

# **Thoughts on Fingerprint Image Quality and Its Evaluation**

**NIST Biometric Quality Workshop II  
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## ***Recap from NEC's Presentation at Previous Workshop (2006)***

- n Positioning quality: a key factor to guarantee common area and matching accuracy**
- n Pattern area: good positioning criteria**
- n Quality to predict accuracy: matcher dependent (algorithm dependent)**
- n NEC quality metrics: better accordance with NEC matchers than NFIQ**

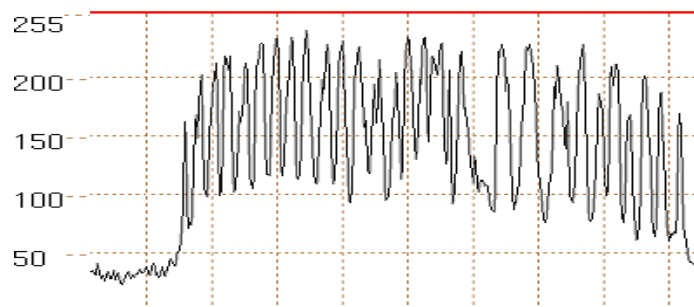
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*Appendix: Improvement on NEC's Predictive Quality Metrics*

# 1. Quality Concepts – Ideal Quality

## What is ideal quality?



### 1) good ridge quality

- dynamic range sufficiently wide
- uniformity evenly distributed density
- linearity gray mid scale reserved
- no saturation (white & black)
- no significant smudge or blur
- sufficient ridge/valley separation

### 2) no problematic background noise

- no leftover fingerprint or stripe pattern
- background lighter than ridge (foreground)

### 3) sufficient size

- excellent for slap/flat matching
- good for latent cognizant

### 4) good positioning and orientation

- pattern area included & fingertip up

### 5) no significant distortion

***Ideal quality Strong image enhancement NOT required***

# 1. Quality Concepts – Poor Quality Samples

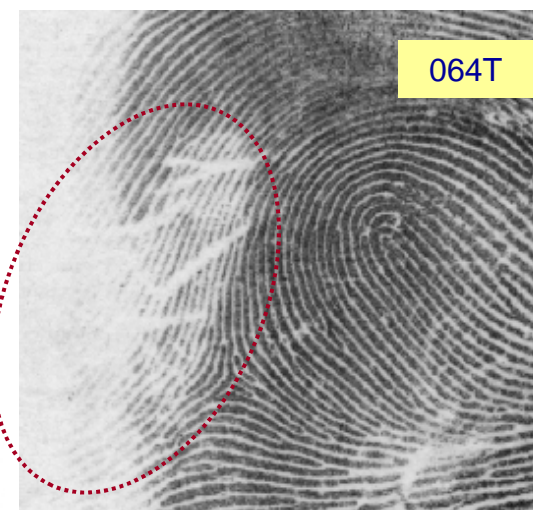
Factors related to ridge quality and background noise

- a) narrow dynamic range
- b) uneven density
- c) white saturation
- d) leftover fingerprint
- e) problematic stripe pattern

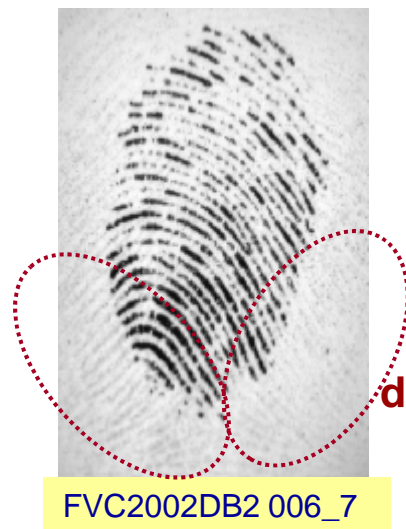
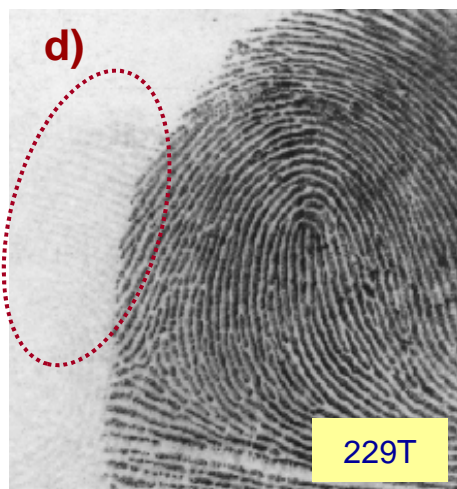
Note: Fingerprint samples are from NIST DB#27 and FVC Data Base.



a), b), c)



a), b)



## 2. Requirements for Quality

“Requirements for quality” depend on:

- target filed – **law enforcement (LE)** or **non LE (NLE)**
- operational requirement - **automatic** or **manual intervening**
- image type (flat, slap, rolled, latent)

two major operational categories:

### **a) fully automatic matching (NLE/LE)**

a1) Positive ID (cooperative & unsupervised - NLE)

a2) Negative ID (uncooperative & supervised - NLE)

a3) Automatic latent (rolled/latent - LE)

### **b) manual intervening operation (LE)**

b1) for latent cognizant rolled print

b2) for latent print

*Note: a1) is not discussed here.*



## 2. Requirements for Quality

### Criteria for quality

- a) criterion for acceptance/rejection at capture
- b) criterion for enrollment or registration
- c) criterion for special search (e.g. latent)
- d) criterion to predict “matching accuracy”

**1) real-time processing required for criterion a)**

**2) quality metrics specific to capture device is effective for real-time processing**

**e.g. positioning & orientation do NOT have to be checked for identification slap (4-slap) capturing**

***Note: Only offline processing for static image is discussed here.***

### 3. Pure Quality and Predictive Quality

***pure quality*** – intrinsic quality of image itself

*matcher independent*

*also Independent of operational needs*

***predictive quality*** - quality for predicting accuracy

*matcher dependent*

**a) predictive quality for auto (PQ\_A)**

*for automatic operation (NLE, automatic latent)*

***compatibility with human examiners NOT required***

***OK to use with non-minutia-based matching***

**b) predictive quality for manual (PQ\_M)**

*for manual intervening operation (LE, manual latent)*

***need to consider compatibility with human***

***examiner's minutia definition***



## 4. Image Enhancement - Quality Metrics

- *Question* -

*Image enhancement – OK for quality metrics?*

### 1) contrast enhancement

- contrast stretch (global & local)
- histogram equalization (global & local)
- sharpening, etc.

**strong tools to cope with “narrow dynamic range” and “uneven density”, etc.**

### 2) ridge enhancement

- filtering – contextual, Fourier, Gabor, Wavelet, etc.
  - (\* based on ridge direction & pitch)
- pore & incipient ridge removal, etc.

**useful to cope with insufficient ridge/valley separation**

## 4. Image Enhancement - Quality Metrics

- *Question* -

*Image enhancement - OK for quality metrics?*

***No, at least for pure quality!***

**image enhancement has side effects such as:**

- 1) increasing background noise
- 2) removing gray intermediate (mid) scale
- 3) creating false (ghost) ridges
- 4) removing true ridges
- 5) **creating false minutiae and missing true minutiae**

# 4. Image Enhancement – Contrast Enhance

Which image is of better quality?

A



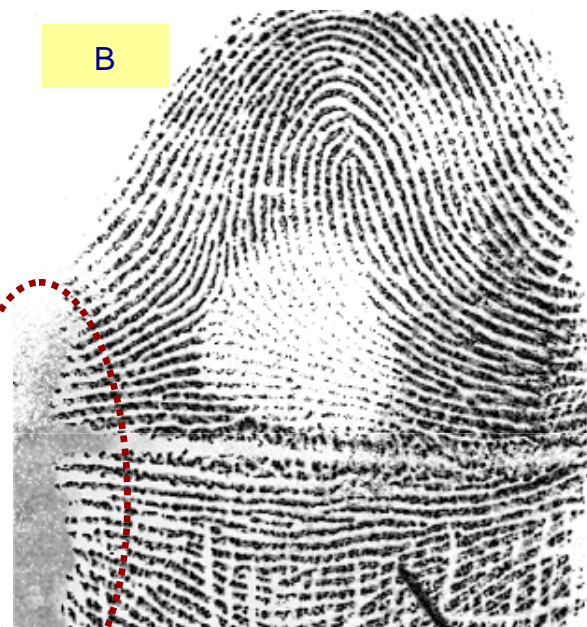
NFIQ=50; NEC=61

dynamic range  
**narrow < wide**

gray mid scale  
**equivalent**

background noise  
**less > more**

B



265T

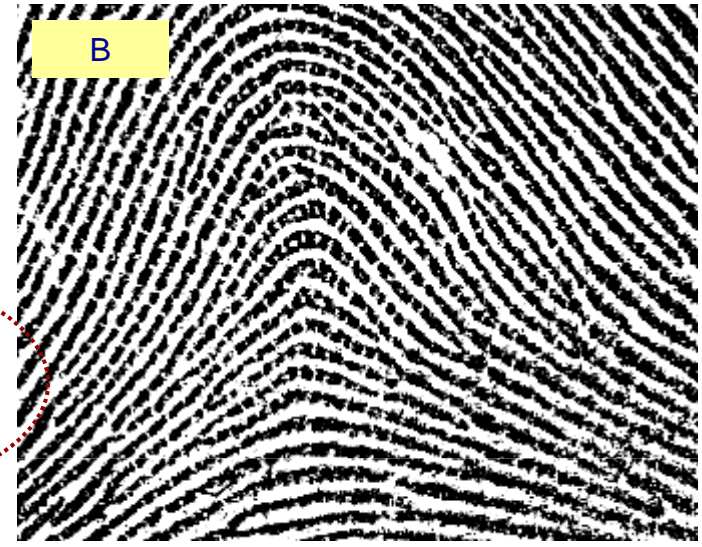
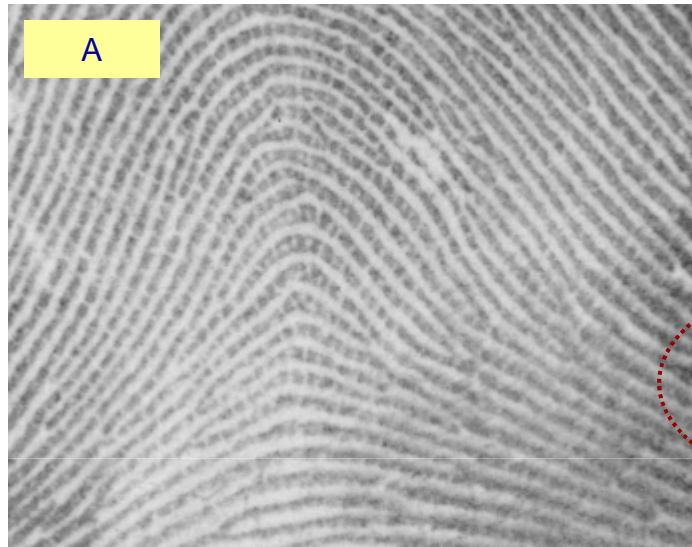
NFIQ=50; NEC=60

Image B: contrast-enhanced image of Image A using local contrast stretch

<p><b>Transformed NFIQ</b> employed for easier comparison</p>		poorer			better	
	NFIQ (original):	5	4	3	2	1
	<b>NFIQ (transformed):</b>	<b>0</b>	<b>25</b>	<b>50</b>	<b>75</b>	<b>100</b>
	NEC quality metrics:	0	-----			

# 4. Image Enhancement – Contrast Enhance

Which image is of better quality?



