

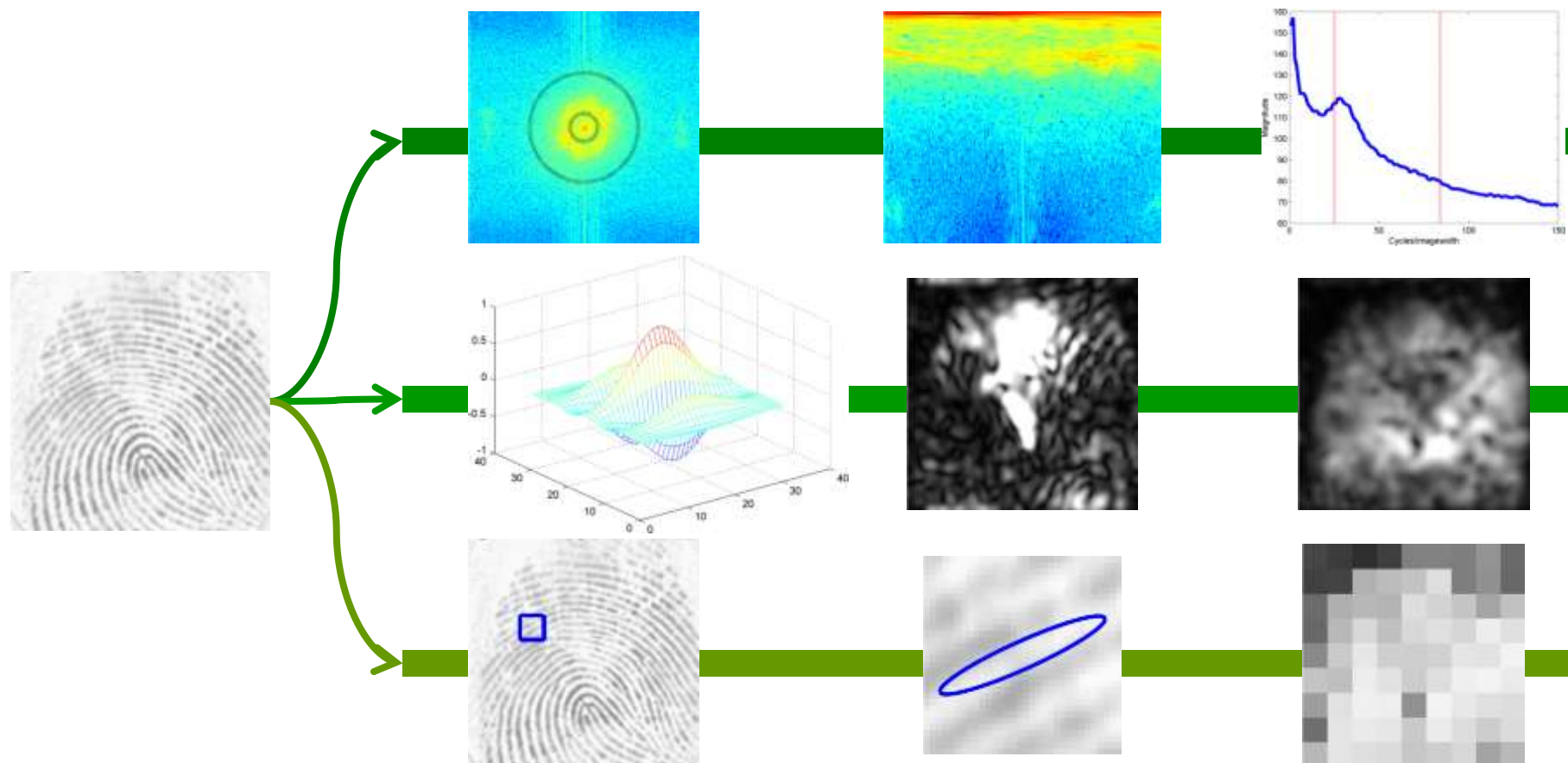
Candidate Features for Quality Assessment

