
Measurement of the information for identification in iris images

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Overview

- Thesis objective
 - Biometric sample quality evaluation
 - Measuring biometric information
 - Approach by Daugman
 - Approach by Adler *et al.*
 - Experimental work
 - Conclusion
-

Biometric Template Uniqueness

- Question to be answered: “How secure is the certain biometric system?”
 - Possible approach: Measurement of amount of information contained in a person’s characteristic
 - Thesis objective: measurement of information for identification in irises
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Biometric Sample Quality Assessment

- Lower quality results in lesser amount of identifying information
- Humans are traditionally believed to be the best quality evaluators
- How reliable is this assumption?

Quality measuring experiments :

- 8 participants
- 7 identification algorithms

Data sets :

- 84 iris images
 - 98 face images
-

Quality Assessment by Participants



Automatic Quality Assessment

$$MS_{i,j} = Q_i Q_j \quad \begin{array}{l} 0 < MS < 1 \\ 0 < Q < 1 \end{array}$$

Match Score
Table

	1	2	3	4
1	1.0	.9	.8	
2		1.0	.7	
3			1.0	
4				1.0

$$\log(MS_{i,j}) = \log(Q_i) + \log(Q_j)$$

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ \vdots & & & \vdots \end{bmatrix} \begin{bmatrix} \log Q_1 \\ \log Q_2 \\ \log Q_3 \\ \log Q_4 \end{bmatrix} = \begin{bmatrix} \log .9 \\ \log .8 \\ \log .7 \\ \vdots \end{bmatrix}$$

Quality Evaluation - Results

- Are humans consistent with each other?

YES

- Are algorithms consistent with each other?

YES

- Are humans consistent with algorithms or other quality measures?

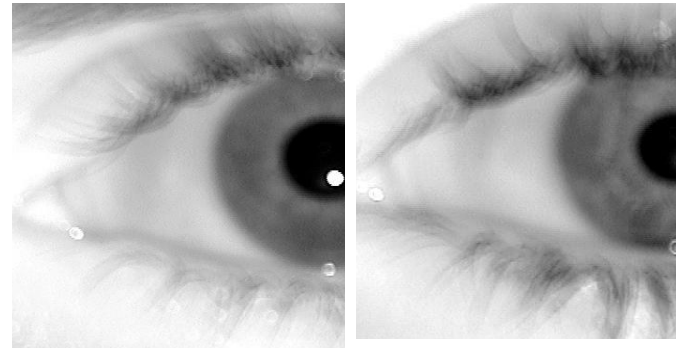
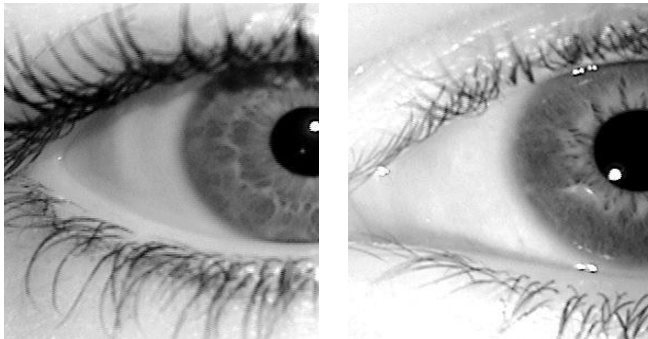
NO

Quality Evaluation – Results (cont.)

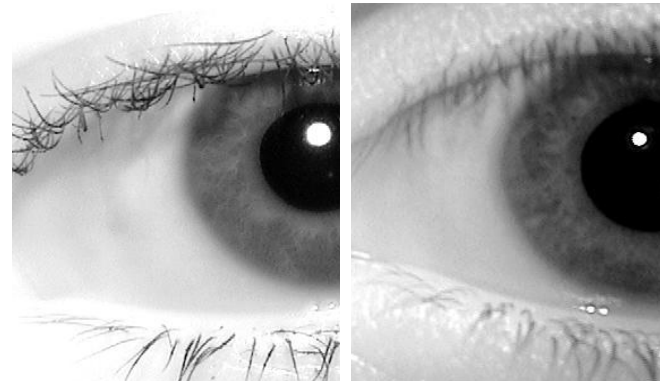
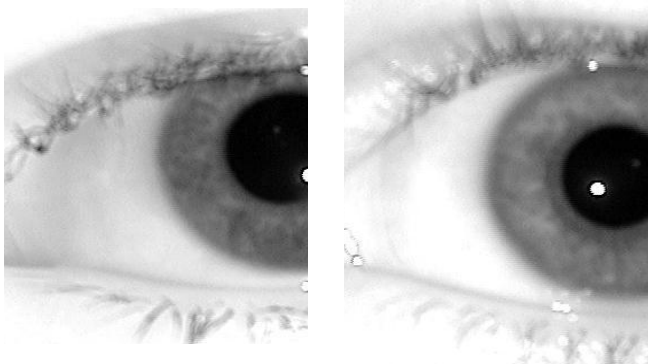
Best Irises

