Spatial Reference Resolution for an Embodied Dialogue Agent

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In dialogue systems, reference resolution is an essential part of the semantic interpretation of an utterance. The resolution task is a variable-assignment problem – given the user’s utterance, we want to connect unknown discourse entities to our system’s internal knowledge base. To illustrate this problem, consider the following example sentences:

(1)
A: Mary likes John.
B: I heard that he gave her a rose.

In order to understand the meaning of these sentences, we must solve two resolution problems. The first sentence introduces two entities to our domain, Mary and John and can be semantically represented as $\text{likes(Mary, John)}$. In the second sentence, we encounter the two pronouns he and her. The task is to then resolve the variables he to John and her to Mary, providing the appropriate semantics of the sentence. [gave(John, rose, Mary)]

Dialogue systems like COMMUNICATOR, TRIPS and TRAINS all include some manner of reference resolution (Byron & Allen 1998). These traditional systems only consider elements of the discourse history and planning state as referents for an ambiguous expression. Extra-linguistic information, such as gesturing or spatial orientation, is ignored. Our work looks to extend this previous work to include spatial context in reference resolution. This is a natural extension of previous work, and holds potential in embedded applications (cars, houses) and robotics.

Spatial Reference Resolution

Context of an ambiguous expression is not limited to just language. We often refer to an item by its spatial location. A host might tell his guest that “the spoons are in the right drawer.” In resolving this reference, the guest must figure out how the description “right” relates to the spatial context and aids in picking out the intended referent.

Using spatial context for reference resolution requires both knowledge of how spatial relationships are organized and also how they are encoded linguistically. Linguistic investigation into spatial reference (Levinson 2003) has yielded three related strategies of encoding spatial reference. To illustrate, consider the following sentences, each referring to a particular door in the world:

(2)
A: Which door should I use?
B$_1$: Use the north door.
B$_2$: Use the back door.
B$_3$: Use the left door.

Traditional resolution algorithms will not be able to resolve these references because the information required to interpret the reference does not come solely from the linguistic context.

In each of the response sentences in Example 2, the speaker has utilized a different, distinct strategy for describing the door. Sentence B$_1$ utilizes the absolute reference strategy. In this scheme, reference is based on an imposed external arrangement (North/South, Above/Below). In B$_2$, intrinsic reference is used, whereby the properties of a grounding item (front/back of the building) are used for reference. Finally, relative reference is utilized, whereby a particular point-of-view anchors the spatial context.

Each strategy places different requirements on reference resolution, but the basic strategy for interpretation remains fairly consistent. First, one must use the spatial description to generate a general search space strategy (right/left, north/south). From that, find all appropriate objects that can be used as a center point for the search (a “grounding mechanism”) in order to define the search space. Finally, search for an item matching other predicates in the description (e.g. “door”, “drawer”) beginning at the grounding item and continuing in the direction indicated by the the spatial phrase.

Computational research in this direction has been limited. There have been investigations into combining spatial information, language and robotics (Moratz & Tenbrink 2006). However, these systems only understand simple commands like “go to the left box.” Spatial reference in this work consists only of robot-centered, relative references and ignore other reference strategies. Additionally, it is not a dialogue agent, as it lacks discourse history and feedback beyond the option to move (success) or not (failure).
Proposed Plan of Research

We are developing a software agent capable of interpreting reference based on a description of the object’s spatial properties. As part of this, we have collected a dialogue corpus of spatial references, allowing us to get insight into the language theory and ensure that our work is based on attested examples.

Corpus Development

Seed Corpus Last summer, we collected a corpus to serve as a seed for this thesis work (Weale & Byron 2007). The corpus consists of two participants arranging a group of boxes into a target configuration. The task elicits the underlying frame of reference strategies by forcing participants to use the spatial properties of the target boxes for reference, as the boxes were identical in all other respects.

This corpus work utilized Quake II, which has been used before in dialogue research (Byron & Fosler-Lussier 2006). This engine allows us to fully capture the spatial context of our dialogue. We can make replayable recordings for sharing, easily create experimental maps and modify the source code (Weale, Mellen, & Byron 2005), facilitating utterance and action synchronization.

The collected descriptions follow the three reference strategies, including phrases like “the north box” or “the front left box”. Thus, this corpus provides a good variety of spatial descriptions under different contexts.

Supplemental Corpus We will extend the corpus to ensure broader coverage and better system generalization. Most importantly, the task environment will now incorporate multiple grounding objects. Additionally, the manipulation of the grounding object will provide support for spatial grouping references or intrinsic reference strategies.

The extension will be implemented this summer. The bulk of the environment creation will be completed by the end of June, and the dialogue collection will be completed by the end of August.

Agent Development

This thesis will focus on learning and implementing an algorithm for resolving references based on a spatial description. For example, given a room containing two tables with three boxes placed on each, the system must be able to resolve references like “the north box” from context.

The spatial reference will require three steps:

Semantic Analysis: From the input utterance, determine the most likely reference strategy and what relationship the description is attempting to describe. For example, “the north box” would become north(?G, [?T: type=box]), a search in the north direction from the center of an undetermined ground for an entity with type box.

Grounding Object Selection: Given the spatial information in conjunction with the initial semantic interpretation, generate referent searches based on available grounding objects from both discourse history and spatial context. If the room contains two tables, this may be north([G=TABLE1(.9)=TABLE2(.1)], [?T:type=box]).

Reference Object Selection: Given the grounding object and relationship, resolve the referent to an object within the search space. In our example, TABLE1 has BOX2 in its north area. Therefore, the most likely relationship found was north(TABLE1,BOX2) and the entire module will return BOX2 as its output.

The corpus data will be used for feature engineering in order to define the context, based on visual saliency of objects, spatial proximity, dialogue recency and description information. We will then train a reference resolution strategy to predict the referent from the context (dialogue and spatial) by using machine learning techniques on these features. This will enable more expandability and flexibility than by simply hand-coding rules.

The output of our module will be the resolution of the description to its denotation in the knowledge-base. The mapping will be saved as part of the discourse context for use in future contexts. In the future, probabilistic weights may be added to the module outputs.

We have used the Quake II engine as our virtual environment. In it, we can simulate embodied technology without the overhead of a robotics lab, allowing for faster and more refined research into spatial language strategies. However, we can broaden the general applicability of our research by simulating noise problems like visual uncertainty, incomplete knowledge and path planning.

A formal dissertation proposal will be completed by the end of May as part of the author’s Ph.D. program. This proposal will detail the timetable of agent development and the interpretation pipeline, which will be based on corpus analysis and current literature in the field.

During the remainder of the summer and the following school year, the agent system will be under active development, based on the seed corpus, feedback from the proposal and supplemental corpus analysis. The anticipated completion date of this project is spring 2008.

References


