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Lifestyle understanding through the analysis of egocentric photo-streams

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*Lifestyle Understanding through the
Analysis of Egocentric Photo-streams*

Estefanía Talavera Martínez

This research has been conducted at the Intelligent Systems group of Johann Bernoulli Institute for Mathematics and Computer Science (Onderzoeksintituut JBI) of the University of Groningen and at the Department of Mathematics and Computer Science of the University of Barcelona.

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To my beloved parents and sister / A mis queridos padres y hermana

Contents

List of Figures	iv
List of Tables	v
1 Introduction	1
1.1 Scope	1
1.1.1 Societal impact	3
1.1.2 Privacy issues	4
1.2 Background	4
1.2.1 Temporal Segmentation	6
1.2.2 Routine Discovery	7
1.2.3 Food Related scene classification	9
1.2.4 Inferring associated sentiment to images	10
1.2.5 Social pattern analysis	11
1.3 Objectives	12
1.4 Research Contributions	12
1.5 Thesis Organization	15
2 Egocentric Photo-streams temporal segmentation	17
2.1 Introduction	18
2.2 Related works	19
2.3 Approach	20
2.3.1 Features	21
2.3.2 Temporal Segmentation	26
2.4 Experiments and Validation	29
2.4.1 Data	30
2.4.2 Experimental setup	32
2.4.3 Experimental results	37
2.4.4 Discussion	40
2.5 Conclusions and future work	41

3	Routine Discovery from Egocentric Images	43
3.1	Introduction	44
3.2	Related works	47
3.2.1	Routine from manual annotation	47
3.2.2	Routine from sensors	47
3.2.3	Routine from conventional images	48
3.2.4	Routine from egocentric images	49
3.3	Unsupervised routine discovery following an outlier detection approach	50
3.3.1	Experiments	52
3.4	Unsupervised routine discovery relying on topic models	58
3.4.1	Experimental Framework and Results	62
3.5	Discussions	73
3.6	Conclusions	74
4	Hierarchical approach to classify food scenes in egocentric photo-streams	75
4.1	Introduction	76
4.1.1	Our aim	76
4.1.2	Personalized Food-Related Environment Recognition	78
4.2	Related works	79
4.2.1	Scene classification	79
4.2.2	Classification of egocentric scenes	80
4.2.3	Food-related scene recognition in egocentric photo-streams	81
4.3	Hierarchical approach for food-related scenes recognition in egocentric photo-streams	82
4.4	Experiments and Results	85
4.4.1	Dataset	85
4.4.2	Experimental setup	89
4.4.3	Dataset Split	90
4.4.4	Evaluation	91
4.4.5	Results	92
4.5	Discussions	95
4.6	Conclusions	98
5	Recognition of Induced Sentiment when Reviewing Personal Egocentric Photos	101
5.1	Introduction	102
5.2	Related works	102
5.3	Sentiment detection by global features analysis	105

CONTENTS

5.3.1	Experimental Setup	107
5.4	Sentiment detection by semantic concepts analysis	109
5.4.1	Sentiment Model	111
5.4.2	Experimental Setup	112
5.5	Discussion and conclusions	114
6	Towards Egocentric Person Re-identification and Social Pattern Analysis	117
6.1	Introduction	118
6.2	Related works	119
6.3	Social Patterns Characterization	120
6.3.1	Person Re-Identification	120
6.3.2	Social Profiles Comparison	122
6.4	Experiments	122
6.4.1	Dataset	122
6.4.2	Experimental setup	123
6.4.3	Results	124
6.5	Conclusions	124
7	Summary and Outlook	127
7.1	Work Summary	127
7.2	Outlook	129
	Bibliography	133
	Summary	147
	Samenvatting	149
	Resumen	151
	Acknowledgements	153
	Research Activities	155
	About the Author	159

