# IM2.VP Visual Information Processing

# Phase II - Achievements

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# IM2.VP - Tasks

IM2.VP.1- Face detection and recognition

• IDIAP (Dr S. Marcel) & EPFL (Prof. J.-Ph. Thiran)

- IM2.VP.2 Multi-source image representation for joint processing and coding
  - EPFL (Prof. P. Frossard)
  - task finished in December 2008

IM2.VP.3 - Multi-person activity discovery

- IDIAP (Dr D. Gatica-Perez & Dr J.-M. Odobez)
- task finished in December 2007

IM2.VP.4 - Head activity analysis and visual Focus of attention (VFOA)

• IDIAP (Dr J.-M. Odobez)

IM2.VP.5 - Recognition and interpretation of body gestures

- ETHZ (Prof. L. van Gool, Dr B. Fasel) & EPFL (Prof. A. Billard)
- task finished in December 2007
- IM2.VP.6 Handwriting recognition and whiteboard data analysis
  - Uni. Bern (Prof. H. Bunke)
- IM2.VP.7 Omnidirectional visual attention

• Uni NE (Prof. H. Hügli)

IM2.VP.8 - Vision-supported speech understanding

• ETHZ (Prof. L. van Gool & Dr B. Pfister)

# Face detection, alignment & recognition

- Frontal and MultiView Face
   Detection System
  - New robust features
  - Pyramid of classifiers for simultaneous non-frontal face detection and pose estimation
  - Information theoretic optimal classifier combination
- Robust-to-illumination Face
   Alignment System
- Generative Models for Face Recognition







# Haar-LBP Features

• We investigated a novel feature set for fast illumination invariant face detection

Combines the advantages of both Haar and LBP



 Similar boosting framework with an augmented Haarlike feature



# Haar-LBP Features

 An HLBP feature compares the LBP counts in 2 subregions (Haar masks)



 Results: more robust to illumination variations compared to Haar and LBP

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### Multi-camera object recognition (P. Frossard)

- Recognition from multiple observation sets
  - Diversity increases performance
  - Multi-camera object recognition
  - Video face recognition
- Graph-based label propagation solution
  - Preserved data and label smoothness
  - •Low-complexity solution
- Distributed solution by consensus





Video sequence



# Sample achievements



Γ	MASC	MSM	KMSM	KLD
	88.88 (1.71)	74.88 (5.02)	83.2500 (3.4)	52.5 (3.95)

TABLE I OBJECT RECOGNITION RATE IN THE MEAN(STD) FORMAT, MEASURED ON THE ETH-80 DATABASE.

#### Classification of multiple observation sets by semisupervised learning

Effrosyni Kokiopoulou and Pascal Frossard submitted to IEEE Transactions on Pattern Analysis and Machine Intelligence, October 2008.

#### <u>3D Face Recognition with Sparse Spherical</u> <u>Representations</u>

Roser Sala Llonch, Effrosyni Kokiopoulou, Ivana Tosic and Pascal Frossard Accepted for publication, Pattern Recognition, April 2009.

Video face recognition with graph-based semi-supervised learning Effrosyni Kokiopoulou and Pascal Frossard Invited paper, IEEE ICME, Cancun, Mexico, July 2009.

#### <u>Graph-based Classification for Multiple Observations of</u> <u>Transformed Patterns</u>

Effrosyni Kokiopoulou, Stefanos Pirillos and Pascal Frossard Proceedings of ICPR, Tampa, Florida, USA, December

#### <u>3D Face Recognition using Sparse Spherical</u> <u>Representations</u>

2008.

Roser Sala Llonch, Effrosyni Kokiopoulou, Ivana Tosic and Pascal Frossard Proceedings of ICPR, Tampa, Florida, USA, December 2008.





