



NFIQ 2.0 – Features for fingerprint quality determination

Martin A. Olsen

Norwegian University of Science and Technology (NTNU)

NIST, Gaithersburg, MD
May 4, 2016



Outline

Introduction

NFIQ 2.0 Quality features

Quality features

Two ground-truth classes

Quality feature example - frequency domain analysis

Actionable feedback

Speeding up NFIQ 2.0

NFIQ 2.0 and WSQ compression

Alignment with international standard

Contact & further information



- ▶ Starting point for features
 - ▶ NFIQ 1.0
 - ▶ ISO/IEC TR 29794-4:2010
 - ▶ Literature



- ▶ Starting point for features
 - ▶ NFIQ 1.0
 - ▶ ISO/IEC TR 29794-4:2010
 - ▶ Literature
- ▶ Implementation of prototype features
- ▶ Hundreds of variations of features; parameter configurations and variations in algorithm steps



- ▶ Starting point for features
 - ▶ NFIQ 1.0
 - ▶ ISO/IEC TR 29794-4:2010
 - ▶ Literature
- ▶ Implementation of prototype features
- ▶ Hundreds of variations of features; parameter configurations and variations in algorithm steps
- ▶ Iterative development to arrive at NFIQ 2.0 feature vector



- ▶ Starting point for features
 - ▶ NFIQ 1.0
 - ▶ ISO/IEC TR 29794-4:2010
 - ▶ Literature
- ▶ Implementation of prototype features
- ▶ Hundreds of variations of features; parameter configurations and variations in algorithm steps
- ▶ Iterative development to arrive at NFIQ 2.0 feature vector
- ▶ Prioritize predictive power and speed of computation



- ▶ Starting point for features
 - ▶ NFIQ 1.0
 - ▶ ISO/IEC TR 29794-4:2010
 - ▶ Literature
- ▶ Implementation of prototype features
- ▶ Hundreds of variations of features; parameter configurations and variations in algorithm steps
- ▶ Iterative development to arrive at NFIQ 2.0 feature vector
- ▶ Prioritize predictive power and speed of computation
- ▶ Workshops central to development of features



- ▶ NFIQ 2.0 is a classifier \Rightarrow features form the basis for prediction
- ▶ Selecting features
 - ▶ Speed of computation
 - ▶ Contribution to predictive performance



- ▶ NFIQ 2.0 is a classifier \Rightarrow features form the basis for prediction
- ▶ Selecting features
 - ▶ Speed of computation
 - ▶ Contribution to predictive performance

$$\begin{aligned}
 \mathbf{Q}_{\text{NFIQ 2.0}} = & \left(Q_{\text{FDA}}^{\mu}, Q_{\text{LCS}}^{\mu}, Q_{\text{OCL}}^{\mu}, Q_{\text{OFL}}^{\mu}, Q_{\text{RVU}}^{\mu}, \right. \\
 & Q_{\text{FDA}}^{\sigma}, Q_{\text{LCS}}^{\sigma}, Q_{\text{OCL}}^{\sigma}, Q_{\text{OFL}}^{\sigma}, Q_{\text{RVU}}^{\sigma}, \\
 & \mathbf{Q}_{\text{FDA}}, \mathbf{Q}_{\text{LCS}}, \mathbf{Q}_{\text{OCL}}, \mathbf{Q}_{\text{OFL}}, \mathbf{Q}_{\text{RVU}}, \\
 & Q_{\text{MU}}, Q_{\text{MMB}}, Q_{\text{COH}}^{\text{rel}}, Q_{\text{COH}}^{\text{sum}}, Q_{\text{AREA}}^{\mu}, \\
 & \left. Q_{\text{MIN}}^{\text{cnt}}, Q_{\text{MIN}}^{\text{com}}, Q_{\text{MIN}}^{\text{mu}}, Q_{\text{MIN}}^{\text{ocl}} \right).
 \end{aligned}$$



- ▶ Global – minutiae count, orientation coherence, ...



