

CHI 2024

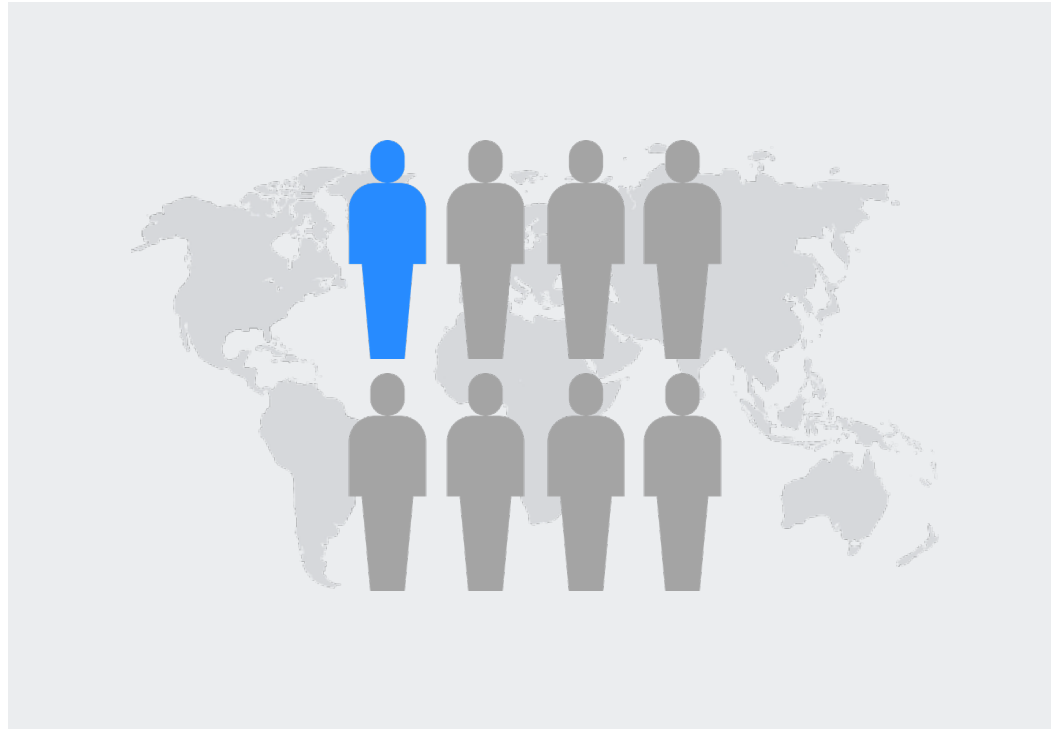
# Exploring Context-Aware Mental Health Self-Tracking Using Multimodal Smart Speakers in Home Environments



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\* Equal contribution

# Mental Health: A Rising Global Concern



**1 in 8 people worldwide  
live with a mental health problem**

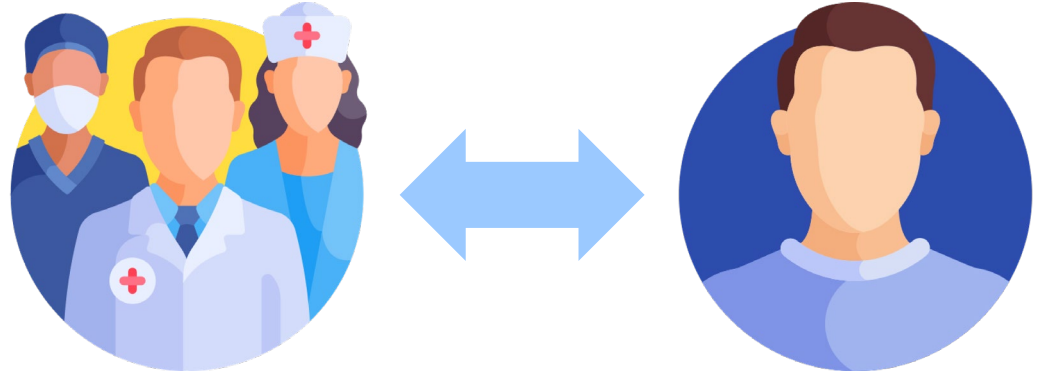
# Self-Tracking: A Method for Mental Health Monitoring