dynamic range  
**narrow < wide**

gray mid scale  
**some > no**

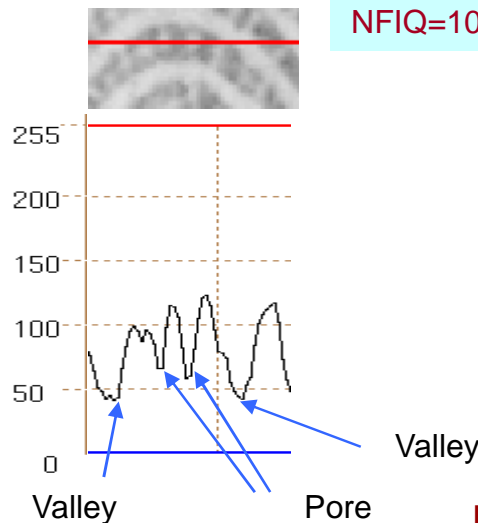
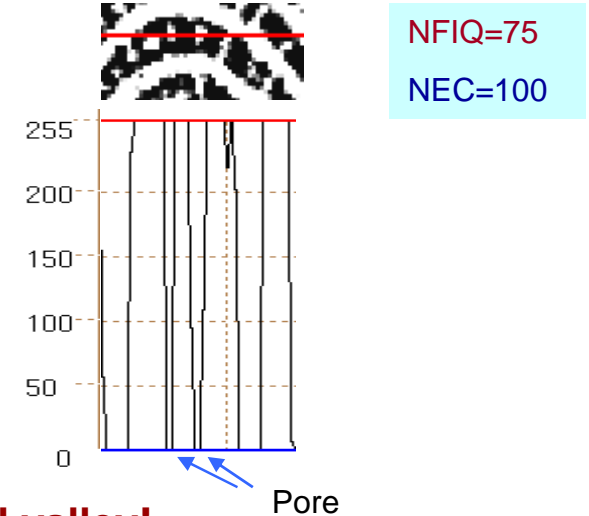


Image B: contrast-enhanced image of Image A using sharpening and contrast stretch



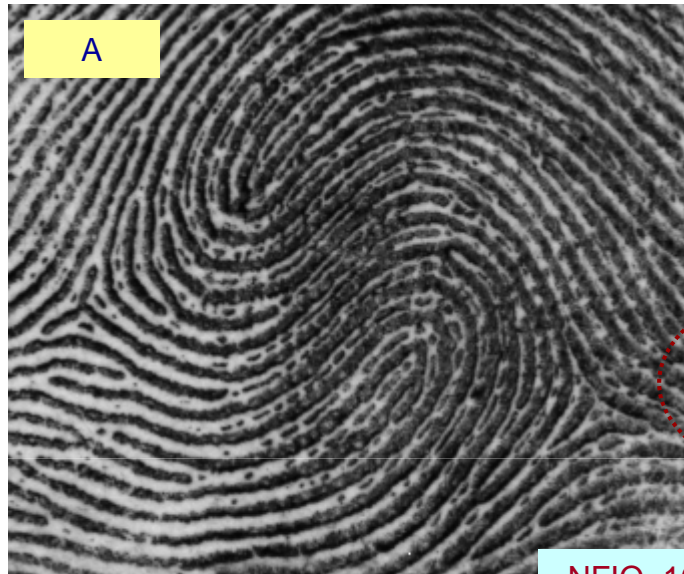
**no difference between pore and valley!**



# 4. Image Enhancement – Contrast Enhance

Which image is of better quality?

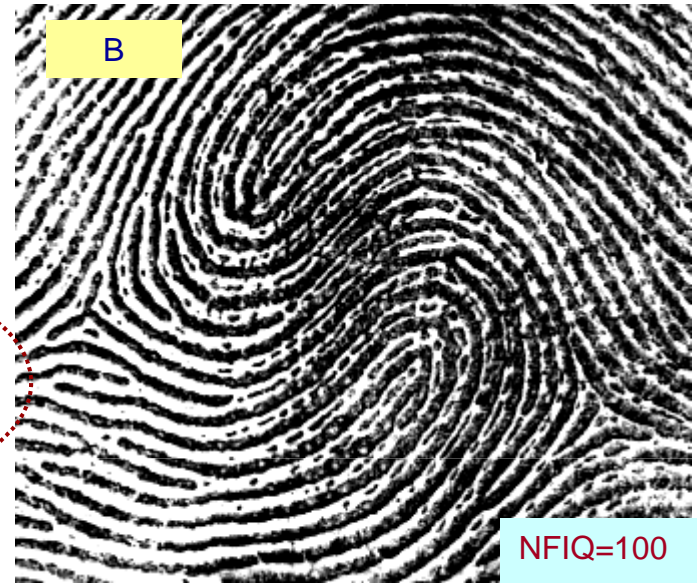
201T



NFIQ=100; NEC=95

dynamic range  
**narrow < wide**

gray mid scale  
**some > no**



NFIQ=100

NEC=100

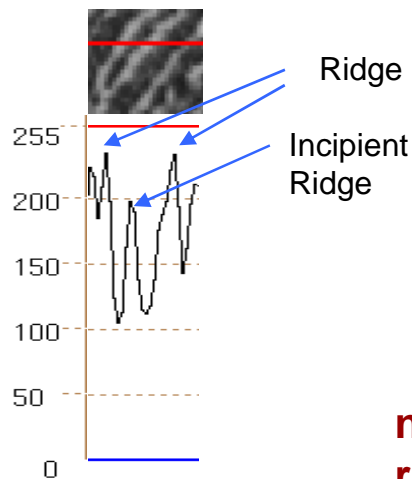
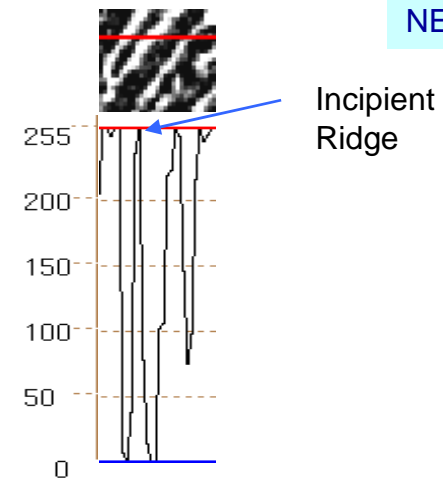


Image B: contrast-enhanced image of Image A using sharpening and contrast stretch

**no difference between incipient ridge and true ridge!**



# 4. Image Enhancement – Contrast Enhance

Which image is of better quality?

A



NFIQ=75; NEC=39

dynamic range  
**narrow** < **wide**

gray mid scale  
**some** > **little**

background noise  
**more** < **less**

white saturation  
**no** > **yes**

B



NFIQ=75; NEC=30

FVC02DB3 001\_3

Image B: contrast-enhanced image of Image A using *light density removing* and contrast stretch



## 4. Image Enhancement – Ridge Enhance

Which image is of better quality?

A



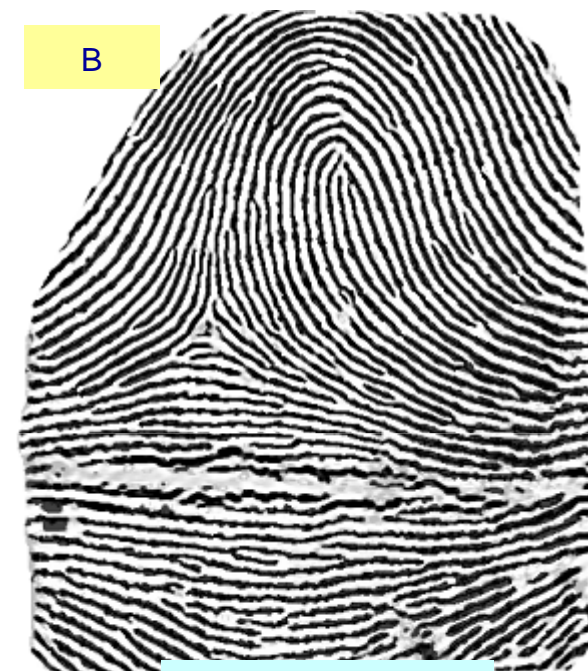
NFIQ=50; NEC=61

dynamic range  
**narrow < wide**

gray mid scale  
**yes > not reliable**

fidelity  
**Yes > No**

B



NFIQ=75; NEC=80

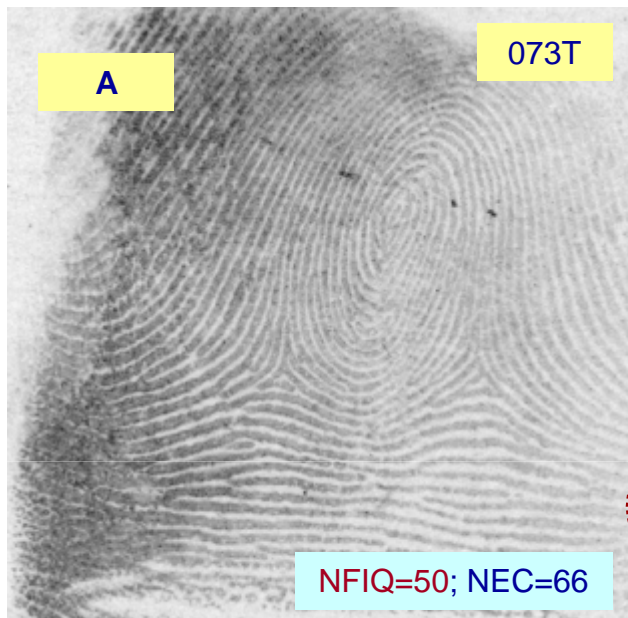
265T

**Image B: ridge-enhanced image of Image A using contextual filtering**

**Ridge-enhanced image does NOT ALWAYS represent the original image.**

# 4. Image Enhancement – Ridge Enhance

Which image is of better quality?

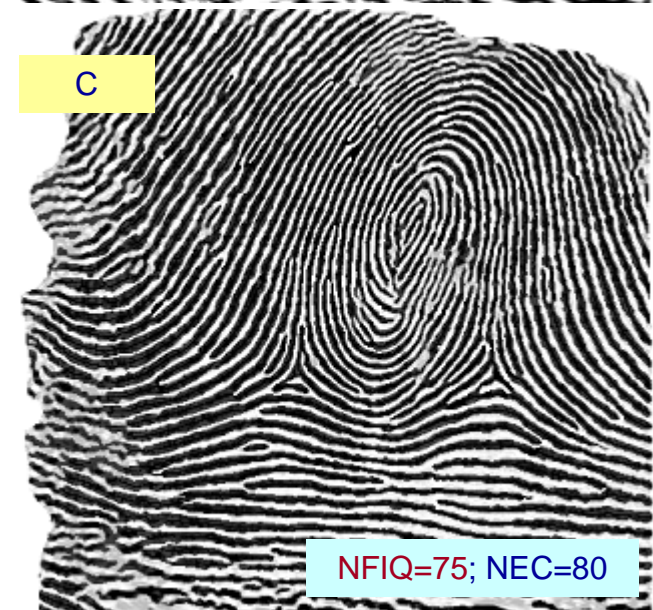
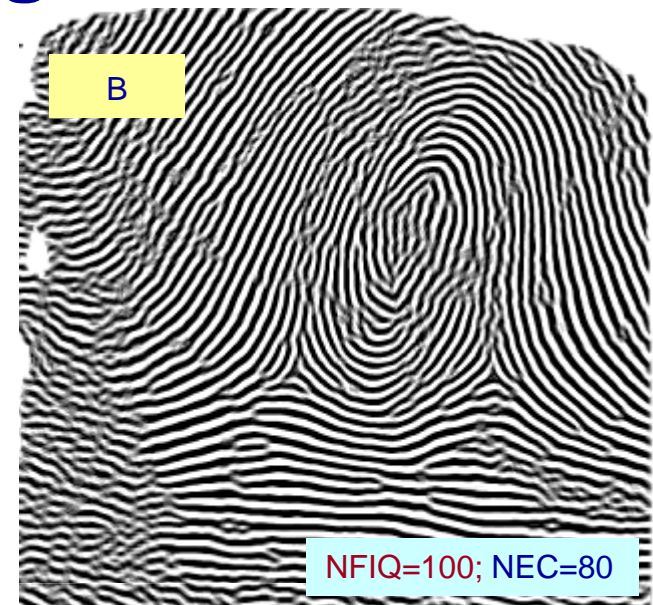


dynamic range  
**narrow < wide**

gray mid scale  
**yes > not reliable**

fidelity

**Yes > No**



**Image B:** ridge-enhanced image using **fixed-width pitch** data ( false ridges)

**Image C:** ridge enhanced image using **variable-width pitch** data (locally estimated)

**Ridge-enhanced image does NOT  
ALWATYS represent the original image.**

## 4. Image Enhancement – Predictive Quality

*Image enhancement - OK for predictive quality?*

### a) predictive quality for auto (PQ\_A)

***no problem to apply any image enhancements***

*reasonable to use equivalent method to matcher*

*(1) false minutia per examiners' definition be useful as long as such "feature" is consistent*