Worst Irises

Human
Selections



Algorithm
Selections



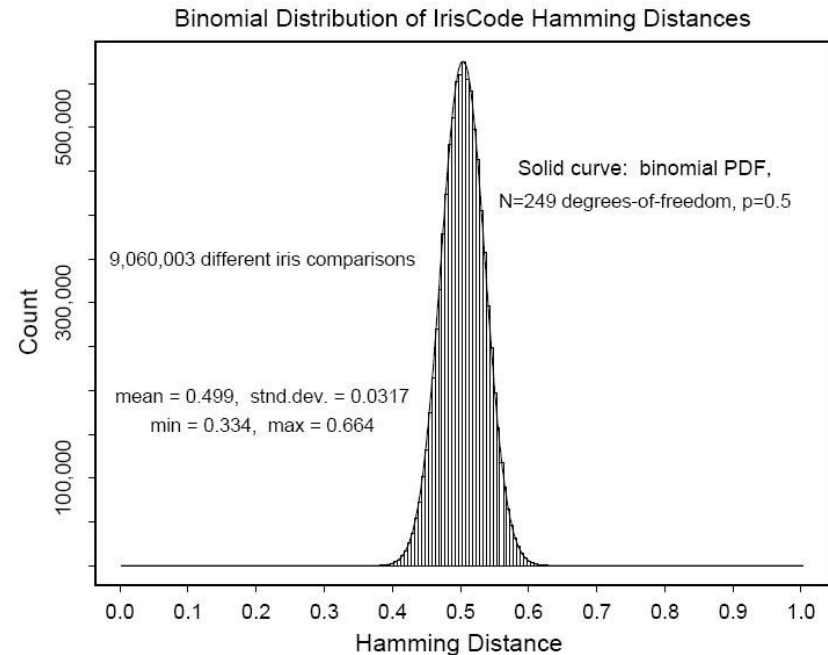
Discrimination Entropy Approach by Daugman

- Assumption:
match score distribution \sim binomial distribution
- Degrees of freedom of the observed distribution:

$$N = p(1 - p)/\sigma^2 = 249$$

- Fractional functional form:

$$f(x) = \frac{N!}{m!(N - m)!} p^m (1 - p)^{N - m}$$



Relative Entropy Approach by Adler et al.

- Biometric information: “decrease in uncertainty about the identity of a person due to a set of biometric measurements”
- $D(p \parallel q)$:
 - p - one person feature distribution
 - q - population feature distribution
 - preferable over H
- Assumption: feature values have Gaussian distribution

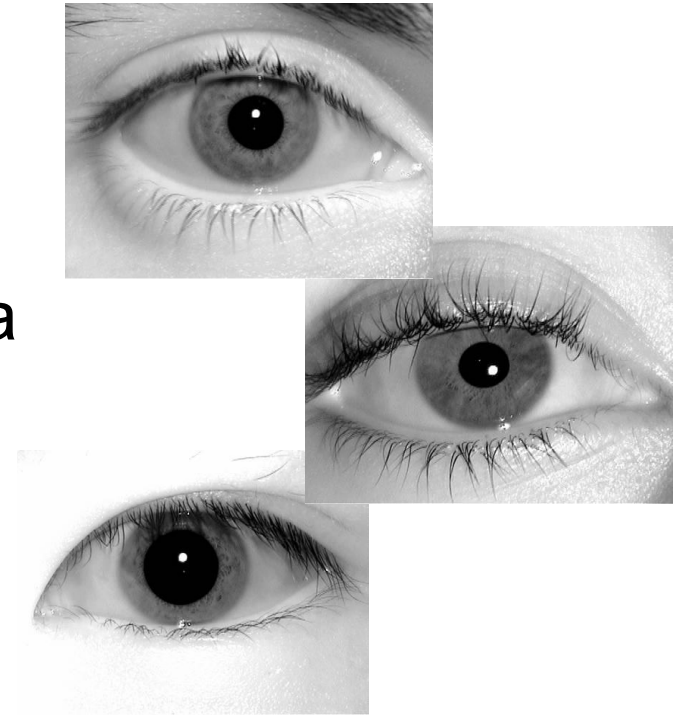
Experimental Setup

- Dataset:

