

**BIOMETRIC  
SYSTEMS**

“Artificial Intelligence” Course  
University of Naples Federico II  
May 13, 2010

## outline

- 1 Introduction to Biometrics
- 2 Fingerprint Recognition
- 3 Face Recognition
- 4 Iris Recognition
- 5 Multimodal Biometric Systems
- 6 Liveness Detection in Fingerprint Scanners

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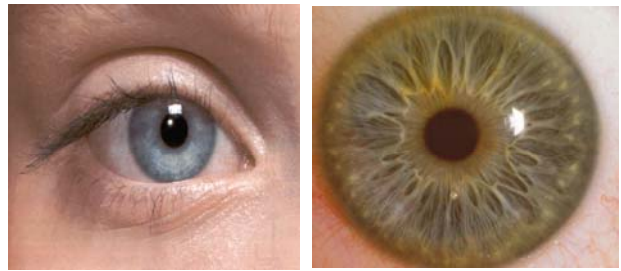
### Iris Recognition

#### “State-of-the-art” Error rates

Modality	Test Label	Test Parameter	FNMR	FMR
Fingerprint	FpVTE 2003	US Government operational data	0.1%	1%
Fingerprint	FVC 2006	Heterogeneous population (young, elderly)	2.2%	2.2%
Face	FRGC 2006	Controlled Illumination, high-resolution images	0.8-1.6%	0.1%
Voice	NIST 2004	Text independent, multi-lingual	5-10%	2-5%
Iris	ITIRT 2005	Indoor environment	0.99%	0.94%
Iris	ICE 2006	Controlled Illumination, broad quality range	1.1-1.4%	0.1%

### Iris Recognition

#### Iris Pattern

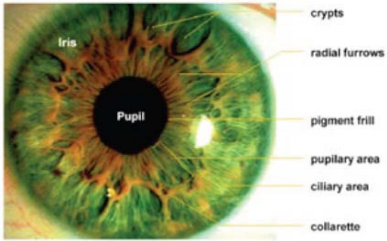


The iris is stable after the first year of life.  
*John Daugman*

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### Iris Recognition

#### Texture of the iris



The iris exhibits a very rich texture. It consists in pectinate ligaments adhering into a tangled mesh revealing striations, ciliary processes, crypts, rings, furrows, vasculature and other features.

*John Daugman*

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### Iris Recognition

#### Iris processing steps

**Enrollment**

```

graph LR
    A[Iris Image Acquisition] --> B[Localization & Normalization]
    B --> C[Image Enhancement]
    C --> D[Feature extraction (Iris Code)]
    D --> E[(Enrolled Database)]
  
```

**Authentication**

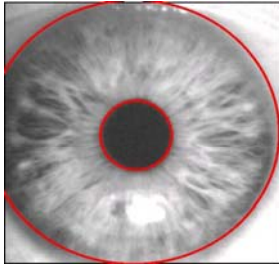
```

graph LR
    F[Iris Image Acquisition] --> G[Localization & Normalization]
    G --> H[Image Enhancement]
    H --> I[Feature extraction]
    I --> J{Compare}
    J --> K[Match Score]
    K --> L[Decision]
  
```

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Iris Recognition

## Localizing Iris



- Localize precisely the inner and outer boundaries of the iris.
- Detect and exclude eyelids if they intrude.

These detection operations are accomplished by an integro-differential operator, which behaves as a circular edge detector:

$$\max_{(r, x_0, y_0)} \left| G_{\sigma}(r) * \frac{\partial}{\partial r} \oint_{r, x_0, y_0} \frac{I(x, y)}{2\pi r} ds \right|$$

- The operator searches for a circular boundary with radius  $r$  and center  $(x_0, y_0)$  such that the change in radial pixel intensity across the boundary is maximum.


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Iris Recognition

## Iris Encoding

The output of the segmentation process is a binary mask that indicates the Iris and non-iris pixels in the image.

**Normalization scheme:** each point in the  $(x, y)$  domain is mapped to a pair of Polar coordinates  $(r, \theta)$ . This results in a fixed size rectangular iris image.



Normalized iris

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Iris Recognition

## IrisCode

**Gabor filters** are then used to extract the textural information from the iris encoding.

$$G(x, y) = e^{-\pi[(x-x_0)^2/\alpha^2 + (y-y_0)^2/\beta^2]} e^{-2\pi i[u_0(x-x_0) + v_0(y-y_0)]}$$

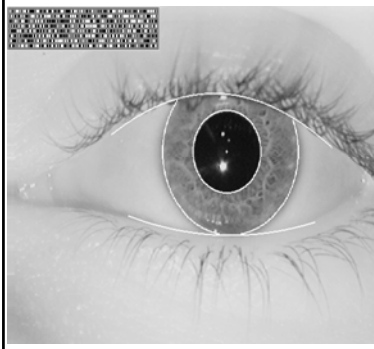
Since the prominence of iris texture changes as one moves away from the pupil, three different Gabor filters are applied to different regions of the normalized iris.