*(2) consistently miss-extracting true (but unstable) minutiae be more favorite than inconsistent extraction*

### b) predictive quality for manual (PQ\_M)

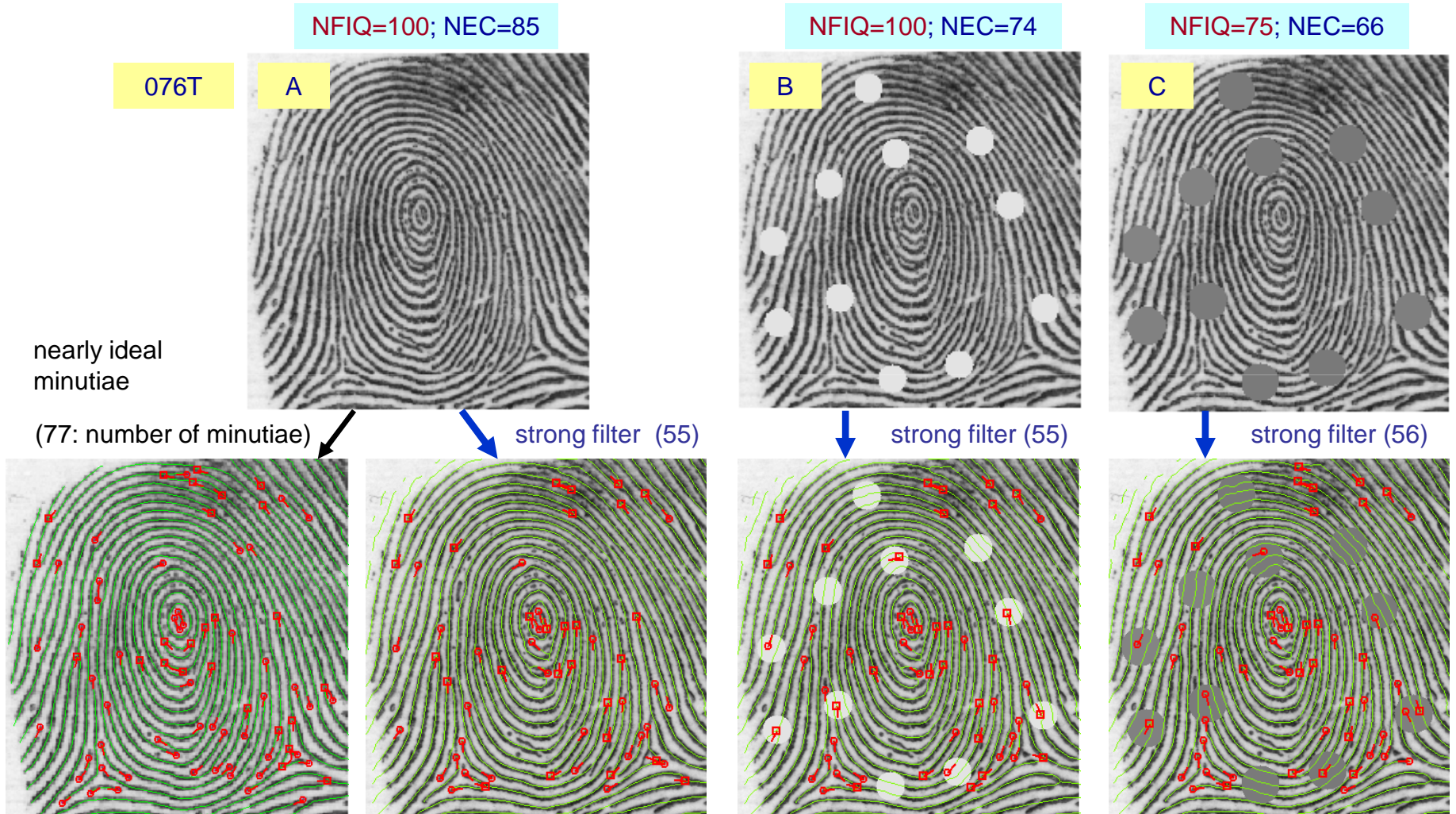
***limited use suggested***

*strong filtering in ridge enhancement tends to create false minutiae or to remove true minutiae*



# 4. Image Enhancement – Predictive Quality

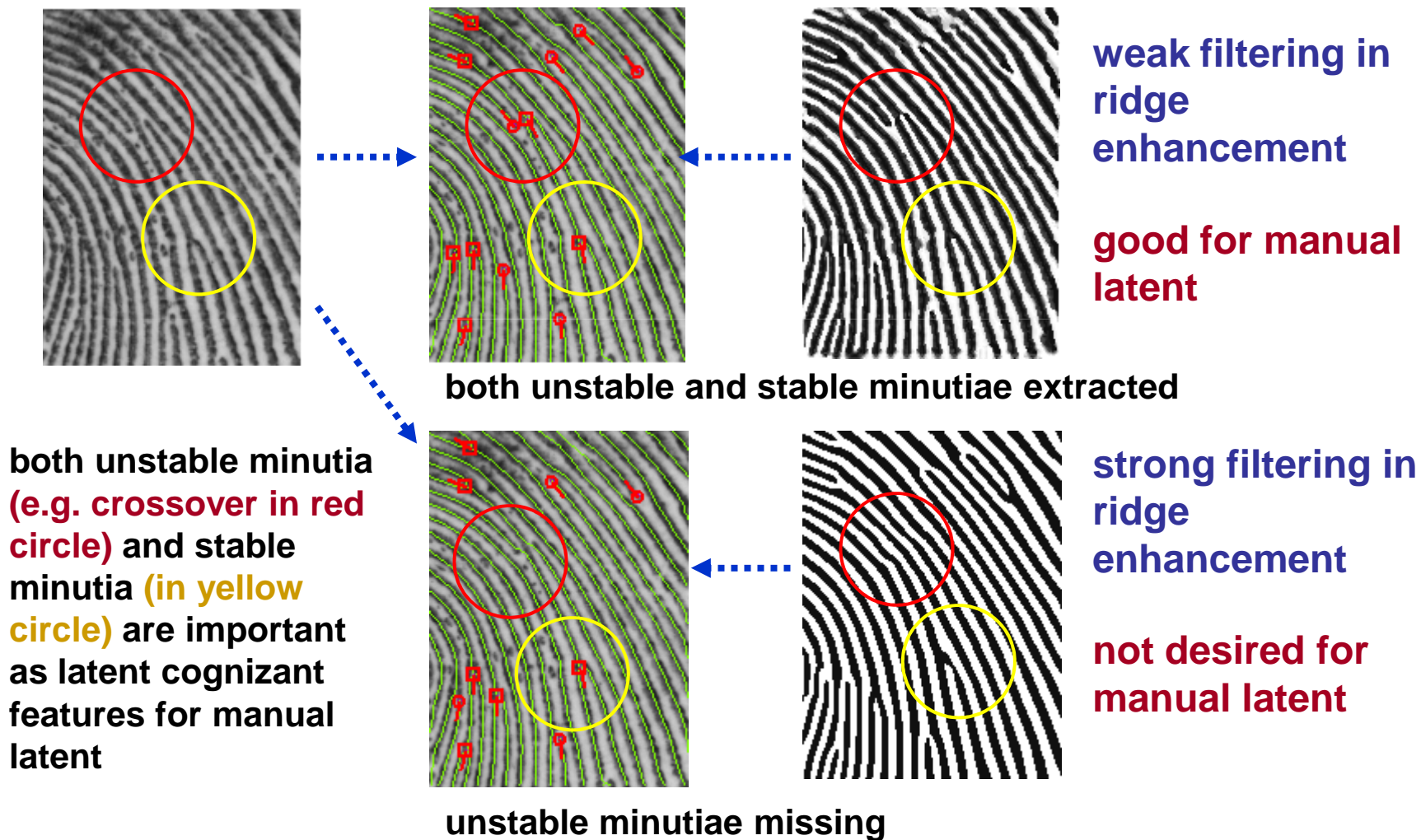
***PQ\_A: strong filtering - robust to low quality image***



***Ignoring unstable minutiae increases consistency of minutia on auto process***

# 4. Image Enhancement – Predictive Quality

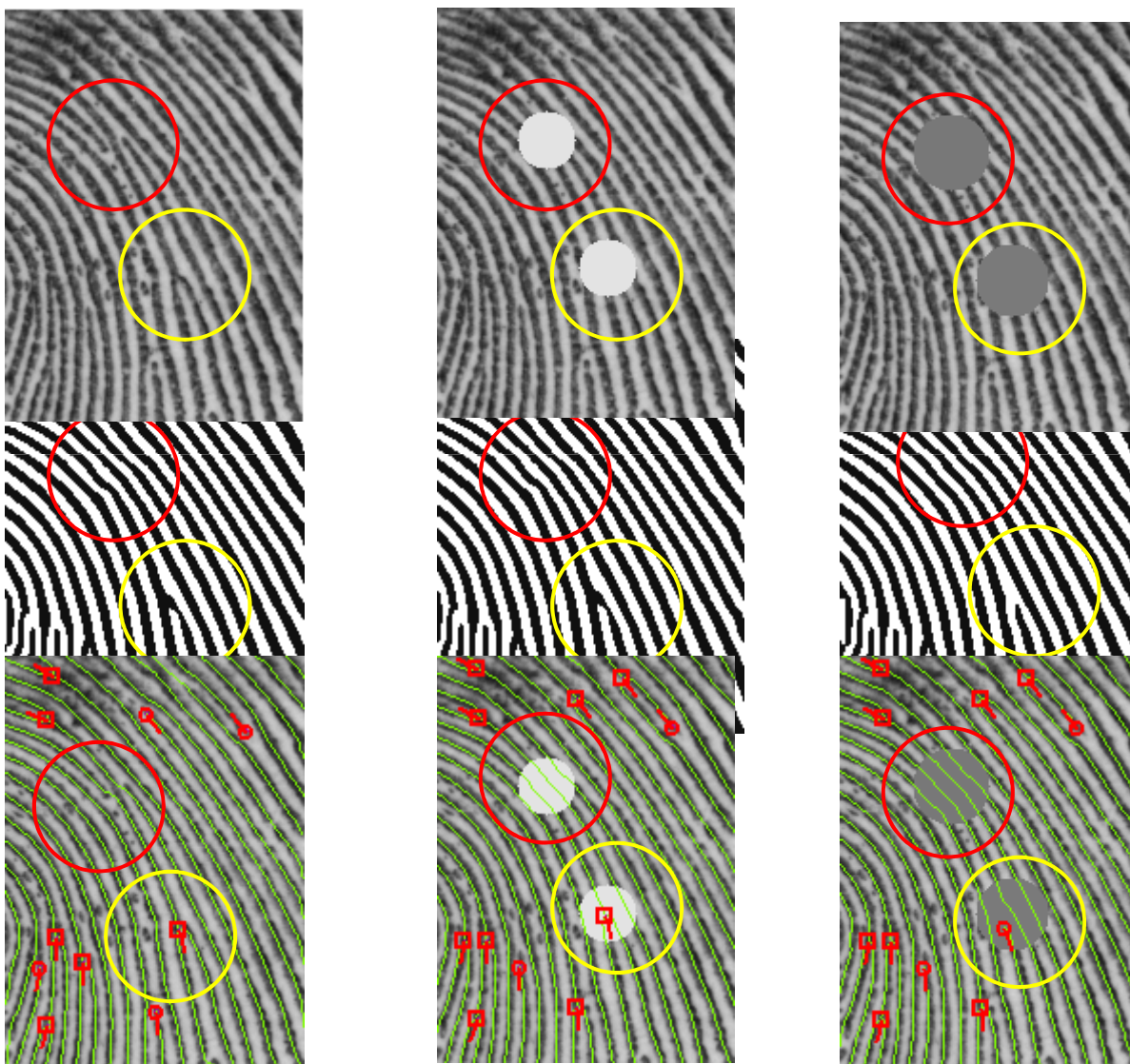
*PQ\_A: strong filtering - not desired for manual latent*