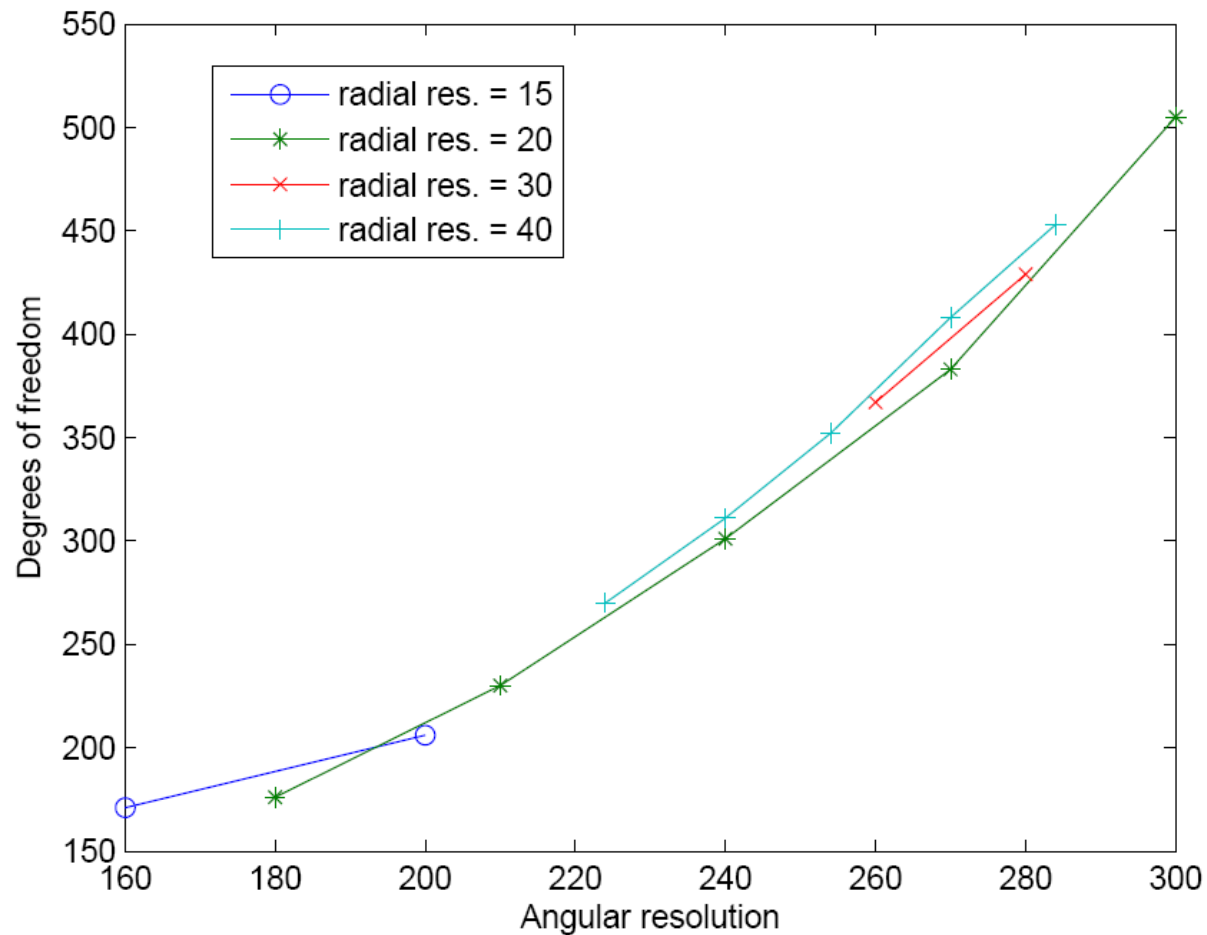
- obtained using L.G. iris camera
- 12 eyes, 30 samples each

- Software:

