





Presenters: Stephanie Homan, Ph.D. & Samantha Weber, Ph.D.

Research groups: B. Kleim, S. Olbrich

Ipsilon Research Meeting, 24 November 2025









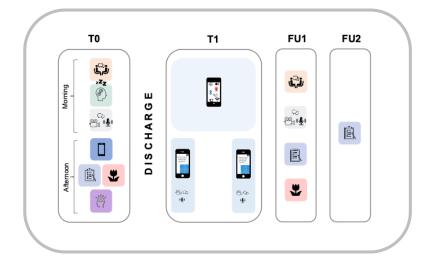




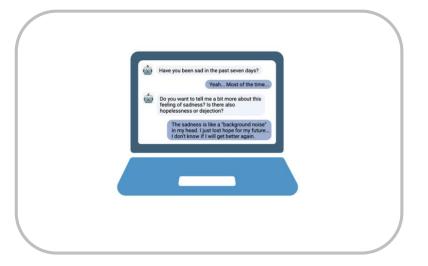
#### DIGITAL TECHNOLOGIES IN SUICIDE RESEARCH

#### **CHATBOTS/AI IN RESEARCH**

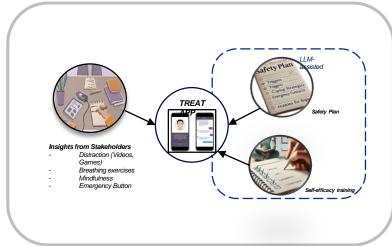
#### **PREDICT**



#### **EXPERIMENT**

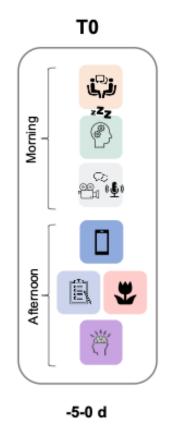


#### **TREAT**



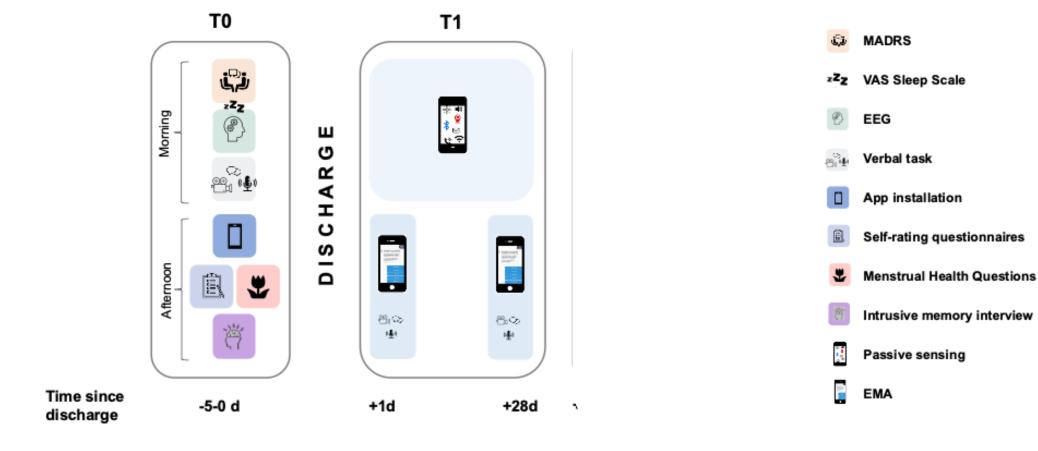


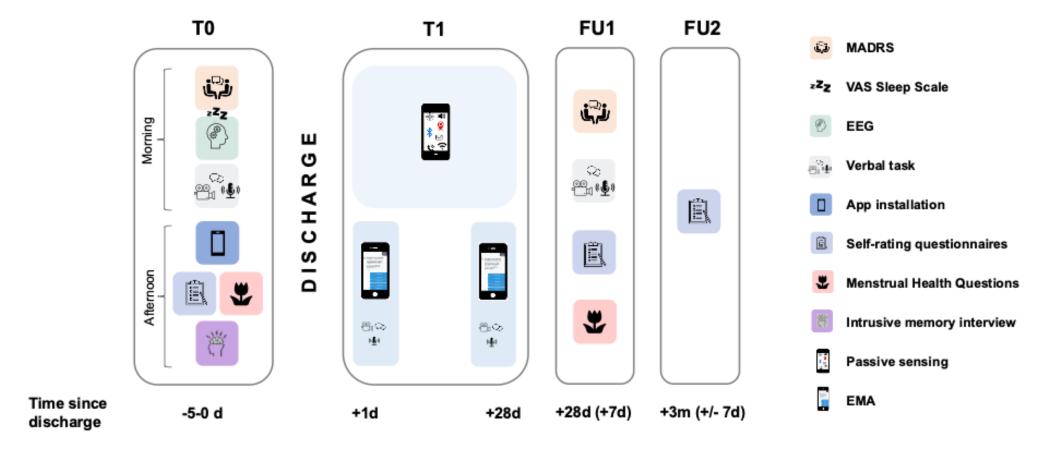
Target population: Transdiagnostic inpatients with current/past suicidal thoughts and behaviors

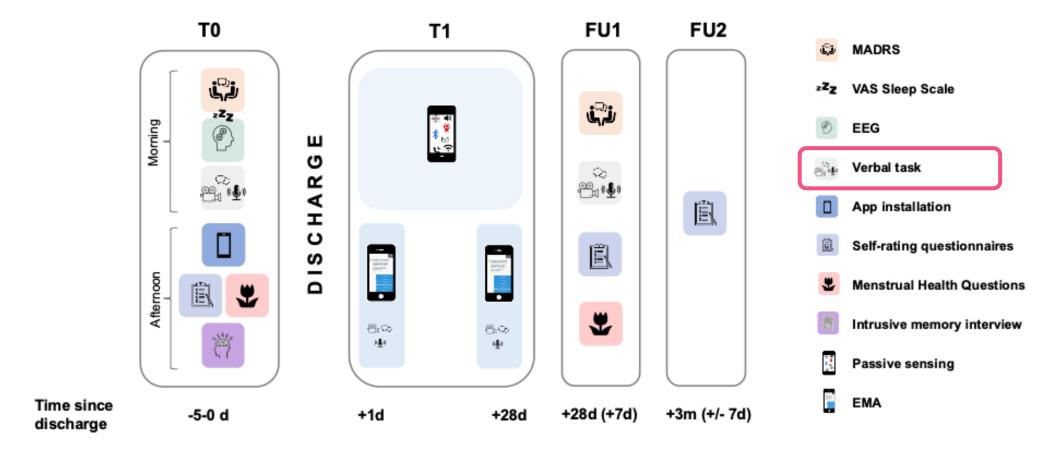


Time since discharge

- MADRS
- zZz VAS Sleep Scale
- Verbal task
- App installation
- Self-rating questionnaires
- Menstrual Health Questions
- Intrusive memory interview
- Passive sensing
- EMA





