A Community-Centered Design Framework for Robot-Assisted Feeding Systems

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ABSTRACT

Robot-assisted feeding (RAF) systems offer enormous potential benefits to community-centered care-giving environments. However, developers of RAF technologies often focus on evaluating their standard transactional functionality, omitting the impact of such technologies in contexts that extend past the interaction of the robot and food receiver. RAF technologies have complex social, cultural and self-identity implications, since a "meal" extends well beyond the simple provisioning of nourishment. To better understand these implications we conducted a contextual inquiry in an assisted-living community with five potential care recipients and five caregivers, as well as interviews with fifteen domain experts including occupational therapists and feeding specialists. Based on our findings from these studies, we developed a new framework for RAF technologies that formulates this vital task as a communitycentered relational service. We then use this framework to qualitatively and quantitatively assess three existing feeding systems and identify areas of improvement. Our work reveals new insights about stakeholders of RAF technologies and provides a roadmap for technology developers to better serve the needs of these stakeholders.

CCS Concepts

•Human-centered computing \rightarrow Empirical studies in accessibility; •Social and professional topics \rightarrow Assistive technologies; People with disabilities; •Computer systems organization \rightarrow Robotic autonomy;

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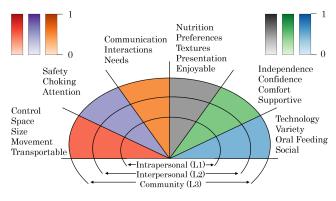


Figure 1: Multi-level Contextual Framework

Author Keywords

Assistive Feeding; Assistive Robotics.

INTRODUCTION

Studies have consistently shown that long-term care recipients who live in the community have far better clinical outcomes than those in institutionalized settings [51]. Nearly 18.7% percent of the non-institutionalized US population live with a disability [21]. Of these, about 12.3 million need assistance with one or more activities of daily living (ADLs), such as feeding, bathing, or dressing. Indeed, needing help with one or more ADLs is the most cited reason for moving to assisted or institutionalized living [54, 1]. Moreover, depression and disability can worsen after ADL assistance is initiated, perhaps due to loss of regular routines that could worsen disability [46]. Sources of support for people with impaired levels of functional capacity in the community are typically family members, which make up an estimated 20% of the adult population in the US [34]. The care network increases in both size and scope as the care recipient's functional capacity declines. Hence, ADL assistance and management concerns many more individuals and communities beyond the person who requires functional support.

Major efforts in the personal robotics community have created robotic solutions to increase independence or prolong independence in self-feeding [22]. Robot-assisted feeding (RAF) in itself poses multifaceted technical challenges. To date, most approaches to defining the RAF design space have focused on technology function and the care recipient, addressing specific sub-tasks in the food acquisition and transfer chain. These RAF systems were considered functional as long as they could pick up a food item and transfer it to a care-recipient. However, assistive robots do not exist in isolation, nor as an isolated unit with their care-recipient. There is an opportunity to design assistive self-care robot technologies that satisfy not only the functional requirements of performing an ADL, but also serve more of a supportive relational role.

We emphasize that this shift from transactional to relational service technology is aligned with the recent movement in human-computer interaction toward increasing the accessibility of technology using social models of disability (e.g., [35, 50]). We aim to understand how personal robotic solutions can be contextually appropriate in an assisted self-care setting, particularly in the case of ADLs, and be seamlessly integrated into a complex home environment. That is, how we can design robot-assisted self-care technology to support the needs of all participants in the home environment and scaffold interactions that promote agency for the robot-assisted individual while also increasing support for other individuals in the support network. We believe that a shift is needed in the assistive robotics for ADL community. This shift would be the first step in community-led ADL management. This, in turn, will promote wider adoption of assistive robotics.

This work offers three key contributions:

- A novel adaptation of a multi-level model from socialecological theory to analyze RAF systems accounting for transactional, relational, community, and social aspects.
- A proposed systematic design framework for evaluating RAF systems informed by our multi-level contextual inquiry with stakeholders.
- Analyzing three state-of-the-art existing RAF systems using our evaluation framework to demonstrate how it can be instrumental in understanding technology design gaps and priorities.

BACKGROUND

Our main tenet is that carefully-constructed, communitycentered design frameworks to develop assistive technology can drive innovations in how we create robotic technologies that are embraced and adopted in homes and living environments. This can be used to improve the community living experience for individuals with disabilities and their caregivers. Frameworks of study surrounding the design process of *public services* within complex social contexts provide us with trialed analytical frameworks. Our study insights, for example, are consistent with frameworks of community living and family centered care; frameworks which have gained increasing attention as a core determinant of care quality since a seminal report by the Institute of Medicine [68]. We take feeding, key among ADLs, as an example. The functional requirement of feeding is to nourish the body with food, but care recipients often reject feeding assistance because of the intricately complex association that food and meals have with self identity, cultural values, family life, and emotional well being [66]. Hence, there are wide ranging factors beyond technical efficacy: relational, emotional, community, and social factors that interact, influence and are influenced by RAF technologies. Design and development of assistive technologies should be informed by these factors and address the physical, cognitive, emotional, and social gaps that remain, even when an individual is functionally assisted with ADLs.

