

# **Images can be regenerated from quantized biometric match score data**

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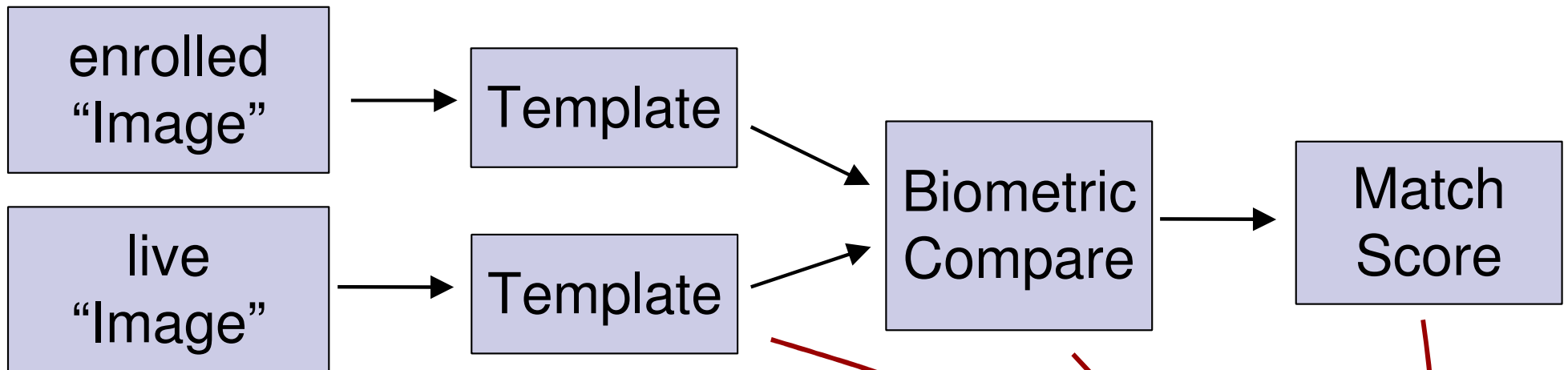
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# Problem: Biometrics security

- *Biometric authentication:*  
identification of individuals using behavioural and/or physiological characteristics:
  - Fingerprint, iris image, face recognition, gait, ...
- Applications:
  - Identity cards and systems (ie. border control)
  - Authentication for login / security
  - Time and attendance
- Biometric systems vulnerabilities?
  - Obviously, they can be exposed to all traditional cryptographic attacks
  - Are they vulnerable to *image based* attacks

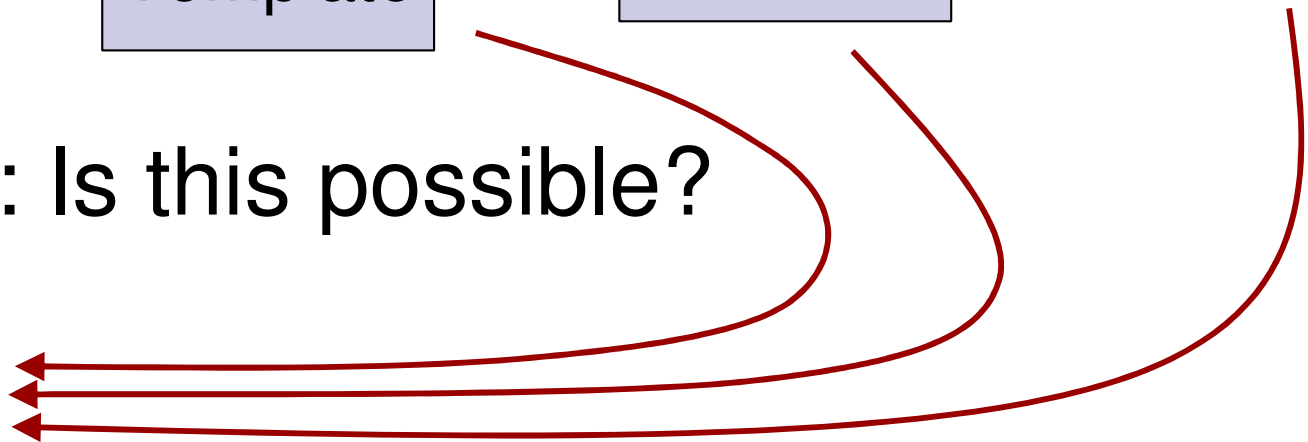
# Images can be **regenerated** ...?