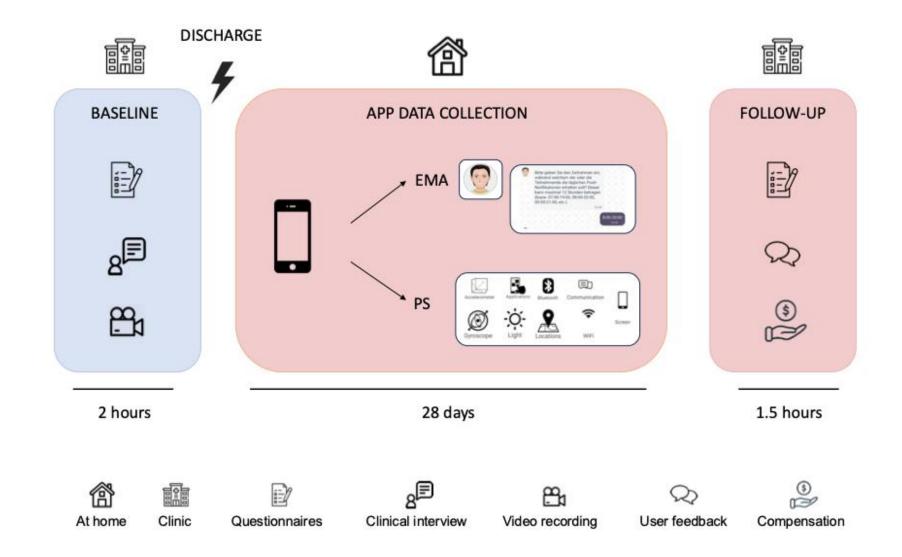


# Non-verbal expressions in suicide risk prediction

- Measurable suicide-risk signals: slowed/monotone speech, longer pauses, reduced facial and movement expressivity.
   (Galatzer-Levy et al., 2021; lyer et al., 2022)
- Multimodal (audio + video + text) models outperform singlemodality approaches. (Alghowinem et al., 2023)
- → Promising **supplementary** markers for suicide-risk detection, but still **early-stage**; larger longitudinal studies and clinical validation are needed. (Parsapoor et al., 2023; Dhelim et al., 2023; Marie et al., 2025; Alghowinem et al., 2023)

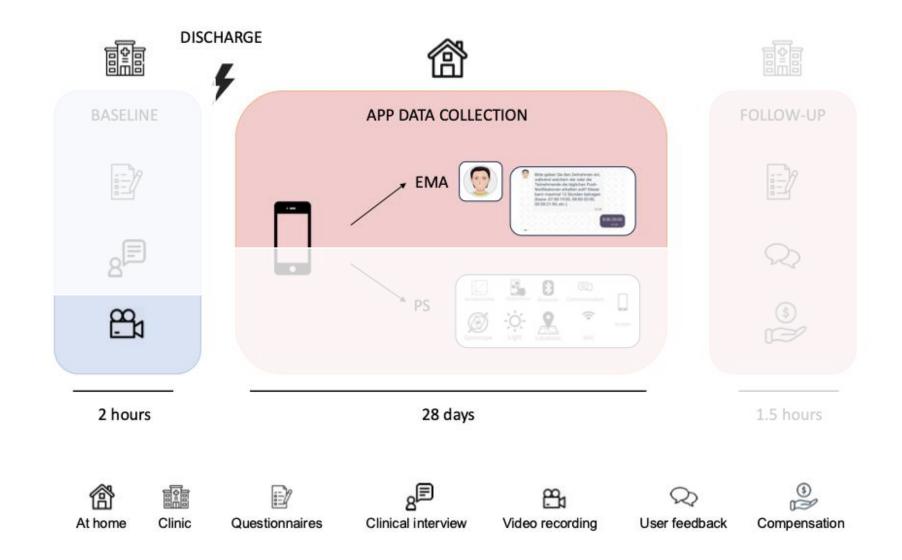






Sels et al., 2021





Sels et al., 2021

#### **Negative Past**

"Can you remember an event that was upsetting or stressful for you?"

#### The linguistic task



1 min.



**Positive Future** 

"Is there an event in the coming days that you are looking forward to?"





1 min.



1 min.



1 min.



1 min.



#### **Negative Future**

"Is there an event in the coming days that you are not looking forward to or even dreading?"

...

#### The linguistic task





1 min.



1 min.



1 min.



1 min.



1 min.



#### **Negative Past**

"Can you remember an event that was upsetting or stressful for you?"

#### **Positive Future**

"Is there an event in the coming days that you are looking forward to?"

#### **Negative Future**

"Is there an event in the coming days that you are not looking forward to or even dreading?"

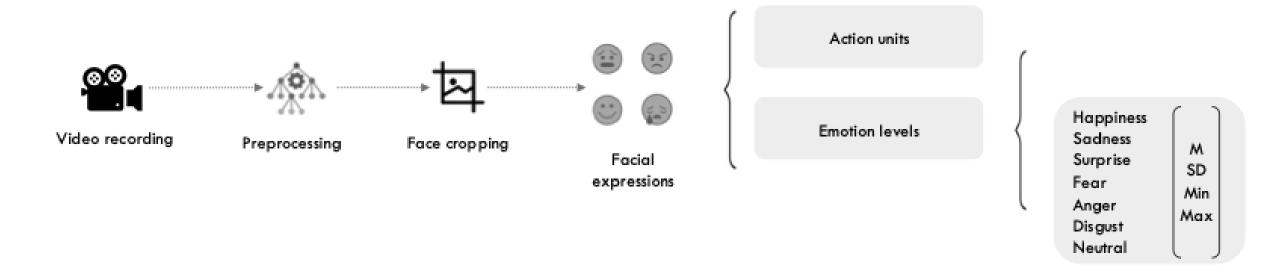
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## Digital biomarker extraction





Victoria Müller



## The sample

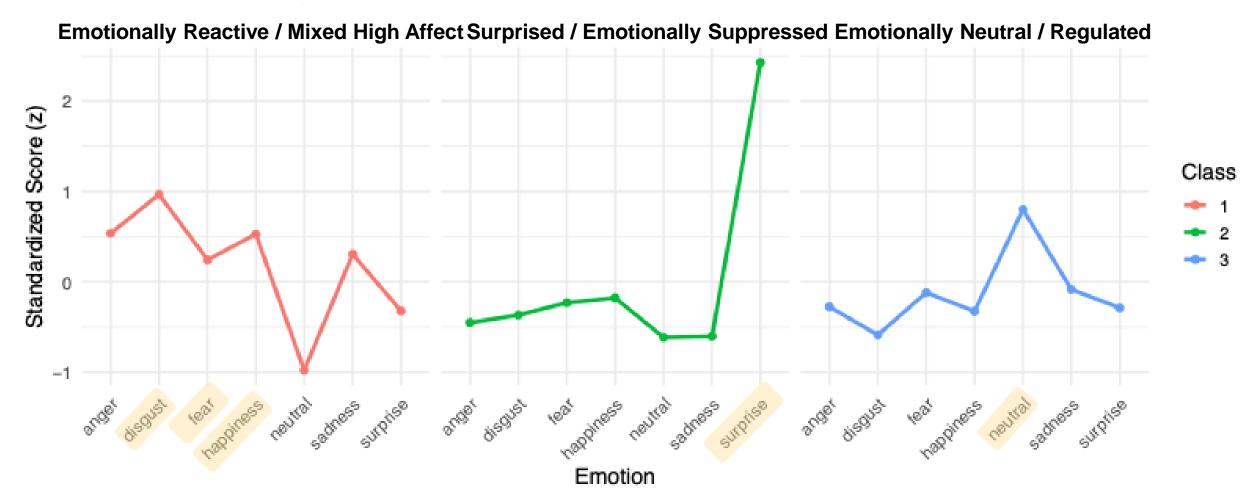
Variable	${f N}$	$N = 38^{1}$
Age	36	33.72 (12.08)
NA		$\overset{\cdot}{2}$
Sex	35	
female		21 / 35 (60%)
male		14 / 35 (40%)
NA		3
Rehospitalized	24	
Yes		10 / 24 (42%)
No		14 / 24 (58%)
NA		14
Severity of suicidal ideation (BSS total score)	23	12.61 (10.68)
NA		15
History of suicide attempts (BSS item 20)	23	
Never		8 / 23 (35%)
Once		4 / 23 (17%)
Twice or more than twice		11 / 23 (48%)
NA		15