The social-ecological model (SEM) [28, 55] provides a conceptual framework to deconstruct complex social contexts like feeding and systematize the analysis, design and evaluation of RAF systems. SEM conceptualizes the social world in four levels of influence. These are: (L1) individual or intrapersonal, (L2) interpersonal, and (L3) community or institutional, and (L4) societal (indicating macro-level societal, policy or regulatory influences). In this paper, we concentrate on the inner three levels (Fig. 1), where L1 involves the care-recipient and caregiver pair, L2 involves family members, and L3 involves the larger social effectors. RAF design decisions involving these spheres will have the greatest impact on adoption and acceptance of robotic applications in the community. We leave out the outermost level (L4), which involves interrogating processes and decision-making at the legislative or national level, as it is outside the scope of our research, but remains critical in adoption of RAF technologies.

To further structure the multi-level approach and focus its use in the context of community-supporting-technology, we inform our design framework with contextual dimensions currently used in evaluating robot-assisted transportation applications at the USDOT [26]. The six contextual dimensions (described below) are helpful in selecting and applying appropriate performance indicators and measures that are integral to community-centered RAF design decisions. Our workflow (described in section 6) analyzes RAF systems per these contexts along the multiple social levels offered by the SEM. The result is a workflow for comparative evaluations for RAF systems in a community-centered approach.

To the best of our knowledge, there are no groups currently framing assistive robotic technologies as a social service intervention. The approach in itself provides broader and layered contexts of analysis, extending beyond previous framings that focus on examining the interaction between the technology and its immediate user. Specifically, the framework provides tools to understand the influence by and on related users in broadening context-layers (care recipients, caregivers, home care personnel, therapists, care providers, social workers, etc). The utility of defining such a framework is demonstrated by using it to differentiate among different RAF solutions. The framework is used to compare several existing RAFs, enabling us, in the process, to identify key properties of the technology and the design space. The comparison highlights areas where existing RAFs provide extensive support and those in which they are deficient, suggesting a future research agenda.



Figure 2: Three RAF systems assessed within the developed framework: (Left) Obi [7], (Middle) PR2 [62], (Right) ADA [36].

RELATED WORK

Robot-Assisted Feeding systems

Several specialized feeding devices for people with disabilities have been developed in the past decade, three of which are shown in Fig. 2. Although there are several commercially available RAF systems in distribution [7, 6, 4, 3, 9, 11, 5, 14, 10] that use a robotic arm to scoop food with a spoon, they have lacked widespread acceptance. There are also RAF systems that were developed in research labs, either as table-mounted robotic arms [25, 71, 69, 59, 65, 64, 41, 43, 47, 75, 85, 44, 78, 77, 87, 86, 74], general purpose mobile manipulators [18, 67, 62], or wheelchair-mounted robotic arms [19, 16, 31, 42, 49, 32, 36].

A comprehensive review of meal assistance robots is given in [57, 76, 45, 23]. Some assistive systems focus less on full autonomy and more on support or adaptation to the user. For example, several systems emphasize tremor cancellation during feeding task [58, 13, 63], while others provide arm support for users who can move their arm but do not have sufficient muscular strength for continued large movements [80, 12, 15]. Park et al. [62] developed a semi-solid food acquisition system for assistive feeding using a general purpose manipulator to scoop yogurt with a spoon. They also developed an anomaly detection framework for assistive feeding using multiple sensing modalities [60, 61]. Herlant [40] developed a robotic system that uses a fork to skewer solid food items. Gallenberger et al. [36] can detect a solid food item specified by a user and pick it up using a fork autonomously. Canal et al. [24] developed an intelligent personalization framework for RAF systems that can adapt to user preferences. McColl and Nejat [53] performed a study with an expressive socially assistive robot Brian 2.1 and older adults during meal time in a care facility and explored ways in which a socially assistive robot can provide cognitive assistance during the feeding activity to analyze user engagement and compliance during meal time.

Contextual inquiries in related assistive technologies

Researchers have also conducted contextual inquiries to analyze other assistive technologies, as well as their opportunities and challenges. Szpiro et al. [73] conducted a contextual inquiry to understand the opportunities and challenges of using mainstream computing devices by people with low vision. Naftali and Findlater [56] performed a contextual session to analyze how users with motor impairments use smartphones in their daily lives. Maciuszek et al. [48] constructed a functional design space of electronic assistive technologies by conducting a study with both the users and designers of the technology. Overall, in the literature, researchers have conducted studies with multiple stakeholders of varied assistive technologies to analyze their functions and scope of usage.