# 4. Image Enhancement – Predictive Quality

*PQ\_A: strong filtering – why robust?*



Strong filtering in ridge enhancement creates stripe patterns even though there is no real ridge information on the input image.

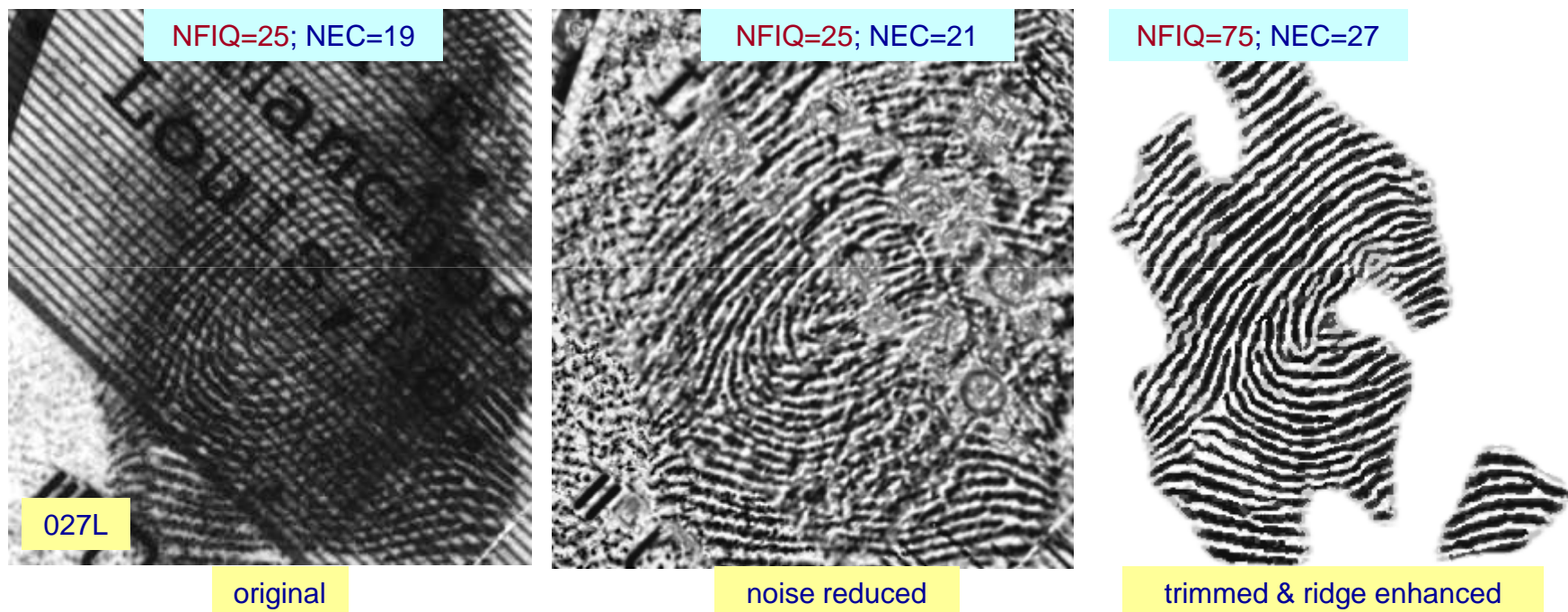
This process also tends to create pseudo ridges (maybe false ridges) or to remove true ridges.



## 4. Image Enhancement – Latent

pure quality for latent print    not practical to be evaluated  
predictive quality for manual latent    dependent on human examiner

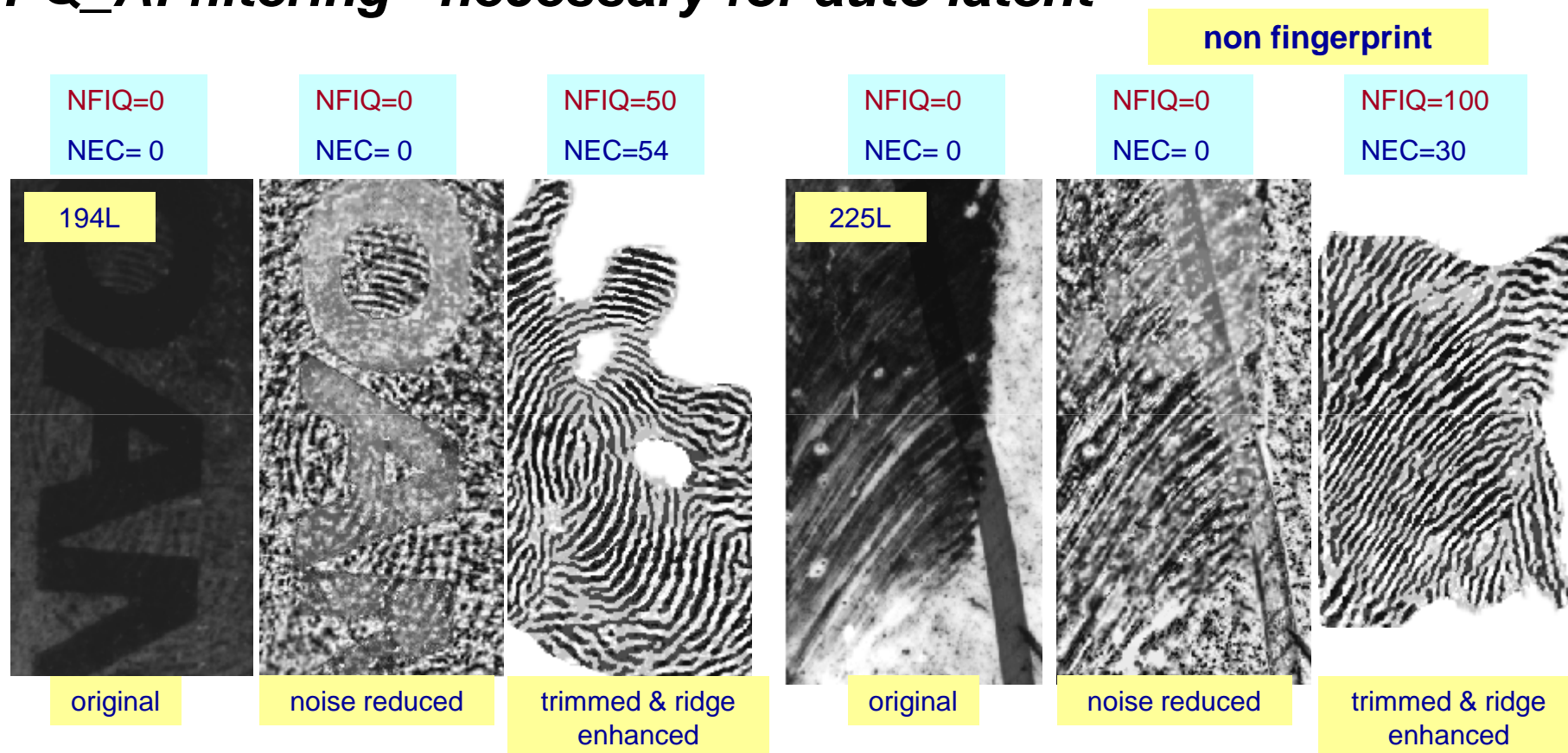
***PQ\_A: noise reduction - necessary for auto latent***



**Some latent prints have severe background noise. Quality metrics as well as matching expectation for those latent prints depend on *noise reduction performance***

# 4. Image Enhancement – Latent

*PQ\_A: filtering - necessary for auto latent*



**filtering**

**robust to noise**

**filtering – tends to create false ridge even from non fingerprint pattern such as smudge**

# 5. Fingerprint Properness Analysis

Fingerprint properness and of matching feature sufficiency, etc. are important for predictive quality metrics.

## 1) fingerprint pattern?

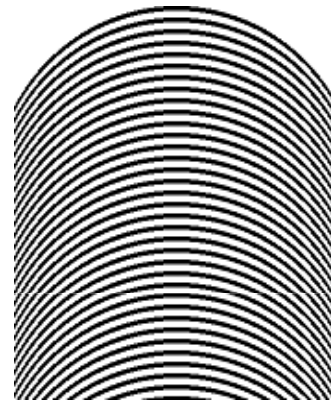


non fingerprint

NFIQ=75

NEC=30

## 2) matching feature sufficient?



no minutia pattern

NFIQ=75

NEC= 0

## 3) rotated?

Accepting rotated images will unnecessary **increase matching cost** and unnecessary **decrease matching accuracy**

NFIQ=100

NEC=33



NFIQ=100

NEC=85





# 5. Fingerprint Properness Analysis

## 4) *distortion?*

FVC2004\_DB1 028\_3



NFIQ=100  
NEC=80

FVC2004\_DB1 028\_4



NFIQ=100  
NEC=69

***difficult to evaluate from static image!***

**<effective countermeasures>**

- 1) identification slap (4-slap) to prevent “intentional” distortion**
- 2) matcher algorithm improvement but, increase in cost involved**

Note: As for positioning and size issues, please see our presentation at 2006 Workshop

# 6. Factors to Degrade Quality

		Possible Causes	Effective Countermeasures	Recapture Useful?
<b>Factors to Degrade Quality</b>	1) poor ridge quality	device performance: 40% capture operation: 30% device problem: 20% nature of skin: 10%	better device supervision periodical replacement nothing	to some extent
	2) background noise	device performance: 50% capture operation: 50%	better device maintenance (clean up)	to some extent
	3) insufficient size	device performance: 20% capture operation: 80%	large area capture device 4-slap (3/2-slap) capture supervision	yes
	4) Improper positioning or orientation	device performance: 20% capture operation: 80%	4-slap (3/2-slap) capture large area capture device supervision	yes
	5) distortion	device performance: 50% capture operation: 50%	4-slap (3/2-slap) capture supervision	limited

# 6. Factors to Degrade Quality

## effective countermeasures

### 1) better device most effective countermeasure

#### identification slap capture (4-slap)

- to consistently capture pattern area
- to solve rotation problem (fingertip up)
- to avoid distortion
- **to avoid wrong finger capture**



Ref. T. Hopper; Identification Flats (NIST Fingerprint Standard Workshop)

*(\*) 3 or 2-slap capturing is also effective*



### 2) better supervision and capture operation

- to reduce background noise (by platen clean up, etc.)
- to capture sufficient size, etc.