## ■ Typical Biometric processing



## ■ *Question:* Is this possible?

**regenerated**  
"Image"



# Traditional wisdom

Most biometric vendors have claimed its impossible or infeasible to recreate the enrolled image.

Reasons:

- templates record features (such as fingerprint minutiae) and not image primitives
- templates are typically calculated using only a small portion of the image
- templates are much smaller than the image
- proprietary nature of the storage format makes templates infeasible to "hack".

# Automatic image *regeneration*

*Question:* is it possible to have generic software to regenerate images from biometric templates?

*Answer:* Yes

*Hill-climbing:* begin with a guess, make small modifications; keep modifications which increase the match score

*Requirement:* access to a biometric server which allows comparison of images to the target

# “Hill-climbing” Algorithm

## ***Preprocessing:***

- *Obtain Local Database (LD) of face images:*  
Images are rotated, scaled, cropped
- *Eigenface decomposition of LD:*  
 $i$ th eigenimage is represented by  $EFi$ .
- *Initial image selection ( $IM_0$ ):*

# “Hill-climbing” Algorithm

***Iterative estimate improvement:*** (for  $i \dots$ )

- Randomly select eigenimage:  $EF_k$
- Iterate for a range of values  $c_j$ :

$MS_j = \text{biometric\_compare}( IM_k + c_j \times EF_k, IM_{targ} )$











- $j_{max} = j$  for which  $MS_j$  is maximum
- $IM_{i+1} = IM_i + c_{j,max} \times EF_k$
- Truncate  $IM_{i+1}$  to image limits (ie. 0 to 255)

# Results

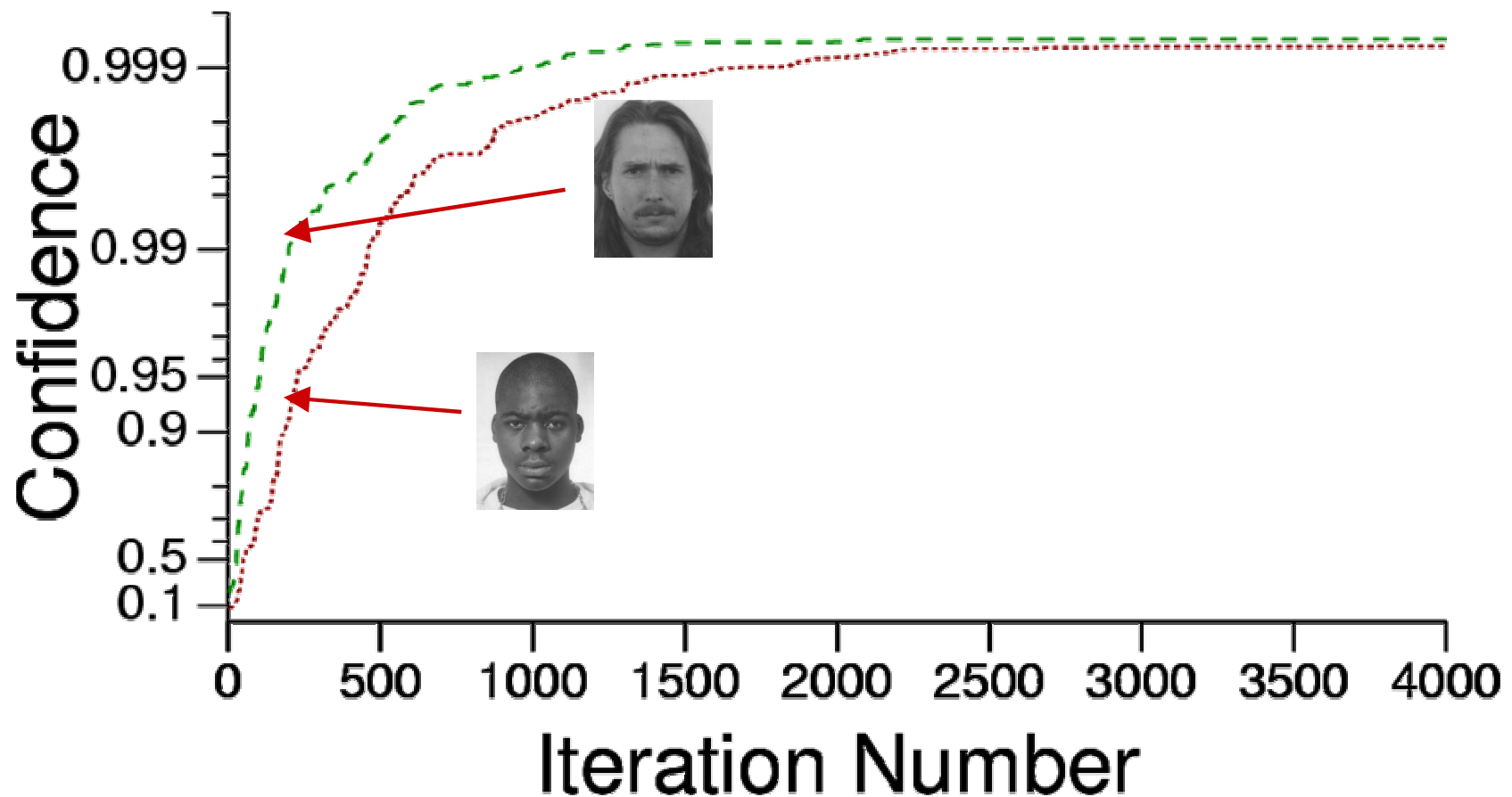
- Tests were performed against three different face recognition algorithms
  - All are recent products by well known commercial vendors of biometric systems.
  - Two of the vendors participated in the 2002 face recognition vendor test
- For all images and all biometric algorithms, the regenerated image compared at over 99.9% confidence



