

A Research Platform for Multi-Agent Dialogue Dynamics *

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Abstract

Dialogue agents are often designed with the tacit assumption that at any one time, there is but one agent and one human, and that their communication channel is exclusive. We are interested in evaluating complications that arise when multiple heterogeneous dialogue agents interact with a human interlocutor, and their communication channel is necessarily shared. To this end we have constructed a multi-agent dialogue test-bed on which to study dialogue coordination issues.

1 Introduction and Motivation

1.1 Spoken Language Interfaces

Spoken language interfaces are the modality of choice for fictionalized accounts of future robots, but have not been a modality of practical consideration until the rather recent computer speed advances and statistical speech recognition breakthroughs of the past few years. Today, given relatively small domains of understanding, automatic speech recognition is at least able to offer itself as an alternative modality, and dialogue systems that take into account the loss of accuracy are able to mitigate its effects.

Thus spoken language interfaces are in the process of becoming technologically feasible, but feasibility aside, what may the best modalities be? The preferred modality of course depends on the tasks at hand, but one ethnographic study (Brumitt & Cadiz, 2001) shows that for some simple and very common tasks, people prefer speech interfaces to any other modality. Although much more is still left to be learned about the preferences of communication modalities, spoken language's high status in human-human communication is likely to be a continuing driving force behind

the preference for spoken language in human-machine communication.

1.2 A Future of Heterogeneous Spoken Dialogue Agents

The recent feasibility and preference for spoken language interfaces has led to an explosion of dialogue agents. VoiceXML alone boasts tens of thousands of applications (Byrne et al., 2004), and is represented primarily only in the telephone sector. The number of users of spoken dialogue agents would be difficult to estimate; but, a single service, AT&T's toll-free directory assistance, alone services 200,000,000 calls per year. It is conceivable that the number of spoken dialogue agent users exceeds even the number of home computer users worldwide.

Spoken dialogue agents, which were only recently specialty products and have their mainstay in computer applications and telephone services, are now found embedded in mobile phones, information kiosks, audio/video equipment, automobiles, toys, personal digital assistants, video games, and robots (Hoge et al., 1999). They are accessible, but the interfaces are almost always integrated into their applications. As the number of appliances with embedded spoken dialogue agents increases, we will be living in environments of multiple heterogeneous dialogue agents.

1.3 Problems with the Communication Channel

Multiple spoken dialogue agents will face some of the same main communication issues that have been active areas of research in the domain of computer networks, namely the issues of message identification, message addressing, channel contention, and session identification.

1.3.1 Message Identification

With multiple spoken dialogue agents in an environment, there is the unintended potential (and some-

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times the need) for dialogue agents to speak to each other. Dialogue agents will undoubtedly misbehave unless they can identify who is speaking. Speaker identification has had some success recently and may be employed to address this issue, but serious issues remain to be resolved, such as its scalability.

1.3.2 Message Addressing

With multiple spoken dialogue agents in an environment, there is an unintended potential that a dialogue agent will mistakenly believe it is the intended recipient of an utterance, whereas in reality the intended recipient may be another dialogue agent, or no dialogue agent at all. Methods employed to identify message addressing use evidence from acoustic, linguistic, and even pragmatic sources of knowledge to determine whether a particular utterance actually addresses the agent or not. Other research focuses on the additional information brought to bear by other communication modalities such as gesture, gaze, and touch. The results of such research, however, are far from optimal, and usually make two rather strict assumptions. One, there is the assumption that the system has deep knowledge of the domain, or that the system's domain is very small. Two, there is the assumption that the agent is the only dialogue agent in its environment, or that other dialogue agents in the environment have domains that are sufficiently different from its own.

1.3.3 Channel Contention

With multiple spoken dialogue agents in an environment, there is the potential that they will speak simultaneously, or that they will interrupt each other. Methods currently employed to resolve these contentions often consist simply of waiting until nobody else is speaking for a second or so, and then to begin speaking until someone interrupts. In environments with one dialogue agent and one human interlocutor, this algorithm is usually sufficient, but studies of larger human group dynamics show that many important subtleties are missing. Agents who do not indicate their desire to speak, or who do not introduce themselves before they speak are likely to not be understood by their listeners, or, at the very minimum, such agents are likely to add to the cognitive load of their listeners.