- ▶ Global – minutiae count, orientation coherence, ...
- ▶ Local – orientation certainty, frequency analysis, ...



- ▶ Global – minutiae count, orientation coherence, ...
- ▶ Local – orientation certainty, frequency analysis, ...
- ▶ Local quality at minutiae locations
- ▶ Mean and standard deviation of local features
- ▶ Histogram of local features (boundaries determined from CDF)



- ▶ Global – minutiae count, orientation coherence, ...
- ▶ Local – orientation certainty, frequency analysis, ...
- ▶ Local quality at minutiae locations
- ▶ Mean and standard deviation of local features
- ▶ Histogram of local features (boundaries determined from CDF)
- ▶ Classifier
 - ▶ Random Forest trained for binary classification
 - ▶ Input: 69 dimensional feature vector
 - ▶ Output: probability of input being Class 1 (high utility) quantized [1, 100]



Name	Capture mode	Type	Number of subjects	Fingers	Number of comparisons per finger	Used for
AZLA	Scanned ink	Operational	240,000	Index and Thumb	120,000 mated. 120,000 non-mated	training + testing
POEBVA	Live scan	Operational	180,000	Index	120,000 mated. 120,000 non-mated	training + testing
VISITIDF	Live scan	Operational	220,000	Index and Thumb	95,000 mated. 120,000 non-mated	training + testing
DHS2	Live scan	Operational	180,000	Index	120,000 mated. 120,000 non-mated	training + testing
IQMI	Scanned ink	Operational	250,000	10 fingers	250,000 mated. 250,000 non-mated	testing
BKA	Live scan	Operational	342,000 images	10 fingers	—	testing
BKA	+ Scanned ink					
SD 29	Scanned ink	Public	209	10 fingers	1912 mated. 35,791 non-mated	testing
FVC 2000 DB1	Live scan	Public	110	8 fingers	—	compliance testing
FVC 2000 DB3	Live scan	Public	110	8 fingers	—	compliance testing
FVC 2002 DB1	Live scan	Public	110	8 fingers	—	compliance testing

► Data from operational sources (Optical sensors)



Name	Capture mode	Type	Number of subjects	Fingers	Number of comparisons per finger	Used for
AZLA	Scanned ink	Operational	240,000	Index and Thumb	120,000 mated. 120,000 non-mated	training + testing
POEBVA	Live scan	Operational	180,000	Index	120,000 mated. 120,000 non-mated	training + testing
VISITIDF	Live scan	Operational	220,000	Index and Thumb	95,000 mated. 120,000 non-mated	training + testing
DHS2	Live scan	Operational	180,000	Index	120,000 mated. 120,000 non-mated	training + testing
IQMI	Scanned ink	Operational	250,000	10 fingers	250,000 mated. 250,000 non-mated	testing
BKA	Live scan	Operational	342,000 images	10 fingers	—	testing
BKA	+ Scanned ink					
SD 29	Scanned ink	Public	209	10 fingers	1912 mated. 35,791 non-mated	testing
FVC 2000 DB1	Live scan	Public	110	8 fingers	—	compliance testing
FVC 2000 DB3	Live scan	Public	110	8 fingers	—	compliance testing
FVC 2002 DB1	Live scan	Public	110	8 fingers	—	compliance testing

- ▶ Data from operational sources (Optical sensors)
- ▶ Training set 6629 images (3295 in Class 0 and 3334 in Class 1)
- ▶ Validation set 99797 randomly selected images
- ▶ External validation on BKA data and FBI data



