



GESTURE RECOGNITION USING A NMF-BASED REPRESENTATION OF MOTION-TRACES EXTRACTED FROM DEPTH SILHOUETTES

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Introduction

• Task :

To recognize full-body gestures from depth-map sequences captured by a monocular depth sensor placed in front of the performer as in Human-Computer Interaction applications.

• Model :

Gesture = a concatenation of atomic sequences of motions and poses of different body parts. • **Proposal** : A novel approach that classifies full-body human gestures using original spatio-temporal features obtained by applying non-negative matrix factorisation (NMF) to an extended depth silhouette representation.

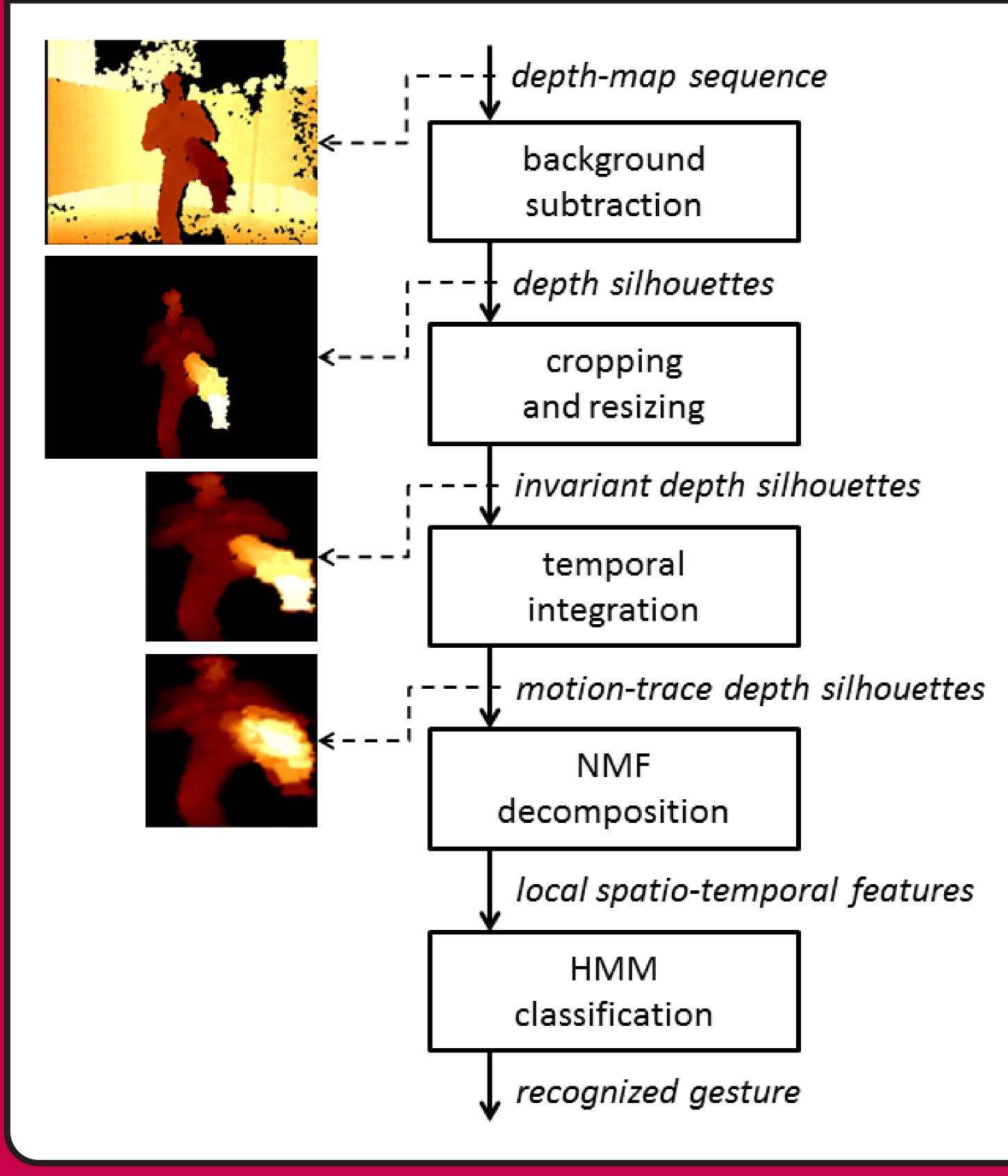
Experimental evaluation

• Dataset :