1.3.4 Sessions Identification

Although communication between humans and agents is improved when those agents introduce themselves before speaking, they cannot introduce themselves before every utterance. A concept of a communication session must be developed.

2 An Experimental Multiple Agent Dialogue System

We have identified four communication issues that will always arise in environments of multiple dialogue agents: message identification, message addressing, channel contention, and identifying communicative sessions. We have engaged in a systematic approach to finding good specific solutions to these problems through experimentation. To this end we have developed a multi-agent dialogue (MAD) system, which can accommodate multiple dialogue agents in a single experimental framework (see Figure 1). The system works both with real robots adapted for the Carmen robot platform (Montemerlo, Roy, & Thrun, 2002), and in a simulated Carmen environment.

The front-end architecture is an instance of the Galaxy-II spoken dialogue system reference architecture (Seneff et al., 1998). We use Sphinx-II (Huang et al., 1993) for automatic speech recognition, Phoenix (Ward, 1994) for context-free grammar parsing, Helios (Bohus & Rudnicky, 2002) for confidence annotation, Ravenclaw (Bohus & Rudnicky, 2003) for dialogue management, ROSETTA (Oh & Rudnicky, 2000) for natural language generation, and Festival (Black, Clark, Richmond, & King, 2004) for text-to-speech rendering. The robots, named Bashful and Clyde (ghosts from Namco Ltd.'s Pac-Man[®]), each have their own Ravenclaw dialogue system. Ravenclaw is a generalized tree-based dialogue management framework that provides the designer of a dialogue management system with mechanisms through which to specify dialogue tasks. Essentially a designer specifies the various actions that must take place in the system (e.g. the action to be taken when the user asks a robot where it is) and the flow of the dialogue.

The back-end consists of programs that use the Carmen set of libraries to communicate with the robots. The libraries currently utilized in our project include those that allow the user to send messages to the robots to get them to move a specified distance in a certain direction, and those that allow the user to set a goal position and then allow the robot to plan a route to that position.

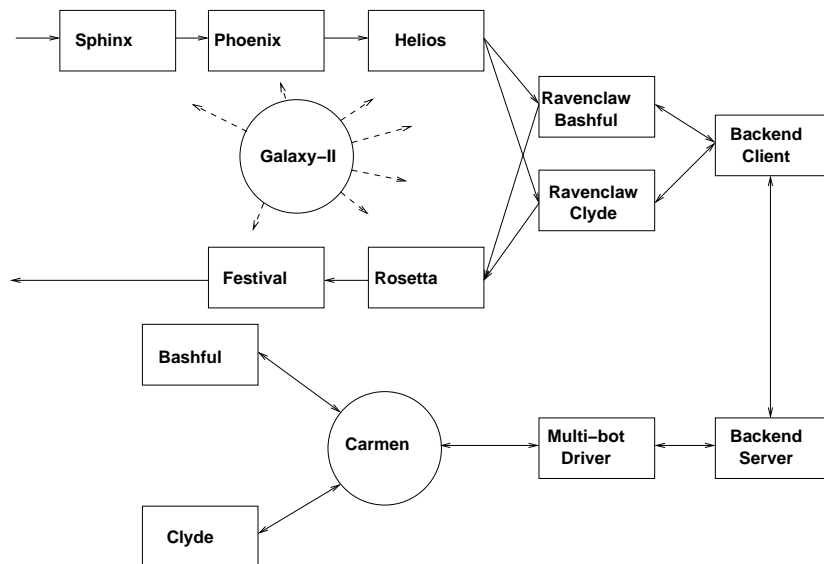


Figure 1: Multiple Agent Dialogue System

3 User Studies

3.1 Goals

We have used the system described above to conduct a preliminary set of pilot studies. We have two goals in performing these studies. The first goal is to establish the *usability* of our system, that is, we ask the question: Can this system be used by a human being to successfully interact with the robots? A large system like the one above can fail to be usable for a variety of reasons: the speech recognition may be too error prone, the speech synthesis may be unintelligible, the pace of the interaction too slow, the robot navigation libraries too unreliable, etc. Our experimentation is designed to show that our system can indeed be used to interact naturally with the robots. The second goal of our studies is to experiment with a very simple mechanism for dealing with multi-agent communication issues. Specifically, we tested a simple strategy for disambiguating the intended addressee of each user utterance.