- ▶ Criteria for two classes of samples in training



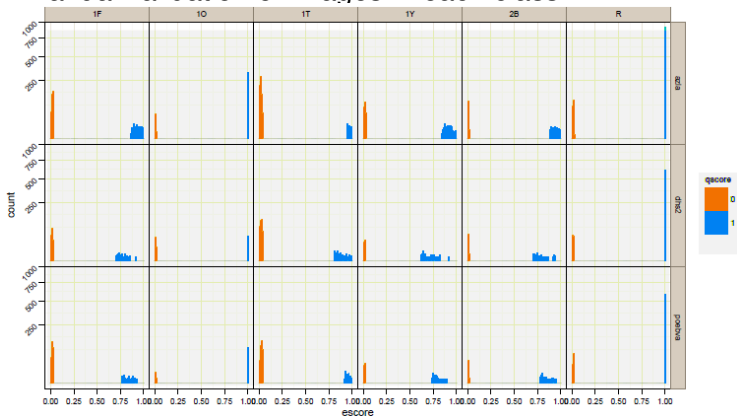
- ▶ Criteria for two classes of samples in training
 - 1 NFIQ=1 ($S_{act} > 0.7$) and S_{gen} in 90th percentile
 - 0 NFIQ=5 ($S_{act} > 0.9$) and $S_{gen} < t$ at $FMR = 10^{-4}$



- ▶ Criteria for two classes of samples in training
 - 1 NFIQ=1 ($S_{act} > 0.7$) and S_{gen} in 90th percentile
 - 0 NFIQ=5 ($S_{act} > 0.9$) and $S_{gen} < t$ at $FMR = 10^{-4}$
- ▶ Manual validation of images in each class



- ▶ Criteria for two classes of samples in training
 - 1 NFIQ=1 ($S_{act} > 0.7$) and S_{gen} in 90th percentile
 - 0 NFIQ=5 ($S_{act} > 0.9$) and $S_{gen} < t$ at $FMR = 10^{-4}$
- ▶ Manual validation of images in each class





Feature importance ranking

	Name	MeanDreaseGini
Q_{FDA}^{σ}	Frequency Domain Analysis_Standard Deviation	140.760
Q_{MIN}^{com}	FingerJet FX OSE COM Minutiae Count	92.089
Q_{MIN}^{ocl}	FingerJet FX OSE OCL MinutiaeQuality	83.027
Q_{RVU}^{μ}	Ridge Valley Uniformity_Mean	69.517
Q_{FDA}^{μ}	Frequency Domain Analysis_Mean	62.229
Q_{MIN}^{cnt}	FingerJet FX OSE Total Minutiae Count	57.565
Q_{RVU}^{σ}	Ridge Valley Uniformity_Standard Deviation	50.946
Q_{LCS}^7	Local Clarity Score_Bin_7	50.688
Q_{LCS}^8	Local Clarity Score_Bin_8	50.100
Q_{FDA}^9	Frequency Domain Analysis_Bin_9	47.844
Q_{COH}^{sum}	ROI Orientation Map Coherence Sum	38.104
Q_{OFL}^2	Orientation Flow_Bin_2	37.172
Q_{LCS}^{μ}	Local Clarity Score_Mean	36.483
Q_{RVU}^5	Ridge Valley Uniformity_Bin_5	35.617
Q_{RVU}^3	Ridge Valley Uniformity_Bin_3	35.139
Q_{AREA}^{μ}	ROI Area Mean	34.932
Q_{OFL}^1	Orientation Flow_Bin_1	33.751
Q_{OFL}^0	Orientation Flow_Bin_0	33.513
Q_{MU}	MU	32.914

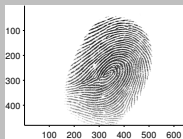


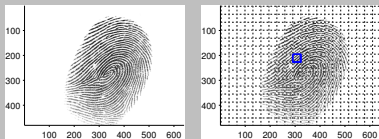
► Q_{FDA} local determination of ridge-valley signature

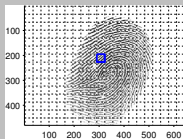
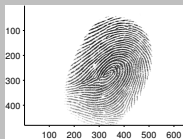
Algorithm 3: fda algorithm