<sup>&</sup>lt;sup>1</sup>Mean (SD); n / N (%)

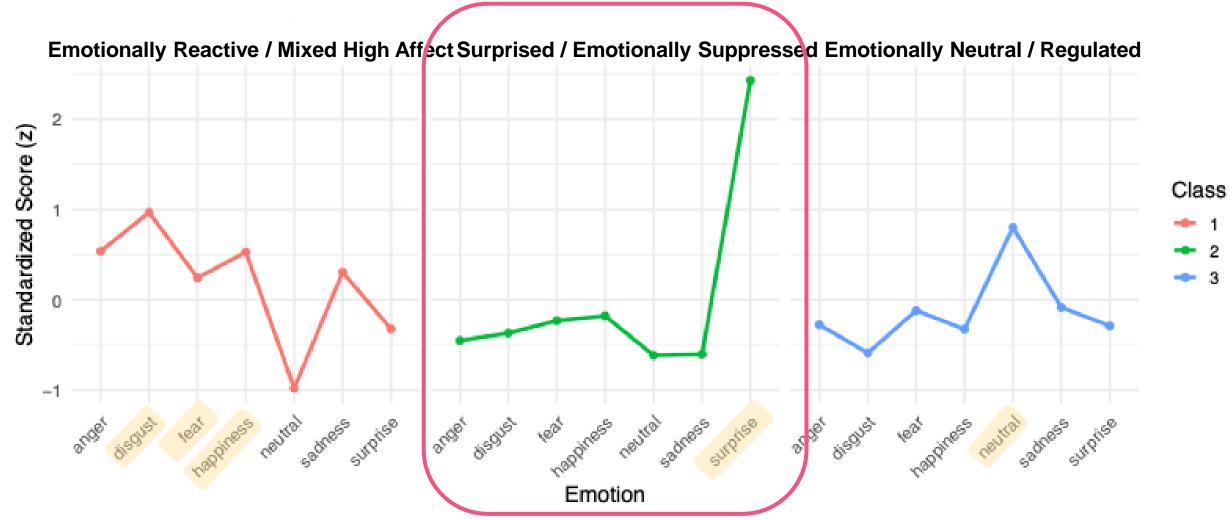
Note. BSS, Beck Scale for Suicide Ideation; Baseline data was not available for all participants.



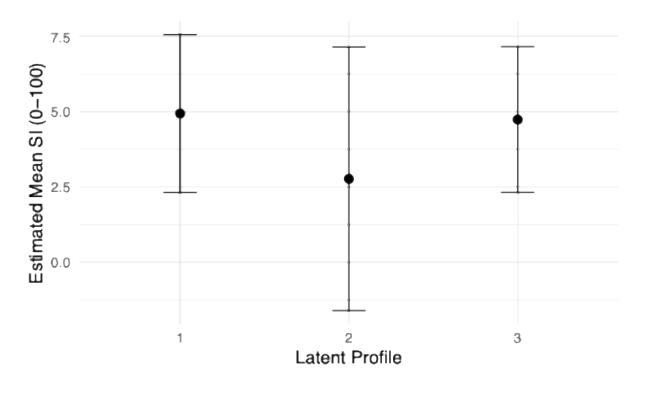
# Are there subgroups that experience unique combinations of emotional intensity?



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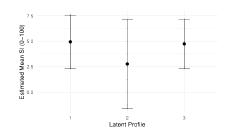


## Are these subgroups related to suicidal ideation post-discharge?



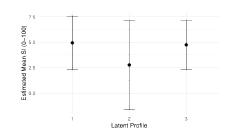
- → No significant differences in suicidal ideation across latent profile classes within the negative past condition.
- → High standard errors suggest imprecise estimates, likely due to limited sample size or variability

## Digital biomarkers as promising correlates for suicidal ideation?



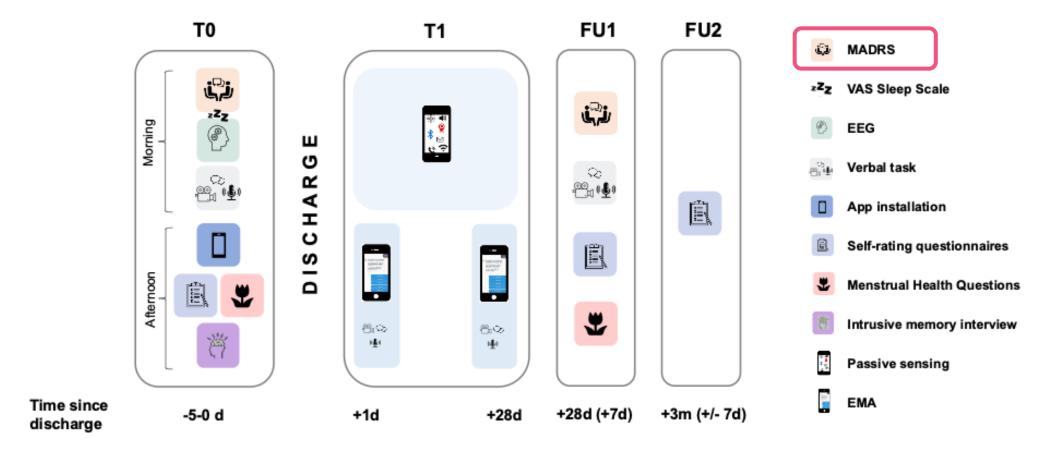
- There is no easy answer.
- Digital biomarkers might have potential for suicide risk prediction and for complementing clinical practice.
- Before we can use these biomarkers in eMental Health, we need to **understand them better** (i.e., include action units, get clinical ratings of facial expressions).

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- Digital biomarkers might have potential for suicide risk prediction and for complementing clinical practice.
- Before we can use these biomarkers in eMental Health, we need to **understand them better** (i.e., include action units, get clinical ratings of facial expressions).
- And: Replication in a larger sample → MULTICAST-PRED





## Using Artificial Intelligence for symptom assessment and monitoring?



## MONTGOMERY and ASBERG Depression Rating Scale (MADRS)

Each item rated from 0 - 6 [0 = no symptoms]

**General**: Hopelessness, dejection, and despair expressed in language, facial expressions, and posture.

**Sadness**: Patient's reports of a depressed mood, including discouragement, dejection, feelings of helplessness, and hopelessness.

**Tension:** *Indefinable feeling of discomfort as well as irritability, restlessness, inner excitement up to anxiety and panic.* 

**Sleep**: Subjective experience of reduced sleep duration or sleep depth.

**Appetite**: The feeling of having less appetite compared to the normal state.

**Concentration**: Difficulties in concentrating

**Lack of Drive**: Difficulties in starting or heaviness in beginning and carrying out everyday activities.

**Emotional Numbness**: The subjective feeling of reduced interest in the environment or activities that previously brought joy. **Pessimistic Thoughts**: Feelings of guilt, inferiority, self-blame,

etc.

**Suicide**: Feeling that life is no longer worth living, natural death would be a relief, thoughts of suicide, and preparation for suicide.