3.2 Addressee Disambiguation Algorithm

We have implemented a simple, first-cut algorithm for disambiguating the intended addressee of each user utterance as follows:

- If an utterance starts with the name of a robot, then that is the robot this utterance is addressed to. We call this form of addressing *explicit addressing*, and the robot being addressed the *explicit addressee*. For example, in the utterance *Bashful, where are you?*, the form of addressing is explicit, and Bashful is the explicit addressee.

- If an utterance does not start with the name of a robot, then the last explicitly addressed robot is being addressed in this utterance. We call this form of addressing *implicit addressing*, and the robot being addressed the *implicit addressee*. For example, if the utterance above is followed by the utterance *Go ten meters north*, the form of addressing is implicit, and Bashful is the implicit addressee.

3.3 Task Description

In our experiment, users were required to navigate the two robots (Bashful and Clyde) through a maze of corridors using only the speech channel to communicate with the robots. The users were not allowed to see the robots and therefore had to rely on spoken dialogue to query the robots regarding their positions in the maze at all points of time.

Specifically, the task involved the following two sub-tasks:

- Find out the initial positions of the two robots in the maze.
- Navigate the robots from their initial positions to the point in the maze marked with an X.

Figure 2 shows a map of the maze used. Participants were provided with a hard-copy of this map (without the initial locations of the robots) and were asked to mark on it the positions they believed the robots were initially. They were also allowed to write on the map during the task if they wished. Users were given a

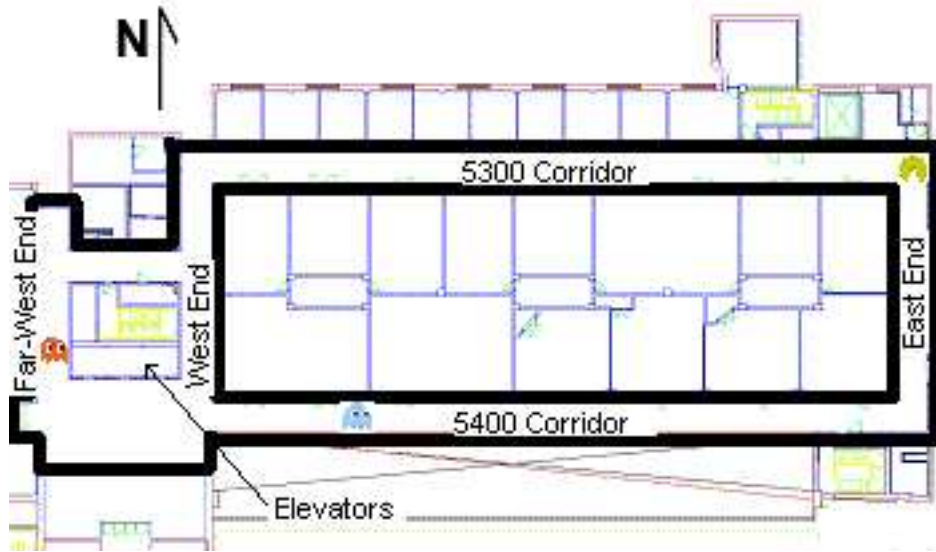


Figure 2: Maze Map

maximum of 30 minutes to finish the task. Note that users were not informed of the addressing mechanism described above since one of the aims of this experiment is to determine if the addressing mechanism can be intuited by the users and, in general, if it makes for naturalistic dialogue.

3.4 Grammar for User Utterances

We used the following simple grammar to parse the user's utterances:

- `HumanReportCommand` → `([RobotName]? report) | ([RobotName])`

`RobotName` can be either `Bashful` or `Clyde`. The user can either utter “Bashful” or “Bashful report” to address `Bashful` explicitly, or just utter “report” to implicitly address the last explicitly addressed robot.

- `HumanLocationQuery` → `([RobotName]? where are you)`

This command can be used to query the robot's location. The user can utter the name of the robot or omit it to engage in explicit or implicit addressing respectively.

- `MoveVector` → `([RobotName]? MOVE [Direction]? [Distance]?)`

This command can be used to direct the

robot to move a certain distance along a certain direction. `Distance` can be any integer distance from 1 to 20 meters, while `Direction` can be either north, south, east or west. Although utterances that do not contain both a direction and a distance are parsed by the grammar, both pieces of information are needed to perform the move. Hence in a situation where the user provided only one or neither of the pieces of information, the dialogue manager would ask the user to supply the missing information.