A STUDY WITH STAKEHOLDERS

The value of this paper lies in connecting systems thinking and assistive robotic design practices while attempting to improve adoption of robots in everyday living. Our design framework proposes a shift towards designing enabling environments that support multiple actors at multiple layers of the social network surrounding a care recipient, and away from a myopic focus on a single user and a fully-automated method to replace human caregivers. The goal of our contextual user study was to better map and understand how social relationships may shape and inform new supportive design objectives and tasks for the RAF system when re-conceptualized as a technology supporting the entire network. We consider this a community-led relational approach. To improve our understanding of the sociology surrounding feeding, we performed a contextual inquiry with experts at various layers: domain experts (occupational therapists, therapeutic recreational specialists and feeding specialists) as well as local needs experts (care recipients and caregivers).

Multi-Level Contextual Inquiry

The goal of our contextual inquiry was to provide insight into all potential stakeholders in the social service the robots may provide. We used a combination of surveys and contextual inquiry to capture a mix of quantitative and qualitative data describing stakeholders' background, preferred methods of feeding (if applicable), attitudes, practices, relationships and information needs, concerning feeding. Our study included 15 experts, and 5 care-recipients with 5 of their caregivers. We describe our methods more specifically below.

Contextual Inquiry in a Community Setting

Care-recipients and caregivers were engaged in a communal meal in their home setting where researchers observed the meal activity. Care-recipients were assisted in a typical fashion by the participant's typical caregiver. The user participants were 2 female and 3 males, all wheelchair users with neck motion ability and showed no evident swallowing problems. The caregivers were 3 female and 2 males, registered nurses and nurse practitioners among their occupations.

This contextual inquiry had three main components:

- 1. Semi-Structured Interview: Understanding the nuances of the feeding task and its challenges
- 2. Activity: Collecting observations while having a meal together
- 3. Introduce idea: Showing a video of our assistive feeding robot and getting their impressions and feedback

We conducted the study by asking participants the questions verbally and then recording responses. One care-recipient had speech impairments and used a tablet to respond to the questions, while others responded verbally. During the study, we used multiple go-pro cameras to record the video and audio which were transcribed later for qualitative analyses.

Inquiry with Experts

We developed an expert survey to gather information from participants who are very familiar with feeding in care settings, and who may be able to influence (to varying degrees) feeding assistance in the community setting. Fifteen experts responded using an electronic questionnaire. The group included occupational therapists (OT), therapeutic recreation specialists, physical therapists, assistive technology professionals, speech-language pathologists (SLP), surgical technologists, and executive director of a non-profit organization focused on independence and technology. Out of the 15 experts, 13 were female, the age ranges from 23 to 63 years of age, with a mean age of 43 ± 12 years. Regarding the experience of feeding someone with mobility impairments, 86.7% of the surveyed experts responded affirmatively. Years of professional experience ranged from 3 to 40 years in their particular occupation, with a mean of 18 ± 12 years. One of our surveyed experts happens to have mobility impairments, so he was surveyed as both an expert and a user. The electronic survey for experts contained twenty open-ended questions as well as twelve Likert-scale questions to gather their impressions and concerns involving a robot for assistance with feeding.

STUDY FINDINGS

In this section, we present the findings of the multi-level contextual inquiry with users, caregivers and experts, as well as our observations during the feeding activity. To organize the collected data, we segregate our analysis by the spheres of influence suggested by the social-ecological model as they would be applied to conducting and evaluating RAF systems. Within each sphere, the participant responses are further interrogated using six contexts of evaluation [26]: technical function, technological robustness, information gaps, usability, user empowerment and social acceptance, and system integration. These contexts provide multiple perspectives to help in mapping the information respondents told us to technology design indicators that would impact the implementation and development of a RAF system. Our analysis of the collected data using the framework as described above, leads us to identifying (within each context of evaluation) indicators that can

be used to evaluate RAF system impact in each sphere. Indicators are the constructs, activities, and behaviors that the RAF technology would operationalize to effect change at the particular sphere. Measures are the specific tools that can be used to evaluate the change. The indicators under each of these evaluation contexts were qualitatively obtained using the stakeholder interviews. Our interviews with stakeholders were analyzed independently by two authors, coded using identified keywords, and then discussed to reach a 100% inter-rater reliability score. In this manner, the community's data derives the evaluation framework that is used in later sections to compare existing RAF systems.