## Support Self-Reflection



Enhance self-awareness  
of mental health

## Help Clinical Decision-Making



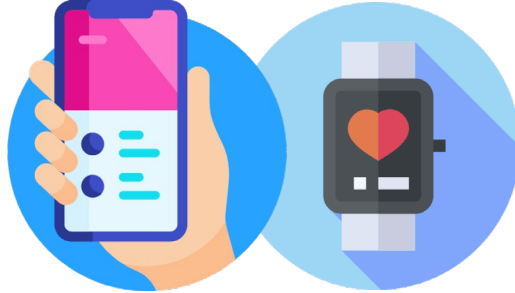
Bridge the information gap between  
healthcare stakeholders and patients

# ESM for Mental Health Self-Tracking

## Diversity of the Experience Sampling Method (ESM) Technologies



**Paper & Pen Method**



**Mobile/Wearable Technology**  
(Wang et al., 2014)



**Smart Speakers in Home Environments**  
(Wei et al., 2021)

# Mental Health Self-Tracking in Home Environments



People with mental health issues  
often stay indoors



Need for mental health self-tracking  
technology in homes is increasing

# Mental Health Self-Tracking with Multimodal Smart Speakers

**Mental Health ESM often requires visual-verbal tasks**

(e.g., Image description task for diagnosing depression or cognitive impairment)

Please describe a given image



Please describe a given image

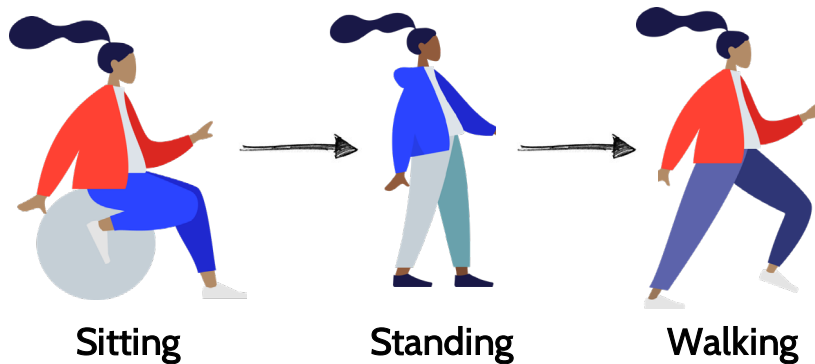


# Opportune Timing for ESM Design in Home Environments

## Identifying opportune moments in previous studies



**Task Breakpoints  
in Mobile/Desktop Environment**  
(Adamczyk et al., 2004)



**User Activity Transitions  
in Mobile Environment**  
(Fischer et al., 2011)

# Opportune Timing for ESM Design in Home Environments

Identifying opportune moments in previous studies

**HCI studies are still to investigate user experiences of context-aware mental health self-tracking using multimodal speakers**



Task Breakpoints  
in Desktop Environment  
(Adamczyk et al., 2004)



Sitting



Standing



Walking

User Activity Transitions  
in Mobile Environment  
(Fischer et al., 2011)



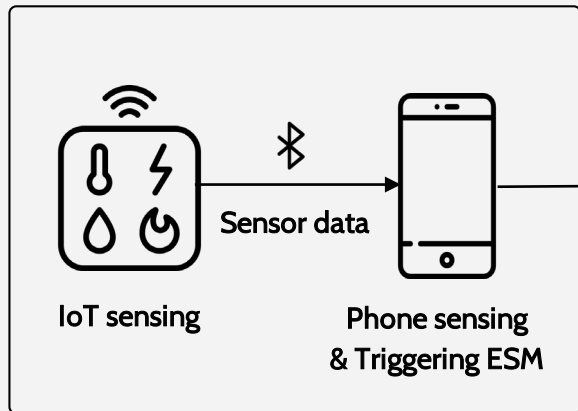
# Context-Aware Self-Tracking System using Multimodal Speakers