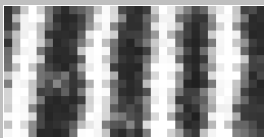
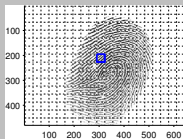
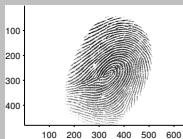
Input: Fingerprint image I
Output: fda quality score Q_{FDA}

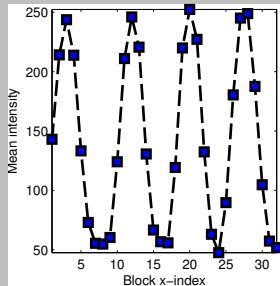
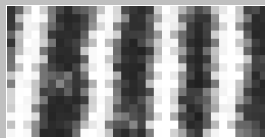
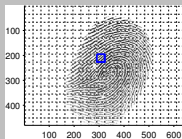
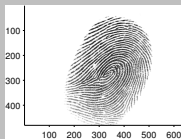
- 1 **for each block** V **in** I **do**
- 2 pad V with 2 pixel around border
- 3 rotate V with nearest neighbour interpolation such that dominant ridge flow is perpendicular to x-axis
- 4 crop V such that no invalid regions are included
- 5 with V obtain the ridge-valley signature T (eq. (11))
- 6 compute the dft of T to obtain the magnitude representation A
- 7 discard the first component of A
- 8 determine F_{max} as the index of the largest magnitude in A
- 9 compute $Q_{\text{FDA}}^{\text{local}}$ of V using A and F_{max} (eq. (12))
- 10 **end**

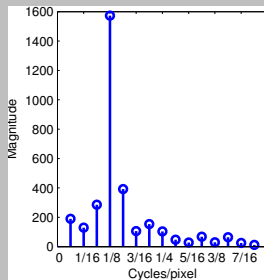
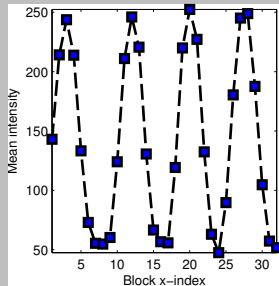
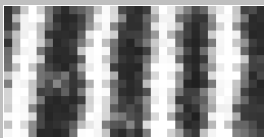
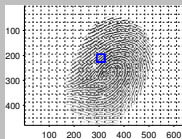
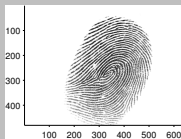


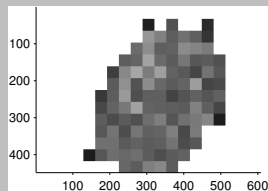
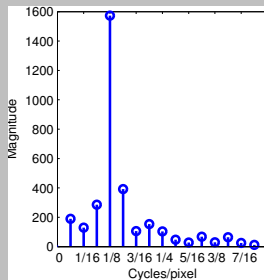
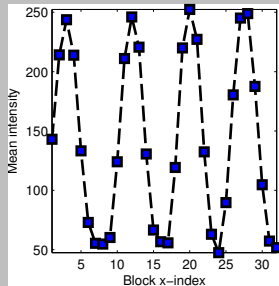
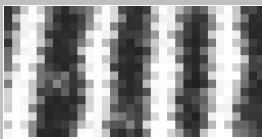
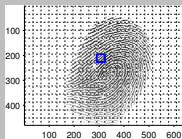
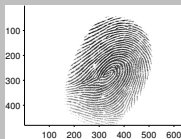