Technical Function:

This evaluation context involves system aspects that directly impact functional metrics related to feeding. Considering the wide range of user capabilities and dietary needs, a RAF system needs to handle multiple tools necessary to manipulate different food types and textures. From our surveyed users, food choices ranged from pre-cut pieces of meats to soft foods such as mashed potatoes and pieces of fruits. Experts such as E6 responded to the nutritional challenges with: "*ensuring adequate caloric intake across a variety of food textures and types that are safe for the patient*". As we move across the different levels of influence food preference may change from a few items or food types at the individual level, while interpersonal and community levels may require the RAF system to handle greater variety.

Caregivers noted the feeding task is physically challenging. This also falls under the context of technical function, as the RAF system could help resolve some of the ergonomic difficulties associated with the task. During our visit, we noticed that the caregivers stood next to the care-recipient holding the bowls in their hands and had to bend or adjust to feed each care-recipient (see Fig. 3). Expert responses also indicated that proper utensil positioning in proximity to mouth could both improve feeding efficiency and achieving proper nutrition, but also maximize care recipient's independence and participation in the feeding process. One of the surveyed caregivers, C5, described her motions during feeding as: "careful smooth and level motions with a spoon,(...) to ensure I have a good grasp and won't drop the food prior to releasing it into the patient's mouth". The surveyed caregivers did not mention ergonomics explicitly, but C1 did comment that the robot needs to "show flexibility" in motion and grasping, perhaps considering her own variety of postures and motions during feeding. These considerations also change as we move between levels, since the system's size and footprint might not be as important in the care recipient's home, whereas positioning the RAF system at a restaurant may be challenging or prohibitive. Several experts also addressed the system's physical qualities. They mentioned the robot size should be small (sleek, compact), it should be quiet, it should not draw unnecessary attention. Other experts mentioned the importance of offering a "guideline for how much space is needed around the client and the robot so the arm doesn't end up bumping into anyone". Safety is of paramount importance. S3 and S4 mentioned an emergency stop button as an important feature for the system. These are relevant in all three spheres.



Figure 3: Observations from caregiver-assisted feeding activity during our contextual inquiry: (Left) Caregiver feeding a care-recipient, (Middle) Care-recipient communicates intent, (Right) Different user interface.

Technology Robustness:

This context addresses how the RAF system would improve safety and reliability by addressing anomalies, uncertainty, and risk. Our participants noted RAF system must adapt to changing environment. At the intrapersonal sphere, a RAF system must adapt to care-recipient positional changes. At the interpersonal level, it must avoid collisions with people or objects within its working space. One of the surveyed caregivers also mentioned robustness in terms of the system maintaining functionality through rough handling and falls. An expert (E3) also commented on robustness regarding the "possibility of (the robot) breaking or needing repairs that could leave individuals stranded if dependent on the robot". Being able to detect these failures and correct them are important indicators in this context. Projecting this indicator to the community sphere, functionality must be robust to environmental complexity (e.g., lighting conditions and background noise). Experts also added comments related to food-intake safety. For example, detecting if a care-recipient may be choking and notifying someone, providing reminders to drink during the meal or to slow down food intake. privacy and security of user-data, relative to data collected or generated by the RAF would be additional viable indicators under this context.

Information Gaps:

Care recipients or caregivers may have functional or situational sensory or cognitive limitations that create gaps in the system's communication with users. The system should provide multiple formats for feedback, and multiple modalities for executing system functions. Even when experts generally suggested voice commands if there are mobility issues, they also acknowledge that some care-recipients may need other methods such as gestures, head or shoulder movements, gaze, or adapt the system's interactions to available AAC (Augmentative and Alternative Communication) technology, or even texting using a tablet as was the case for one of our participants with speech impairment (see Fig. 3). At the interpersonal sphere, potentially less structured environments may require redundant formats, for example, between verbal commands and button pushes. Analogously, if the RAF system includes vocal output interactions, those might become bothersome to third parties at the interpersonal sphere.

Usability:

In the *usability* context, the RAF addresses indicators that impact customization and user preferential aspects. At the intrapersonal sphere, most participants discussed necessary adaptations to care-recipient preferences and needs. *Customization capabilities to adapt to user preference* is a fundamental indicator within the usability context. Caregivers and experts highlighted the importance of *bite size tuning*. Seven experts mentioned 'bite size' as an important consideration while feeding someone. S1, S3, and S4 mentioned that food needs to be cut, while S5 prefers to have the food cut as he eats. S3 said that she needs smaller pieces "*to avoid difficulties chewing/swallowing*". The importance of honoring the preferences of a user was also mentioned by caregivers. We highlight C1's response in particular: "*we notice he wants his mouth to be full*".