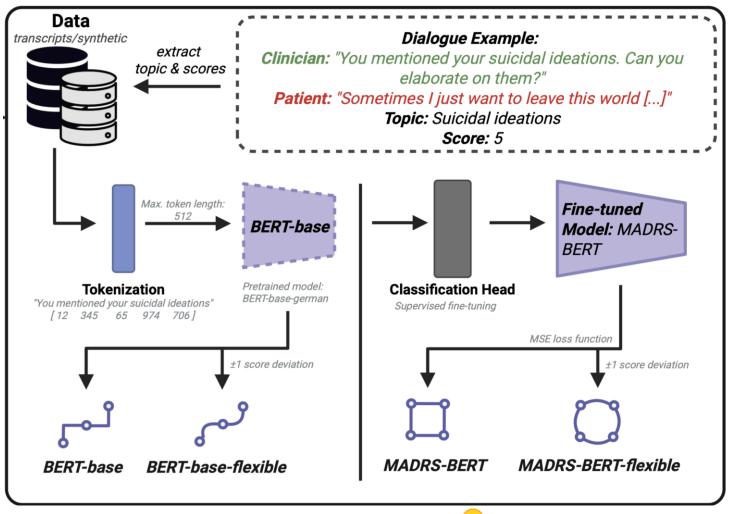


**Trained Personnel** 





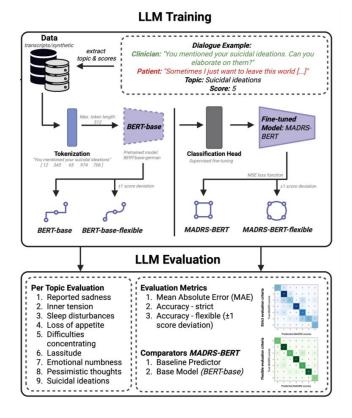
# Assessing depressive Symptoms using AI – LLM Training and Evaluation LLM Training





google-bert/bert-base-german-cased

## Assessing depressive Symptoms using AI – LLM Training and Evaluation



## Table 1 | Comparison of *MADRS-BERT* and Baseline Predictor (Mean Regression Model) Performance Across MADRS items

MADRS Item	Mean	MAE ↓ (±std)		
	<b>MADRS Score</b>			
		Baseline predictor	MADRS-BERT	
Reported sadness	3.0	1.7	0.9 (±0.04)	
Inner tension	3.0	1.5	<b>0.7</b> (±0.15)	
Sleep disturbances	2.9	1.7	0.9 (±0.16)	
Loss of appetite	2.8	1.8	0.8 (±0.08)	
Difficulties concentrating	2.9	1.7	0.8 (±0.22)	
Lassitude	2.8	1.8	0.8 (±0.21)	
Emotional numbness	2.8	1.8	1.0 (±0.26)	
Pessimistic thoughts	2.9	1.6	0.8 (±0.14)	
Suicidal ideations	2.9	1.7	0.8 (±0.16)	
Total	2.89	1.70	0.83	

The table reports the Mean Score, and Mean Absolute Error (MAE) for the baseline predictor and the fine-tuned model (*MADRS-BERT*) across all nine MADRS items. The baseline predictor assigns the mean MADRS score per topic as the predicted value, serving as a naive statistical reference. MAE quantifies the prediction error, with lower values indicating better performance. Bold numbers highlight the best results.

MAE < 1 is comparable with our clinician performance

Weber et al. 2025 npj Digital Medicine 08.12.2025

## Assessing depressive Symptoms using AI – LLM Training and Evaluation

Table 2 | Performance metrics of the fine-tuned *MADRS-BERT* and *BERT-base* models under strict and flexible criteria for accuracy

	MADRS-BERT		BERT-base		
MADRS Item	Accuracy ↑ [%] Flexible	Accuracy ↑ [%] Strict	Accuracy ↑ [%] Flexible	Accuracy ↑ [%] Strict	
Reported sadness	80 (±0.03)	40 (±0.07)	29 (± 0.04)	14 ( ± 0.03)	
Inner tension	88 (± 0.06)	49 (±0.10)	25 (± 0.04)	12 ( ± 0.07)	
Sleep disturbances	82 (±0.08)	44 (±0.09)	30 ( ± 0.07)	17 (±0.07)	
Loss of appetite	79 (±0.04)	43 (±0.12)	33 (± 0.06)	20 ( ± 0.07)	
Difficulties concentrating	83 (±0.08)	40 (±0.14)	31 (±0.06)	15 ( ± 0.06)	
Lassitude	86 (± 0.07)	46 (±0.16)	31 (±0.09)	19 ( ± 0.08)	
Emotional numbness	80 (±0.12)	35 (±0.11)	33 (±0.11)	20 ( ± 0.08)	
Pessimistic thoughts	85 (± 0.07)	41 (±0.10)	26 ( ± 0.04)	14 ( ± 0.05)	
Suicidal ideations	83 (± 0.10)	44 (±0.10)	32 (±0.05)	17 (±0.04)	

Mean and standard deviation of accuracies across five folds. Strict evaluation for accuracy considers exact score predictions, while flexible evaluation allows a deviation of ±1 from the actual score. Bold numbers highlight the best results.







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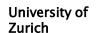


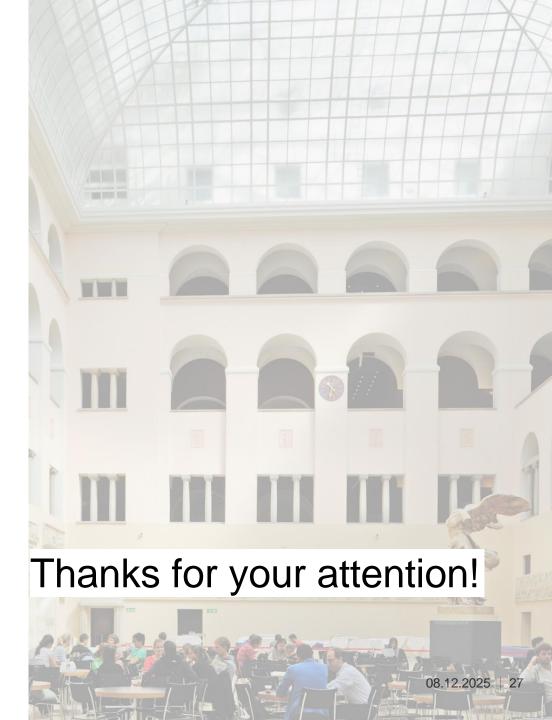
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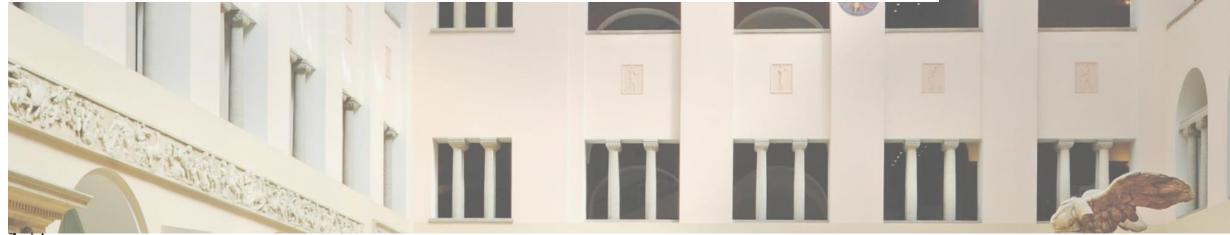




