#### **Face Detection and Bi-Modal Authentication**

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### Outline

- Introduction and Background
- Face Detection + demo
- Bi-Modal Authentication + demo
- Discussions

# **Introduction and Background**

- Technologies developed at IDIAP of potential interest:
  - Face detection and tracking,
  - Biometric authentication (Face + Speech),
- Demonstrators:
  - Bi-modal authentication demo since 2003 (versions  $\alpha$  and 1),
  - Face detector added in 2004 (version 2),
  - Software: based on Torch library (www.torch.ch)
  - Interface: QT
  - Platform: Linux
- Time-line:



# A step forward ...

- A step forward: move IDIAP demonstrators (running Linux) to Win XP,
- Partnership with HEVs (Technical University of Valais),
- TheArk/IdeArk is funding a project between IDIAP and HEVs,
- The goal of the project is to develop two applications:
  - A face tracker (fast face detector)
  - A Bio-Login and its User Manager (BLUM)

# **Partners & Persons involved**

- IDIAP (4 persons):
  - S. Marcel (Face detection and authentication),
  - J. Mariethoz (Speech authentication),
  - Y. Rodriguez (Face detection),
  - F. Cardinaux (Face authentication)
- TheArk: F. Bagnoud,
- IdeArk: F. Crittin,
- HEVs (4 persons):
  - D. Gabioud (Project Manager),
  - J-P. Gehrig (Face tracker and BLUM),
  - F. Dessimoz (Bio-Login and BLUM),
  - G. Maître (Quality control & Performance evaluation)

### Outline

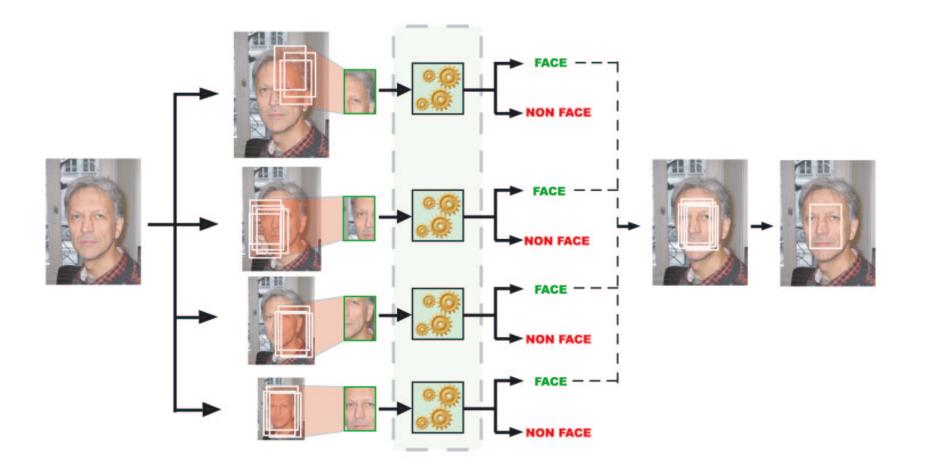
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## **Face Processing: Definitions**

- Face localization: locates human faces in images at different positions, scales, orientations and lighting conditions with the assumption that the image contains one and only one face,
- Face detection: generalization of face localization to multiple faces,
- Face tracking: keep trace of similar detected/localized faces within a video sequence.

### **Face Detection: Overview**

• Frontal face detection:



it is so fast that it provides the illusion of tracking

## **Face Detection: Demonstration**

- Details:
  - Programming: Torch, C++ and QT
  - Platform: Win XP
  - Device: a Logitech camera (QuickCam Zoom, 4000 Pro, Notebook Pro, Orbit Sphere)



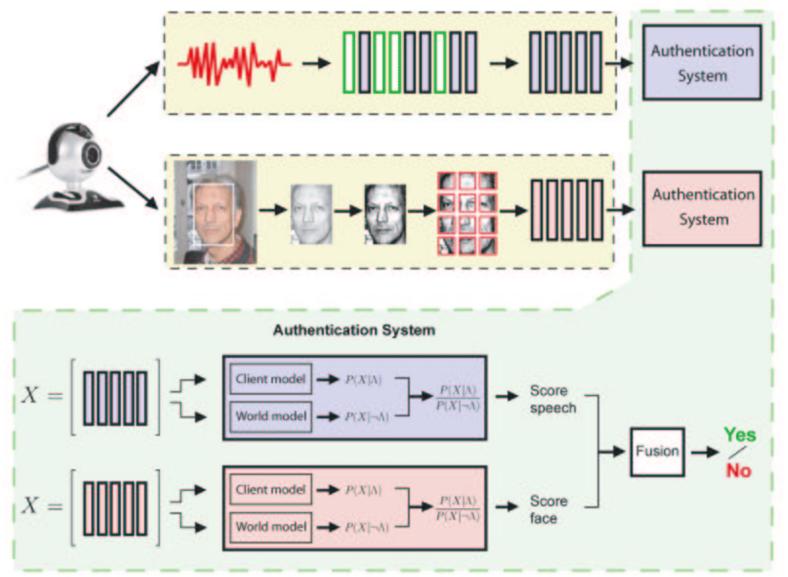
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## **Bi-Modal Authentication: Definitions**

- Identification: identify a person based on the face/speech of a person. This face/speech has to be compared with all the registered persons (one-to-many matching),
- Authentication (or verification): is concerned with validating a claimed identity based on the face/speech, and either accepting or rejecting the identity claim (one-to-one matching),
- Recognition: general topic that includes both identification and authentication.