Although users were not presented with the above grammar, they were informed that they could ask “Where are you?” and that they could tell the robot to move between 1 and 20 meters in one of the four directions.

3.5 Robot Responses to User Utterances

The Ravenclaw dialog system used in this platform requires the designer to specify the response of the system for every parse-able user utterance. Responses may include speech output, back-end actions taken by the system, or a combination of the two. Following is a description of the system responses for each of the three families of user utterances described above.

- **Response to `HumanReportCommand`:** Robots responded to this family of utterances by saying *Bashful here*, or *Clyde reporting*, etc. This dialog helps the user initiate a communication channel with a robot.

- **Response to HumanLocationQuery:** Robots responded to this family of utterances by specifying where in the maze they were. Each part of the map a robot could be in was pre-assigned a name as shown in figure 2. The system’s back-end mapped the robot’s absolute (x, y) coordinates obtained from the CARMEN robot API to the corresponding area name. The system also computed the approximate distance from the closest end of the area (east or west end for an area that is longer along the east–west axis than along the north–south axis, and north or south end for north–south oriented areas). A typical reply to a HumanLocationQuery is *I am now in the fifty three hundred corridor, about five meters from the east end.*
- **Response to MoveVector:** The addressed robot responded to this family of utterances by first making sure it had a value for both the distance and the direction components. If one or both values were missing, the system engaged the user in a follow-up dialog by asking, for example, *How far do you want me to go east?* or *In which direction do you want me to go five meters?* The user could reply to these questions by saying, for example, *Five meters* or *Move north five meters.* Once both values were provided, the system used the robot’s current (x, y) coordinates to compute the destination position, and then used CARMEN’s autonomous navigation API to move the robot to the new position. At the same time, the robot would inform the user exactly how far it was going and in which direction as a confirmation. For example, the robot would say *Going five meters toward the north.*

3.6 Other Details

For this pilot study, we used simulated robots in a simulated environment. These robots were initially placed at the positions shown in Figure 2 for each participant in this user study. Participants used a single head-mounted close-talking microphone to speak to both robots, and the speech from both the robots was routed through a single set of speakers. To help the user to distinguish between the speech from the two robots, we used a male voice to synthesize the speech from Bashful, and a female voice to synthesize the voice from Clyde.

3.7 Results

We ran the experiment with 6 different participants. Table 1 summarizes the results of the experiment. Every participant could correctly identify the

Table 1: *Pilot-study results*

Participant #	Task Success	Time Taken (mins)	Addressing Mechanisms Used
1	Both	28	Only explicit
2	One	21	Both forms
3	Both	28	Both forms
4	None	18	Both forms
5	One	20	Only explicit
6	One	12	Only explicit

approximate initial positions of the robots on the map. For each participant, we measured task success by noting the positions of the robots at the end of the experiment.

During the experiment we noted what addressing mechanisms, explicit or implicit, the participant was using in their utterances. Three participants used only the explicit form of addressing; that is, each of their utterances was prefaced with the name of the robot. We asked each of these participants at the end of the experiment whether they realized that they could engage in implicit addressing, or if they simply chose not to. All three replied that they did not realize that implicit addressing was possible. The remaining three participants used both forms of addressing.

3.8 Analysis

When asked after the experiment how the interaction felt, every participant replied that they found both the dialogue and the pace of the interaction naturalistic. These reports established that our implemented system can be used successfully to interact with robots. Furthermore, the fact that every participant understood the explicit addressing mechanism and half the participants understood the implicit mechanism implies that our simple addressee disambiguation algorithm is easy to understand and makes for natural dialogue. A snippet of typical dialogue between a user and the robots follows.

USER: Bashful where are you?
 BASHFUL: This is Bashful. I am now in the fifty three hundred corridor, about three meters from the west end.
 USER: Go twenty meters east.
 BASHFUL: Going twenty meters toward the east.
 USER: Clyde?
 CLYDE: Clyde reporting.
 USER: Where are you?
 . . .