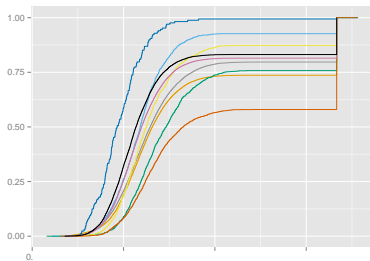


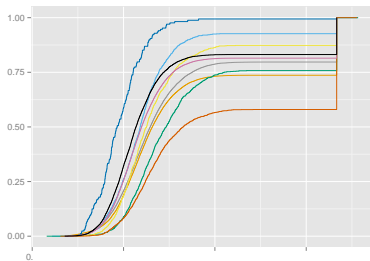








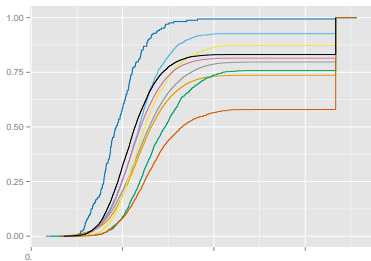




$$B_{FDA} = \{ -\infty, 0.26800, 0.30400, 0.33000, 0.35500, \\ 0.38000, 0.40700, 0.44000, 0.50000, 1.00000, \infty \}.$$



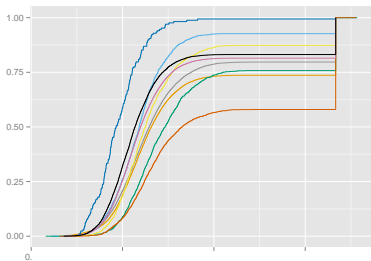
Cumulative density



$$B_{FDA} = \{ -\infty, 0.26800, 0.30400, 0.33000, 0.35500, \\ 0.38000, 0.40700, 0.44000, 0.50000, 1.00000, \infty \} .$$

- ▶ Local quality values \Rightarrow fixed length feature vector
- ▶ Mean, std.dev., 10 bin histogram \Rightarrow 12-dimension feature vector

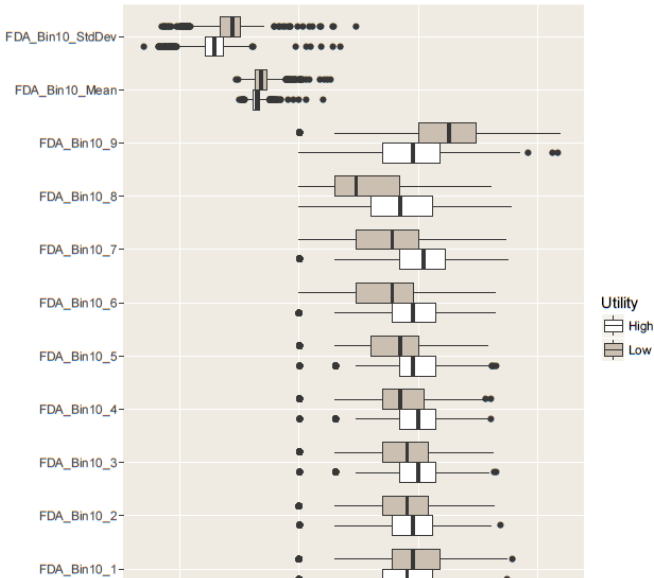
Feature value



$$B_{\text{FDA}} = \{ -\infty, 0.26800, 0.30400, 0.33000, 0.35500, \\ 0.38000, 0.40700, 0.44000, 0.50000, 1.00000, \infty \}.$$

- ▶ Local quality values \Rightarrow fixed length feature vector
- ▶ Mean, std.dev., 10 bin histogram \Rightarrow 12-dimension feature vector

$$Q_{\text{FDA}} : Q_{\text{FDA}}^{\mu}, Q_{\text{FDA}}^{\sigma}, Q_{\text{FDA}}^1, \dots, Q_{\text{FDA}}^{10}$$



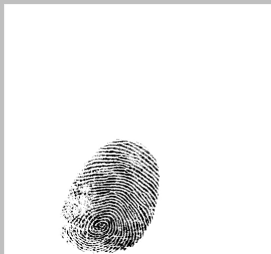


- ▶ Request \Rightarrow near frame rate quality assessment (10 Hz)





