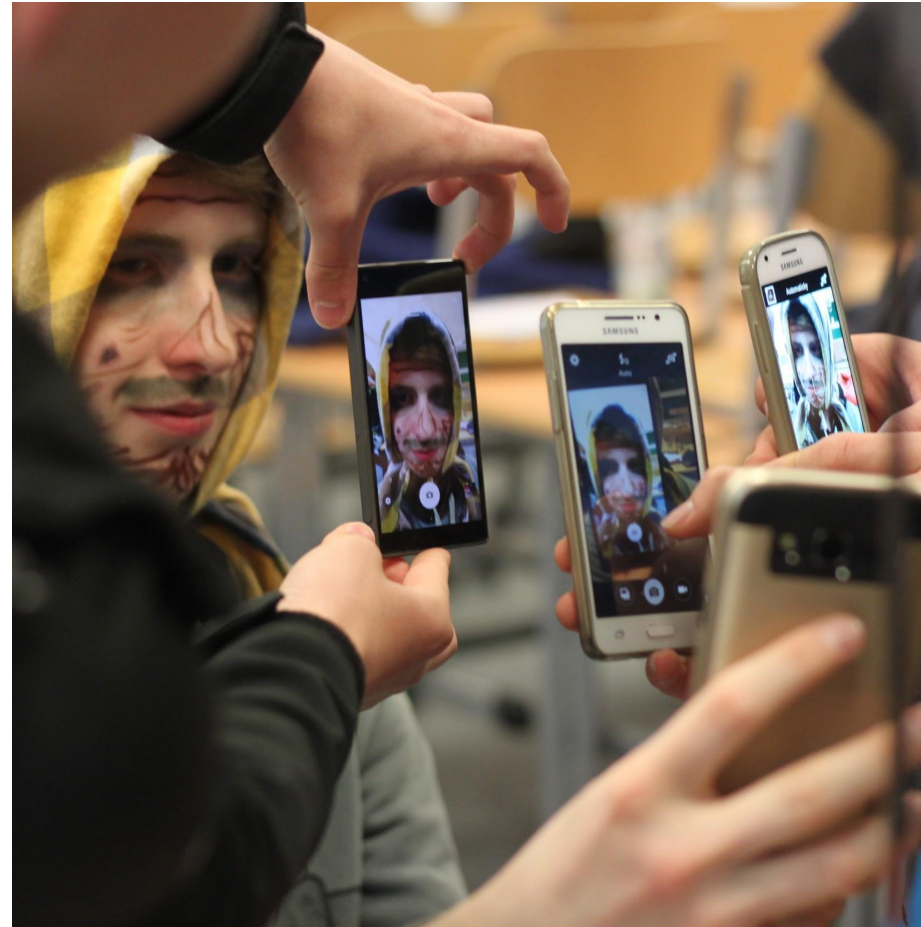


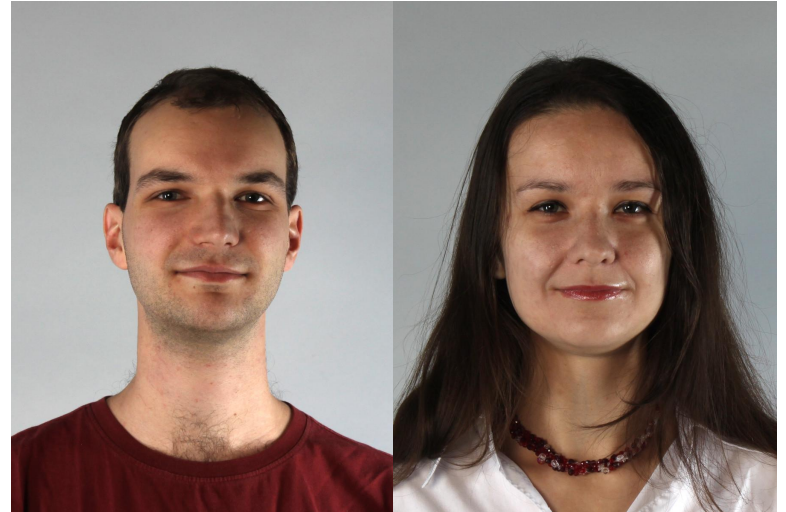
Photo © 2016 Dominika Krejčí, InterSoB

# Seminar task

Exploring face matching in OpenBR

## Seminar task

- Inspect what influences face recognition in OpenBR
  - <http://openbiometrics.org/>
- Use faces of the teachers to experiment
  - Compare images to “photoshopped versions” to determine what changes matter most
  - Be creative and playful!