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Quality Features For NFIQ 2.0

- Requirements
 - Based on publically available algorithms
 - Standardized interface (inputs and outputs)



Implemented Quality Features

- More than 30 features identified and tested on multiple datasets
 - NFIQ
 - Implemented from ISO/IEC TR 29794-4
 - Frequency Domain Analysis
 - Local Clarity Score
 - Orientation Certainty Level
 - Orientation Flow
 - Radial Power Spectrum
 - Ridge Valley Uniformity
 - Gabor filter
 - Ridge line count
 - Gabor (Shen et al., Quality Measures of Fingerprint Images, 2001)
 - Minutiae count in region of interest
 - FingerJetFX

Feature Example

Orientation Certainty Level

- See NFIQ 2.0 project page at http://www.nist.gov/itl/iad/ig/development_nfiq_2.cfm
- ISO/IEC 29794-4:2010
- Block wise approach

```
function [orientationCertaintyLevel] = compOcl(im, mask)
    allfun = inline('all(x(:))');
    [rows cols] = size(im);
    eblksz = ceil(sqrt(sum(v1sz.^2)));
    blkoffset = ceil((eblksz - blkksz)/2);
    mapsize = fix(([rows cols] - (eblksz - blkksz))./blk)
    maskBseg = false(mapsize);

    ocls = zeros(mapsize);
```

```
end
```

```
}
```

OCL input parameters		
Name	Default	Description
I	-	Input image
B_h	32	Block height in pixels
B_w	32	Block width in pixels
$mask$	-	Segmentation mask

Orientation Certainty Level

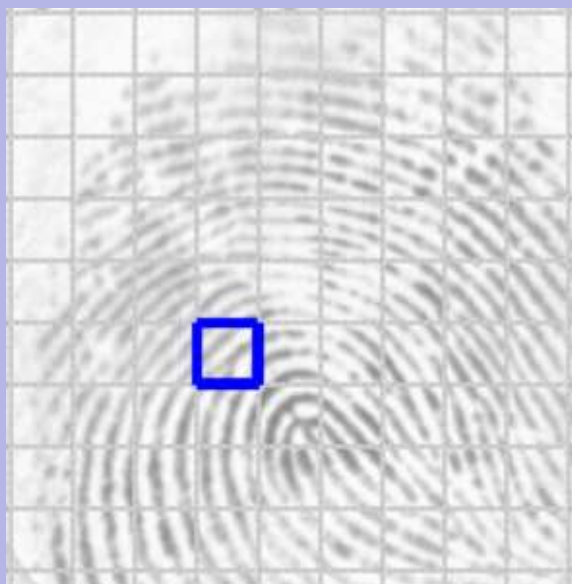
Algorithm → Covariance

1. Compute the intensity gradient of each block

$$[dx \quad dy] = \text{gradient}(B)$$

2. Compute the covariance matrix from the gradients

$$C = \frac{1}{N} \sum_N \begin{Bmatrix} dx & dy \end{Bmatrix} = \begin{bmatrix} a & c \\ c & d \end{bmatrix}$$

