

# **ESR12. Al for Dementia Care**

### 2<sup>nd</sup> Doctoral Seminar

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## Introduction: AI for Dementia Care

# What's dementia?

 Syndrome in which there is a deterioration in cognitive functioning beyond what might be expected from normal ageing [1]

### Why dementia?

- Behavioural changes strongly correlated with the degree of functional and cognitive impairment [2].
   Behavioral and Psychological Symptoms of Dementia (BPSD): agitation, aberrant motor behaviour, anxiety, irritability, depression, apathy, delusions, changes in sleep or appetite [2].
- One of the major causes of dependency among older people [3]



Data source: WHO [1]

[1] World Health Organization https://www.who.int/news-room/fact-sheets/detail/dementia (accessed April 25, 2022)

[2] Joaquim Cerejeira, Luisa Lagarto, and Elizabeta Blagoja Mukaetova-Ladinska. "Behavioral and psychological symptoms of dementia". In: Frontiers in neurology 3 (2012), p. 73.

[3] Global status report on the public health response to dementia. World Health Organization (2021)





# Al for behaviour analysis from unobtrusive sensor data

Goal: Development of AI methods for **measuring** the **behaviours** of care home residents with **dementia** using **unobtrusive sensors (depth maps)** 

In order to:

- 1. Detect and **measure behavioural changes** indicative of dementia (BPSD) in the mid and long term (weeks, months, years)
- 2. Provide **assistance with ADLs** for people with dementia in the short term (seconds, minutes)





RQ1. How and what behaviours can be measured RQ2. AI methods and data RQ3. Measuring behaviour changes in the long term

RQ4. Providing assistance with ADLs in the short term

BPSDs taxonomy and measurement methods

BPSDs change detector

**ToiletHelp** 









# Goal

Step-by-step guidance for people with mild dementia in the toilet using a depth camera

# RQ4. Providing assistance with ADLs in the short term

- 1. How can the model be designed to **capture** the toilet procedure in such a way that it is possible to detect deviations indicative of the need for assistance?
- 2. How **interaction** with the user must be designed to be **effective** for people with dementia?
- 3. How can the effectiveness of the interaction be **measured** for the target group?







# What was done until the last Doctoral Seminar in Aachen?

- 1. Focus groups with health professionals: How the system should communicate with people with dementia?
- 2. Prototype developed with handwashing and acknowledgement module
- 3. Validation in the lab: functional testing
  - 98.5% avg. accuracy in action recognition
  - Interaction: 100% in fixed scenarios, 8/10 correct in open scenarios
- 4. New visualizations and audio

### 2 publications:

"RITA: A privacy-aware toileting assistance designed for people with dementia" by Irene Ballester, Tamar Mujirishvili and Martin Kampel, In *Proceedings of the 15th EAI International Conference on Pervasive Computing Technologies for Healthcare,* 2021

"Automated vision-based toilet assistance for people with dementia" by Irene Ballester and Martin Kampel, AHFE 2022 - 13th International Conference on Applied Human Factors and Ergonomics, July 24-28, 2022, New York, USA

+ collaboration in the DIANA project





# What has been done since then until now?

Further improvement of the prototype and evaluation of the system with **real data:** 

- Evaluation of the toilet action recognition module on 50 sequences with 20 people with dementia
- 2. Validation of the **interaction module** involving 30 older adults and 14 care staff members



All this work to be re-submitted:

Ballester I., Gall M. Brandstötter B., Kampel M. "ToiletHelp: an assistive technology to guide people with dementia in the toilet", ACM Journal on Interactive, Mobile, Wearable and Ubiquitous Technologies (IMWUT) Deadline: 15 May

VISUAAL



### Methodology

Action is detected correctly if: H1 < fID\_detected – fID\_gt< H2

### **Results**

Toilet Module evaluated against human annotators for action recognition with 20 people with dementia and 50 sequences, with 182 actions:

- Hospital group accuracy: **85%**
- Day center accuracy: **70%**
- Overall accuracy: 81%