- Masek and Kovesi (2003)
- open-source iris recognition software



Discrimination Entropy: Experimental Results

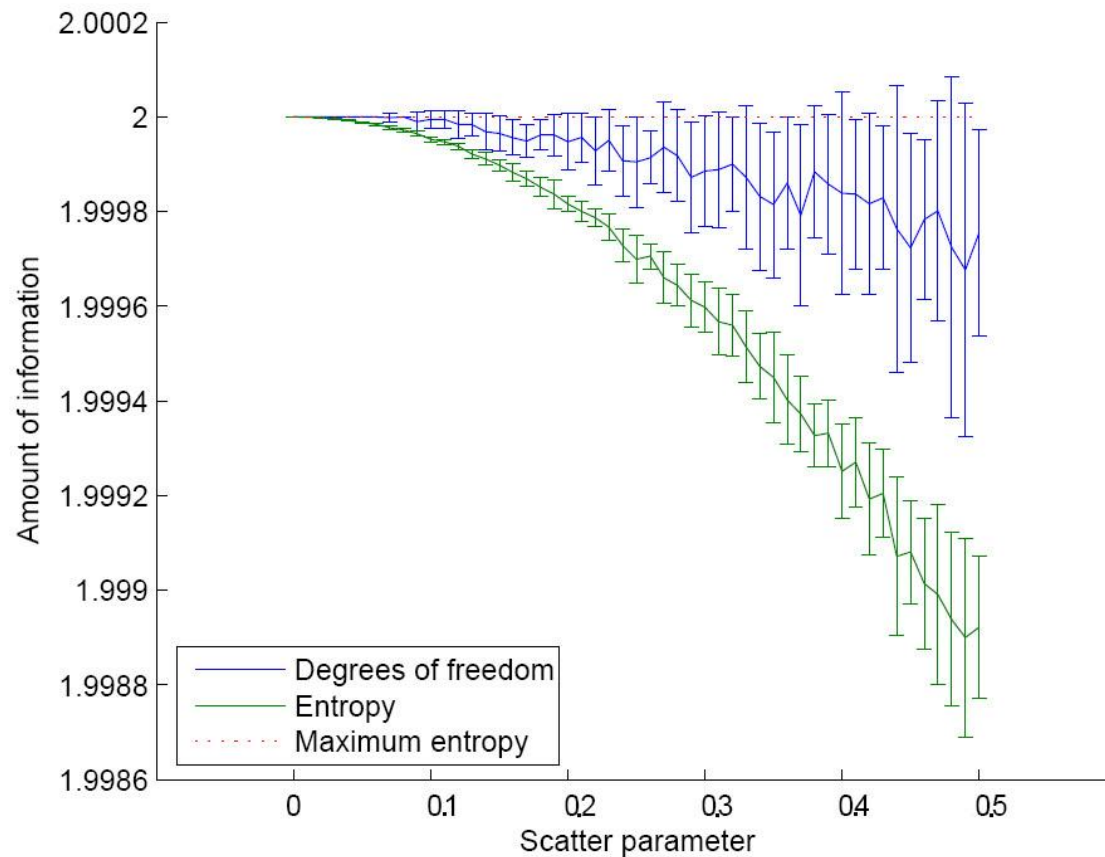


Discrimination Entropy: Statistical Analysis

- Discrimination entropy vs. $H(p)$
 - Procedure:
 - Define a template as a binary string of length **M**
 - Assign a probability to each template (2^M values)
 - Calculate $H(p)$ of the template
 - Calculate HD distribution and fit the binomial curve
 - Compare the results
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Discrimination Entropy: Statistical Analysis - Scheme 1