# 7. Quality Metrics Evaluation

evaluation method - different per type of quality metrics

- pure quality metrics evaluation

*no straightforward method*     **not good for contest**

*recommendation*

***specific criteria be evaluated by specific algorithms***

- predictive quality metrics evaluation

*matcher dependent*      **tied up with matcher**

*recommendation for contest*

***RRG(99.9): Rejection Rate to Guarantee 99.9%  
quality and matcher integrated evaluation***

# 7. Quality Metrics Evaluation

## pure quality metrics evaluation

FVC2002DB3 22\_6

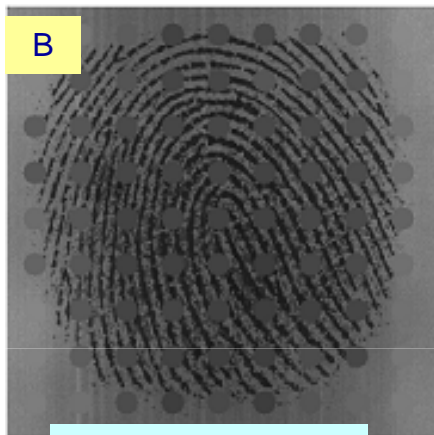
FVC2002DB3 22\_2

A



NFIQ=75; NEC=28

B



NFIQ=50; NEC=26

*Which image is of better quality?*

A) good ridge quality but fingertip only

B) poor ridge quality but pattern area exits

*difficult to define overall quality rating!*



*Can you rank these images?*

# 7. Quality Metrics Evaluation

## pure quality metrics evaluation

*specific criteria be evaluated by specific algorithms*

- 1) ridge quality    specific check tool  
*simple method suggested    NIST (or public domain) open source*
- 2) background noise    specific check tool  
*simple method suggested    NIST (or public domain) open source*
- 3) size    specific check tool  
*simple method suggested    NIST (or public domain) open source*
- 4) positioning and orientation    specific check tool  
*good candidate for contest*  
*reference (correct) data needed (manual coded) for contest*
- 5) distortion    difficult to check  
*difficult to evaluate from static image*

# 7. Quality Metrics Evaluation

## predictive quality metrics evaluation

*1) PQ\_A evaluation is relatively Simple.  
This is discussed here.*

*2) However, PQ\_M evaluation is difficult.  
Method for pure quality evaluation is also practical for  
PQ\_M evaluation.*



# 7. Quality Metrics Evaluation

recommended criteria for PQ\_A metrics evaluation

## **RRG(X): Rejection Rate to Guarantee X% Accuracy**

**Given:**

- 1) A set of fingerprint images*
- 2) Its accuracy is less than X% (e.g. 99.9%)*

**Question:**

*How much proportion of the poorest quality images need to be rejected in order to guarantee X% (e.g. 99.9%) accuracy?*

RRG(X) – straightforward criteria to evaluate the predictive quality (PQ\_A) metrics and matching performance at the same time

# 7. Quality Metrics Evaluation

## RRG(99.9) - recommended evaluation method

### 1) contestants provide three programs

- a) a quality program to produce quality metrics (e.g. Q: 0-100)
- b) a feature extraction program to produce templates
- c) a matching program to produce score (e.g. 0-9999)

### 2) NIST conducts test at the NIST facility

- a) produces quality metrics for search and file ( $Q_{\text{search}}$ ,  $Q_{\text{file}}$ )
- b) TAR (or first rank hit) considered for simplicity
- c) determine RRG(99.9) as follows

reject a mate if ( $Q_{\text{search}} < Q_{\text{th}}$ ) or ( $Q_{\text{file}} < Q_{\text{th}}$ );  $Q_{\text{th}} : Q_{\text{threshold}}$

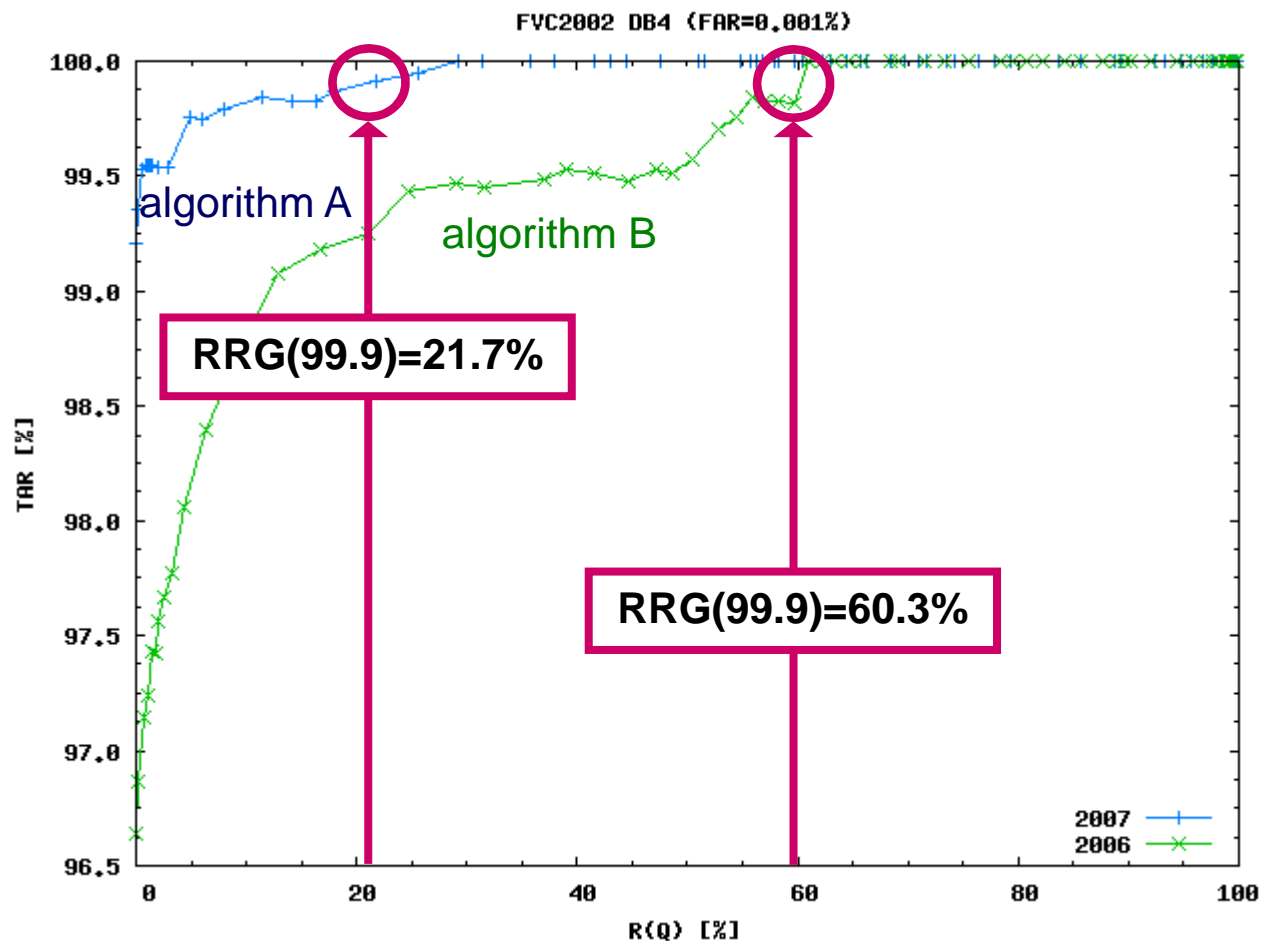
calculate  $R(Q_{\text{th}})$  - percentage the rejection (a function of Q)

find RRG(99.9) – minimum R to achieve 99.9% accuracy

*Note: 99.9% (instead of 100%) is recommended as target accuracy in order to avoid undesired side effect from the exceptional data*

# 7. Quality Metrics Evaluation

## Sample of RRG(X): R(Q) vs. TAR



straightforward method to compare different algorithms

# 8. Conclusion and Suggestion

## 1) image enhancement for pure quality

- shall NOT be used or limited to moderate contrast enhancement

## 2) image enhancement for predictive quality

- for PQ\_A (automatic) no restriction
- for PQ\_M (manual) limited use suggested

## 3) evaluation for pure quality

- specific algorithm be developed by NIST (or public domain)
- not appropriate for contest

## 4) evaluation for predictive quality (PQ\_A)

- RRG(99.9) criteria suggested
- appropriate for contest with proprietary matcher

## 5) practical solution for negative identification system

- identification slap capture (4-slap capture)



# Appendix: Improvement on NEC's Predictive Quality Metrics (1/13)

## NEC quality metrics      PQ\_A

- n Rated on a 0-100 scale, where 0 is the lowest quality and 100 is the highest quality
- n **Nonlinear combination of four independent indices**
  - n ridge quality with its area size
  - n high-confidence minutiae count
  - n positioning quality for common area
  - n distortion tolerance

Note: Appendix is prepared by Amane Yoshida from his research.

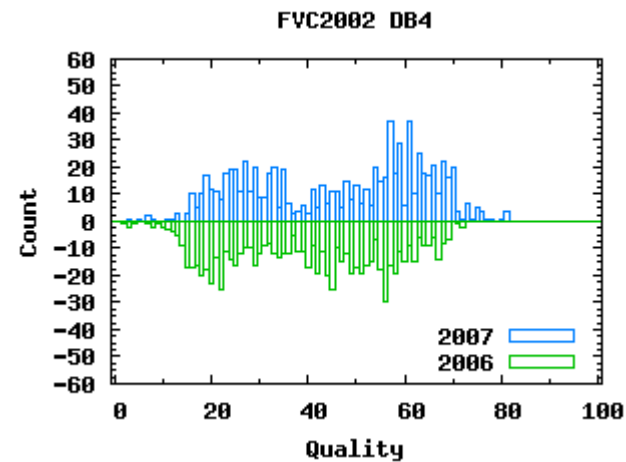
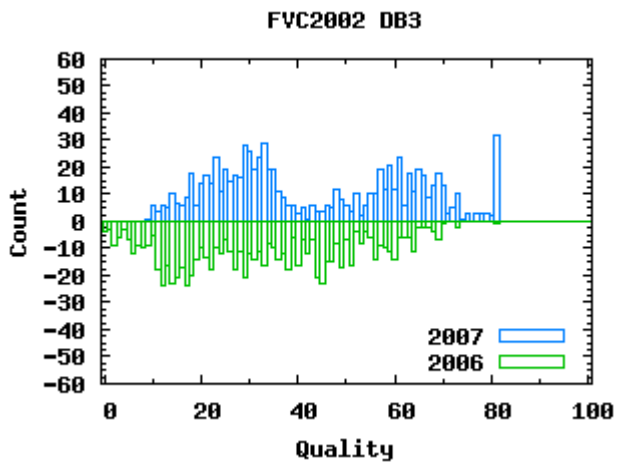
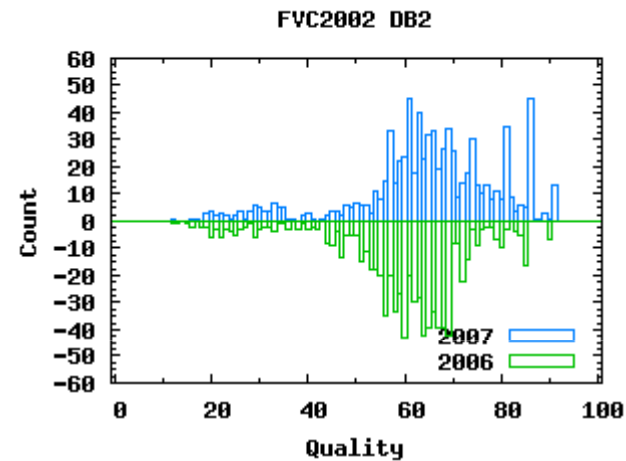
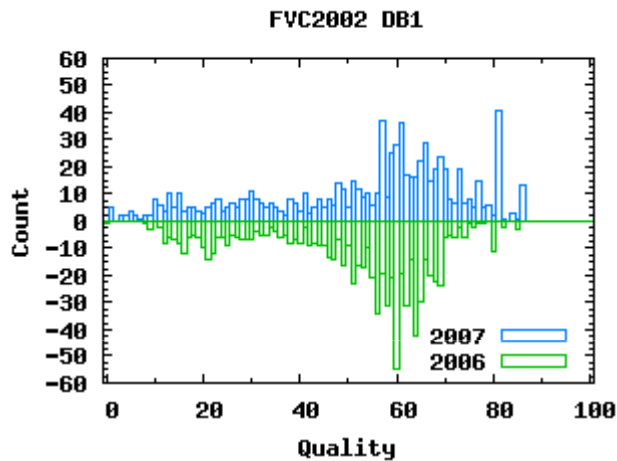
# Appendix: FVC2002 (2/13)