List of Figures

1.1	Illustration of collected photo-streams	2
1.2	Wearable camera - Narrative Clip.	5
1.3	Examples of wearable cameras	6
1.4	Illustration of the temporal segmentation of a collected photo-stream	7
1.5	Illustration of behaviours that describe the routine of a person	8
1.6	Illustration of food-related daily habits	9
1.7	Illustration of a camera user reviewing his or her collected events, being affected by their associated sentiment.	10
1.8	Pipeline for the analysis of social patterns	11
2.1	Example of temporal segmentation of an egocentric sequence	18
2.2	General scheme of the SR-Clustering method	21
2.3	Graph obtained after calculating similarities of the concepts of a day's lifelog and clustering them	23
2.4	Example of the final semantic feature matrix obtained for an egocentric sequence	24
2.5	Example of extracted tags on different segments	25
2.6	General scheme of the semantic feature extraction methodology.	26
2.7	Change detection by the different algorithms implemented	28
2.8	Different segmentation results obtained by different subjects	33
2.9	LCE and GCE of the manual segmentations	34
2.10	Correlation of the LCE and GCE among sets	35
2.11	LCE and GCE of the manual segmentations - excluding the camera wearer segmentation	36
2.12	Correlation of the LCE and GCE among sets - excluding the camera wearer segmentation	37
2.13	Examples of different segments and the top 8 found concepts	38
3.1	Example of images recorded by one of the camera wearers.	44
3.2	Pipeline of the proposed model.	50
3.3	Average number of images per recorded egocentric photo-stream. We give the number of collected days per user between parenthesis.	53

3.4	Histograms showing the occurrence of activities throughout the days	56
3.5	Visualization of the obtained classification results	57
3.6	Illustration of the proposed Topics-based model	58
3.7	Illustration of a photo-stream/document described by proportion of topics	60
3.8	Average number and variance of egocentric images per recorded photo-stream for the 7 users	63
3.9	Example of selected images throughout some of the recorded photo-streams of User1.	63
3.10	Number of <i>Routine</i> and <i>Non-Routine</i> days for each user (U) in the <i>EgoRoutine</i> dataset.	64
3.11	Example of given photo-streams, sample images at several time-slots, their representative topics, and the concepts that compose them.	71
3.12	Affinity matrix (DTW) and the later discrimination as Routine or Non-Routine related days (SpClust) of collected days by users 3 and 7	72
4.1	Examples of images of each of the proposed food-related categories present in the introduced EgoFoodPlaces dataset.	77
4.2	The proposed semantic tree for food-related scenes categorization.	84
4.3	Total number of images per food-related scene class.	86
4.4	Illustration of the variability of the size of the events for the different food-related scene classes.	87
4.5	Visualization of the distribution of the classes using the t-SNE algorithm.	88
4.6	Mean Silhouette Score for the samples within the studied food-related classes	88
4.7	Confusion matrix with the classification performance of the proposed hierarchical classification model.	94
4.8	Examples of top 5 classes for the images in the test set	95
4.9	Illustration of detected food-related events in egocentric photo-streams	97
5.1	Examples of Positive, Negative and Neutral images.	106
5.2	Architecture of the proposed method combining global and semantic features	107
5.3	Examples of the automatic event sentiment classification	109
5.4	Sketch of the proposed method for semantic concepts analysis	110
6.1	Architecture of the proposed model	118
6.2	Samples of the clusters obtained from recorded days	121
6.3	Obtained social profiles as a result of applying our method	125
7.1	Future directions of research	130

List of Tables

1.1	Comparison of some popular wearable cameras.	6
2.1	Table summarizing the main characteristics of the datasets used in this work:	30
2.2	Average FM results of the state-of-the-art works on the egocentric datasets	39
2.3	Average FM score on each of the tested methods using our proposal of semantic features on the dataset presented in (Poleg et al., 2014).	40
3.1	Description of the collected Egoroutine dataset by 5 users.	52
3.2	Summary of the labelling results for the Egoroutine dataset.	53
3.3	Performance of the different methods implemented for the discovery of routine and non-routine days.	55
3.4	Total number of recorded days and collected images per user.	62
3.5	Summary of the agreement among the 6 individuals that labelled the collected photo-streams into Routine or Non-Routine related days.	64
3.6	Results of the proposed pipeline and baseline models	67
3.7	Results of the proposed pipeline for the best setting of the parameters	68
3.8	Example of detected concepts in a given recorded day by User 1	68
3.9	Comparison between our previous work and the model here proposed	72
4.1	Food-related scene classification performance.	93
4.2	Classification performance at different levels of the proposed semantic tree for food-related scenes categorization.	93
5.1	Different image sentiment ontologies.	103
5.2	Description of the UBRUG-EgoSenti dataset.	108
5.3	Performance results achieved at image and event level.	108
5.4	Examples of clustered concepts based on their semantic similarity, initially grouped following the distance computed by the WordNet tool.	111
5.5	Parameter-selection results	113
5.6	Test set results	114

6.1	Average Precision, Recall, and F-Measure result for each of the tested methods on the extended test-set composed by egocentric images. . .	123
6.2	This table shows the social behavioural traits obtained from the detected social interactions for the different camera wearer.	124