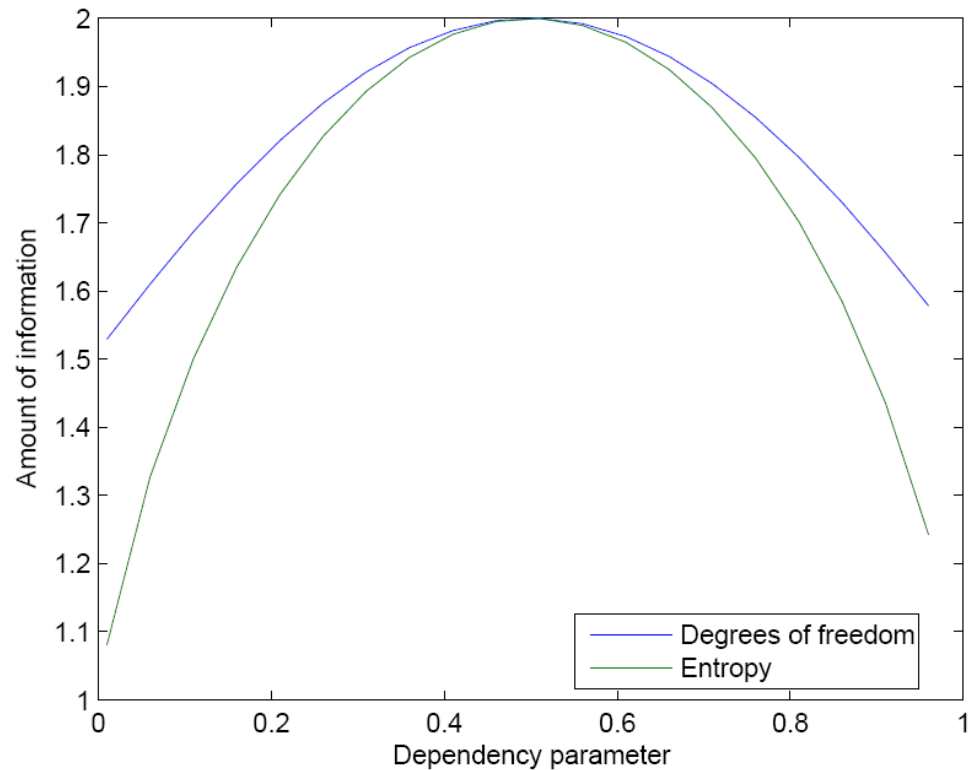
Differently scattered probabilities



Discrimination Entropy: Statistical Analysis - Scheme 2

Templates with varying dependencies between their bits

$$\begin{bmatrix} P(00) & P(01) \\ P(10) & P(11) \end{bmatrix} = \frac{1}{2} \begin{bmatrix} d & 1-d \\ 1-d & d \end{bmatrix}$$