Regarding timing for transferring the bite, the surveyed caregivers agreed that they *detect care recipient gestures* (open their mouths or nod head), or *wait for a positive response after querying* if they are ready for another bite. This translates to two indicators that are dependent on *properly detecting whether mouth is open or not*. When considering the interpersonal or community level, determining when to transfer the bite might become more dynamic to account for interactions such as participating in a conversation. Also, the mouth detecting feature needs to be more robust to detect only the face and mouth of the care recipient in an environment with multiple people.

Additionally, eight experts mentioned 'pacing' as an important consideration while feeding someone. S2 says that it should be up to the user to select the speed of the robot when it is feeding him. To highlight this point, S3 thinks that using a robot feeding would be faster, while S4 feels sometimes the caregivers hurry and expects "... the robot would adjust more to our time". Experts such as E9 described the assisted feeding activity as "time consuming, you can't really do something else at the same time", whereas E12 addresses the challenges with more empathetic perspective: "having patience to allow natural eating pace". This translates to the robotic feeding system allowing the caregiver some flexibility to focus their attention on other tasks or activities, which may include companionship and social interactions with the care recipient. Bite sequence is also an important consideration. S3 and S4 mentioned that

sometimes they choose what they want in their next bite, and sometimes they do not. This issue also relates to how much each user can participate during the meal, maximizing their independence and making the activity a positive experience, and "*more than a nutritional exchange*" (E2). Additional usability indicators (important in all assistive robotic systems) impact learnability, memorability (ease of training and retaining training on use of the technology). Usability also prevents potential user error. At the institutional level, one indicator could be the ability to identify patient trends (such as tracking decline in function or skill).

User Empowerment and Social Acceptance:

There are additional emotional and social supports that the RAF system can facilitate associated with eating as a daily activity that are addressed through this context. Several experts addressed the user's comfort level with phrases such as "preserve a patient's dignity" (E12), "person being fed may or may not enjoy having a person feeding them" (E6), and sometimes the user may "prefer a family member feed (them) because is more enjoyable" (E13). E3 commented:

Oftentimes meals are a time for socialization and eating. Often for people with disabilities that socialization piece is taken away because the caregiver may be so focused on feeding and not facilitate that social piece or they may be in a rush to finish feeding.

From the users' perspective, S3 and S4 mentioned feeling "embarrassed" when asking someone else to help them eat. S2 takes it further by imagining a social interaction where he meets someone in a coffee-house and the robot assists him with eating a donut. One of the caregivers (C3) also mentions the potential for increasing the care-recipient's self-esteem. On the topic of independence, the surveyed experts and caregivers vastly agree on that aspect being one of the greatest potential benefits of adopting a robotic feeding system. S5 also mentioned feeling like a child when he is being fed. This was also addressed by an expert (E9): "a degree of infantilization that can occur. It's such a personal thing to have someone shoveling food in your mouth, really represents a loss of independence across the board". Some experts also mention that the presence of an interaction-less robotic feeding system may result in potential social isolation. C2 explained that when a caregiver feeds a care recipient there are interactions that would go missing with a robotic system: "when we feed, we talk, we laugh, we ask how they are doing!". Experts further expand this issue by assigning the caregivers a role of companionship. E2 said:

When self feeding is possible in older patients, then the eating process can shift to one where the caregiver is present as a social partner/companion around a meal rather than having the experience be one that feels tied to a dependent relationship

As a part of the experts' survey, several Likert-scale questions addressed concerns or potential benefits of an assistive robotic system:

1. I am **delighted** to think of the INDEPENDENCE assistive robots can add to the life of a person with a disability.

- 2. I am **delighted** to think of the POSSIBLE TASKS assistive robots can perform in the life of a person with a disability.
- 3. I am **concerned** that individuals with disabilities would feel UNEASY if they had to use the help of robots to eat.
- 4. I am **concerned** that caregivers would feel UNEASY if they were given a job where they had to use robots.
- 5. I am **concerned** that robots would be a BAD INFLUENCE in the life of a caregiver.
- 6. I am **concerned** that robots would be a BAD ADDITION in the life of a person with a disability.
- 7. I feel it is **possible** for individuals with disabilities to DE-PEND TOO MUCH on robots.

The results are presented in Fig. 4. Overall, experts strongly agreed with the idea of increased independence and greater number of tasks that a person with a disability could accomplish using a robotic feeding system. Regarding the concerns of including robots in the process of feeding, experts were not concerned with the inclusion of technology for either stakeholder, care-recipient or caregiver. Experts also disagreed to consider adding a robotic system as a bad influence to caregivers or a bad addition to the care-recipients' lives. We averaged the Likert-scale responses across experts and used a t-test to compare them against a neutral statement, also used as null hypothesis (Likert score = 3). Normality assumption was checked using Skewness and Kurtosis. Using the experimental data, we found statistical significance for all the questions except (4) and (5) with p < 0.05.