## Our System:



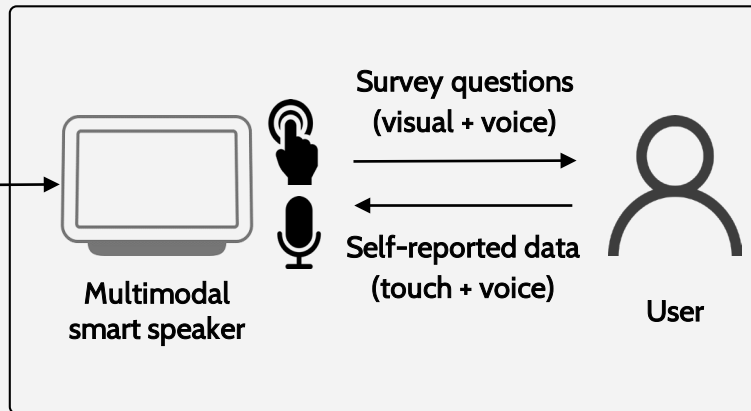
### Home environment

#### ① Context-Aware ESM Scheduling (based on user context transitions)



Wake-up  
keyword

#### ② Multimodal ESM Survey (via a multimodal smart speaker)



# Context-Aware Self-Tracking System using Multimodal Speakers

Our System Prototype:



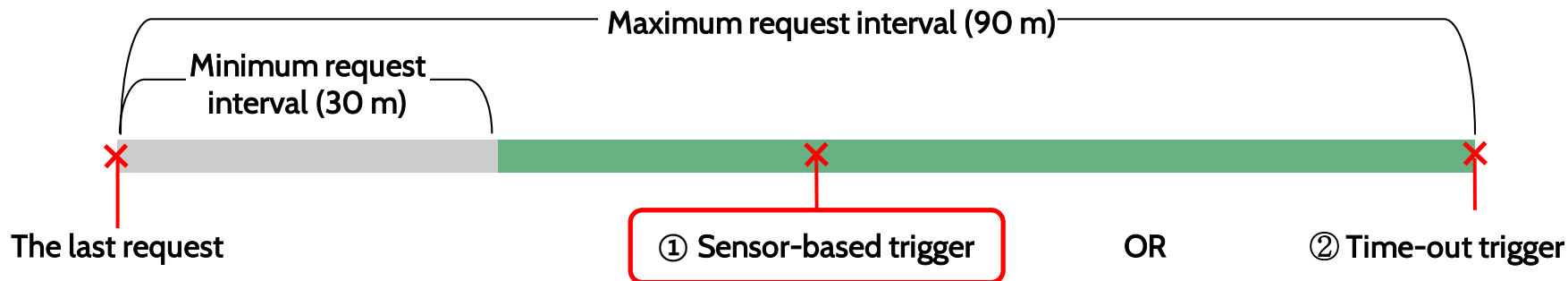
①-1 IoT Sensor  
: Collect CO2 data

①-2 Speaker triggering app  
with wide-angle lens  
: Collect noise, brightness, and # of people