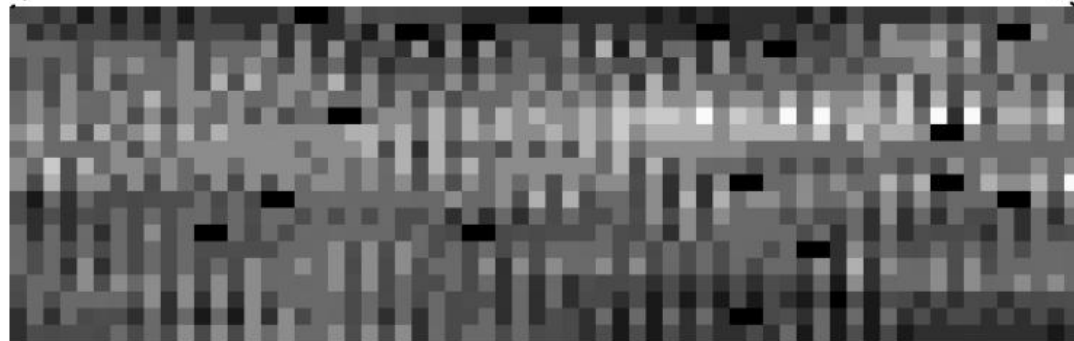


Relative Entropy: Feature Entropy

Pupil



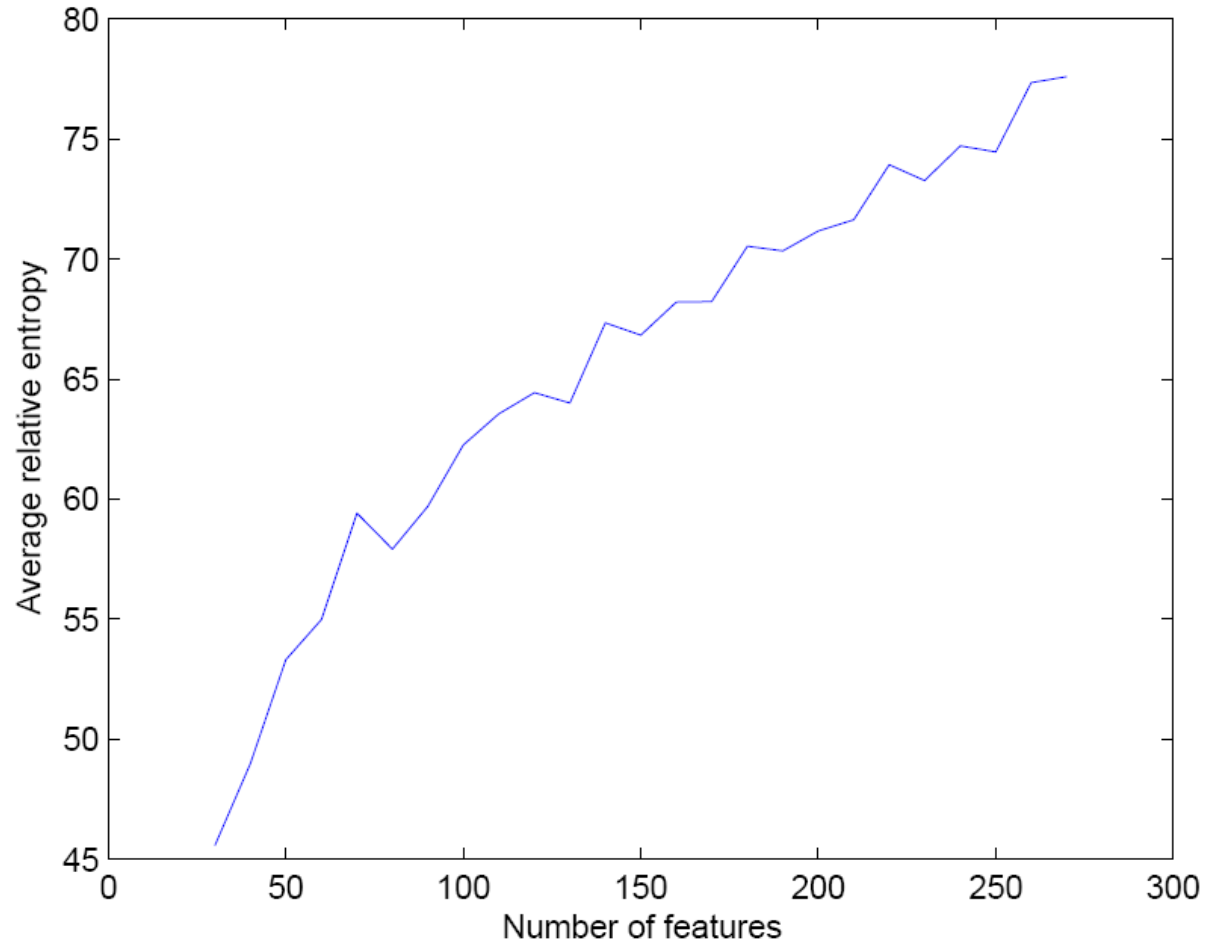
Sclera



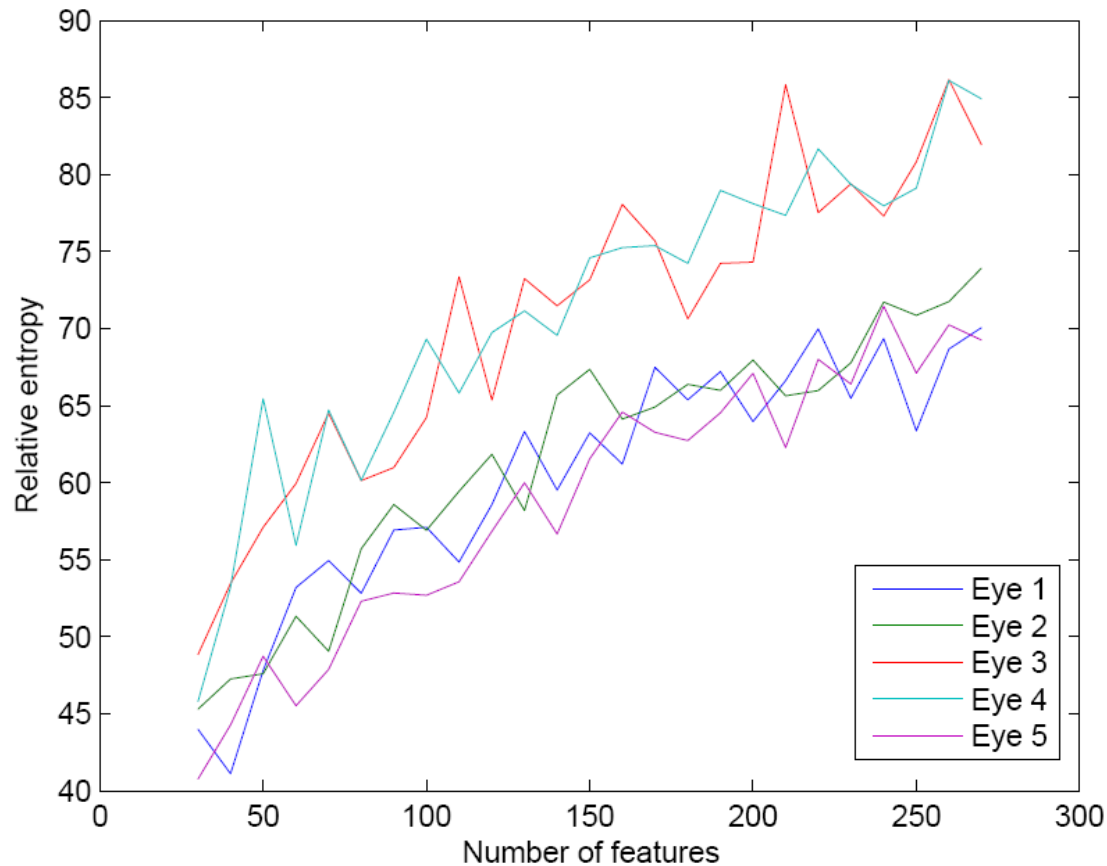
Relative Entropy: Iris Template Entropy

Template dimensions :

