Collaborative Problem Solving in Human-Robot Interaction

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1 Introduction

1.1 Problem Statement

Collaborative Problem Solving is a multidisciplinary problem in Human-Robot Interaction which brings together much of the current work in robotics, natural language, human-computer interaction, and artificial intelligence. Human and robot participants must be able to understand each other without impediment to the task, must understand each other's roles and intentions in the task (intention recognition), and must agree on steps to be performed to complete the task (collaborative problem solving and joint intention). The robots must also be careful not to take actions which might endanger the human participants (safety). Speech output can be accompanied by facial expressions or gestures, such as looking at and/or pointing to the object being referenced. This an example of multimodal interaction.

1.2 Area

Human-Agent Interaction is an extension of Human-Computer Interaction to interactions with software that exhibits some form of agency, potentially embodied and situated in a virtual world. Human-Robot Interaction is an extension of Human-Agent Interaction to interactions with embodied, situated agents in the physical world. [GS07] gives a survey of the field of Human-Robot Interaction.

Social Human-Robot Interaction studies robots which interact by social means with humans and each other, including gestures, facial expressions and speech. [Fon03] gives a survey of Social Human-Robot Interaction.

Human-Robot Collaboration studies social interaction between humans and robots in the context of a shared task to be performed. [BWB08] gives a survey of Human-Robot Collaboration.

1.3 Overview

In this report we will cover several projects in the area of collaborative problemsolving in human-robot interaction between a single robot and single human participant. We will in particular focus on approaches to natural language generation, intention recognition and the use natural conversation in shared planning and execution of a task. In Section 2 we will look at Human-Robot communication in general. In Sections 3,4, 5 we will look at different types of communication. In Sections 6, 7,8 we will look at how the collaborative problem-solving aspect comes together. Finally in Section 9 we will discuss and summarise.

2 Human-Robot Communication

At the core of Human-Robot Communication, is the notion of having multiples *modes* of interaction or communication. These interactions can be generated from or interpreted to an internal grammar by the robot system. The 'sentences' in the grammar are operated on by a dialog engine, which interacts with the task planning system and the robot's internal model of the world.

In terms of communication from the robot to the human, there is some leeway in which input and output modalities are provided to the robot. The Leonardo project[?] focuses on an expressive robot capable of many facial and body gestures and without any natural language generation or speech synthesis capabilites. The JAST project [TODO] combines two industrial arms with the iCat expressive robot face, and features speech generation and recognition. [vB04] describes the Lino and iCat robots and the principles used to animate them. Museum guide robots such as INDIGO typically feature speech but not gesture modes of communication [TODO] although the Robotinho project does use gestures [TODO].

3 Natural Language Processing and Generation

In general, Speech Recognition and Generation are treated as solved problems in human-robot interaction, with some leeway for the inevitable interpretation errors from either participant. Speech is then treated in terms of text. Natural Language Processing is a well established field dealing with processing the text into some more abstract model. Natural Language Generation is the counterpart of Natural Language Processing. It refers to the production of human text or speech from some internal data structure specific to the agent. In the context of this review the most important aspects are communicating the actions to be performed and the objects they should be performed on. In fact, generating descriptions of objects and their location is easy; the difficult part is deciding which information to include or not in the description. If there is not enough information, the description can be ambiguous or confusing for the human listener; if the information is too specific, it may seem to the listener as though the extra information was included for a specific reason even if it is irrelevant to the task. In [FGI⁺09] the JAST project has investigated different strategies for referring expressions and their impact on both task performance and user satisfaction. In [BKS⁺09] the GIVE challenge evaluates different strategies for guiding a user in an online game to move through an environment and perform actions. [BG08] find that task performance metrics do not correlate with metrics that measure how "human-like" the output of an NLG system is.

4 Gesture Recognition and Generation

When humans communicate in a shared task they will often use gestures as well as speech. These may serve several purposes: to convey intention or emotion, to direct the other participant's actions or attention, or to directly execute a step in the plan for the task. For example, a pointing gesture could direct the other participant's attention to a tool or area, or it could tell them to move themselves to the location pointed to.

