

# Instructions for ACL-2016 Proceedings

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## Abstract

This should be a 6-8 page conference paper with appendices, if relevant. Good reports from last year: 1 and 7

## 1 Introduction

*from coursework spec:* main research or technical question addressed

Socially assistive robots (SARs) are a crucial part of the future of many sectors, for example, in education or healthcare (Gunson et al., 2022). Especially the latter depends on technology advancements as it is facing numerous obstacles in the future, such as increasing spendings and a growing percentage of older people. A serious lack of healthcare workers is already occurring, with 10 million more healthworkers needed worldwide by 2030 (Cooper et al., 2020; WHO, 2023). SARs can pose a solution to the problem, as they are able to support healthcare in various ways, such as encouraging older people to keep living independently for longer or reducing caregiver burden (Cooper et al., 2020).

These scenarios require SARs to be able to handle multi-party interactions as it is likely that more than one person will interact with the system. Compared to handling dyadic interactions, handling multi-party conversations includes more complex challenges, such as Speaker Recognition, Addressee Recognition, Response Selection (summarised in “who says what to whom”) and turn-taking (Addlesee et al., 2023; Johansson and Skantze, 2015).

*Include here what exactly we examined about turn-taking*

In this work, we propose a model trained on multi-party human-human conversation data. We collected the data from recordings of special “Who

wants to be millionaire?” episodes where two candidates collaborated to answer the host’s questions.

*Include results here.*

## 2 Background

*from coursework spec:* literature review / related work, including a critical analysis of the field, and commentary on applicability of the technologies and methods used in emerging technologies and application areas

### 2.1 Socially Assistive Robots

For healthcare, as well as for any other sector, the difficulty of successfully designing SARs lies in creating robots that can effectively converse with humans and adhere to social norms (Moujahid et al., 2022). The more expressive a robot is, the more it will be perceived as intelligent, conscious and polite (Moujahid et al., 2022). To achieve such a positive perception, multiple parts need to be combined into one conversational system, such as the ability to carry out visually grounded as well as task-based dialogues, to perceive and discuss its environment and to chit-chat (Gunson et al., 2022).

The SPRING project conducts research on a SAR robot that is deployed in an eldercare hospital reception area (Addlesee et al., 2020). The conversational system is deployed on humanoid ARI robot produced by Pal Robotics (Robotics, 2023). ARIs capabilities can be extended with custom AI algorithms, in the case of SPRING-ARI a visual perception system, a dialogue system, and a social interaction planner (Addlesee et al., 2020). While the SPRING-ARI system successfully demonstrates that task-based, social and visually grounded dialogue can be combined with physical actions, it still lacks the ability to handle

conversations with more than one person simultaneously (Addlesee et al., 2020).

## 2.2 Multi-party Human Robot Interaction

As stated above, the endeavour to create conversational systems becomes considerably more difficult when dealing with multi-party interactions (Addlesee et al., 2023). Especially turn-taking poses a central problem. It is defined as follows:

The rules of turn-taking organize the conversation into turns, during which one of the participants has the right to speak while the others agree to listen (Żarkowski, 2019)

In dyadic conversations, there are only two roles a participant can take: speaker or listener, hence it is clear when and to whom the turn is yielded. In multi-party conversations, participants can take multiple roles, therefore turn-taking needs to be coordinated (Johansson and Skantze, 2015). Humans signal their intents mostly through gaze, but also through pauses, prosody, and body positioning (Żarkowski, 2019). To copy this behaviour, earlier models for conversational systems relied on silence time-outs to coordinate turn-taking, however, this approach is found to be too simplistic (Skantze, 2021). Instead, mimicking human turn-taking behaviour better by using a combination of verbal and non-verbal cues leads to robots that are better perceived (Moujahid et al., 2022).

*State exactly the gap that we will fill - whatever that will be*

## 3 Data Collection

- (Laura) Talk about multi-party data collection
- (Aron and Katie) How we collected our data, describe the intents we used to label the data
- (Aron and Katie) Cohen's Kappa for our data collection method: probably need to annotate a couple of transcripts twice for reporting on this

## 4 Design and Implementation

- *from coursework spec:* design and implementation of the system: components and architecture
- Jack's System Graph

## 4.1 Natural Language Understanding

- Very short description of RASA
- Refer to the intents described in 3 Data Collection
- Different versions of the model: show difference in F-Score, Confusion Matrix, ... (Laura: look up in literature what is used, what should be used)

## 4.2 Dialogue Management

- Clearly explain 2 parts (State-Machine and NN)
- State-Machine: high-level control, handles things we have no data for, eg. pauses
- NN: Report on differences between RNN and LSTM and why the choice for the LSTM has been made
- (optional): Comparison to RASA rule-policy

## 4.3 Natural Language Generation

## 5 Evaluation

*from coursework spec:* evaluation of the system and presentation of the results

### 5.1 Methodology

### 5.2 Experiment Layout

### 5.3 Results

## 6 Conclusion

### 6.1 Ethical Reflection

## 7 Future Work

*from coursework spec:* suggestions for future work

## Acknowledgments

The acknowledgments should go immediately before the references. Do not number the acknowledgments section. Do not include this section when submitting your paper for review.

## References

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## **A Supplemental Material, Appendix**