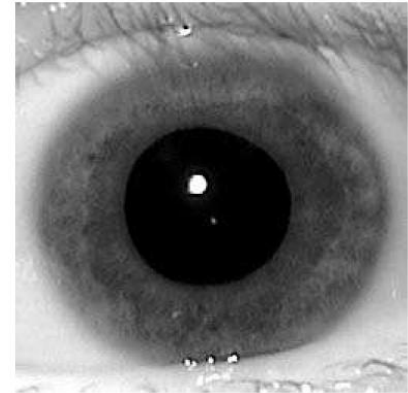
20 x 480



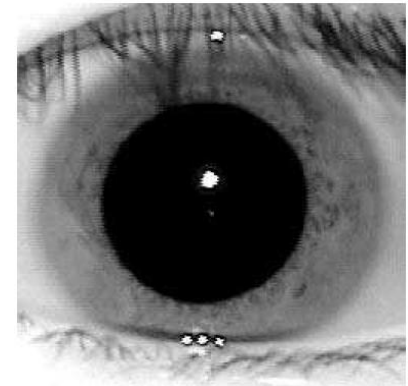
Relative Entropy: Iris Template Entropy (cont.)



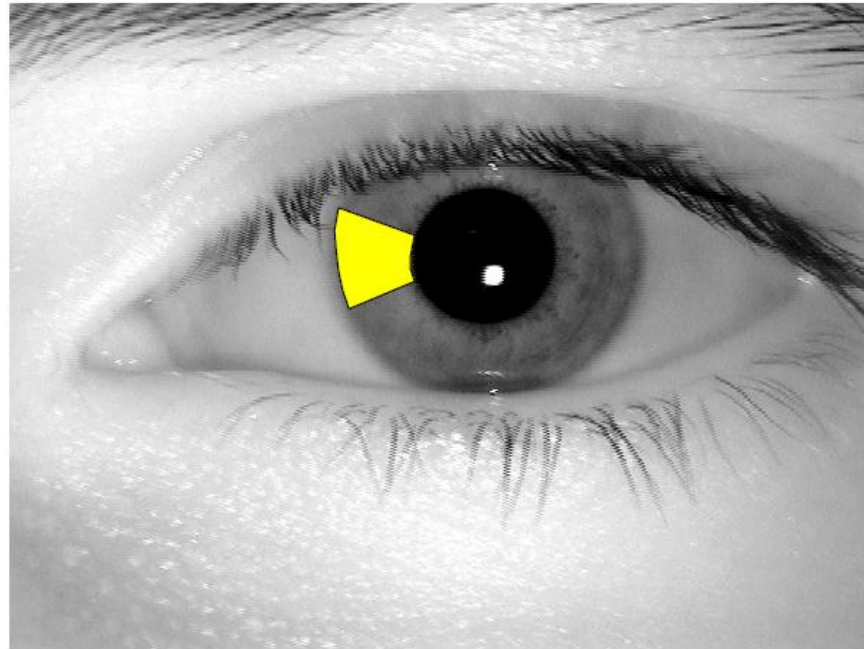
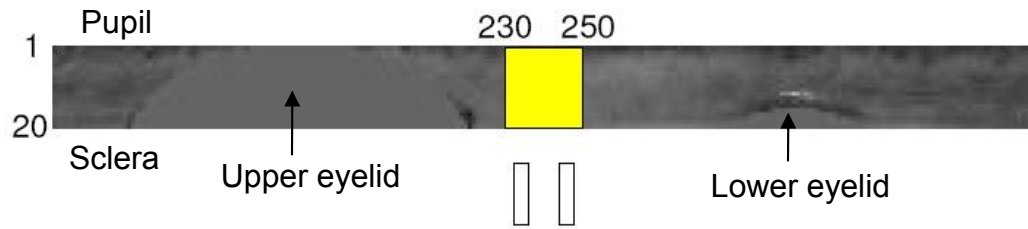
Higher information content