System Integration:

This context encompasses how the robotic system can be integrated with other systems in day to day life. For example, the robotic system would need to share power resources with the electric chair. This leads to managing resources and maybe incorporating low-battery warnings and integrating with other assistive devices and alarm systems (for safety) to notify caregivers when anomalies are detected. Using this perspective, it is important to consider the impact when dealing with different levels of influence. To use the battery level as an example, in an intrapersonal level battery life might not even be considered while feeding since charging could be done at the same time. This may not be the case in the community level, where both wheelchair and RAF system share the battery resource and a minimum battery level might be required to consider having a meal in a restaurant. From our observations during the feeding activities, S5 was already using multiple assistive devices and technologies based on sip and puff.

A WORKFLOW FOR EVALUATION

We developed the following workflow for testing RAF technologies based on goal-driven contexts. First, any assistive technology needs to define the problem. In our case, the problem is to develop a robotic system that can assist people with feeding as a long-term care solution. The next step is identifying stakeholders. Based on the problem and the stakeholders, the system needs to be developed and evaluated.

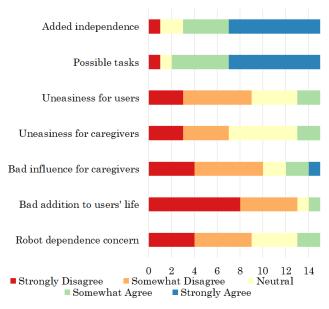


Figure 4: Experts' responses to Likert-scaled questions.

Identifying Stakeholders

This includes understanding different demographic factors that this technology attempts to address, including breakdowns by axes considering mobility profile (as defined in the framework), gender, race, environment, location, socioeconomic status and educational level. Usual stakeholders for RAF are people with upper-mobility impairments with conditions such as cerebrovascular diseases like strokes [79], Parkinson's, arthritis, multiple sclerosis [17], spinal cord injuries [52], bilateral amputations, and many others which can render individuals unable to eat on their own accord.

Evaluation Indicators

The goal for a RAF system is to increase the care-recipient's independence and maximize their participation during the activity while still providing adequate nutrition. Based on the problem definition and the stakeholders, the evaluation criteria can be specified from among the indicators given below. These indicators hinge on our observations during the feeding activity of the contextual inquiry as well as feedback from care-recipients, caregivers, and experts. We have come up with the following evaluation indicators:

Technical Function.

The evaluation of this context involves performance indicators related to bite acquisition and transfer efficiency as well as safety.

- *Range of food items:* This tells us what kind of food items can the RAF system successfully feed a person.
- *Bite acquisition success rate:* This tells us how successful a bite acquisition attempt by a robot is, for a variety of food items.
- *Bite transfer success rate:* This tells us how successful a bite transfer attempt by a robot is, for a variety of food items.

- *Face detection:* This is important for the robot to be able to detect where the mouth is.
- *User interface:* This tells us if a user interface is available for a robotic system.
- *Reachability and Mobility:* The robotic system should have a big enough workspace to be able to reach and pick food up from a range of plate positions and should be mobile such that the user can eat food wherever they want to, in their own homes, with their family, or even in a restaurant.
- *Collision Detection*: Safety is of paramount importance. The robotic system should be able to detect collisions and stop immediately in case of unexpected collisions.
- *E-stop:* The robotic system should have multiple redundant features such as emergency-stops and/or other modalities such as speech to command the robot to stop whenever they want.

Technology Robustness.

The evaluation related to this context mainly addresses the capability of a RAF system to be robust to perturbations.

- *Collision Avoidance:* Ability to adapt its trajectory to avoid colliding with unexpected static or dynamic objects or body parts in its path.
- *Visual Servoing:* Ability to adapt its motion to the movements of the care-recipient's neck/head or movements of the plate or bowl during the course of a meal.
- *Anomaly detection:* Ability to detect anomalies whether it is task, software, hardware, or environment-based.
- Anomaly correction: Ability to correct task-based anomalies such as bite-acquisition and bite-transfer failures.

Information Gaps.

The evaluation related to this context mainly addresses the robot-user communication of information.

- *User interface suitability:* Depending on the target subpopulation, a robotic system needs to be equipped with a user interface with a suitable modality that can be used to communicate.
- *Robot communication interface:* It is equally important for a robotic system to communicate what it is doing at a particular feeding phase so that care-recipient feels more connected to it. This could be especially important for the blind community.

Usability.

The evaluation related to this context focuses on the customization and usability aspects based on inputs from our stakeholders.

- *Ease of bite transfer:* This tells us how easy a bite transfer attempt for a robotic system is.
- *Bite transfer fluency:* One of the factors of a fluent bite transfer, which requires fluent human-robot interaction, is proper bite timing. Proper bite timing is important for an

enjoyable eating experience so the care-recipient does not feel rushed or is kept waiting.