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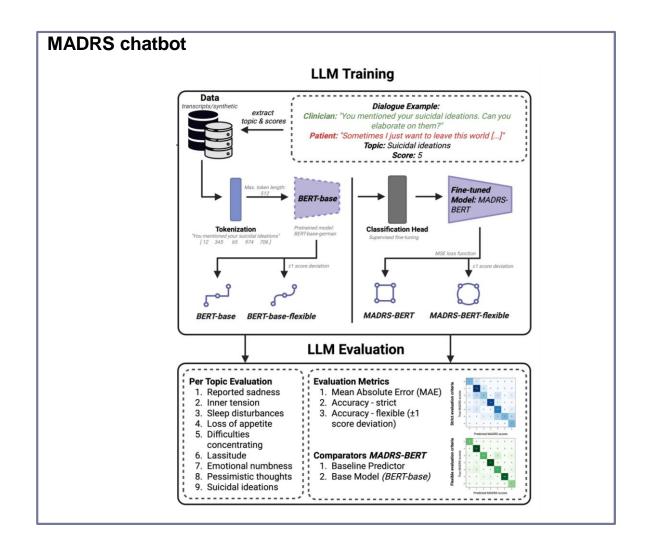
#### **Vision for MULTICAST TREAT**

Aim: Developing a digital intervention for people with suicidal thoughts and behaviors that ...

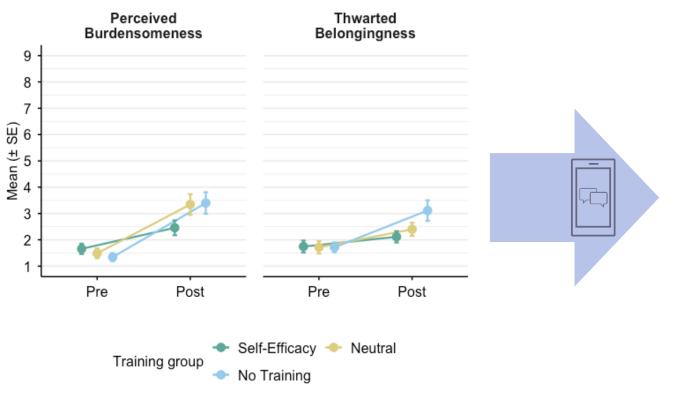
- ... covers different phases of symptom severity
- ...builds on stakeholder input
- ...includes modules that are evidence-based (safety plan) and innovative (self-efficacy training)
- ...leverages modern technologies to promote engagement

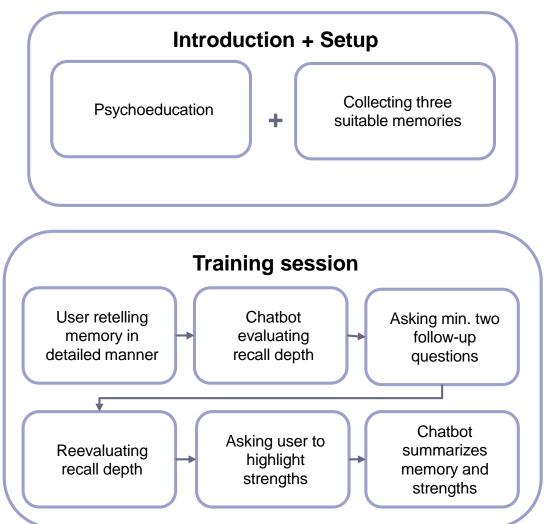


## **MUTLICAST TREAT: Preliminary work**



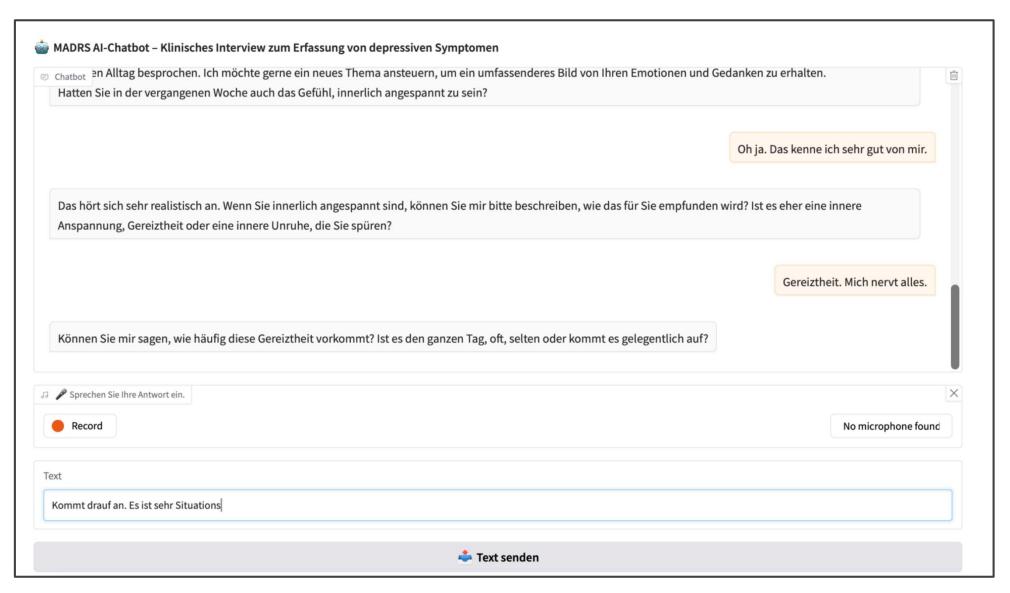
## **Autobiographical self-efficacy training**





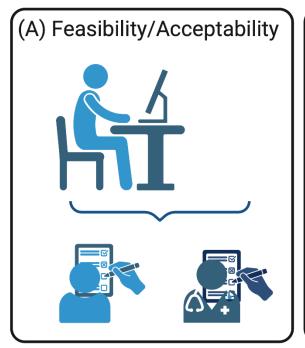
→ Rule-based structure + LLMassisted evaluation and text generation.

#### **MADRS Chatbot**

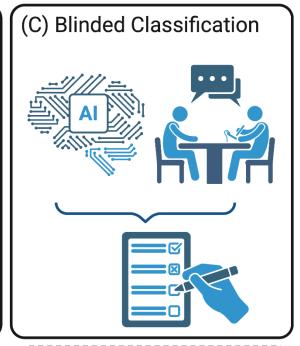


University of Zurich

#### MADRS Chatbot: Feasibility & Acceptibility Trial at PUK







#### Patient & Clinicians<sup>1,2</sup>

Acceptability & Feasibility Perceived Ease of Use Perceived Session Alliance Perceived Enjoyment Perceived Usefulness

Semi-structured Feedback

What was good? What to improve?