A subset of the Huawei/3DLife 3D human reconstruction and action recognition Grand Challenge database (http://mmv.eecs.qmul.ac. uk/mmgc2013/):

# of performers	8
# of different gestures	8

Overview of our system



of gesture repetitions \mathbf{b} per participant

• Evaluation procedure :

- *leave-one-participant-out* cross-validation (8 folds),
- adjustment of motion feature dimension (PCA, NMF),
- model hyper-parameter tuning (HMM),
- metrics : F-measure uniformly averaged over all gestures and over all folds.
- Reference systems :
 - Angles deduced from main body joints [1]
 - PCA-representation of main body joints trajectories [2]
 - PCA-representation of motion trace, adapted from [3]

Classification results

NMF representation process

To achieve the proposed **NMF representation process**, we have to perform two consecutive steps:

Best classification results in F-measure on 8 gesture classes with HMM classifiers

Approaches				
[1]	[2]	[3]	proposed	
78%	89%	89%	91%	
(Q=6)	$(w{=}6, d{=}30, Q{=}2)$	$(w{=}5, d{=}40, Q{=}2)$	$(w{=}6, k{=}128, Q{=}7)$	
(w: window size in samples, d and k feature dimension				
(resp. PCA and NMF), Q : number of HMM states)				

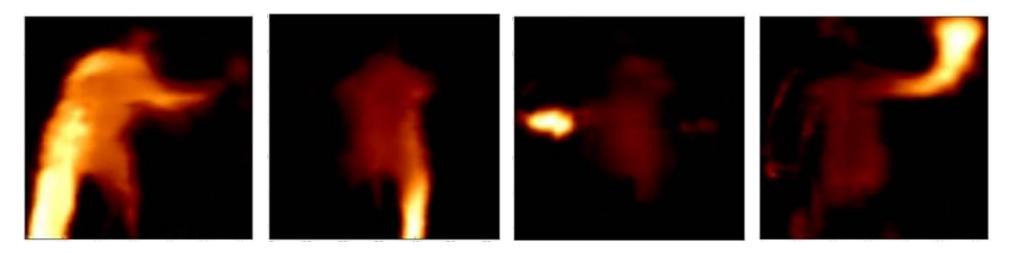
• The method [1] only based on global pose features is outperformed by the other methods. that incorporate motion dynamics and a decomposition process through their feature representation.

• Moreover the use of a NMF decomposition technique is more suitable to represent spatio-temporal motion features compared to PCA as the performance of our system is superior to the other reference systems.

Conclusion

• Our system obtains better performance than reference systems based on static features, PCA decomposition and/or 3D body joint positions.

- learning step : to create the **local spatio-temporal feature** dictionary, W (randomly initialized), using the entire training set.
- representation step : to project each incoming motion-trace on the dictionary components to obtain the activation vectors which are used as motion features.



NMF component examples from the created local spatio-temporal feature dictionary.

- This highlights the importance of considering a gesture as a concatenation of atomic sequences of motions and poses of different body parts.
- G.Th. Papadopoulos, A. Axenopoulos and P. Daras, Real-time Skeleton-|1| tracking-based Human Action Recognition Using Kinect Data, in Proc. of the 20th International Conference on MultiMedia Modeling, 2014.
- [2] A. Masurelle, S. Essid and G. Richard, Multimodal Classification of Dance Movements using Body Joint Trajectories and Step Sounds, in International Workshop on Image and Audio Analysis for Multimedia Interactive Services WIAMIS, 2013.
- [3] R. Muñoz-Salinas, R. Medina-Carnicer, F.J. Madrid-Cuevas and A. Carmona-Poyato, Depth silhouettes for gesture recognition, in Pattern Recognition Letters, vol.29, no.3, pp.319-329, 2008.