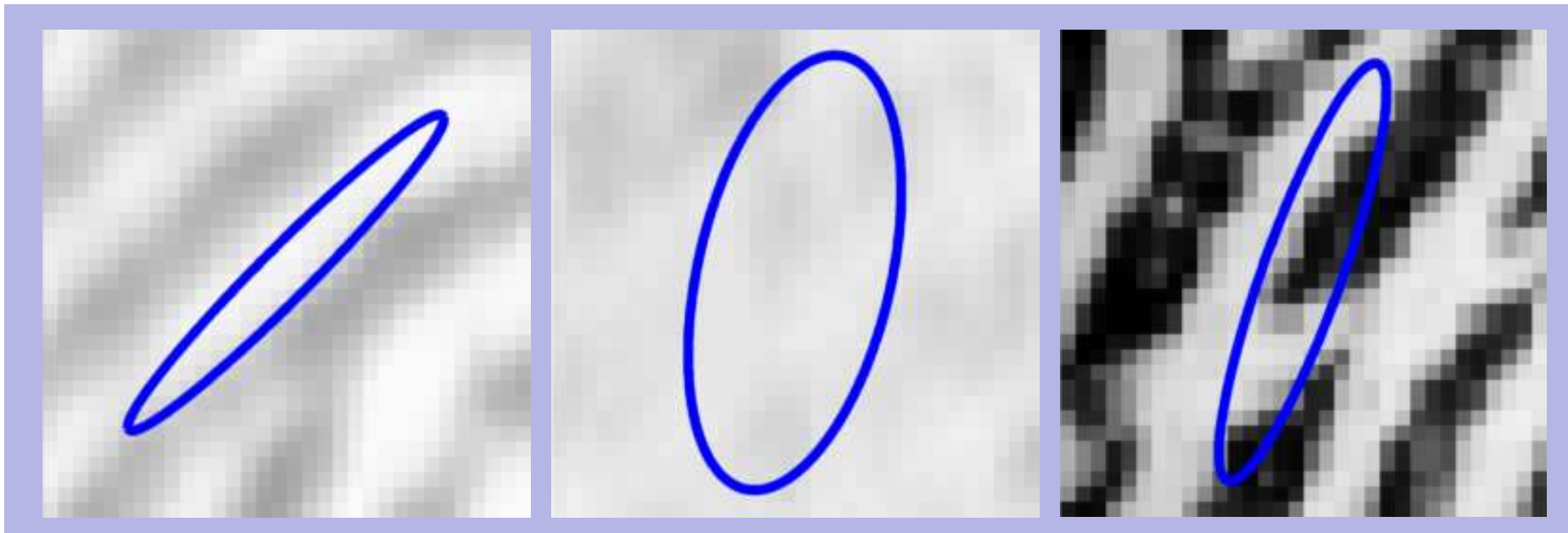


Orientation Certainty Level

Algorithm → Eigenvalue computation

3. Compute the eigenvalues to obtain *OCL* for each block

$$\lambda_{min} = \frac{a + b - \sqrt{(a - b)^2 + 4c^2}}{2}$$
$$\lambda_{max} = \frac{a + b + \sqrt{(a - b)^2 + 4c^2}}{2}$$



Data collection for wet/dry detection

Overview

- 5 fingerprint sensors (optical)
- 33 subjects
- 4 impressions/finger/sensor
- 6600 images total
- 4 types of treatment
 - No treatment
 - alcohol-dried
 - crème-moisturized
 - water