#### Clinicians only<sup>1,2</sup>

Professionalism Factual Accuracy **Empathy** 

Acceptability & Feasibility Perceived Ease of Use Perceived Session Alliance Perceived Enjoyment Perceived Usefulness

#### **Clinicians only**

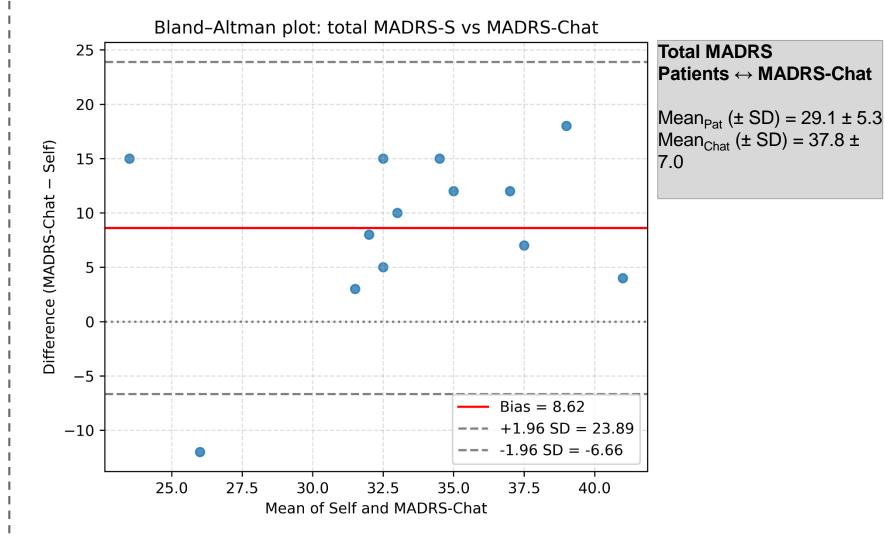
Al-led Intervies versus **Human-led Interviews** 

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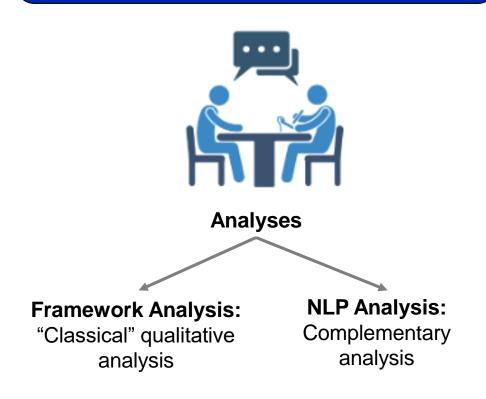
#### Stakeholder

## Quantitative Online Questionnaire N = 30

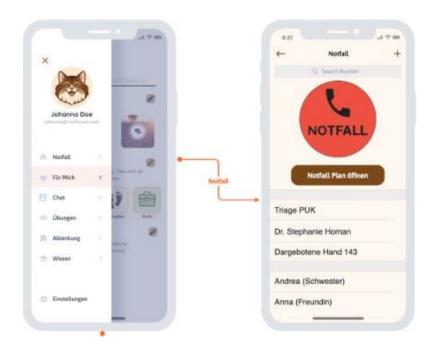
Subje	ctive Experience of Suicidal thoughts	and behavi	our		
1.	Let's think about the last seven days: H	low often di	d you expe	erience su	icidal
	thoughts?				
	□ Never □ 1-3 times □ 4-6 times □	daily 🗆 se	veral time	es a day	
		,		,	
2.	Please indicate how often you used the	strategies	in the list i	n the last	five suicidal
	crises.				
		1-2	3-4	Every	I don't
		times	times	time	use this
					strategy
	Reach out to family members or		1	1	1
	friends				
	Reach out to a mental health professional			1	1
	Reach out to a crisis service (e.g.,	+		-	+
	Dargebotene Hand)			1	1
	Nothing	+	-	-	+
	Use a specific technique. Please	+		+	+
	specify;			1	
	specify.			1	1
		-1	1	1	1

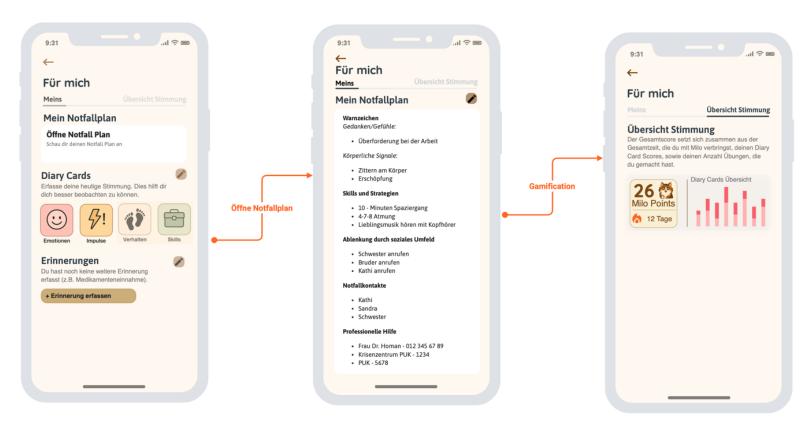
**Analyses:** Descriptive statistics

## **Qualitative**Stakeholder Interviews N = 14



#### **Outlook TREAT App based on Stakeholder Input**





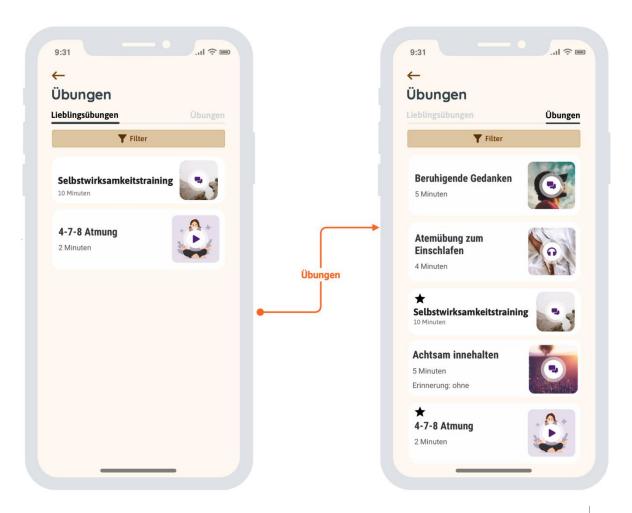
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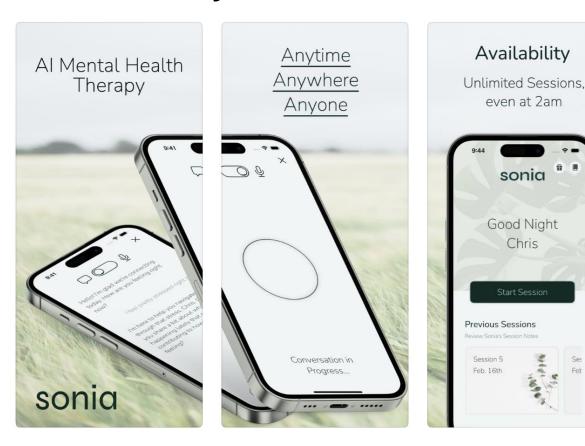
## **Outlook TREAT App based on Stakeholder Input**

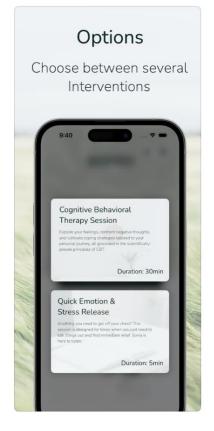






## **Real-time Risk-Detection System for Chatbots**





"General purpose models interrupt a conversation when detecting sensitive content (e.g., suicidal ideations).