### **Bi-Modal Authentication: Overview**



## **Bi-Modal Authentication: demonstration**

- Details:
  - Programming: Torch, C++ and QT
  - Platform: Win XP
  - Device: a Logitech camera (QuickCam Zoom, 4000 Pro, Notebook Pro, Orbit Sphere)
- Two applications:
  - Bio-Login User Manager: Creates a new account (Enrollment)



- Bio-Login: Login using your face and your voice (Test)



#### **Discussions**

- Questions ?
- More details ?

#### **More details**

• Bi-Modal authentication

### **Bi-Modal Authentication: Framework**

- Face (64x80 pixels): Discrete Cosine Transform (DCT) based features and Gaussian Mixture Models (GMMs)
  - features: sequence  $(X_1^F = \{\mathbf{x}_1...\mathbf{x}_F\})$  of DCTmod2 frames  $\mathbf{x}_f \in \mathbb{R}^{18}$  (15 DCT coeff. -3 first coeff. +3  $\Delta_x$  +3  $\Delta_y$ )
  - models: diagonal GMMs ( $\lambda^{face}$ ) with 512 gaussians (18'944 param.)
  - score:  $\Lambda_C^{face}(X_1^F) = \log P(X_1^F | \lambda_C^{face}) \log P(X_1^F | \neg \lambda_C^{face})$
- Speech (8 KHz): Linear Freq Cepstral Coefficient (LFCC) features and GMMs
  - features: sequence  $(Y_1^S = \{\mathbf{y}_1...\mathbf{y}_S\})$  of LFCC frames  $\mathbf{y}_s \in \mathbb{R}^{33}$ (16 basis +  $\partial$  + log energy  $\partial$ )
  - models: diagonal GMMs ( $\lambda^{speech}$ ) with 200 gaussians (13'400 param.)
  - score:  $\Lambda_C^{speech}(Y_1^S) = \log P(Y_1^S | \lambda_C^{speech}) \log P(X_1^S | \neg \lambda_C^{speech})$
- Fusion:  $P(X, Y|C) = w \cdot \Lambda_C^{face}(X) + (1-w) \cdot \Lambda_C^{speech}(Y)_{\mathrm{Se}}(X) + (1-w) \cdot \Lambda_C^{speech}(Y)$

. Sébastien Marcel, February 2005 – p.16/2

### **Bi-Modal Authentication: Enrollment**

- Let us note:
  - the parameter set for client C as  $\lambda_C$ ,
  - the parameter set describing a generic client as  $\neg \lambda_C$ ,
  - $X = {\mathbf{x}_t}_{t=1}^T$  as the set of feature vectors  $\mathbf{x}_t$  supporting the claim.
- 1 train a world model  $\neg \lambda_C$  from a large dataset by Maximum Likelihood,
- 2 adapt a client model  $\lambda_C$  from  $\neg \lambda_C$  using client data by Maximum A Posteriori,
- given a claim for client C's identity and a set of feature vectors X supporting the claim, we find an opinion on the claim using:

$$\Lambda_C(X) = \log P(X|\lambda_C) - \log P(X|\neg\lambda_C)$$
(1)

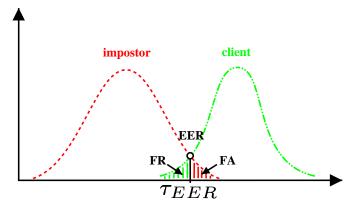
where  $P(X|\lambda_C)$  is the likelihood of the claim coming from the true claimant and  $P(X|\neg\lambda_C)$  is the likelihood of the claim coming from an impostor,

#### Fusion

- Goal: merge outputs of face and speech experts (2 or more) into a feature vector  $[\Lambda^1(X),...,\Lambda^n(X)]$  and try to classify it as a client or an impostor,
- Classifiers: Linear, MLP (Multi-Layer Perceptrons), SVM (Support Vector Machines),
- Fusion produces an opinion  $\Lambda^*(X)$  that might be used for final decision.

# **Performance Evaluation**

- the verification decision is then reached as follows: given a threshold  $\tau$ , the claim is:
  - is accepted when  $\Lambda(X) \geq \tau$  ,
  - is rejected when  $\Lambda(X) < \tau$ .
- Select a threshold to take the final decision:



- False Rejection (FRR): when the system rejects a client,
- False Acceptance Rate (FAR): when the system accepts an impostor,
- Half Total Error Rate ( $HTER = \frac{FRR + FAR}{2}$ ): unique measure,
- the decision threshold  $\tau$  chosen on a development data set.



results (in terms of HTER) on XM2VTS, BANCA and IDIAP databases:

	XM2VTS (LP1)	BANCA (Mc)	BANCA (P)	IDIAP
Face	1.67	5.77	18.96	7.61
Speech	1.14	4.32	12.29	3.15
Fusion	0.48	4.32	9.99	1.49

- XM2VTS can be found at http://www.ee.surrey.ac.uk/Research/VSSP/xm2vtsdb/,
- BANCA can be found at http://www.ee.surrey.ac.uk/banca/.