In a robot, gesture recognition will be performed as a step after the visual interpretation of a scene from video data.

4.1 Hand-over of objects

Extending an arm holding an object towards the other participants signals the intention to hand over the object while also being the first step of the execution of this intention. $[KSS^+06]$ and $[HRK^+08]$ explore handing-over gestures and approaches that are comfortable and recognisable to the human user.

4.2 Intent Recognition

[NDK⁺05] looks at classifying human gestures and recognising intent, and find that the context as well as the gesture itself need to be taken into account. The JAST robot can infer the intent of the human user when they pick up pieces and correct them when they pick up the wrong piece [RKF⁺08].

5 Multimodal Interaction

When performing a task, humans have many simultaneous modes of interaction: speech, facial expressions and stances, gaze direction, gestures, direct physical contact. It may be beneficial or even necessary to consider multiple modes at once to understand the overall meaning.

In the JAST project, the generation of referring expressions and the use of gestures were both studied and evaluated for their effectiveness in communicating with the human participant [TODO]. [Van05] gives an algorithm for the generation of multimodal referring expressions. In the Leonardo project, speech output was avoided entirely in favour of an expressive body and facial design. [TODO] describes the MultiML language for representing multimodal actions in a dialogue.

6 Collaborative Problem-Solving

[LT00] describes the TRINDI dialog engine toolkit. It is based on the notion of a shared or individual *information state* which is updated by the *dialog moves* of the participants. The tookit defines the basic data structures and some dialog moves, but the precise information and the choice of dialog moves can be selected according to the task required. Simple examples of dialog moves are asking or answering a question. The TRINDI toolkit has been used in the JAST project.

[BA05] propose a different system which includes dialogue moves (which they call *interaction acts*) which are used to negotiate, accept, or reject changes to

the shared problem-solving state, such as deciding to focus on a particular subproblem or adopt a certain solution. In addition, there are wrapped in *grounding* acts which handle turn-taking in conversation, requests for acknowledgement, and requests for clarification.

Joint Intention refers to the state of affairs where several participants share a common goal and a common plan for achieving that goal. In order to reach this situation, both participants must continually communicate their intentions as the execution of the task progresses. In the above system, joint intention is achieved by negotiating every change to a shared problem-solving state.

In the Leonardo project, the human participant sets the goal and teaches the robot the steps required [TODO]. For example, the robot is taught to press a button, and then is given the task of pressing several buttons.

In the reverse direction, in the JAST project, the robot participant sets the goal and teaches the human the steps required [TODO]. The human is taught sub-tasks which are then combined into a larger task.

[FKHB06] describes NASA's HRI/OS, developed as part of the Robonaut robotic astronaut project. HRI/OS allows for a wide range of interactions between humans and robots, from remote teleoperation to local collaboration. Robots using this system can request help from humans or other robots when they are unable to complete a task by themselves.

7 Learning

In terms of working on a shared task, the Leonardo project focuses on the human teaching the robot how to participate in a task. The idea is that active tutelage can be much more effective than relying on blind experimentation by the robot or complex, brittle a priori knowledge in getting the robot to perform a new task.

8 Mixed-Initiative Interaction

When both participants contribute towards the goal, this is termed a Mixed-Initiative system. [Gui96] describes an early model of mixed-initiative communication based on exchanging information and deductions between agents until a conclusion is reached. [BHL04] describe how Leonardo can suggest it takes the initiative or request help from its human partner. [BA05] describe a model for negotiating shared goals and plans between participants, and use it to analyse a planning discussion between two human participants. [FA07] describe an architecture for implementing a similar model in an agent.

9 Discussion

Collaborative Problem-Solving is a dynamic and promising area of Human-Robot Interaction research, with applications in space[TODO], medical and elderly care[TODO], and any context in which it is desirable to have autonomous robot participants perform tasks in a dangerous environment in collaboration with human participants, such as military applications[TODO] and urban search

and rescue[TODO]. Collaborative problem-solving, mixed-intiative and multimodal interactions allow for natural integration of robot members in a team without requiring special training of the human participants, and as the robot becomes more aware of intent and social cues, it is regarded less as a tool and more as a participant[TODO].

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