Data collection for wet/dry detection

Examples

- Objective measurement of skin moisture level



62.5%



84.6%



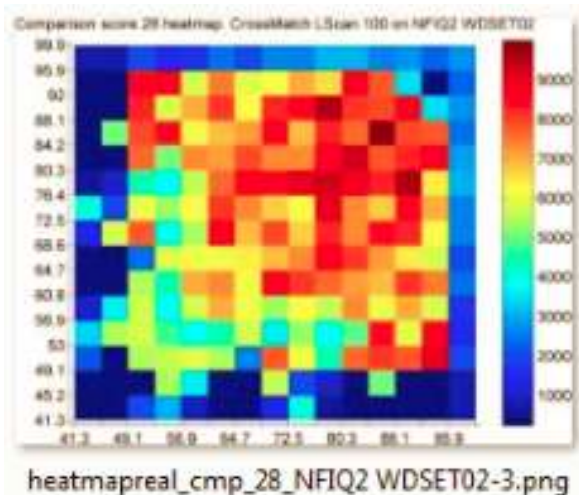
99.9%



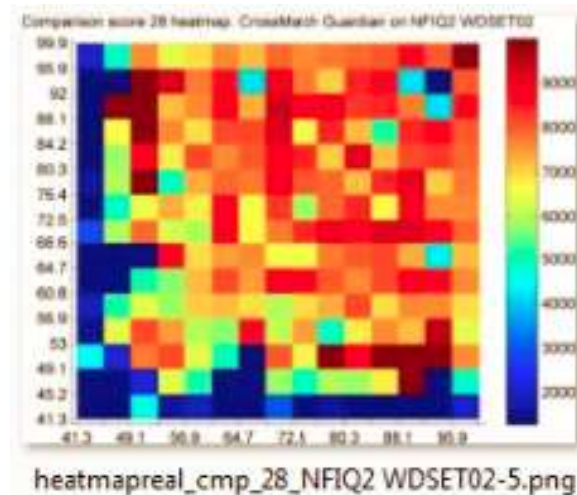
Observations on wet/dry impressions

- Wet fingerprints are generally handled well by recent sensors
- Dry fingerprints cause degradation in comparison score

Older generation sensor



New generation sensor



See master thesis by Marek Dusio (DTU) – to be published in June 2013

NFIQ 2.0 Lite

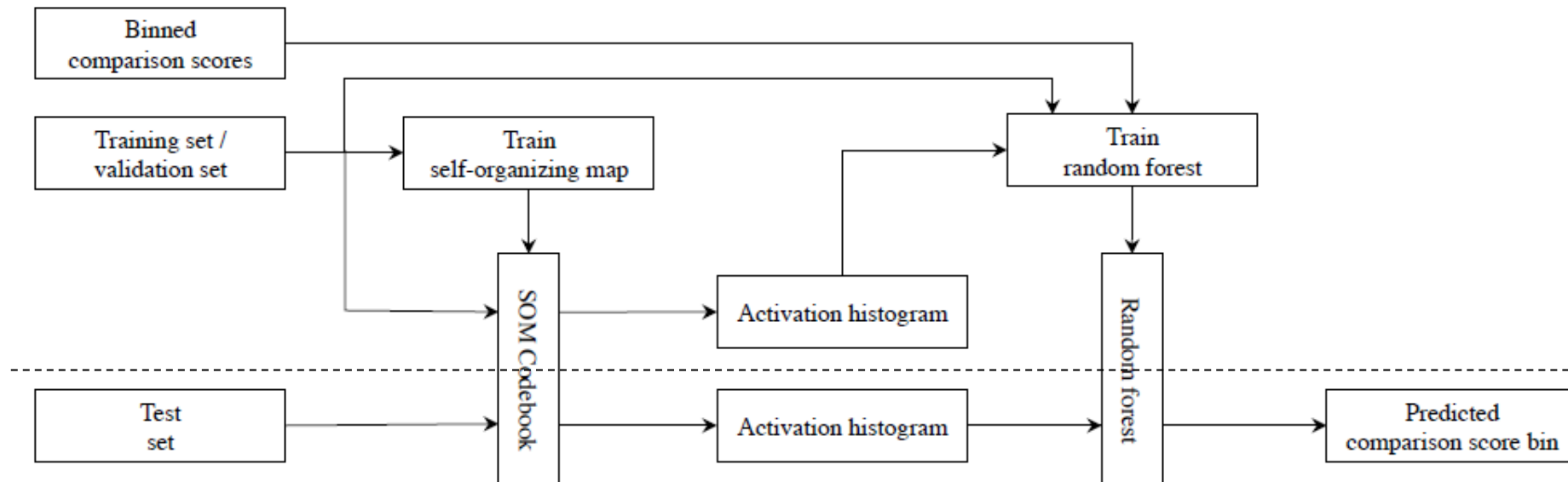
- Motivation:
 - Execution speed of feature extraction is important in some applications (even though processors are getting faster and faster). Aim at 125–150 ms for inclusion of quality assessment into auto-capture loop of sensors. – IBPC '12
- Potential solution:
 - Pre-compute a lookup table which can speed up the quality assessment