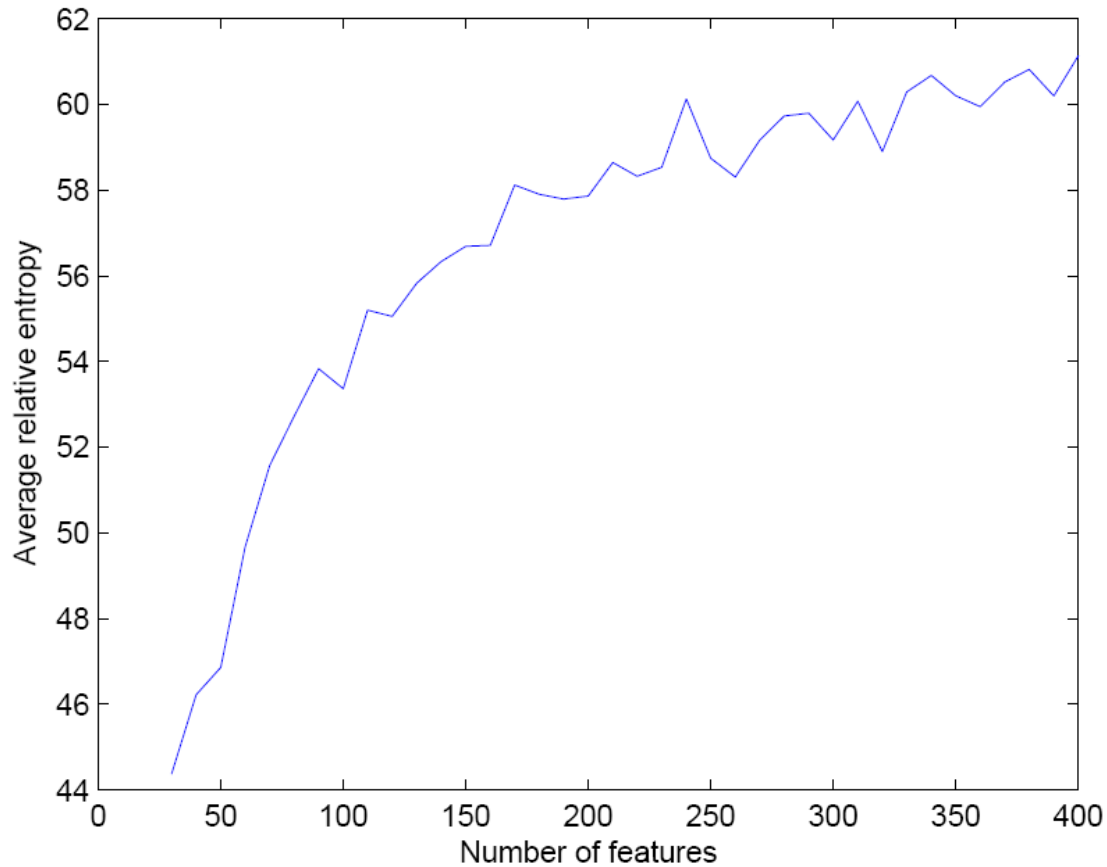
Lower information content



Relative Entropy: Iris Region Entropy



Relative Entropy: Iris Region Entropy (cont.)



Conclusion

- Discrimination entropy:
 - too idealistic
 - does not measure identification information
 - Relative entropy:
 - more appropriate measure of biometric information
 - the most informative: inner circles closer to the pupil
 - converges due to neighbor pixel dependencies
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Thank you!

Questions?