## n Accuracy Improvement

	Speed		TAR at FAR=0.01% [%]			
	Match	FE	DB1	DB2	DB3	DB4
<b>SDK H4 (2007)</b>	<b>H-equiv.</b>	<b>Slow</b>	<b>99.73</b> <b>(+0.09)</b>	<b>99.93</b> <b>(+0.18)</b>	<b>99.07</b> <b>(+0.69)</b>	<b>99.80</b> <b>(+1.09)</b>
SDK H3 (2006)	H-equiv.	Slow	99.64	99.75	98.38	98.71
SDK H2	H-equiv.	H-equiv.	99.45	99.79	95.18	97.38
SDK H	See NISTIR7151		99.02	99.68	92.13	96.36

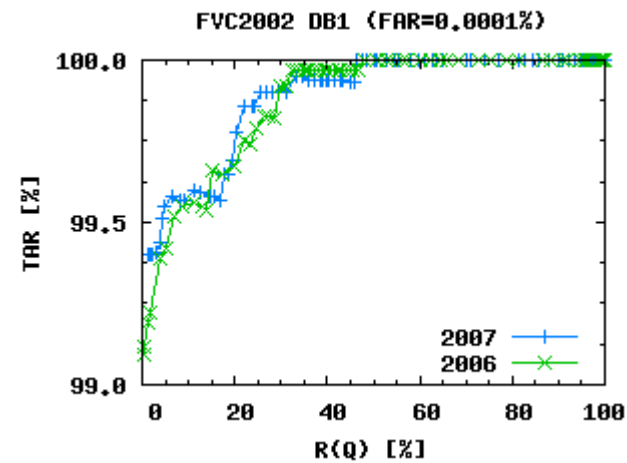
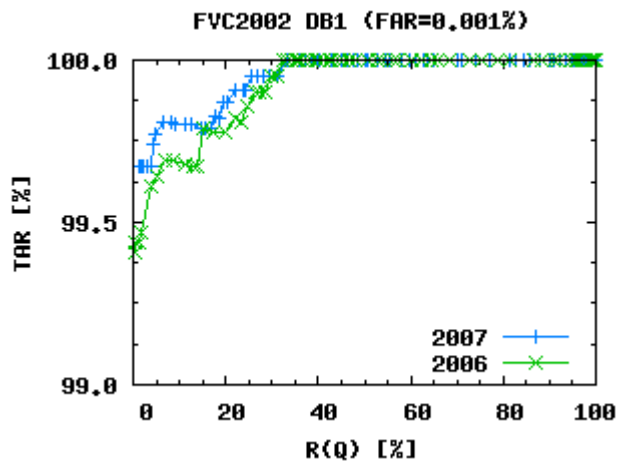
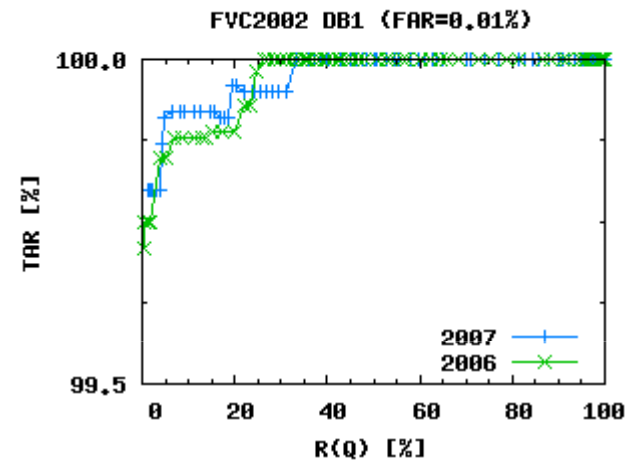
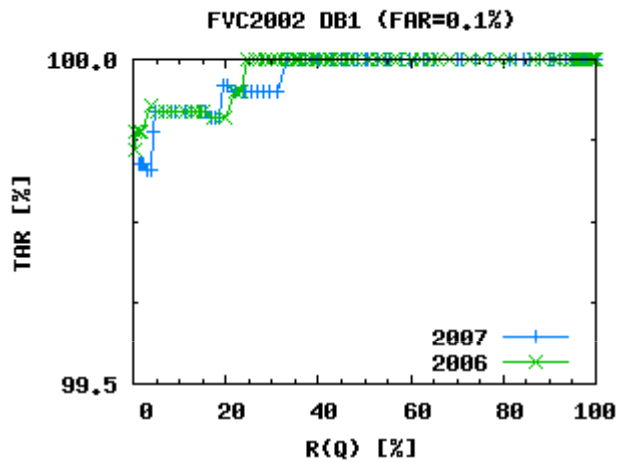
# Appendix: FVC2002 (3/13)

## n Quality Distributions



# Appendix: FVC2002 (4/13)

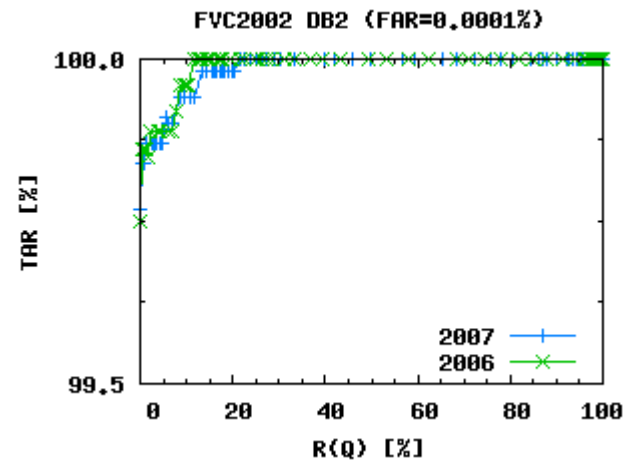
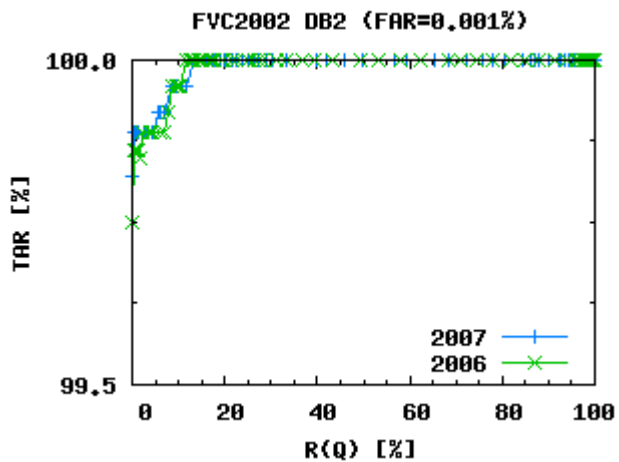
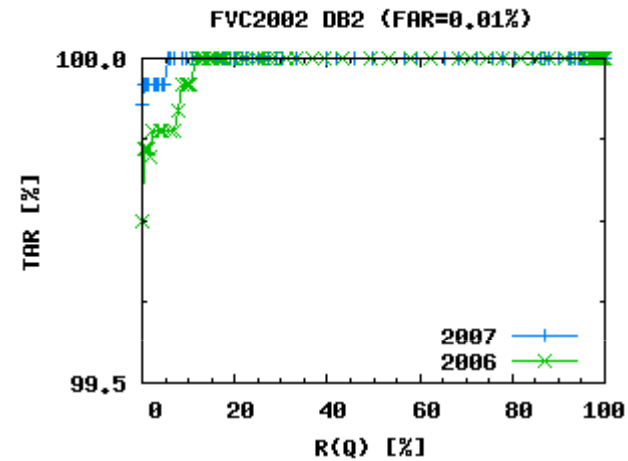
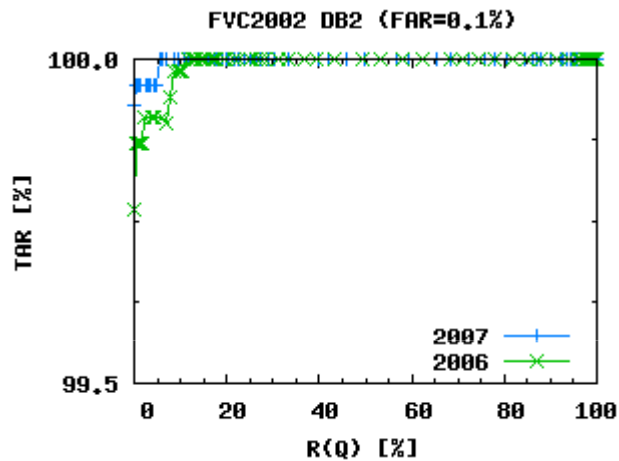
## n DB1: RRG comparison over varying FAR