NFIQ 2.0 Lite

Machine Learning approach

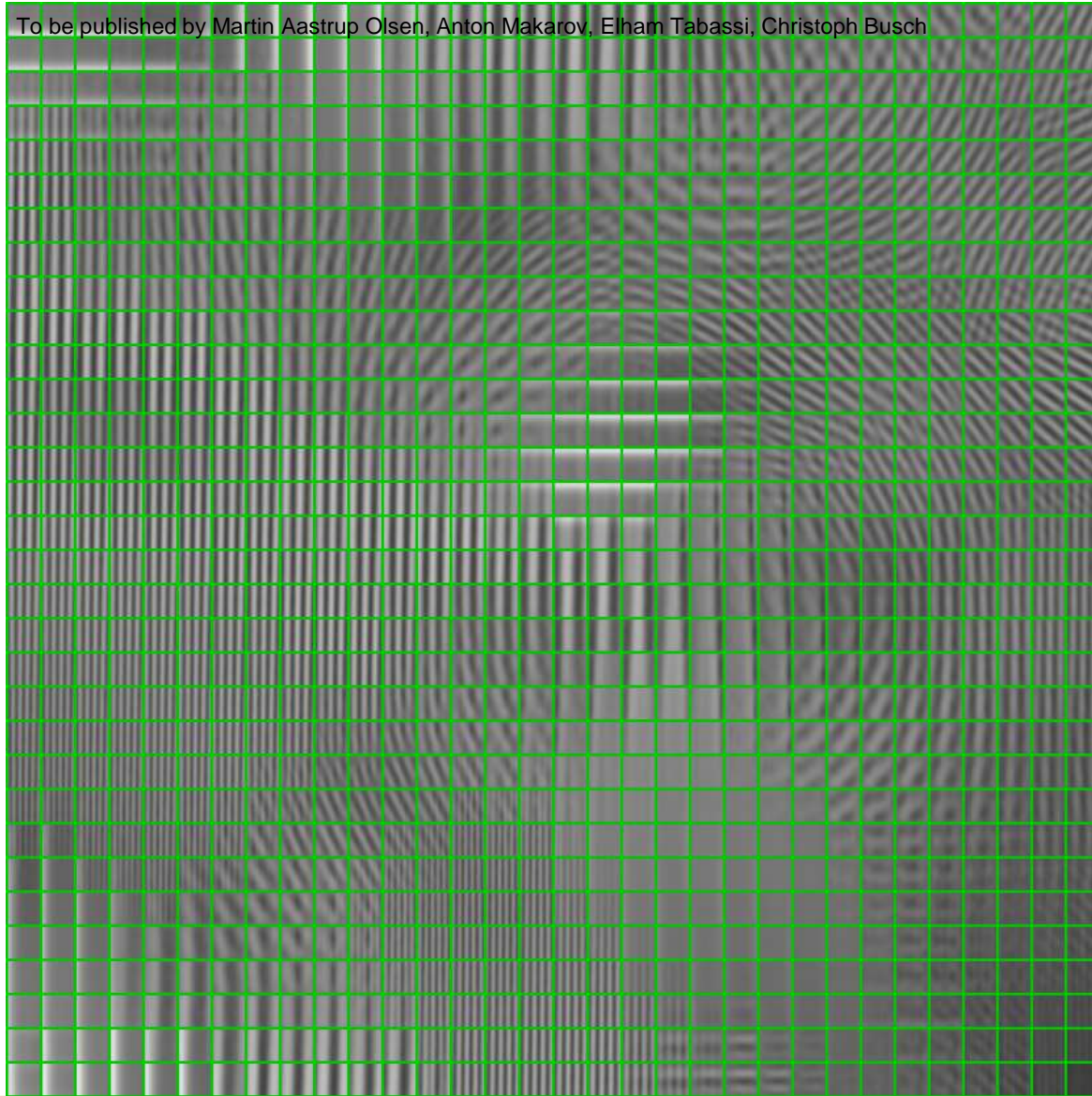
- Two stage process
 - Clustering using Self-Organizing Map
 - Prediction using Random Forest