## Multimodal multiparty joint VFOA recognition



#### Evaluation on IM2/AMIDA datasets

5 hours of natural meeting - 55% VFOA recognition rate – improvement due to context modeling > 10%

S. Ba and J.M. Odobez, Recognizing Visual Focus of Attention from Head Pose in Natural Meetings, IEEE Trans. SMC, Feb 2009.
 S. Ba and J.M. Odobez, Multi-Person Visual Focus of Attention from Head Pose and Meeting Contextual Cues, PAMI, second revision, 2009.

### Multimodal multiparty VFOA recognition



### Multimodal multiparty VFOA recognition

### Notice

- Liveliness, difficulty of data
- person 3 focus changes according to context
   (between looking at person1, slide, standing person)
- slide changes favor looking at slides
- person 4 erroneous VFOA estimation (mainly due to head pose estimation problems)



## **Robust 3D Facial Actions Tracking**

# Goals : tracking head pose and facial actions on high resolution video sequences

- not limited to near frontal poses
- investigate head activities in natural conversation

### Approach

- 3D deformable mesh model
- likelihood modeling of different features
- Structural (salient) features

local, illumination and view invariant
model learned from virtual samples
=> stable representation

 Appearance facial texture dense, illumination sensitive online learned view-based templates
 => good for frame-to-frame smooth tracking

### Evaluation

- Head pose, Boston Univ. database: 3° for pan angle; better performance than state of the art trackers

- stability on long sequences (interviews)



[Lefevre 2009] S. Lefèvre and J.M. Odobez, Stucture and Appearance Features for Robust 3D Facial Actions Tracking, Best student paper award at the IEEE ICME 2009 conference.

## Monocular Body Pose Estimation (ETHZ)



### Speeded-Up Robust Features (SURF)



H.Bay, T. Tuytelaars, L. Van Gool, ECCV 2006 **465 citations** (google scholar)

## Online Handwriting Recognition (Prof. H. Bunke)

- Recognition of Whiteboard Notes:
   Online, Offline and Combination
  - Evaluation of Features
  - BLSTM and HMM Recognizers
  - Combining Online and Offline Information
- > Text vs. Non-Text distinction in Online Handwritten Documents. (in progress)
  - Segmentation followed by Classification
  - Classification of Individual Strokes



M. Liwicki and H. Bunke. *Recognition of Whiteboard Notes – Online, Offine and Combination*. World Scientific, 2008





## Online Handwriting Recognition (Prof. H. Bunke)

- > Writer Identification and Verification
  - Evaluation of confidence measures for HMM and GMM based systems
  - Evaluation of feature selection methods
- Ensemble Methods for Handwritten Text
   Line Recognition
  - Improved recognition via combinations of automatically generated recognizers
- Semi-Supervised Learning for Handwriting Recognition (in progress)
  - Investigating self-learning for HMM and NN-based recognizers

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R. Bertolami and H. Bunke. Hidden Markov model based ensemble methods for offine handwritten text line recognition. *Pattern Recognition*, 41(11):3452-3460, 2008

A. Schlapbach, H. Bunke, A Writer Identification and Verification System Using HMM Based Recognizers, *Pattern Analysis & Applications*, 10(1): 33-43, 2007





# Visual Attention in Omnidirectional



MAIN RESULTS (May 2007-May 2009):

- Development of static visual attention algorithm on the sphere with direct application to omnidirectional images:

- Development of dynamic visual attention for detecting spots of attention in omnidirectional video

<u>References:</u>

«Visual Attention on the Sphere», IEEE TIP, November 2008,

« Dynamic Attentive System for Omnidirectional Video», PCS, May 2009



### IM2.VP - conclusions

- IM2.VP developed and provided new methods and tools for visual information analysis
  - > Face image processing
  - > VFOA
  - > Gesture
  - > Object recognition
  - Handwriting
- Largely applied to IM2 corpus and other reference datasets
- Contributed in enriching the IM2 corpus
- Integrated in multimodal analysis (IM2.MPR)
- > Impressive publication record
  - > Journal papers (IEEE TIP, IEEE PAMI, etc)
  - > Conference papers
- > Spin-off companies
  - > Kooaba
  - > nViso
  - > ...