## Seminar task (tips)

- What can influence identification/age estimation?
  - Distance between eyes/mouth/nose/...
  - Light/colour differences (think CV Dazzle)
  - Wrinkles, hair style, general “smoothness”
  - “Transplanting” eyes/parts of other faces
- What is necessary to avoid face detection?
  - Deleting/covering an eye/mouth/...
  - Multiple eyes/mouths/...
  - Colour changes, wrong distance ratios

## Seminar task (examples)



# OpenBR invocation (prepared VM)

- Face recognition/comparison

```
br -algorithm FaceRecognition -compare me.jpg you.jpg
```

Approximately: similarity < 2 is different people, similarity > 3 is the same person

- Age estimation

```
br -algorithm AgeEstimation -enroll me.jpg meta.csv
```

```
cat meta.csv
```

- Gender estimation

```
br -algorithm GenderEstimation -enroll me.jpg meta.csv
```

```
cat meta.csv
```

- Documentation

<http://openbiometrics.org/docs/tutorials/#face-recognition>

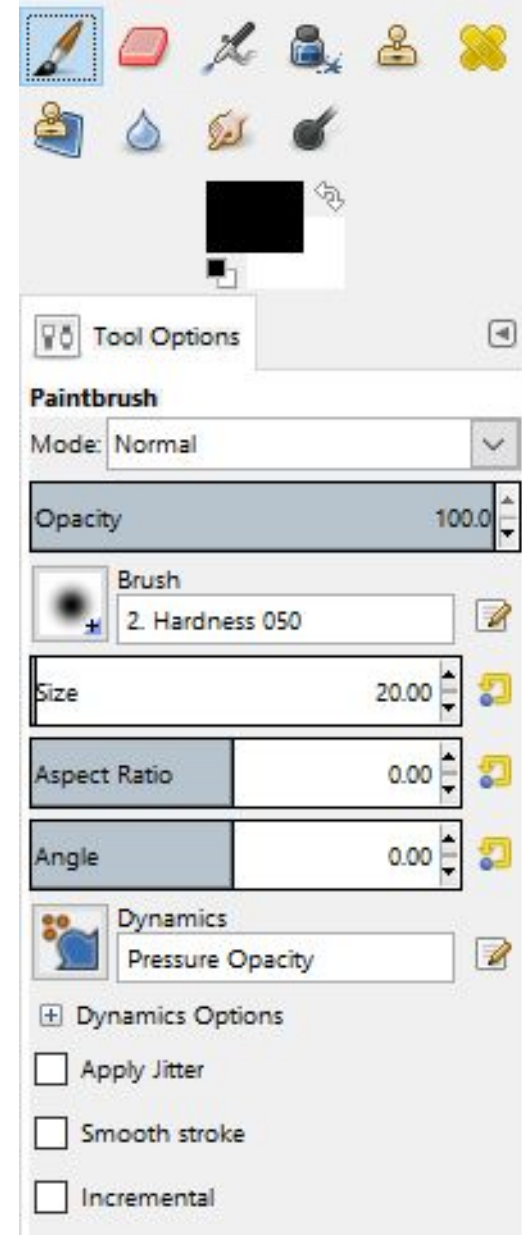
## Prepared VM (details)

- .ova file at O:\pv181\pv181-biometrics1
- Import PV181-biometrics1.ova to VirtualBox
  - Import appliance (don't create a new machine)
- Boot the system
  - Ubuntu 16.04, login: 'vagrant', password: 'vagrant'
  - Images available in Documents
  - Everything necessary is already installed
  - You are sudo, in case you want to add something
- It's build with [Vagrant](#)
  - Vagrantfile available in study materials, if interested



# GIMP basics

- Paintbrush tool
  - Shape, opacity, size
  - Mode (normal, darken, saturation, ...)
- Clone tool
  - Select source with Ctrl
- Smudge tool
- Others as you see fit...
- You may want single-window mode
  - Windows > Single-Window Mode

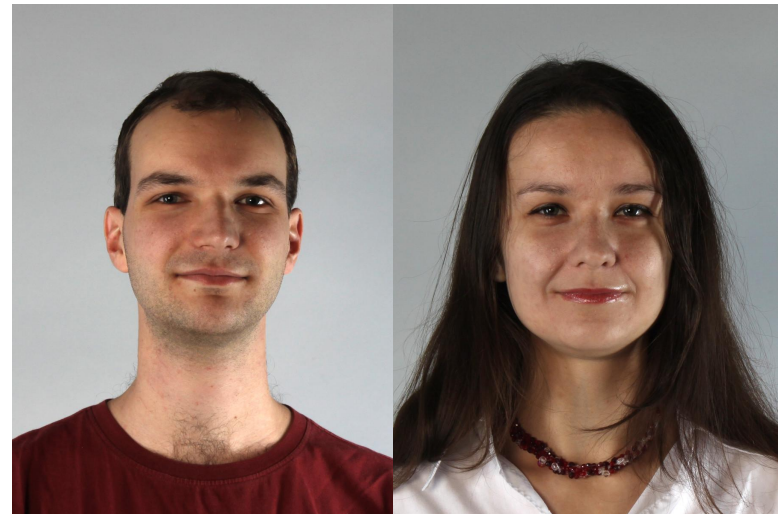


# Homework

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 homework vault **a single ZIP file** with
  - Report (PDF),  
see next slide
  - Used adjusted images
- Due date: 6. 12. 2017



# 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