To be published by Martin Aastrup Olsen, Anton Makarov, Elham Tabassi, Christoph Busch

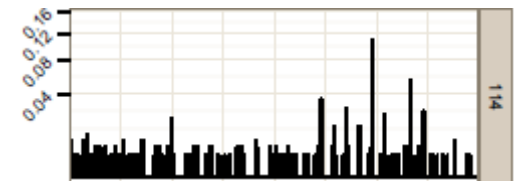
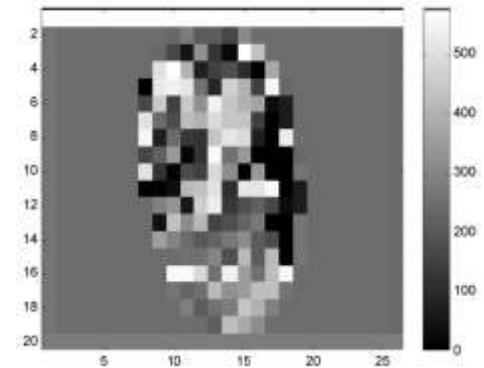


SOM example

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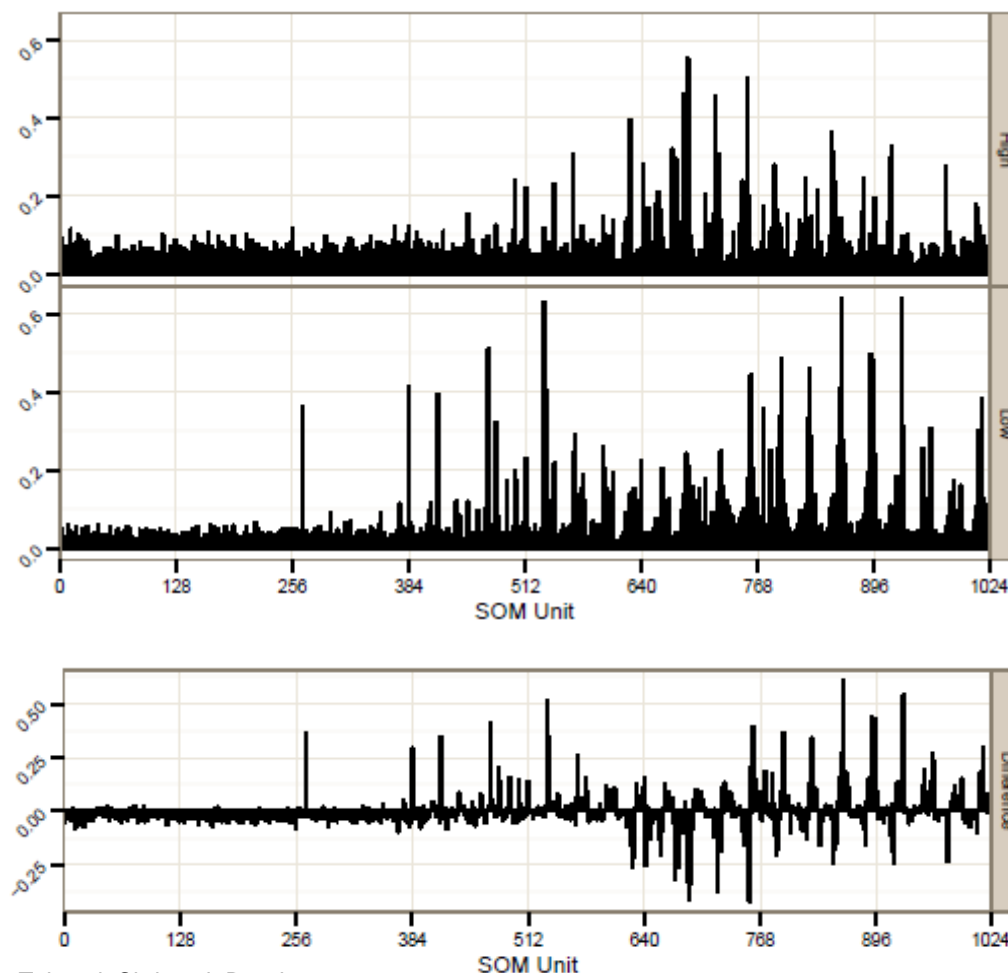