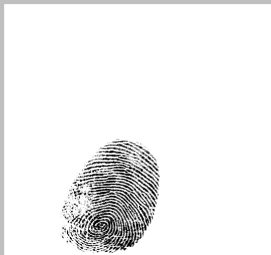
- ▶ Request \Rightarrow near frame rate quality assessment (10 Hz)
- ▶ Slap sensors provide large finger images



800 \times 750 pixel sensor output reproduced at 25% scale



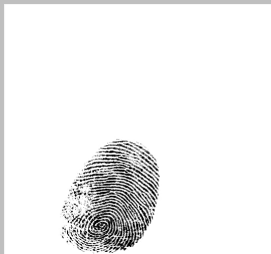
- ▶ Request \Rightarrow near frame rate quality assessment (10 Hz)
- ▶ Slap sensors provide large finger images
- ▶ Removal of near constant area
- ▶ No processing of background area blocks



800 \times 750 pixel sensor output reproduced at 25% scale



- ▶ Request \Rightarrow near frame rate quality assessment (10 Hz)
- ▶ Slap sensors provide large finger images
- ▶ Removal of near constant area
- ▶ No processing of background area blocks
- ▶ Avoid removing low quality fingerprint areas



800 \times 750 pixel sensor output reproduced at 25% scale



$330 \times 286 = 94380$
(15.7%)



$$330 \times 286 = 94380 \\ (15.7\%)$$



$$330 \times 286 - (10 \times (32 \times 32)) = 84140 \\ (13.9\%)$$



- ▶ Demand for actionable feedback from quality algorithm
- ▶ More than a quality score - helps to answer the why
- ▶ Provide information \Rightarrow improve quality at recapture



- ▶ Demand for actionable feedback from quality algorithm
- ▶ More than a quality score - helps to answer the why
- ▶ Provide information \Rightarrow improve quality at recapture
- ▶ Finger image receives low quality score - why?





- ▶ Demand for actionable feedback from quality algorithm
- ▶ More than a quality score - helps to answer the why
- ▶ Provide information \Rightarrow improve quality at recapture
- ▶ Finger image receives low quality score - why?
 - ▶ Unintended interaction with sensor
 - ▶ Pre-processing error, e.g. segmentation
 - ▶ Sensor failure





- ▶ Demand for actionable feedback from quality algorithm
- ▶ More than a quality score - helps to answer the why
- ▶ Provide information \Rightarrow improve quality at recapture
- ▶ Finger image receives low quality score - why?
 - ▶ Unintended interaction with sensor
 - ▶ Pre-processing error, e.g. segmentation
 - ▶ Sensor failure
- ▶ NFIQ 2.0 research kit offers actionable feedback





- ▶ Demand for actionable feedback from quality algorithm
- ▶ More than a quality score - helps to answer the why
- ▶ Provide information \Rightarrow improve quality at recapture
- ▶ Finger image receives low quality score - why?
 - ▶ Unintended interaction with sensor
 - ▶ Pre-processing error, e.g. segmentation
 - ▶ Sensor failure
- ▶ NFIQ 2.0 research kit offers actionable feedback
 - ▶ empty image ($\mu > 250$)
 - ▶ uniform image pixel intensity ($\sigma = 1.0$)
 - ▶ no or few minutiae detected ($N_{min} < 5$)
 - ▶ small foreground area ($N_{fgnd} < 50000$)