After completing the experiment, participants were asked to provide feedback on any aspect of the experiment. Every participant felt that the robots do not always provide as much feedback as they could. For example in the current design when a robot is asked to go further than it can, they do not report this inability. Participants also expressed satisfaction at having two robots to work with instead of one. They felt that the pace of the interaction would have been too slow if there was only one robot, since robots take a long time to move from one point to another. Participants also felt that the set of commands that the user can issue was limiting. When the robot was stuck against an unknown obstruction, the participants felt that more exploratory commands such as *What can you see?* would have been very useful to make progress toward the goal. Some participants felt that descriptions of the locations as spoken by the robots were sometimes unintuitive. For example, when the robot said *I am in front of the elevators about 3 meters from the east end*, the robot was referring to the east end of ‘in front of the elevators’. Since “in front of the elevators” is not normally the type of area with clear boundaries, the subject naturally thought that “east end” was the east end of the map. Further research is necessary to determine how to describe the current location of a robot such that the description is maximally intuitive from a human’s point of view.

4 Conclusions

Heterogeneous interface agents cannot act in concert, achieving a globally optimal interface strategy by understanding or predicting each others’ behavior. Some research groups have taken the approach that constructs an aggregated spoken dialogue front-end for a community of under-specified agents. The Speech Graffiti Personal Universal Controller (Harris, 2004), which was designed explicitly with multi-agent control in mind, is such an aggregating system. This system severely limits the expressive power of natural language, however, and any aggregating spoken dialogue front-end will potentially sacrifice the integration of domain knowledge into the dialogue.

In order to directly address heterogeneous multi-agent communication problems, we have established an understanding of the issues and a platform for experimentation in that domain. The platform, with a few simple strategies, has yielded some interesting and relatively positive results in a small pilot study.

References

Black, A., Clark, R., Richmond, K., & King, S. (2004). *The festival speech synthesis system*. [http://](http://www.cstr.ed.ac.uk/projects/festival/)

- www.cstr.ed.ac.uk/projects/festival/.
- Bohus, D., & Rudnicky, A. (2002, November). *Integrating multiple knowledge sources for utterance-level confidence annotation in the CMU Communicator spoken dialog system* (Tech. Rep. No. CMU-CS-02-190). Pittsburgh, Pennsylvania: School of Computer Science, Carnegie Mellon University.
- Bohus, D., & Rudnicky, A. I. (2003). Ravenclaw: Dialog management using hierarchical task decomposition and an expectation agenda. In *Eurospeech*. Geneva, Switzerland.
- Brumitt, B., & Cadiz, J. J. (2001). ‘let there be light’ examining interfaces for homes of the future. In *Proceeding of interact 2001* (pp. 375–382). (<http://research.microsoft.com/research/coet/Homes/INTERACT2001/paper.pdf>)
- Byrne, B., Rngelsma, J., Ferrans, J., Jablokov, I., Jackson, E., Kruger, S., JimLarson, Marchand, R., Oshry, M., Rehor, K., & Scholtz, B. (2004). *Voicexml forum - faq*. (<http://www.voicexml.org/faqs.html>)
- Harris, T. (2004). *The speech graffiti personal universal controller*. Unpublished master’s thesis, Carnegie Mellon University, Pittsburgh.
- Hoge, H., Burchard, B., Comeyne, R., Diehl, F., Fischer, V., Hakkinen, J., & Marasek, K. (1999). *Market analysis*. (http://www.speecon.com/public_docs/D11v2.2-market_final.doc)
- Huang, D., Alleva, F., Hon, H. W., Hwang, M. Y., Lee, K. F., & Rosenfeld, R. (1993). The Sphinx-II speech recognition system: An overview. *Computer, Speech, and Language*, 7(2), 137–148.
- Montemerlo, M., Roy, N., & Thrun, S. (2002). *Carnegie mellon robot navigation toolkit*. <http://www-2.cs.cmu.edu/~carmen/>.
- Oh, A. H., & Rudnicky, A. (2000, May). Stochastic language generation for spoken dialogue systems. In *Anlp/naacl workshop on conversational systems* (pp. 27–32).
- Seneff, S., Hurley, E., Lau, R., Pao, C., Schmid, P., & Zue, V. (1998). Galaxy-II: A reference architecture for conversational system development. In *Proceedings of the international conference on spoken language processing*. (<http://www.sls.csail.mit.edu/sls/publications/1998/icslp98-ga%laxy.pdf>)
- Ward, W. (1994, September). Extracting information from spontaneous speech. In *Proceedings of the international conference on spoken language processing*.