Dataset	% of successfully recognized actions												
	Next to the toilet bowl		Sat down on the toilet bowl		Stood up the toilet	from bowl	Next the bas	to sin	Total				
	N actions	Acc.	N actions	Acc.	N actions	Acc	N actions	Acc.	N actions	Acc.			
S1 (N=1)	6	33%	10	90%	10	90%	7	71%	33	76%			
S2 (N=1)	10	70%	10	100%	10	90%	10	100%	40	90%			
S3 (N=1)	10	100%	10	80%	10	100%	6	100%	36	94%			
Hospital (N=3)	26	73%	30	90%	30	93%	23	92%	109	87%			
Day center (N=17)	17	59%	19	95%	19	68%	18	56%	73	70%			
All (N=20)	43	67%	49	92%	49	84%	41	76%	182	80%			



# Evaluation of the interaction module

### Methodology

Install the prototype in a semi-public toilet and ask caregivers and older adults about their opinion via questionnaires

### Conclusions

End-users (30 older adults, 17 in DC1 and 13 in DC2)

- Increased feeling of safety and independency and not feeling feel afraid nor annoyed
- Participants with dementia affirm they understand the instructions
- Changes in rating as the user uses the system more times

Caregivers (14 professionals, 4 in DC1 and 10 in DC2)

- Positive rating in terms of being useful for end-users and reducing caregivers' workload
- Positive validation of the interaction modalities

Poster accepted at: Ballester I., Gall M., Kampel M. "Design of a Toileting Assistive System for People with Dementia", WISH Symposium at the ACM CHI Conference on Human Factors in Computing Systems, April 2023





# What's next?

- Rewrite IMWUT paper: add F1-score, improve clarity of explanations
- (Draft of) next steps
  - 1. Improved action recognition module
    - Enhancing the recognition accuracy of the actions "next to the toilet" and "next to the basin" by computing orientation towards the object of interest
    - Enabling learning to adapt to individual rhythms

### 2. Interaction effectiveness evaluation

Measure how effective the interaction is

• Next paper: CHI 2024 (Submission deadline: September 2023). IMWUT/UbiComp 2024? (Submission deadline: November 2023).















# RQ1. How and what behaviours can be measured

- 1. Which BPSDs can be measured from data collected continuously by unobtrusive sensors?
- 2. Which **metrics** (e.g., frequency, intensity and duration) can be used to quantify each of those behaviours?

# RQ2. AI methods and data

- 1. Which **AI methods** are the most suited to model the different behaviours?
- 2. What **type of data** is best suited for modelling each behaviour?
- 3. How can AI methods for behaviour measuring be designed to allow **inter-resident variabilities** to be taken into account in the model to be able to generalize to other subjects?





# **BPSDs Taxonomy and AI methods**

# What has been done?

### **Preliminary analysis**



### Literature research and draft of potential BPSDs

BSPDs	Metrics
<ul> <li>Twisted day-night rhythm</li> <li>Sundowning phenomenon</li> <li>Daytime sleepiness</li> <li>Sleep restlessness</li> </ul>	Active time while being in the room • During day time • During night time Walking speed
Decreased general activity level Increase in incident falling frequency Eloping attempts Wandering (?)	<ul> <li>Time in the room</li> <li>During day time</li> <li>During night time</li> <li>Time on bed</li> <li>Number of falls</li> <li>Time sitting, laving, standing</li> </ul>

and mainly.... struggling with DATA COLLECTION





## **BPSDs Taxonomy and AI methods**

# What's next? (I)

- 1. Ethical approval. Expected: May 2023
- 2. Start of data collection in a care facility in Austria. Planned: May 2023
  - 10 participants already accepted (with and without dementia)
  - Depth maps in single rooms, cognitive status (MMSE), medical and care documentation
- **3. Measuring BPSD using depth maps.** (In collaboration with PD Dr. med. Thomas Münzer) Goal: Research whether BPSD can be measured to detect dementia in care facility residents
  - Develop pre-processing methods for false detections, lost tracking, > 1 person in the room
  - Develop taxonomy: behaviours (such as gait speed [4]) + metrics + measuring methods
  - Define and perform statistical analysis (cofounding factors [4], Poincaré Plot [5])
     Paper target deadline: Early-Mid 2024. e.g., Journal of Biomedical and Health Informatics, The Lancet Digital Health, Journals of Gerontology

[4] Liu, Y., Zhang, G., Tarolli, C. G., Hristov, R., Jensen-Roberts, S., Waddell, E. M., ... & Katabi, D. (2022). Monitoring gait at home with radio waves in Parkinson's disease: A marker of severity, progression, and medication response. *Science Translational Medicine*, 14(663), eadc9669.