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Iris Recognition

## Comparison



- **Two IrisCodes** of eyes are compared by using Hamming Distance (HD) in order to detect the fraction of their bits that disagree.

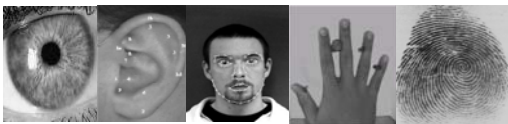
$$HD = \frac{\|(I_1 \otimes I_2) \cap M_1 \cap M_2\|}{\|M_1 \cap M_2\|}$$

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Multimodal Biometric Systems


## Fusion in Biometrics

A combination of multiple *pieces of the evidence*



### Why Multi-Biometrics?

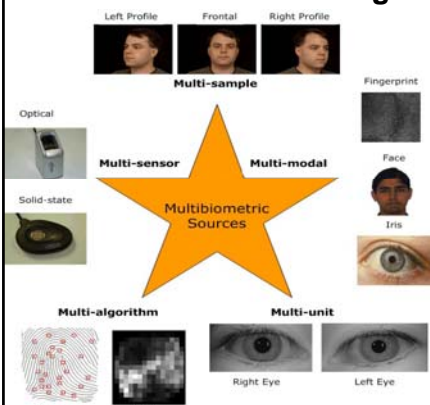
- **Performance Improvement**, error rates are not zero
- **Noisy Data**, e.g., low quality images
- **Intra-class Variations**, e.g., different samples of the same person
- **Distinctiveness**: Inter-class Similarity
- **Non-Universality**
- **Spoof Attacks**, e.g., artificial fingers



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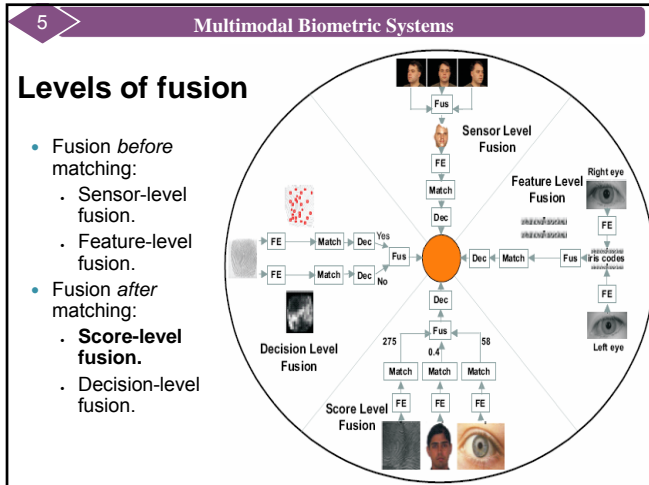
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Multimodal Biometric Systems

## What to integrate?



- multi-Sensor
- multi-Sample
- multi-Algorithm
- multi-Instance
- multi-Modal

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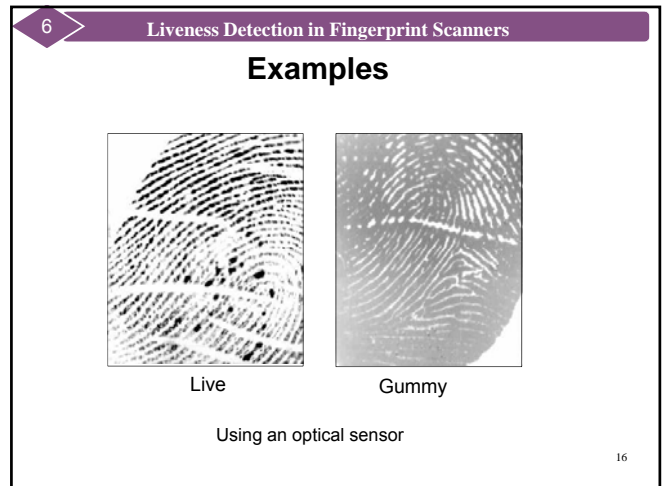
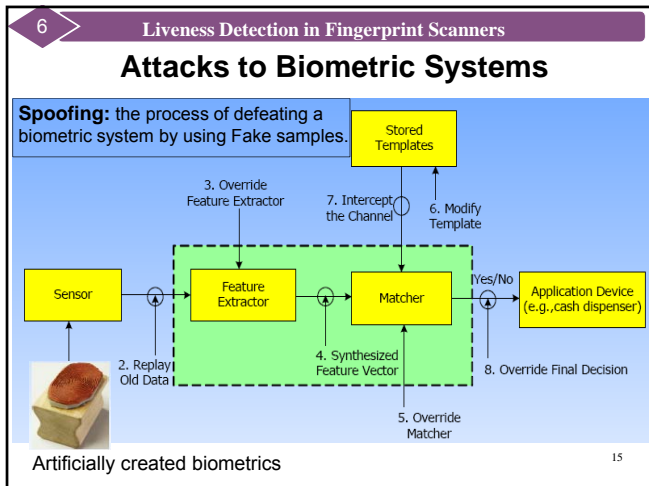


6 **Liveness Detection in Fingerprint Scanners**

### Biometrics and Liveness

- Is Biometrics a reliable, secure solution?
- What are the **threats** to biometric systems?
- How can we make biometric systems more secure?
- Since intruders will introduce a large number of spoofed biometrics, Liveness will enhance performance of a biometric system.
- Liveness detection reads claimant's physiological signs of life.