Blocksize: 24 x 24 px
SOM units: 32 x 32



SOM activation histogram

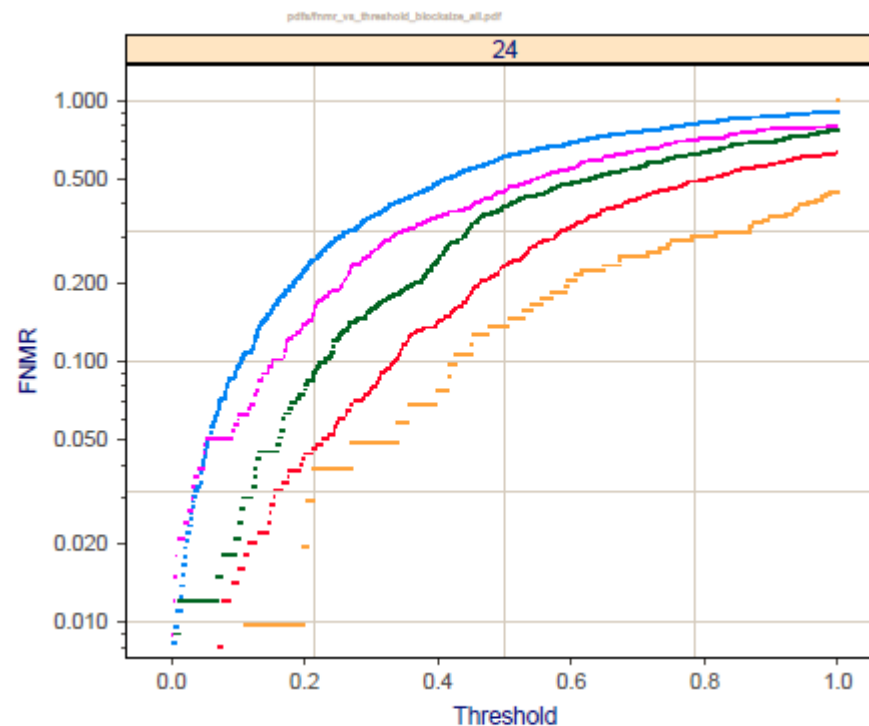
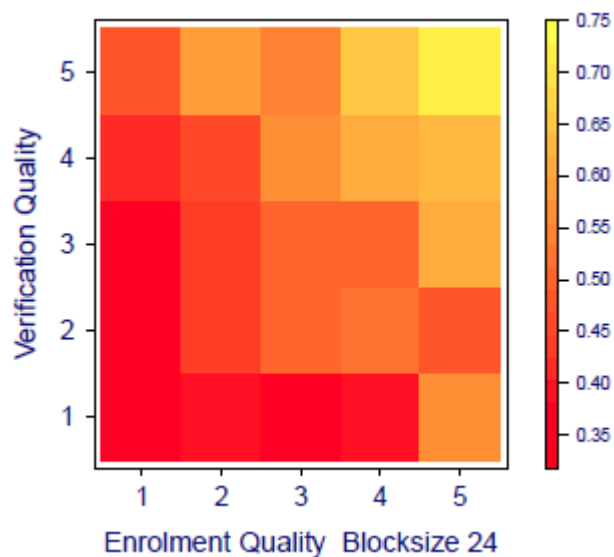
- 64 images with highest comparison score
- 64 images producing false non-match at $\text{FNMR} = 10^{-4}$
- Difference between histograms.



To be published by Martin Aastrup Olsen, Anton Makarov, Elham Tabassi, Christoph Busch

Performance indication

- FNMR vs. score threshold for each level of quality





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