[5] Urwyler, P., Stucki, R., Rampa, L., Müri, R., Mosimann, U. P., & Nef, T. (2017). Cognitive impairment categorized in community-dwelling older adults with and without dementia using in-home sensors that recognise activities of daily living. *Scientific reports*, 7(1), 1-9.





# What's next? (II)

### 4. Video Anomaly Detection for Risky Behaviour of People with Dementia

Method: Apply and develop OCC [6,7] or/and unsupervised [8] autoencoders and/or GAN approaches

### **1. Behaviours of People with Dementia in the Toilet**

- Challenge: large unlabelled data of people in the toilet
- Potential solution: use ToiletHelp to label normal sequences

### **2.** Behaviours of Residents in their room (TBD)

Abnormal behaviours "labelled" from care documentation

### Paper target deadline: End 2023. ECCV2024? ICIP 2024?



[6] Schneider, P., Rambach, J., Mirbach, B., & Stricker, D. (2022). Unsupervised anomaly detection from time-of-flight depth images. In *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition* (pp. 231-240).

[7] Mishra, P. K., Iaboni, A., Ye, B., Newman, K., Mihailidis, A., & Khan, S. S. (2023). Privacy-protecting behaviours of risk detection in people with dementia using videos. *BioMedical Engineering OnLine*, 22(1), 1-17.

[8] Zaheer, M. Z., Mahmood, A., Khan, M. H., Segu, M., Yu, F., & Lee, S. I. (2022). Generative cooperative learning for unsupervised video anomaly detection. In *Proceedings of the IEEE/CVF* Conference on Computer Vision and Pattern Recognition (pp. 14744-14754).











### **BPSDs change detector**

# Goal

# Detect and measure behavioural changes indicative of dementia in the mid and long term







# RQ3. Measuring behaviour changes in the long term

- 1. How can **changes** in behaviours be identified and measured?
- 2. How are **time horizons** defined for the detection of behavioural changes in the context of dementia through data collected by unobtrusive sensors?
  - 1. How long does an individual have to be observed for their statistically "normal" behaviour (routine) to be modelled?
  - 2. How long does a **trend have to be consistent over time** to be considered a change in behaviour statistically?
  - 3. How can the model be designed to account for **variability** that depends, for example, on the day of the week or the year's season?





# BPSDs change detector

# How?









### **BPSDs change detector**

# How?

- Anomaly detection for time series data
   Anomaly if (prediction reality) > ε
- Forecasting methods using RNNs (LSTM), Transformers

# **Evaluation**

Compare changes detected with changes in MMSE or events recorded in the care documentation

Paper target deadline: Summer 2024. ICML 2025? PerCom 2025?

# $D \rightarrow \mathbf{X}(\mathbf{n})$



LSTM Autoencoder

Input Sequence

Reconstructed Sequence

Figure extracted from Trinh, H. D., Zeydan, E., Giupponi, L., & Dini, P. (2019). Detecting mobile traffic anomalies through physical control channel fingerprinting: A deep semi-supervised approach. *IEEE Access*, *7*, 152187-152201.





### Thesis contribution

1. Taxonomy of BPSDs and methods to measure them via unobtrusive monitoring.

2. Method for the identification and measurement of dementiarelated behavioural changes.

**3. Assistive system** to support people with dementia in using the toilet.













# Estimated timeline

isuA

	2021			2022			2023				2024				2025	
	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1
ToiletHelp - Interaction with people with dementia			Pervasive Health Paper													
Collaboration with DIANA project		MidTerm Deliv.						End Project Deliv.								
ToiletHelp - Functional Evaluation				AHFE Paper												
Proficiency Evaluation					Research Proposa Subm.	n I Present.	/									
ToiletHelp - Evaluation									IMWUT Paper							
ToiletHelp - Improved recognition + interaction evaluation										CHI Paper						
Video Anomaly Detection for people with dementia											CV paper					
Measuring BPSDs using depth maps													Medical journal paper			
Automated change measurement in behaviour														ICML/Per Com Paper		
Secondments					Alicante						TBD					
Thesis writing																Defense
AL				Phl ir	D Ser Aacl	minar hen		٦	loday	Pl	nD Se in Viei	minar nna				