- *Bite size adaptability:* Ability to adapt how much food it picks up in one bite cycle according to a care-recipient's preference.
- *Bite timing adaptability:* Ability to adapt when to feed a care-recipient based on their preference so that the entire meal is a positive experience for the care-recipient.
- *Bite sequence adaptability:* Ability to feed food items in certain order based on a care-recipient's preference or their nutrition requirements to make the entire meal an enjoyable experience.

User Empowerment and Social Acceptance.

The evaluation related to this context addresses user empowerment and how likely it is for them to socially accept this technology in longer term. We base these indicators on Technology Acceptance Models (TAMs) [29, 38, 33, 30, 82, 83, 84, 27, 70, 81, 39, 72].

- *Perceived Usefulness:* Be useful such that the care recipient's subjective probability using the technology will increase his or her performance
- *Perceived Ease of Use:* Degree to which the care-recipient expects that using the robotic system would be free of effort for the feeding task.
- *Attitude:* A care-recipient's positive or negative feelings (evaluative affect) about using the robotic feeding system.
- *Intention to Use:* A care-recipient's intention to perform a specific behavior to use the robotic system for feeding tasks.
- *Perceived Enjoyment:* Extent to which the activity of using the robotic system is perceived to be enjoyable in its own right, apart from any performance consequences that may be anticipated.

System Integration.

The evaluation related to this context addresses the compatibility of the robotic system with other systems in the home so that proper integration is achieved.

- *Onboard electronics:* Onboard sensing, computation, and networking capabilities so that it can feed people with minimal changes to the in-situ environment of the care-recipient and can communicate with other networked devices if needed.
- Anomaly monitoring capability: Integration with an alarm system to indicate low levels of battery, choking hazards, or anything else that needs immediate attention such as irregular bib placement, irregular hydration etc.
- *Integration with other assistive devices:* Ability to integrate with other assistive devices that are present in a carerecipient's home such as a wheelchair or an autobed [37].

EVALUATING ROBOT-ASSISTED FEEDING SYSTEMS

In this section, we describe some of the existing robotic feeding systems and evaluate them according to our framework. We categorize these assistive robotic feeding systems into three categories: 1) table-mounted robotic arms, 2) general purpose mobile manipulators, and 3) wheelchair-mounted robotic arms and analyze one representative robot from each of these categories.

Comparing existing robot-assisted feeding systems

Table-mounted Robotic Arms: Obi

Out of the myriads of table-mounted robotic arms for feeding mentioned in Section 3, here we take a closer look at *Obi*. Obi is a commercially available table-mounted portable assistive feeding device (See Fig. 2, left) and weighs approximately 7 lbs. It can detect collisions and can feed up to 4 hours with one charge. It has minimal autonomy but has a teaching mode by which its trajectories can be pre-programmed for later use. Obi has two user controls to command food delivery including multi-directional food capture, food re-positioning, and portion control, and to choose among four food bowls. It uses a washable spoon to scoop food from the food bowls.

General Purpose Mobile Manipulators: PR2

Out of the general purpose mobile manipulators for feeding mentioned in Section 3, here we focus on PR2. PR2 is a 32-DoF general-purpose mobile robot manipulator [20] with an omni-directional mobile base, a 1-DoF telescoping spine, and two 7-DoF back-drivable arms for developing a RAF system [62]. The robot holds a bowl and a utensil and the system can accept commands from a user via a graphical user interface. The system can scoop a spoonful of food by locating the highest food location in a bowl using a headmounted RGBD camera, feeds a user by estimating the face and mouth pose using a wrist-mounted camera, and can also wipe the spoon [62]. While running the tasks, the system runs a multimodal execution monitor to detect anomalous behaviors using 6 different sensors [61].

Wheelchair-mounted Robotic Arms: ADA

Out of the wheelchair-mounted robotic arms for feeding mentioned in Section 3, we focus on our autonomous robotic feeding system, ADA [36]. Our system consists of a 6-DoF JACO robotic arm [2] mounted on a powered ROVI wheelchair [8] to mimic similar setups used in real homes. The robotic arm has a wrist-mounted camera and has 2 fingers equipped with tactile sensors that grab a fork. We used tactile sensing to control the end effector forces during skewering and to detect if food acquisition was successful as well as if the fork hits something unexpectedly to improve safety. Depending on what food item a user selects, the robot perceives the food item on the plate using perception algorithms [36] and decides the best strategy to pick it up. Once the algorithms decide the strategy, the robot arm servoes to it using feedback from the visual modality, acquires the bite using feedback from the haptic modality [18], and then feeds it to a person by detecting the face and servoing to it using feedback from the visual modality.