② Multimodal speaker  
: Provide voice and touch interactions

# Context-Aware ESM Scheduling

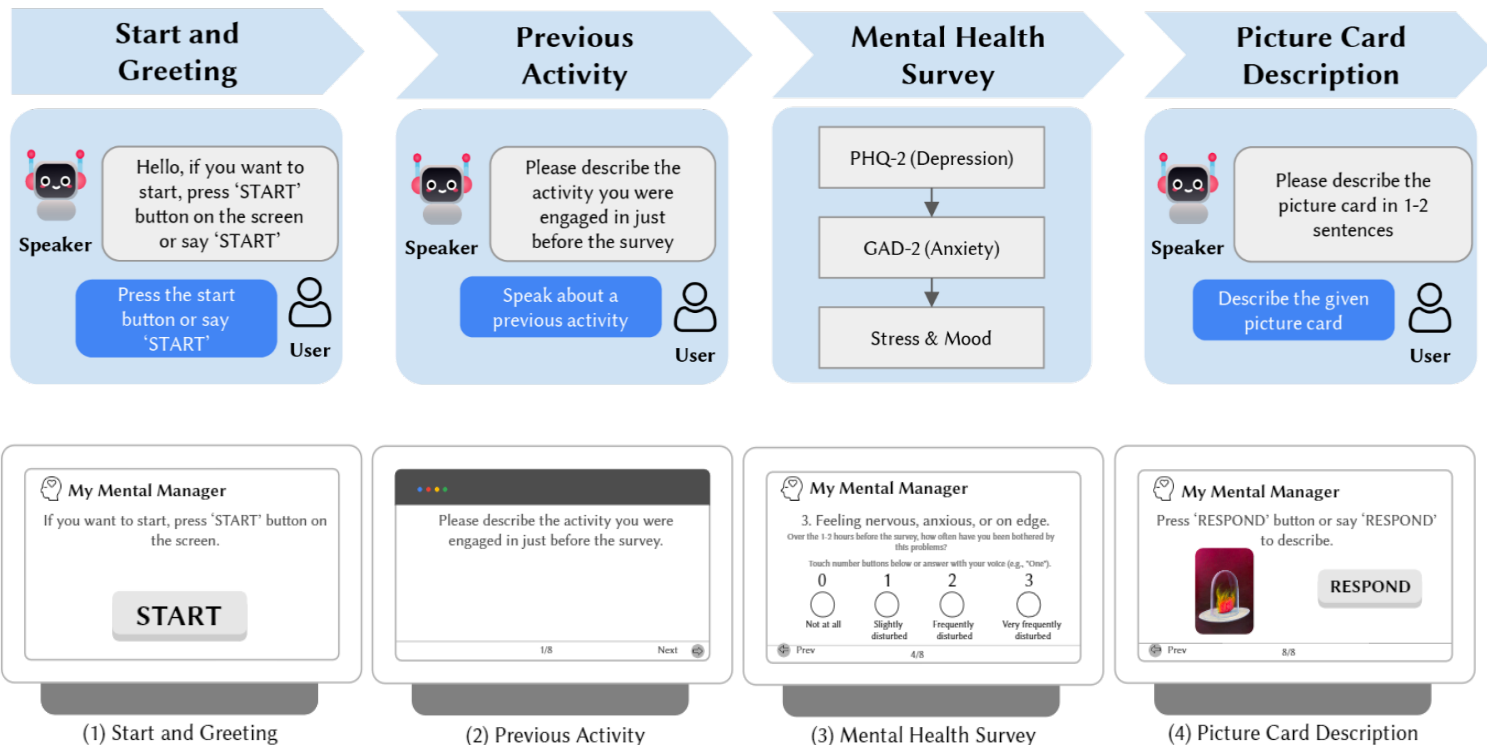
Determine opportune moments for ESM requests in home environments



- Detect user context transitions using sensors:
  - **Auditory channel availability** using *Noise Sensor*
  - **Proximity to smart speakers** using *Light Sensor, CO2 Sensor, Camera*

# Multimodal ESM Survey

## ESM Task Steps and User Interface



## Research Questions

1. How do users perceive proactive mental health self-tracking using multimodal speakers?
2. How do ESM compliance rates change across different context transitions?
3. What are the preferred interaction modalities for responding to multimodal speakers?

# Field Study Methods

## Participants (N=20)

- Recruitment criteria
  - People who were diagnosed with at least mild depression (a PHQ 9 score of 5 or higher)
  - People who had private spaces at home or were single-person households
  - People who spent a minimum of 5 hours daily in their room, excluding sleep time



## Research Questions

1. How do users perceive proactive mental health self-tracking using multimodal speakers?
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# RQ1: Overall User Experience of Proactive Mental Health Self-Tracking using Multimodal Smart Speakers

Positive Aspects: Proactive system helped users to gauge mental health status



“I don’t usually get a chance to ask myself these questions (related to mental health). But every hour or two, the system asks you how you’re feeling or how stressed you are, and *it gives me more opportunities to think about whether I’ve just gotten stressed.*” - P18

“Before, I had no idea about my moods. But when I got a chance to think about it (through the survey), I was like, *‘I see ... what was happening’ and could relieve negative emotions.*” - P19



# RQ1: Overall User Experience of Proactive Mental Health Self-Tracking using Multimodal Smart Speakers

Positive Aspects: Human-like factors made users engaging to ESM requests



“I was more focused on the question because the speaker asked questions verbally. Also, there’s only one question on the screen. It makes me concentrate on each question.” - P12

“I felt like it’s a person because the timing was not exactly regular. It’s usually unpredictable when someone will contact you. So, the timing of the speaker talking to me made me feel like a person.” - P13

# RQ1: Overall User Experience of Proactive Mental Health Self-Tracking using Multimodal Smart Speakers

Negative Aspects: Machine-like interaction style led to boredom



“The questions and pictures are repeated over and over again. *As the experiment progressed, I felt bored because the system became more habitual and predictable.*” - P15

“I think it was annoying to keep asking the same questions over and over again. So, *there was a decrease in the sincerity of responses.*” - P17

## Research Questions

1. How do users perceive proactive mental health self-tracking using multimodal speakers?
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## RQ2: ESM Response Rates across Different Context Transition