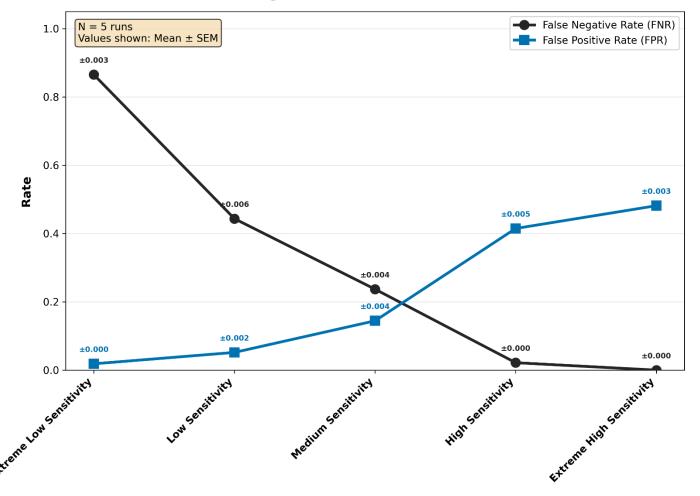
Aborting the conversation - when the user is in a critical state but also when a user might not be at acute risk - may leave them felt left alone, which -therapeutically speaking – isn't optimal in either case"

## **Real-time Risk-Detection System for Chatbots**

### **LLM-based High Risk Detection**

- 200 real user conversation excerpts where general purpose models aborted conversation
- 4 clinicians labelled data according to High Risk vs. Low Risk
- LLM-based classifier with varying sensitivity threshold to investigate the unevitable trade-off between false-positives (FP) and falsenegatives (FN).

#### **False Negative & False Positive Rates**

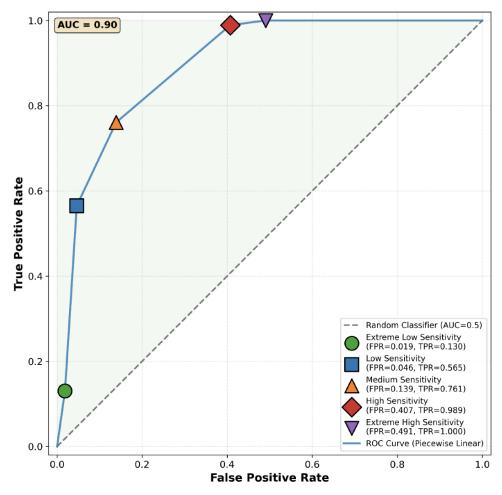


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#### **ROC Curve: FPR vs TPR Trade-off Analysis**













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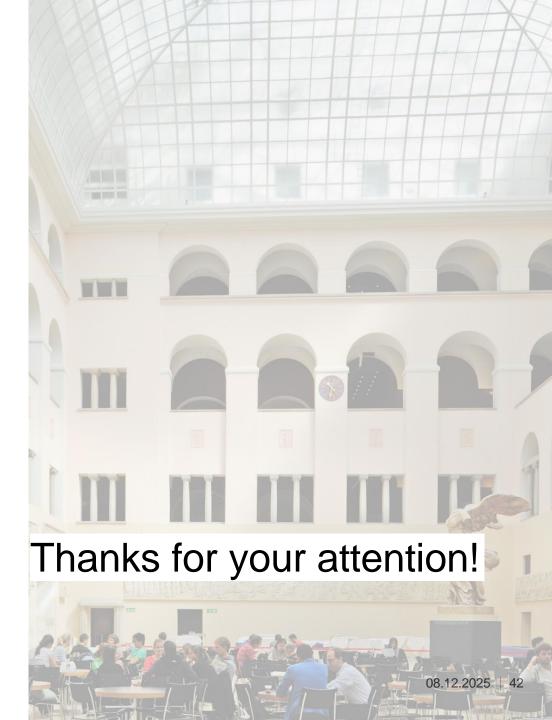
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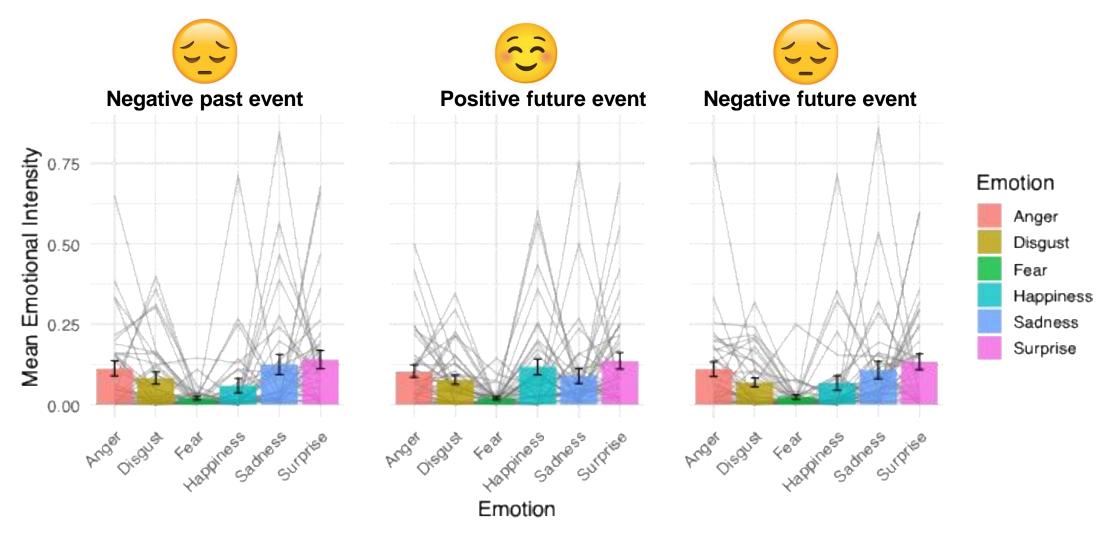
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# Can digital biomarkers accurately reflect the type of emotional content in a video?