Evaluation Contexts	Indicators	Obi			PR2			ADA		
		L1	L2	L3	L1	L2	L3	L1	L2	L3
Technical Function	Range of food items [†]				•					
	Bite acquisition success rate	-	-	-	_	-	-		•	
	Bite transfer success rate	-	-	-	_	-	-		•	
	Face Detection	X	X	X						
	User interface availability									
	Reachability and mobility $^{\otimes}$		•				•			
	Collision detection									
	E-stop									
Technology Robustness	Collision avoidance	X	X	X			•			
	Visual servoing	X	×	X	X	X	X			
	Anomaly detection *	X	X	X					•	•
	Anomaly correction *	X	X	X					•	•
Information Gaps	User interface suitability °		•		•				•	
	Robot communication interface	X	X	×						
Usability	Ease of bite transfer	-	-	-	-	-	-			
	Bite transfer fluency \diamond		•						•	
	Bite size adaptability [‡]							X	X	X
	Bite timing adaptability [‡]		•	•	•					
	Bite sequence adaptability [‡]				X	X	X			
User Empowerment and Social Acceptance	Perceived usefulness *			-			-			-
	Perceived ease of use			-			-	-	-	-
	Attitude *			-			_			_
	Intention to use *	-	-	_	_	_	-			_
	Perceived enjoyment			-	_	-	—	-	-	_
System Integration	Onboard electronics									
	Anomaly monitoring capability Integration with other assistive devices	X X	X X	X X	×	×	×	×	×	×
					r					· ·

Table 1: Evaluation of three existing robot-assisted feeding systems: darker colors and larger circles indicate higher ratings

[†] Obi can scoop a variety of food items; the PR2 system can handle yogurt, cereal, cottage cheese, watermelon, fried rice, and mixed fruit; ADA can handle 16 different solid food items including salad leaves.

[®] Obi is portable but table-mounted and difficult to setup in a new environment; PR2 is mobile but it is difficult to move it everywhere a user goes.

PR2 system can address a wider range of anomalies compared to ADA.
Obi uses buttons (requires hand movement); PR2 uses GUI (requires some arm movement); ADA uses speech (requires voice).

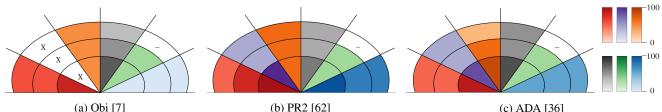
 $^{\diamond}$ ADA can autonomously trigger feeding when the mouth is open.

[‡] These systems cannot adapt autonomously. They can be triggered by a user and thus, can adapt to the pace and portion as needed. ADA has the capability to detect and select a particular food item on a plate autonomously, and can thus follow a user-specified bite sequence.

* For ADA, evaluations were done using a robot video.

Evaluation

Based on the evaluation indicators developed in the metrics subsection, we performed an evaluation of the three robotic systems: Obi, PR2, and ADA. Detailed results are given in Table 1 and Fig. 5. Note, not all the measures were available for these robotic systems publicly and thus, for the unavailable information, we put a '-' instead on the table. Note, though Obi is commercially available, some of the known barriers to its user acceptance as reported by one of its distributors are due to minimal autonomy. For example, the most frequent problem is that people 'bite' the spoon hard which makes it fall off and then the user can not reattach it by themselves due to their impairments.



(a) Obi [7] (b) PR2 [62] (c) ADA [36] Figure 5: Summary comparison: we used the average for each level from Table 1. Darker colors imply higher ratings.

DISCUSSION

Evaluating RAF systems from a social-ecological approach presents a number of limitations. First, it is not obvious how to facilitate data collection and assess the reliability of data within the four levels of influence (in particular the larger scale ones). An exciting research challenge will be to understand the heterogeneous technological competencies and technology adoption rates among the various levels. As a sensing technology, RAF systems may make detailed quantitative data available for questions for which only qualitative data has traditionally been available, and data analysis could displace statistical approximations in examining certain questions at the macro-levels. Exploring the possibilities for evaluating RAF systems as multi-leveled through the SEM lens is the first step towards developing a universal reporting system that will provide comparable data related to each level of the model. The next challenge will be developing ways to synthesize multi-level information to draw conclusions that are broad enough for generalization yet specific enough to be useful to policy and community stakeholders. In addition, using this evaluation framework, we have identified 3 research milestones that our system, ADA, needs to achieve in the future: a) User Empowerment and Social Acceptance: Testing with end-users, b) Technology Robustness: Anomaly detection and correction, and c) Technical Function: "Range of food items". These were chosen not only based on the scores and where we lack compared to other systems based on these scores but these were also validated from user, caregiver and/or expert responses through our previous contextual inquiry.

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