ESM response rates were lower in the time-out trigger condition and morning

| Trigger type          | Num. responses | Num. requests | Response rate |
|-----------------------|----------------|---------------|---------------|
| Maximum time interval | 1,502          | 2,815         | 53.4%         |
| CO <sub>2</sub>       | 164            | 272           | 60.3%         |
| Human                 | 364            | 549           | 66.3%         |
| Light                 | 157            | 206           | 76.2%         |
| Noise                 | 14             | 21            | 66.7%         |
| Total                 | 2,201          | 3,863         | 57.0%         |

| Time of day             | Num. responses | Num. requests | Response rate |
|-------------------------|----------------|---------------|---------------|
| Dawn (2:00~7:59)        | 35             | 64            | 54.7%         |
| Morning (8:00~13:59)    | 549            | 1049          | 52.3%         |
| Afternoon (14:00~19:59) | 767            | 1388          | 55.3%         |
| Night (20:00~01:59)     | 850            | 1362          | 62.4%         |
| Total                   | 2,201          | 3,863         | 57.0%         |

## RQ2: ESM Response Rates across Different Context Transition

Responded more to ESM in the afternoon and night than in the morning

| Predictors              | B (SE)       | z-statistic | 95% CI for odds ratio |            |       | p-value |
|-------------------------|--------------|-------------|-----------------------|------------|-------|---------|
|                         |              |             | Lower                 | Odds ratio | Upper |         |
| (Intercept)             | -0.04 (0.20) | -0.20       | 0.66                  | 0.96       | 1.41  | 0.84    |
| Time of day             |              |             |                       |            |       |         |
| Dawn (2:00–7:59)        | 0.50 (0.31)  | -1.61       | 0.90                  | 1.64       | 3.00  | 0.11    |
| Afternoon (14:00–19:59) | 0.24 (0.09)  | 2.58        | 1.06                  | 1.27       | 1.53  | 0.01    |
| Night (20:00–1:59)      | 0.43 (0.10)  | 4.54        | 1.28                  | 1.54       | 1.86  | <0.001  |
| Trigger type            |              |             |                       |            |       |         |
| CO <sub>2</sub>         | 0.60 (0.15)  | 3.95        | 1.35                  | 1.80       | 2.42  | <0.001  |
| Human                   | 0.92 (0.12)  | 7.98        | 2.01                  | 2.52       | 3.16  | <0.001  |
| Light                   | 1.24 (0.19)  | 6.65        | 2.39                  | 3.44       | 4.95  | <0.001  |
| Noise                   | 0.35 (0.49)  | 0.72        | 0.55                  | 1.42       | 3.71  | 0.47    |

## RQ2: ESM Response Rates across Different Context Transition

Responded more to ESM when users were near the speakers (CO<sub>2</sub>, Human, Light)

| Predictors              | B (SE)       | z-statistic | 95% CI for odds ratio |            |       | p-value          |
|-------------------------|--------------|-------------|-----------------------|------------|-------|------------------|
|                         |              |             | Lower                 | Odds ratio | Upper |                  |
| (Intercept)             | -0.04 (0.20) | -0.20       | 0.66                  | 0.96       | 1.41  | 0.84             |
| Time of day             |              |             |                       |            |       |                  |
| Dawn (2:00–7:59)        | 0.50 (0.31)  | -1.61       | 0.90                  | 1.64       | 3.00  | 0.11             |
| Afternoon (14:00–19:59) | 0.24 (0.09)  | 2.58        | 1.06                  | 1.27       | 1.53  | <b>0.01</b>      |
| Night (20:00–1:59)      | 0.43 (0.10)  | 4.54        | 1.28                  | 1.54       | 1.86  | <b>&lt;0.001</b> |
| Trigger type            |              |             |                       |            |       |                  |
| CO <sub>2</sub>         | 0.60 (0.15)  | 3.95        | 1.35                  | 1.80       | 2.42  | <b>&lt;0.001</b> |
| Human                   | 0.92 (0.12)  | 7.98        | 2.01                  | 2.52       | 3.16  | <b>&lt;0.001</b> |
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| Noise                   | 0.35 (0.49)  | 0.72        | 0.55                  | 1.42       | 3.71  | 0.47             |

## Research Questions

1. How do users perceive proactive mental health self-tracking using multimodal speakers?
2. How do ESM compliance rates change across different context transitions?
3. What are the preferred interaction modalities for responding to ESM requests?