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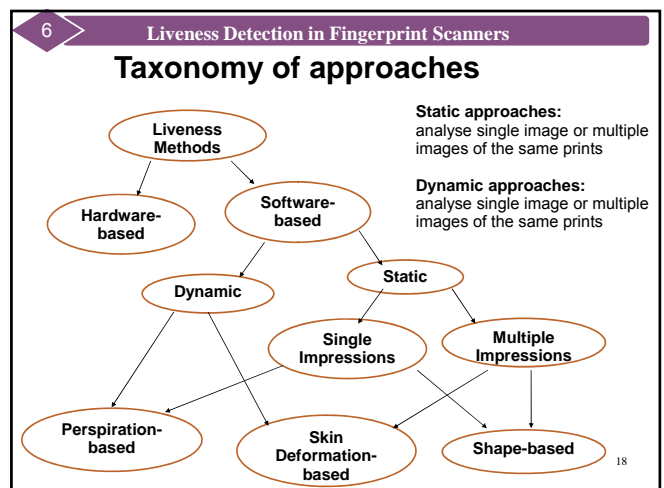


6 **Liveness Detection in Fingerprint Scanners**

### Analysis of the attack

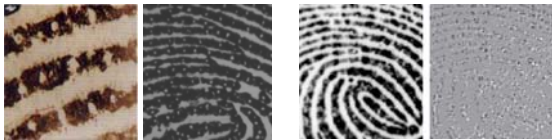
- 1) Latent fingerprint dusted visible on the side of mug.
- 2) Finished transparency showing negative image.
- 3) Finished board (mold).
- 4) Gelatine spread on the board as a thin layer.
- 5) Press the gelatine gently on the pad with the finger.

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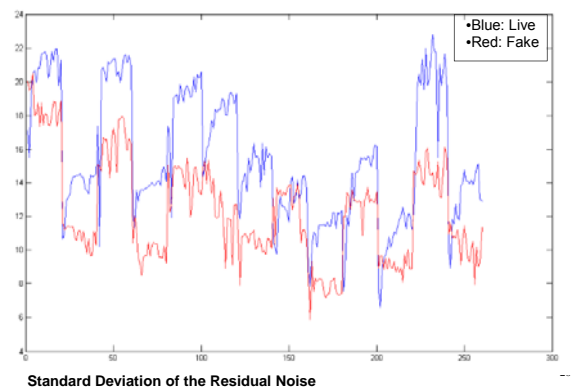
### Static Algorithm

- The fingerprint image is transformed into a *ridge signal*.
- The algorithm is based on static features derived from a single fingerprint image.
  - Pore spacing.
  - Residual noise of the fingerprint image.
  - Intensity-based features.
  - Texture analysis.



Over 1000 dpi    1000 dpi sensor    Original and noise residual images 19

### Searching for discriminant features



Standard Deviation of the Residual Noise

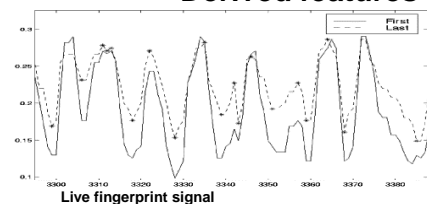
### Dynamic Algorithm

#### West Virginia Perspiration Method:

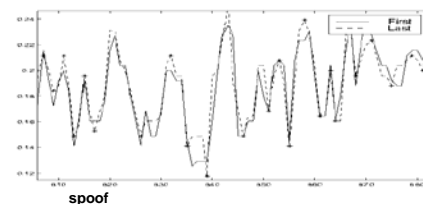
- The algorithm analyse the perspiration over time as sign of life.
- Fingerprint images are processed and ridge signals are obtained.
- Signal amplitude is proportional to the moisture along the traversed ridges.
- Peaks relate to the moistest and valleyss to driest regions.
- In live fingers, perspiration starts around the pores and spread along the ridges creating a distinct signature of the process.

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### Derived features



Constant periodic Peaks and rising valleys for the live signal.



spoof

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### References

- K. Choi H. Choi, R. Kang and J. Kim, "**Aliveness detection of fingerprint using multiple static features**", *World Academy of Science, Engineering and Technology*, 28: 157-162, 2007.
- S. Schuckers, "**Spoofing and Anti-spoofing measures**", *Information Security Technical Report*, 7(4): 56-62, 2002.

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