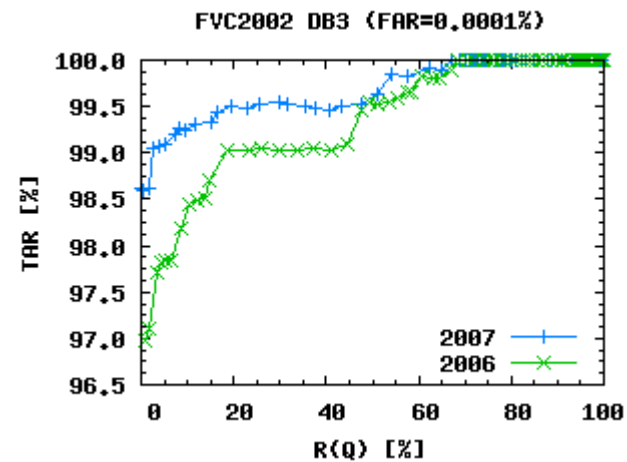
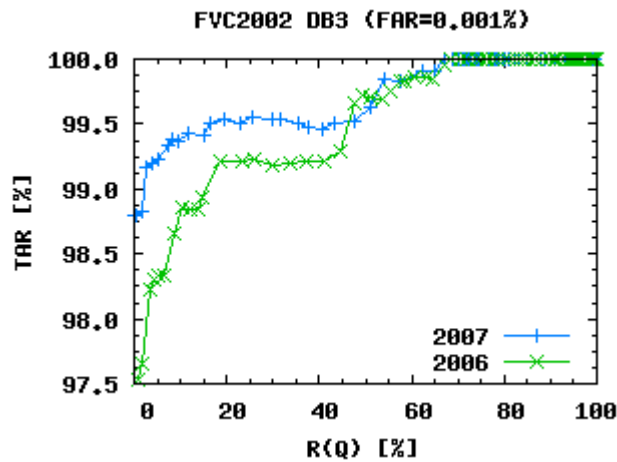
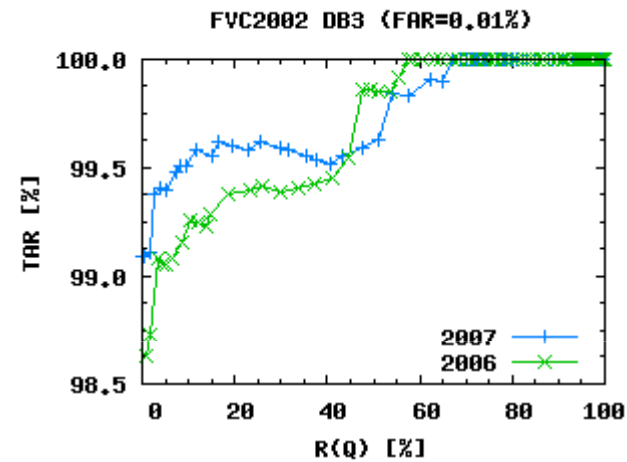
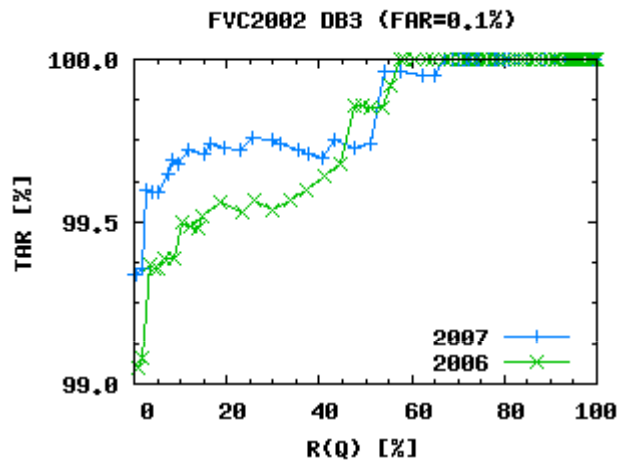
# Appendix: FVC2002 (5/13)

## n DB2: RRG comparison over varying FAR



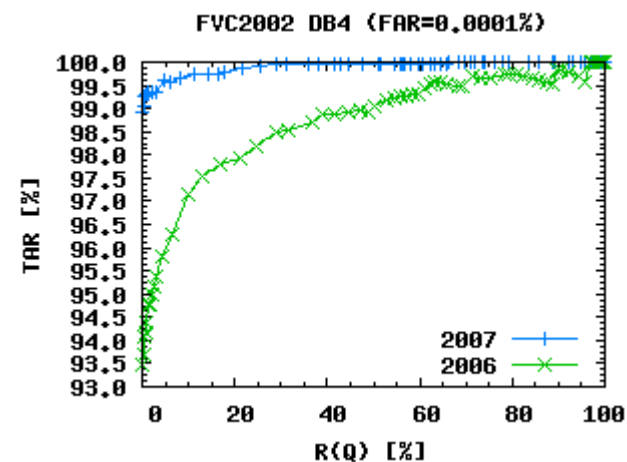
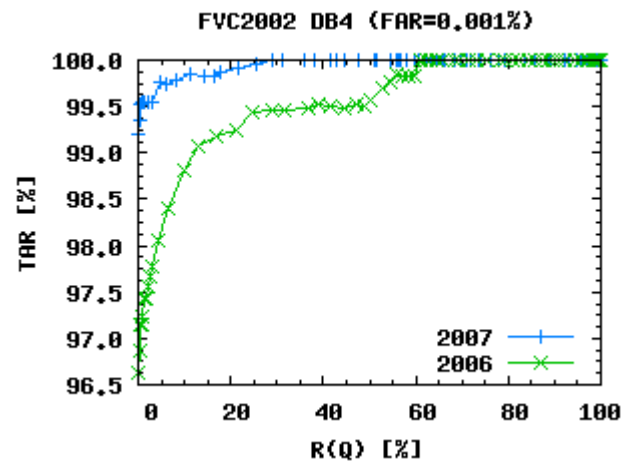
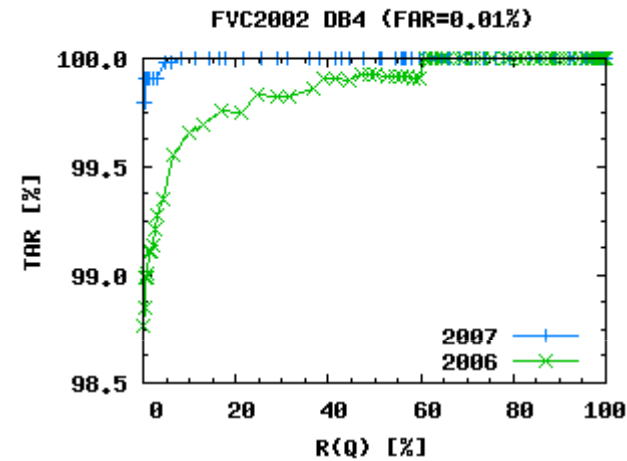
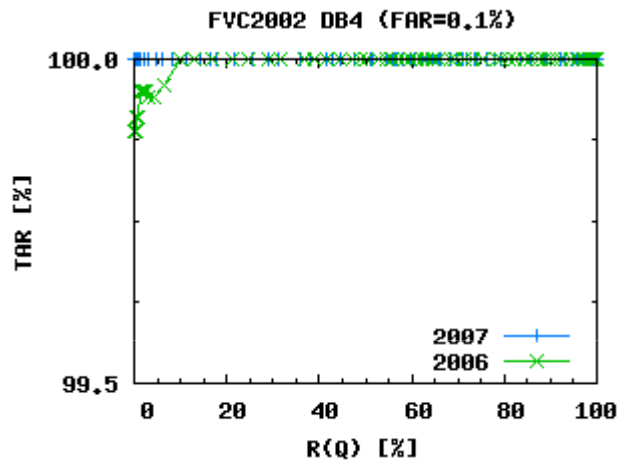
# Appendix: FVC2002 (6/13)

## n DB3: RRG comparison over varying FAR



# Appendix: FVC2002 (7/13)

## n DB4: RRG comparison over varying FAR



# Appendix: FVC2004 (8/13)

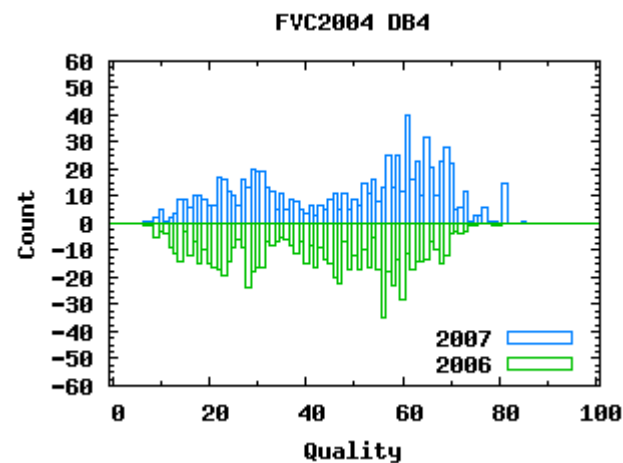
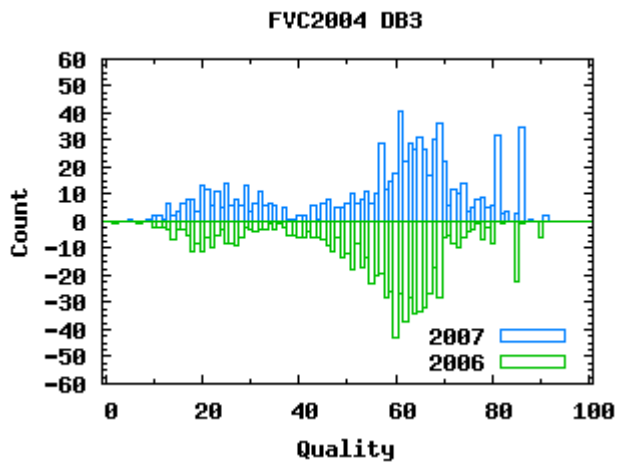
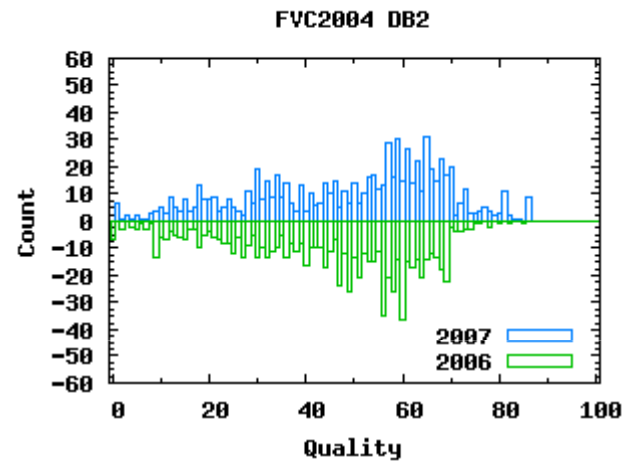
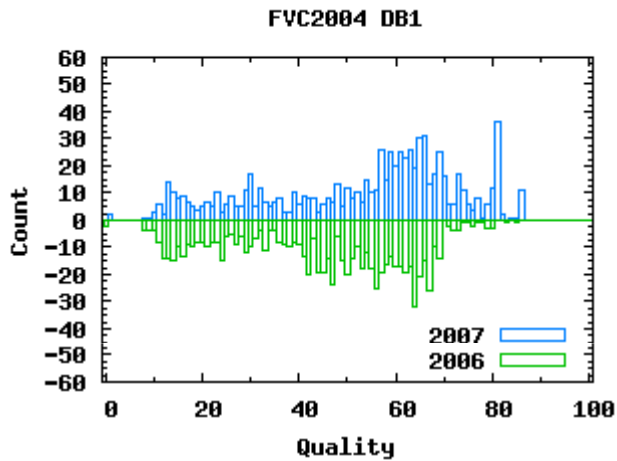
## n Accuracy Improvement

	Speed		TAR at FAR=0.01% [%]			
	Match	FE	DB1	DB2	DB3	DB4
<b>SDK H4 (2007)</b>	<b>H-equiv.</b>	<b>Slow</b>	<b>96.68</b> <b>(+0.93)</b>	<b>97.13</b> <b>(+0.58)</b>	<b>99.14</b> <b>(+0.07)</b>	<b>99.46</b> <b>(+0.69)</b>
SDK H3 (2006)	H-equiv.	Slow	95.75	96.55	99.07	98.77
SDK H2	H-equiv.	H-equiv.	95.66	95.09	98.70	97.96
SDK H	See NISTIR7151		93.63	94.88	97.79	97.02