# Thank you!

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# ViBRANT Workshop



Part of <u>EICS 2023</u>, the 15th ACM SIGCHI Symposium on Engineering Interactive Computing Systems at Computational Foundry, Swansea, Wales, UK - June 27–30, 2023

#### **Important Dates**

Submission deadline: 29.05.2023

Notification of acceptance: 09.06.2023

Planned workshop day: 27.06.2023 (half-day)

## **Call for Papers**

- 1. (Out-of-proceedings) research highlights and work-inprogress (2-page extended abstract)
- 2. Archival original work (published by Springer LNCS)
- video-based technology for health monitoring
- fair and privacy-aware systems
- integration of engineering issues in the design process of video-based AAL
- legal requirements and privacy issues for data collection and processing in care
- user acceptance criteria for video-based AAL
- GDPR requirements for video-based AAL solutions
- best practices of interdisciplinary collaborations between legal, technical, social, health sciences



### **Actions evaluated**

- 1. The person is about to sit
- 2. The person sat down
- 3. The person stood up
- 4. The person is about to wash their hands

### Inclusion criteria:

- Participant sits on the toilet
- Only one person in the room
- The room is empty at the beginning and the end of the sequence

### **Description of participants:**

- 6 male, 14 female
- Age range: 60-97
- 19 participants with dementia, 1 without
- 6 participants use walking aids







# Action is detected correctly if:

# H1 < fID\_detected – fID\_groundtruth < H2

- **fID\_detected**: frame in which the system detects the action
- **fID\_groundtruth**: mean of the frames labelled by 3 annotators: mean(An1, An2, An3)

### How to calculate the thresholds H1, H2?

- H1= uncertainty\_action + min\_system\_delay H2= uncertainty\_action + max\_system\_delay
  - Uncertainty\_action = SD(An1, An2, An3) for each action-patient
  - System\_delay = frames the system waits to ensure a robust recognition





### Success assessment: Sensor vs. annotators

Example: S1 (Hosp), Action=Sitting down

H1 < fID\_detected – fID\_groundtruth < H2

	An1	An2	An3	Avrg = GT	Std Dev	Sensor	Sensor-GT	H2-H1	Success?
R 1	408	413	419	413.3	5.5	400	-13.3	35.0	
R 2	343	340	349	344.0	4.6	344	0.0	33.2	
R 3	455	453	458	455.3	2.5	400	-55.3	29.0	
R 4	436	433	437	435.3	2.1	440	4.7	28.2	
R 5	458	448	455	453.7	5.1	440	-13.7	34.3	
R 6	442	436	439	439.0	3.0	428	-11.0	30.0	
R 7	452	452	461	455.0	5.2	424	-31.0	34.4	
R 8	432	432	455	439.7	13.3	436	-3.7	50.6	
R 9	456	456	460	457.3	2.3	456	-1.3	28.6	
R 10	368	363	373	368.0	5.0	372	4.0	34.0	
Average	425.0	422.6	430.6	426.1	4.9	414.0	-12.1	33.7	Accura
Std. Dev.	40	40.0	39.4	39.6		34.9	18.6	6.5	90%





