Taking Turns Talking About Text in a Reading Tutor that Listens

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Abstract

In this paper we report on ongoing work on turn-taking in Project LISTEN's Reading Tutor (Mostow & Aist CALICO 1999). Project LISTEN's Reading Tutor listens to children read aloud and helps them learn to read. The Reading Tutor's repertoire of turn-taking behaviors includes not only alternating turns, but also backchanneling, interrupting, and prompting.

Turn-taking in conversation

Human-computer spoken dialog relies on taking turns: knowing not just what to say, but when to say it. Humans take turns in many different ways: alternating with another speaker, backchanneling, taking another turn if there is a pause in the conversation, interrupting, and pausing. Spoken dialog systems began by assuming a very strict model for turntaking: alternating turns. Telephonebased dialog systems were among the first to allow the user to interrupt, or "barge-in". As human-computer spoken conversation extends into new domains and new social roles, computers should be able to take turns using the full, rich set of turn-taking behaviors that humans engage in.

A Reading Tutor that Listens

Project LISTEN's Reading Tutor represents a special case of spoken dialog: A single task - teaching reading but on text from any domain. The Reading Tutor allows users to type in and narrate their own stories, which can then be used to teach children to read (Mostow & Aist USPTO 1999). The Reading Tutor uses such recorded human speech for its expressiveness and warmth. Turn-taking behaviors are generated using on a combination of user-added resources such as story narrations, and lexical resources such as word recordings, recordings of syllables. recordings of letters, and other recorded phrases. prompts and Turn-taking behaviors selected using are а combination of reading-specific heuristics - such as "sound out a word only if it has four or fewer phonemes" - and taskindependent heuristics such as "take a turn if the user falls silent for an extended period of time."

Turn-taking in the Reading Tutor

Following Sacks et al. (1974), Cassell et al. (1994), Thorisson (1996), and Ward

(1996), we describe turn-taking actions as heuristic rules with left-hand preconditions based on dialog state and right-hand actions consisting of multimodal output.

We divide the turn-taking state using two binary variables:

USER-TURN Is the user taking a turn? SYSTEM-TURN Is the system taking a turn?

We use two variables instead of a single "whose turn is it?" variable for two reasons.

1. Separate variables allows us to represent our uncertainty about whether the user is taking a turn separately from whether the system is taking a turn.

2. In human-human conversation, turns *typically* overlap. The assumption that one participant is speaking at a time "reflects ideology more than practice" (Tannen 1994, page 62). Thus machine interlocutors should model overlap, including cases where both participants take extended turns cooperatively or competitively.

This two-variable framework lets us represent a wide variety of turn-taking Of particular note behaviors. are prompting. backchanneling, and interrupting. The Reading Tutor prompts in response to extended student inactivity. The Reading Tutor backchannels in response to a slight pause. Finally, the Reading Tutor can interrupt to correct a student mistake. We operationalize the concept "interrupt to correct a mistake" as follows:

If USER-TURN=true and SYSTEM-TURN=false and the user has misread an important word and continued to the next word, then cough and underline a word. We introduced these turn-taking behaviors and the architecture underlying them into the Reading Tutor beginning in 1996-97. We have used a variety of methodologies to evaluate these turntaking actions, and improve them.

To evaluate and improve the robustness of the turn-taking architecture, we tested Reading Tutor revised the in progressively more difficult environments. We started with adult users in the laboratory and ended up with children using the Reading Tutor on their own in a classroom without the researchers present, supervised by a teacher who was teaching the rest of the class.

To evaluate and improve the quality of the interactions, we invited critique of videotaped interaction by an expert in reading instruction. The expert was asked to view videotape of the Reading Tutor interacting with a student and judge the Reading Tutor's responses.

To explore the local effects of on the dialog. we introduced systematic variation in the Reading Tutor's responses as an "invisible experiment" (Mostow & Aist 2000), similar to varying dialog choices when using reinforcement learning to learn dialog policies (Singh et al. 1999). For example, to look for the local effect of interrupting, we modified the Reading Tutor to sometimes interrupt when its heuristic for interrupting applied, and sometimes remain silent. Analyzing what happens next in the dialog - for example, whether the student speaks again, clicks Go to move on, or does something else - allows us to quantify the local effects of conversational actions with respect to what would have happened if another action had been taken instead.

Conclusion

We have described some of the progress that spoken dialog systems have made towards taking turns like humans, enumerated some contributions made towards that goal within Project LISTEN's Reading Tutor, and described some of the evaluation methodologies appropriate to conversational behaviors.

What remains to be done? To expand turn-taking capabilities into novel areas, the field of dialog research needs to include user-system dialogs with varied social roles. Many systems focus on some form of information seeking (train timetables, movie times) or assistance in constructing a plan (travel agent, military logistics). We have looked at a less explored area in human-computer spoken dialog: tutoring. Other social roles - an adversary or a partner in a game, or a computerized salesperson that tries to maximize company profit rather than end-user satisfaction - may offer new insights. Much interesting work remains to build dialog systems that take turns as well as humans.

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