# RQ3: Interaction Modality Preferences based on User Context

Most users preferred to respond with GUI over VUI for multiple choice questions

Reasons for Using a GUI



Limitation of VUI & Familiarity with GUI

Reasons for Using a VUI

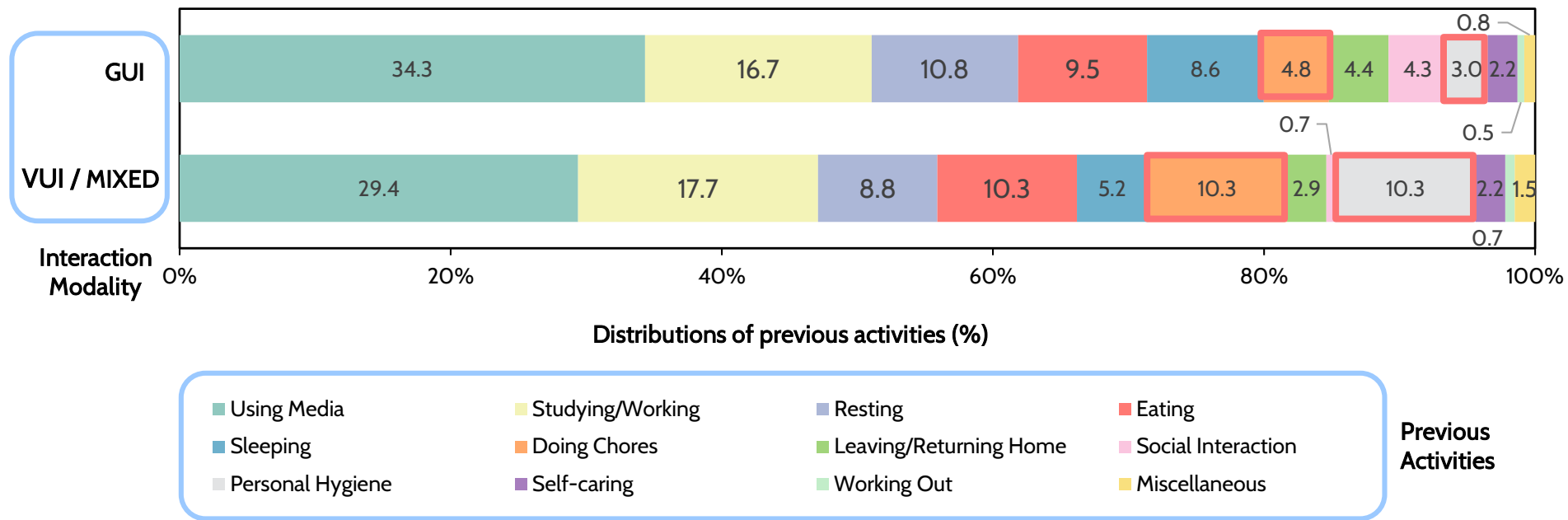


Situations when hands are occupied



# RQ3: Interaction Modality Preferences based on User Context

VUI/MIXED was more frequent than GUI in doing chores and personal hygiene



# Summary of Key Findings

- RQ1: Overall User Experience
  - **Proactive self-tracking** can increase self-reflection regarding mental health
  - **Human-likeness** helped users engaging in answering mental health questions
- RQ2: ESM Response Rates across Different Context Transition
  - **User response rates** improved when ESM are requested in **context transitions**
- RQ3: Interaction Modality Preferences based on User Context
  - **Users' previous contexts** influenced their **interaction modality selection**

# Discussion

## Context Awareness for Modality Selection and Adaptation



Detect user contexts and **adaptively select** an appropriate **interaction modality**

## Sensor Selection in Home Environments



Use sensors that are capable of **detecting** multiple users

# Design Implications

## Consideration for ESM System Design



Consider **context-sensing**  
in the **multimodal ESM** interaction design

## Consideration for Engaging ESM Interaction Design



**Vary the tone and content** as in context-  
tailored adaptations

# Exploring Context-Aware Mental Health Self-Tracking Using Multimodal Smart Speakers in Home Environments

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## Takeaway Notes

- **Context-awareness** improves user compliance of ESM surveys and makes user feel it like human
- HCI studies continue to investigate **context-tailored adaptations** for making user engaging in ESM