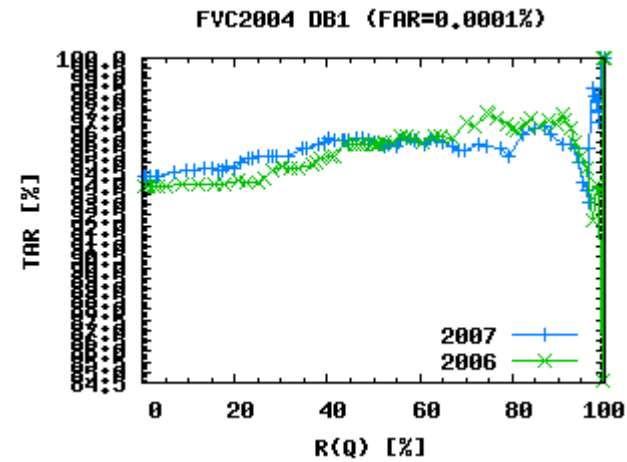
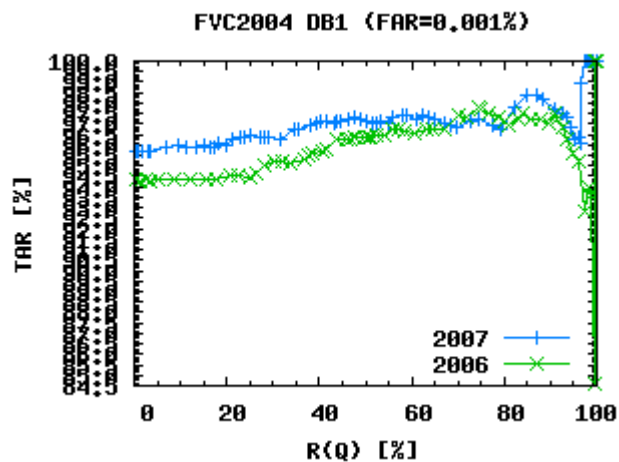
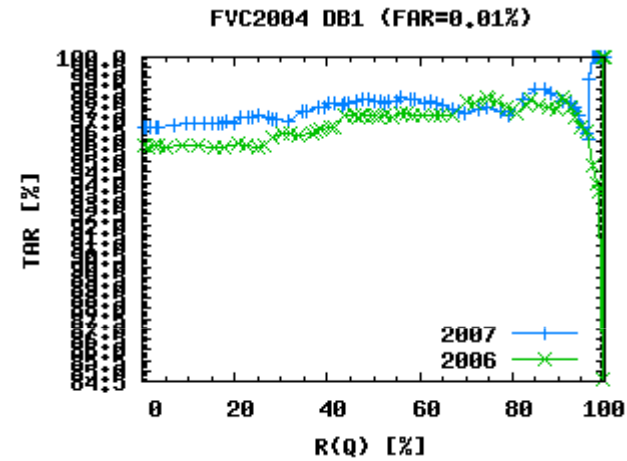
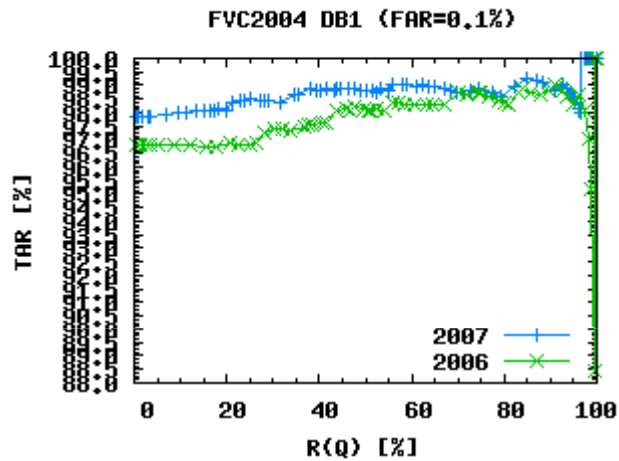
# Appendix: FVC2004 (9/13)

## n Quality Distributions



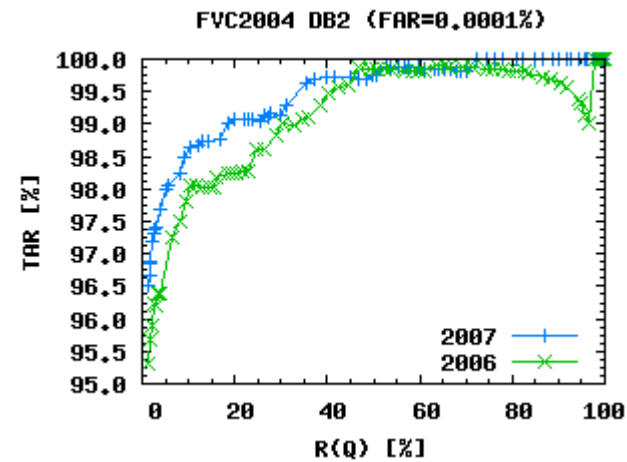
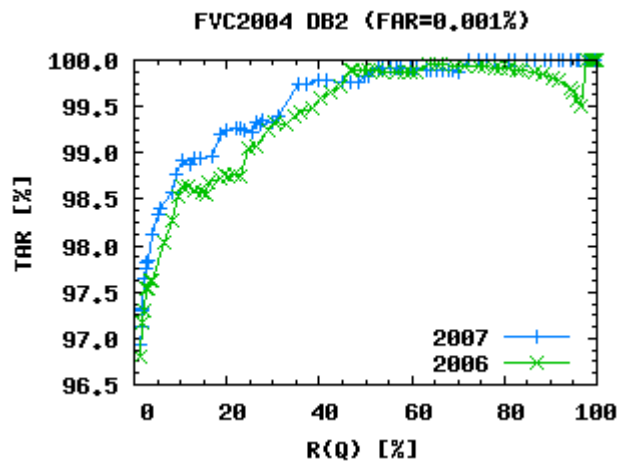
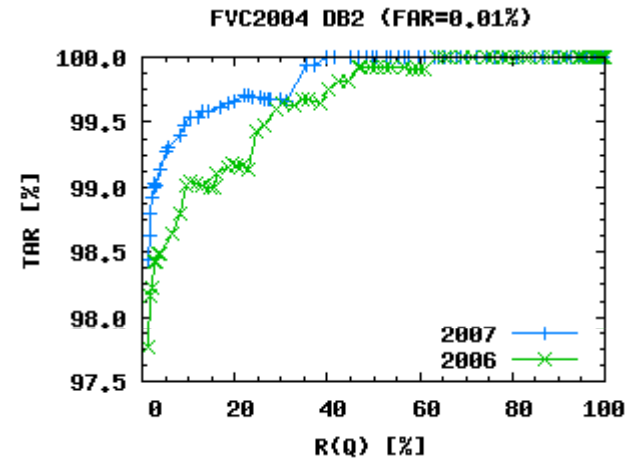
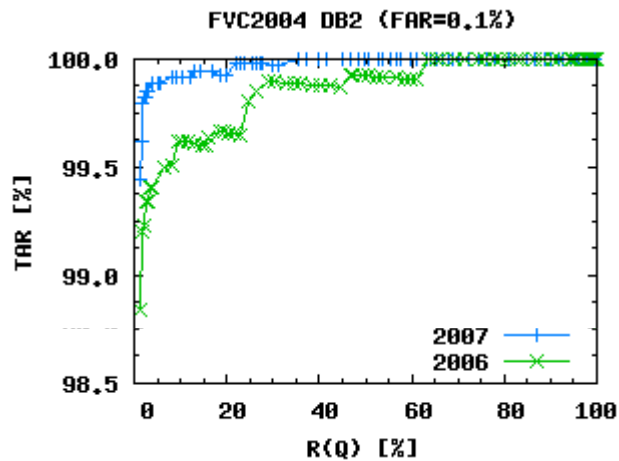
# Appendix: FVC2004 (10/13)

## n DB1: RRG comparison over varying FAR



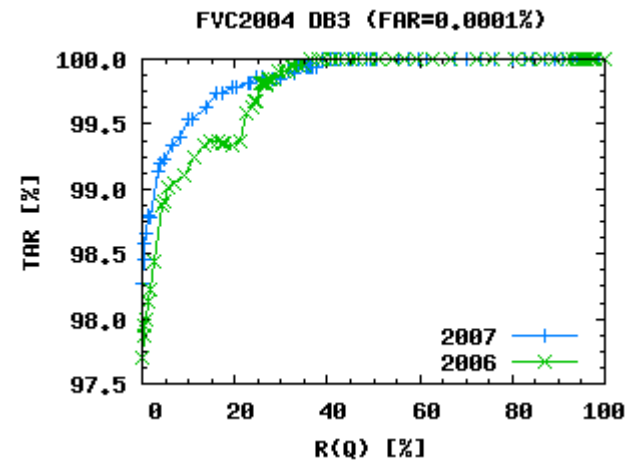
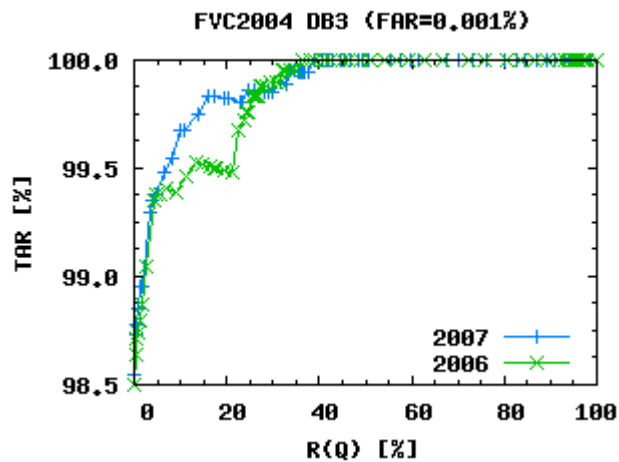
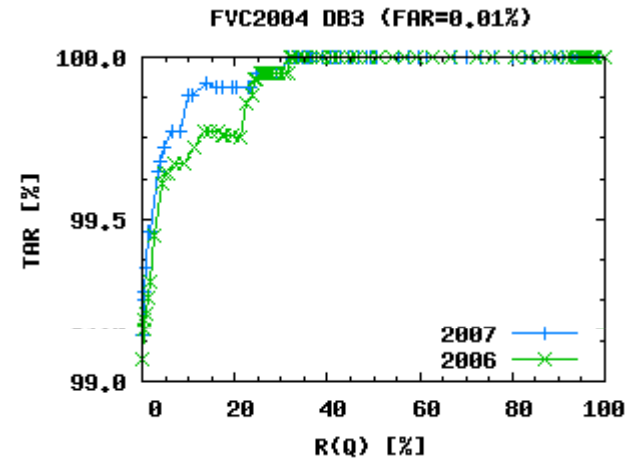
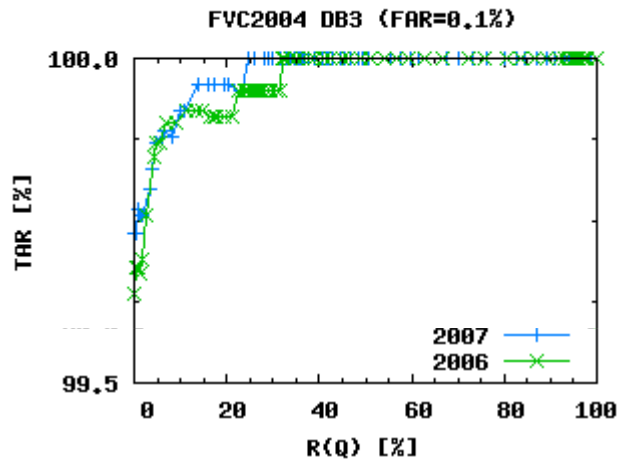
# Appendix: FVC2004 (11/13)

## n DB2: RRG comparison over varying FAR



# Appendix: FVC2004 (12/13)

## n DB3: RRG comparison over varying FAR



# Appendix: FVC2004 (13/13)

## n DB4: RRG comparison over varying FAR