Homan et al., in prep

## 1.1 PREDICT: PREDICTING SUICIDAL THOUGHTS AND BEHAVIORS





PUK Zurich, N = 100





UPK Basel, N = 50



NYU & Langone Hospital, *N* = 50

Screened

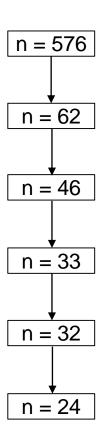
Inclusion

BL completed

EMA phase completed

FU1 completed

FU2 completed

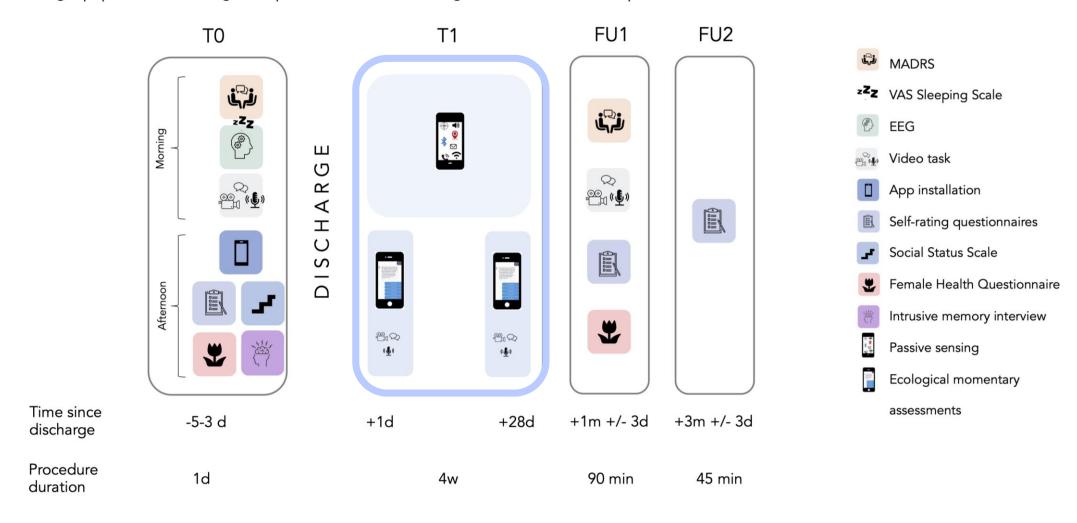


n = 64

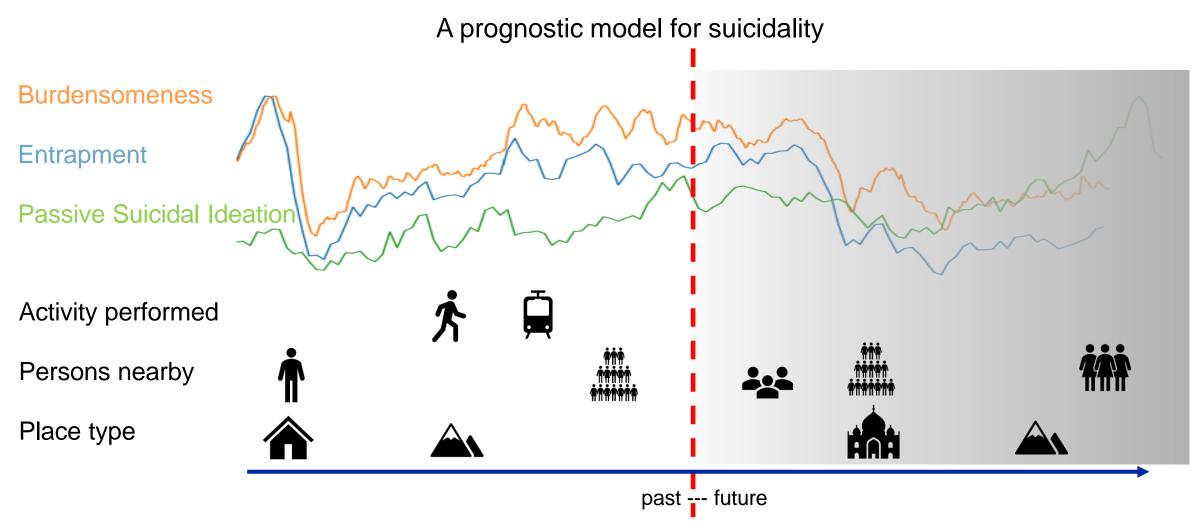
## 1.2 PREDICT: PREDICTING SUICIDAL THOUGHTS AND BEHAVIORS

## Integrating passive sensing and EMA data

Target population: Transdiagnostic patients with suicidal thoughts and behaviors at hospital admission.



# 1.2 PREDICT: PREDICTING SUICIDAL THOUGHTS AND BEHAVIORS Integrating passive sensing and EMA data



# 1.2 PREDICT: PREDICTING SUICIDAL THOUGHTS AND BEHAVIORS Integrating digital health data

# Performance in different scenarios

- Educational level, Income, age, gender
  - Baseline depression
    - History of suicide

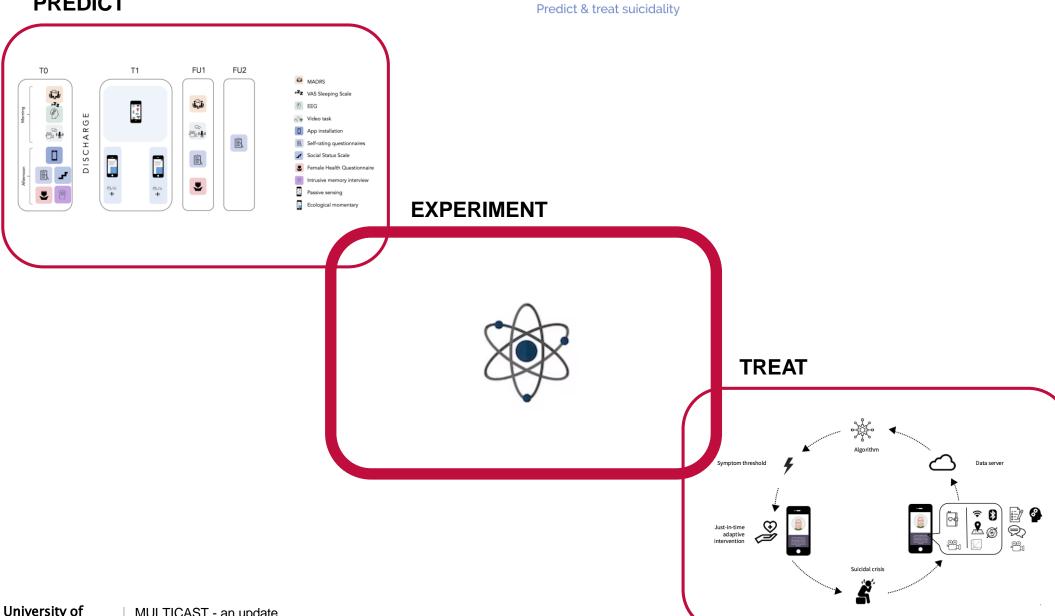


# Clinical Actionability

- Is symptom deterioration predicted enough in advance to intervene?
- What variables can be predicted reliably for the next day?

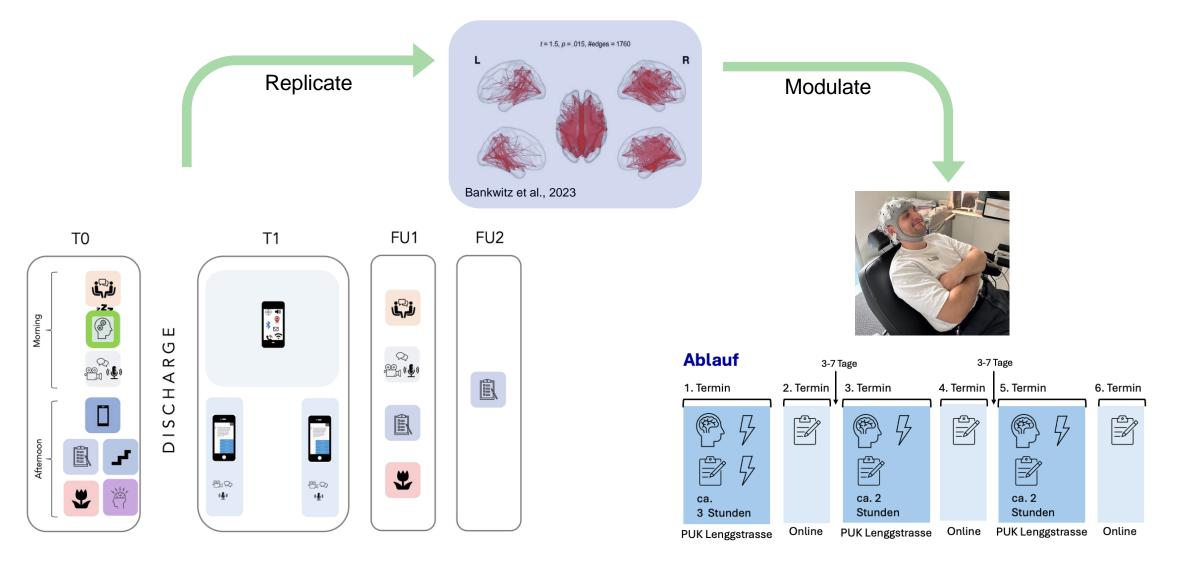


#### **PREDICT**



### 2.1 FROM PREDICT TO EXPERIMENT

# Integrating and modulating brain activity patterns





#### **PREDICT**