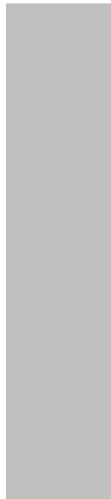


empty image
few minutiae

$$\mu > 250$$
$$N_{min} < 5$$

uniform image intensity
small foreground

$$\sigma = 1.0$$
$$N_{fgnd} < 50000$$



NFIQ 2.0 = 89

$\mu = 177$

$\sigma = 99$

$N_{min} = 60$

$N_{fgnd} = 117337$



empty image
few minutiae

$$\mu > 250$$
$$N_{min} < 5$$

uniform image intensity
small foreground

$$\sigma = 1.0$$
$$N_{fgnd} < 50000$$



NFIQ 2.0 = 21

$$\mu = 220$$

$$\sigma = 64$$

$$N_{min} = 40$$

$$\uparrow N_{fgnd} = 36887$$



empty image
few minutiae

$\mu > 250$
 $N_{min} < 5$

uniform image intensity
small foreground

$\sigma = 1.0$
 $N_{fgnd} < 50000$



NFIQ 2.0 = 1

$\mu = 196$

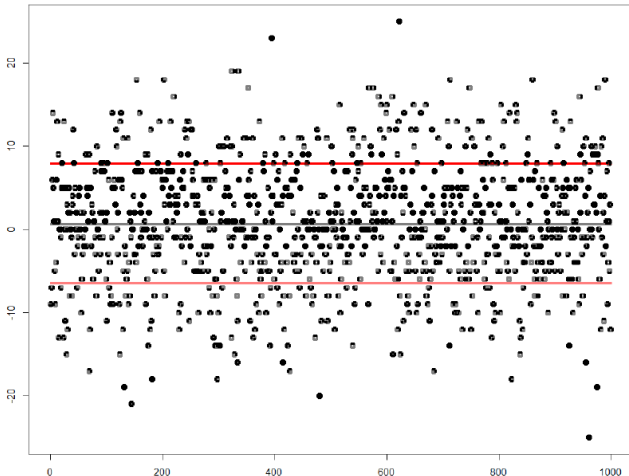
$\sigma = 79$

↑ $N_{min} = 0$

↑ $N_{fgnd} = 16262$

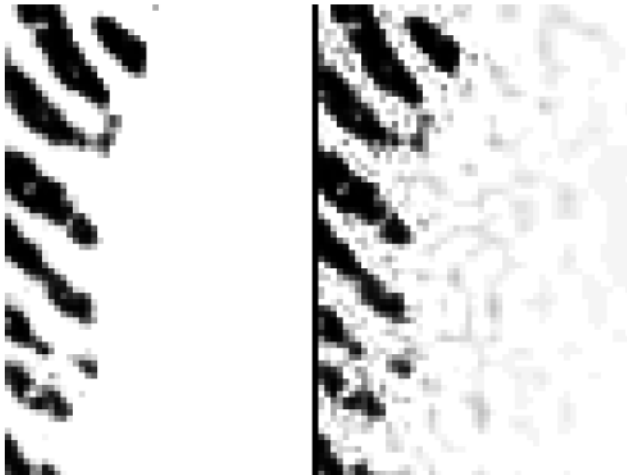


- ▶ Deviation between uncompressed and WSQ compressed (factor 8). 1000 images, MCYT 330 DP.





- ▶ Fingerprint boundary artifact at WSQ compression (factor 8). Gamma adjusted.





- ▶ Standardization of features a priority throughout NFIQ 2.0 development



- ▶ Standardization of features a priority throughout NFIQ 2.0 development
- ▶ 29794-4 – biometric sample quality – finger image data
 - ▶ current status is 3rd Committee Draft
 - ▶ progression to Draft International Standard in May 2016
 - ▶ projected release as International Standard in 2017



- ▶ Standardization of features a priority throughout NFIQ 2.0 development
- ▶ 29794-4 – biometric sample quality – finger image data
 - ▶ current status is 3rd Committee Draft
 - ▶ progression to Draft International Standard in May 2016
 - ▶ projected release as International Standard in 2017
- ▶ NFIQ 2.0 effectively a reference implementation of 29794-4 at this point
 - ▶ Open source, publicly available



Thanks for your attention

Martin A. Olsen

Contact: martin.olsen@{cased.de; ntnu.no}

NFIQ 2.0 nist.gov/itl/iad/ig/development_nfiq_2.cfm

Prototype quality features share.nbl.nislabs.no/public