# Results

	Initial Image	Iteration 200	Iteration 600	Iteration 4000	Target Image
A					
B					

# Results: Confidence vs. iteration

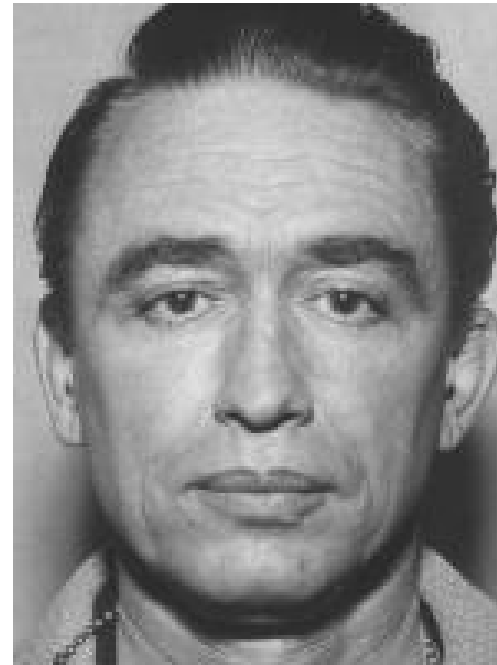


*Confidence* is the probability of correct verification for a given match score

# Improved regenerated image



Average of 10  
Best Estimates



Target Image

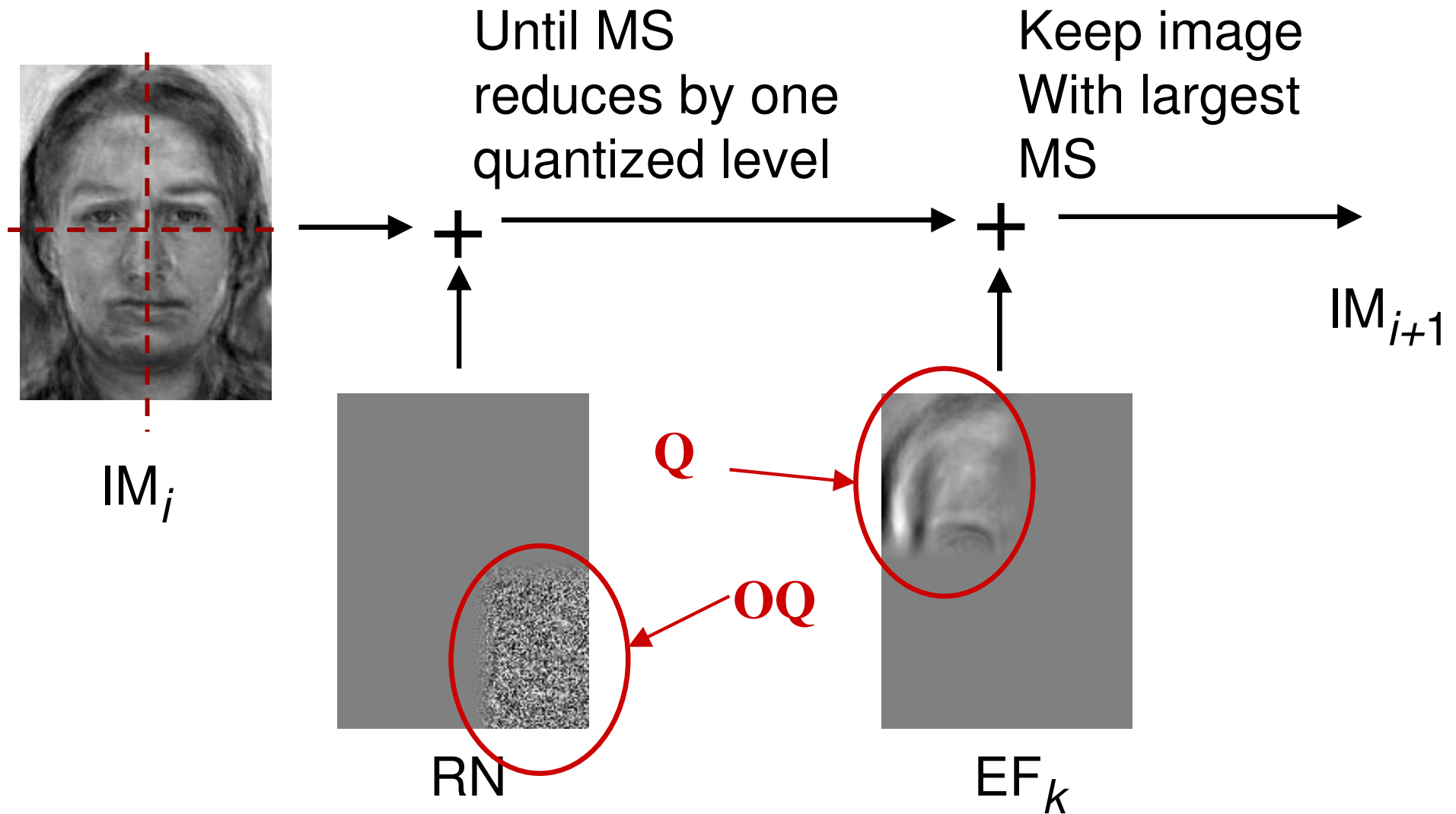
# Protection:

## According to BioAPI

- “...allowing only discrete increments of score to be returned to the application eliminates this method of attack.”
- Idea: most image modifications will not change the match score
- **This work:** We modify the “hill-climbing” algorithm to work with quantized data

Source: BioAPI, version 1.1, p.21, <http://www.bioapi.org>

# Modified “hill-climbing”



# Modified “hill-climbing”

***Iterative estimate improvement:*** (for  $i \dots$ )

- Select eigenimage,  $EF_k$
  - Select quadrant  $Q$ . Opposite quadrant is  $OQ$ .
  - Generate image  $RN$ : noise in  $OQ$  and zero elsewhere.
  - Calculate amount of  $RN$  to reduce the  $MS_i$  by one quantization level.
- New