Evaluation of the interaction module		18	1st time runs			All test runs		
End-users in DC1 (St. Gallen, CH)		Yes	Neutral	No	Yes	Neutral	No	
	Q1. The TH makes me feel safer.	11	3	2	22	3	2	
N = 17 older adults attending 1-2	Q2. The TH makes me feel more independent.	11	0	5	19	3	5	
days/week the day care centre	Q3. The TH makes me feel afraid.	1	0	15	1	0	26	
days, week the day eare centre	Q4. I find the TH annoying.	1	1	14	1	3	23	
• 11 women. 6 men	Q5. I would like to have the TH in my private bathroom	. 2	1	13	5	1	21	
,	Q6. I have understood all the instructions.	14	2	1	25	2	1	
<ul> <li>Dementia severity (6 measured by MMSE, 11 measured by MoCA) :</li> <li>Mild: 11</li> </ul>	Q1. The TH makes Q2. The TH makes Q3. The TH makes Q4. I find th me feel safer. me feel more me feel afraid. TH annoying independent.	e Q5. I ç. have priva	would like to the TH in my te bathroom.	Q6. I under instru	have rstood al ictions.	l the		
Moderate: 5	σ 6-	_				Answ Ne	<b>er</b> utral	
						No		
<ul> <li>Normal cognitive functioning: 1</li> </ul>		_				Yes	5	
<ul> <li>Mean age = 78,6. SD = 7.2, max. = 86, min. = 60</li> </ul>	<sup>2</sup> 2-2-10-10-10-10-10-10-10-10-10-10-10-10-10-	-	н.					

Mean score of affinity for new • technologies = 1.5(0-2)





# test run

### **Evaluation of the interaction module**

### End-users in DC2 (Coimbra, PT)

N = 13 older adults attending the day centre daily

- 9 women, 4 men
- No cognitive status reported
- Age distribution
  - 71-80 years: 3 participants
  - 81-90 years: 6 participants
  - +90 years: 4 participants
- Mean score of affinity for new technologies = 4.2 (SD=0.9, scale=1-5)

### (1=I do not agree at all, 10= Totally agree)

	1st tim (N=	e runs 13)	2-4 tim (N=	e runs =8)	All test run (N=21)		
	Mean	SD	Mean	SD	Mean	SD	
Q1. The TH would make me feel safer.	8,5	2,7	9,3	0,7	8,8	2,2	
Q2. The TH would make me feel more independent.	8,5	4,2	8,8	0,9	8,6	3,3	
Q3. The TH would make me feel comfortable.	8,3	2,5	8,9	0,8	8,5	2,0	
Q4. The TH would make me feel afraid.	2,2	2,5	1,3	0,7	1,8	2,1	
Q5. I find the TH annoying.	1,4	0,9	1,3	0,5	1,3	0,7	
Q6. I would prefer to be assisted by the TH rather than by a caregiver.	5,7	3,5	6,0	3,6	5,8	3,5	
Q7. I would like to have the TH in my private bathroom.	7,9	2,9	8,5	1,9	8,1	2,6	





### **Evaluation of the interaction module**

Care staff in DC1 and DC2		Total (	N=14)	DC1 (N=4)		DC2 (N=10)	
		Mean	SD	Mean	SD	Mean	SD
N=14 care staff (N=4 from DC1 and N=10 from DC2)	Q1. How useful would you rate the use of the TH for older adults with cognitive impairment?	8,1	1,7	7,5	0,6	8,4	2,0
All women	Q2. How useful would you rate the use of the TH for reducing the workload for healthcare workers?	8,1	1,9	6,3	1,3	8,7	1,6
<ul> <li>Affinity for technology (scale: 1-5)</li> <li>Total: mean=4.1, SD=0.9</li> </ul>	Q3. How adequate do you find the interaction modalities used in the TH to guide the user?	8,2	1,3	7,5	1,3	8,5	1,3
<ul><li>DC1: mean=3.3, SD=0.8</li><li>DC2: mean=4.4, SD=0.7</li></ul>	Q4. Is the vocabulary in the TH adequate to guide the user?	8,8	1,3	8,5	1,9	8,9	1,0
DC1 DC2	Q5. Is the tone used in the TH adequate to guide the user?	8,6	1,2	8,3	1,7	8,7	1,1
Aged         Work experience         Aged         Work experience           21-30 y.         0         <5 y.	Q6. Are the videos used in the TH adequate to guide the user?		0,9	8,8	1,0	8,3	0,9
41-50 y.       1       10 to 16 y.       0 $41-50$ y.       3       10 to 16 y.       0 $51-60$ y.       3       16 to 20 y.       1       51-60 y.       1       16 to 20 y.       2 $61-70$ y.       0       >20 y.       3       61-70 y.       1       >20 y.       4	Q7. How intrusive do you think the TH is compared to a caregiver guiding the person with dementia?	4,9	3,1	7,7	1,5	4,1	3,0