$MS_i = \text{biometric\_compare}(IM_i, IM_{targ})$

$MS_{NI} = \text{biometric\_compare}(IM_i + n \times RN, IM_{targ})$

# Modified “hill-climbing”

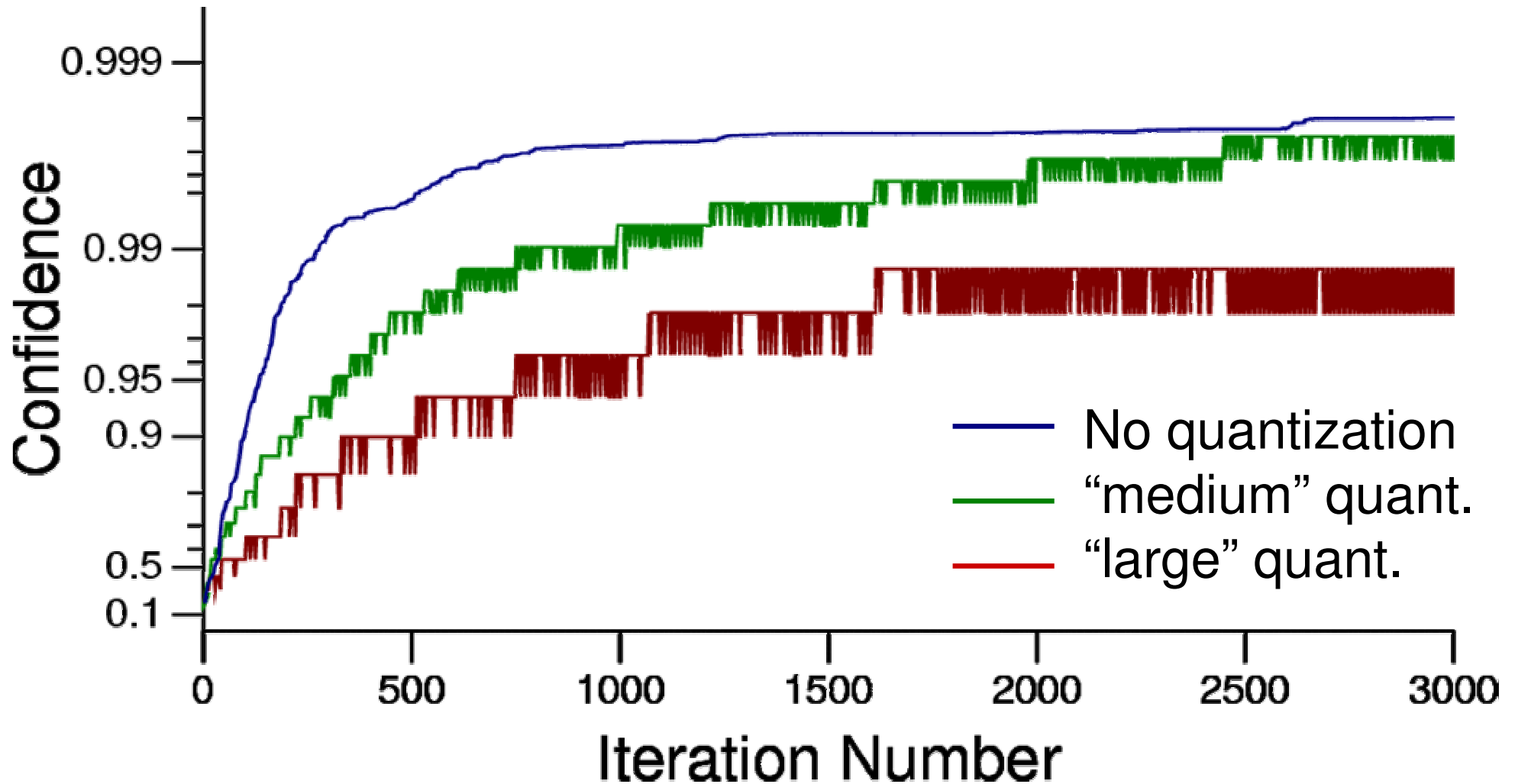
***Iterative improvement (continued ...)***

- Randomly select:  $EF_k$
- Iterate for a range of  $c_j$  using quadrant  $Q$

$MS_j = \text{biometric\_compare}( IM_k + c_j \times EF_{k,Q}, IM_{targ} )$

- $j_{max} = j$  for which  $MS_j$  is maximum
- $IM_{i+1} = IM_i + c_{j,max} \times EF_{k,Q}$
- Truncate  $IM_{i+1}$  to image limits (ie. 0 to 255)

# Results: modified “hill-climbing”





# Modified “hill-climbing”

- Discrete match score means less information is available
  - algorithm takes longer
- Image regeneration works because biometric algorithms “sum up” matching characteristics
  - Changes in quadrants are “independent”
  - We degrade image in one quadrant so that match score is in most informative range

# Discussion

Images can be regenerated from biometric templates

- will fool biometric algorithm
- visually reflect important features
- The BioAPI recommendation of using quantized match scores does not provide complete protection

# So what?

Approaches shown are:

- Time consuming
  - needs 40,000 biometric comparisons
- Doesn't produce great images
  - Neither fingerprint / facerec. images look much like the originals

# Implications:

- Image regeneration *is* possible
- Smarter people can probably figure out better and faster ways to do it
- Look alike image could be used to
  - masquerade as target
  - Identify target person

# Some privacy/security implications:

## Biometric Data on ID documents:

- Not an issue for Face Rec. (holders photo is already on the document)
- However, countries may use fingerprint / iris template.

## Security agencies may allow searches against watch list:

- Primary agency does not want to distribute images
- However, another agency may access these images through regeneration from match scores

# Final thought

- There is a tendency to use results from cryptography in biometrics security
- However, biometrics images are **not** random data
- Such correlations can probably be exploited to in many biometric systems