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Towards a computational model of frame of reference alignment in Swedish dialogue

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Abstract
In this paper we examine how people negotiate, interpret and repair the frame of reference (FoR) in online text based dialogues discussing spatial scenes in Swedish. We describe work-in-progress in which participants are given different perspectives of the same scene and asked to locate several objects that are only shown on one of their pictures. This task requires participants to coordinate on FoR in order to identify the missing objects. This study has implications for situated dialogue systems.

1. Introduction
Directional spatial descriptions such as “to the left of green cup” or “in front of the blue one” require the specification of a frame of reference (FoR) in which the spatial regions “left” and “front” are projected, for example “from where I stand” or “from Katie’s point of view”.

A good grasp of spatial language is crucial for interactive embodied situated agents or robots which will engage in conversations involving such descriptions. These agents have to build representations of their perceptual environment and connect their interpretations to shared meanings in the common ground (Clark, 1996) through interaction with their human dialogue partners.

There are two main challenges surrounding the computational modelling of FoR. Firstly, there are several ways in which the viewpoint may be assigned (Levinson, 2003) – intrinsic (assigned by the landmark object of the description, e.g. “the blue cup to the left of the red cup” relative to the orientation of the red cup; extrinsic (an external viewpoint such as superimposed grid structure or cardinal directions N, S, E and W), e.g. “the blue cup in C2” or “the blue cup to the N of the red cup”; or relative (with reference to a conversational participant or object in the scene), e.g. “the blue cup to the left of the red cup (from where I stand)” relative to Participant 1.

The second challenge is that the viewpoint may not be overtly specified and must be recovered from the linguistic or perceptual context. Such underspecification may lead to situations where conversational partners fail to accommodate the same FoR leading to miscommunication.

There are a number of factors that affect the choice of FoR, including: task (Tversky, 1991), personal style (Levelt, 1982), arrangement of the scene and the position of the agent (Taylor and Tversky, 1996; Kelleher and Costello, 2009; Li et al., 2011), the presence of a social partner (Durran et al., 2011), the communicative role and knowledge of information (Schober, 1995), but very little work has investigated choice of FoR from a dialogic paradigm, in which participants can work through potential misunderstandings together using processes of repair.

We are interested in how participants align their spatial representations in dyadic text dialogues when they perceive a scene from different perspectives. How do they identify if a misalignment has occurred, and what strategies do they use to get back on track?

2. Method

Task Using 3D modelling software we designed a virtual scene depicting a table with several mugs of different colours and shapes placed on it. As shown in Figure 1, the scene includes three people on different sides of the table. The people standing at the opposite side of the table were the avatars of the participants (the man = P1 and the woman = P2), and a third person at the side of the table was described to the participants as an observer “Katie”. Our earlier study (Dobnik et al., 2014) shows that participants prefer to assign relative FoR to a neutral landmark that is not conversational participants. Therefore, in this experiment we introduce Katie to fulfil this role.

Each participant was shown the scene from their avatar’s point of view (see Figures 2 and 3), and informed that some of the objects on the table were missing from their picture, but visible to their partner. Their joint task was to discover the missing objects. The objects that were hidden from each participants are marked with their ID in Figure 1.

The task ensures all possible FoR assignments: (i) the mugs on the table have handles which means that they have orientation and can assign intrinsic FoR; (ii) the surface of the table grounds the extrinsic FoR; and (iii) the conversational participants and the observer Katie can assign relative FoR. There are mugs of different colours which gives rise to linguistic data on how reference to individual objects is expressed in terms of the amount of linguistic information
in a description, over several turns of dialogue. We are investigating this in another related line of work.

**Procedure** Each participant was seated at their own computer and separated so that they could not see each other or each other’s screens. Communication was through an online text based chat tool (Dialogue Experimental Toolkit, DiET, (Healey et al., 2003)), which records each key press and associated timing data. Participants were instructed that they should chat to each other until they found the missing objects or for at least 30 minutes. Following completion of the task participants were debriefed about the nature of the experiment.

### 3. Summary of results

In the pilot study (Dobnik et al., 2015), we recorded and annotated in detail two dialogues in English. The native language of the first pair was Swedish while the second pair were native British English speakers. The first dyad took approximately 30 minutes to find the objects and produced 157 turns in total. The second dyad discussed the task for a little over an hour, during which they produced 441 turns.

The pilot study suggests that there is no general preference of FoR in dialogue but the choice is related to the communicative acts of particular dialogue or conversational games (a sequence of dialogue moves centred towards a particular goal (Kowtko et al., 1992; Pulman, 1997)) at specific points in the dialogue. There is also evidence that participants align their FoR locally over a sequence of turns, but not globally; at points of misunderstanding it may be prudent to shift FoR in order to get the conversation back on track. We isolate several conversational games where the dynamics of the FoR assignment appears to be linked to other properties of interaction between the agents, for example whether they are focusing on a particular part of the scene or whether they are identifying individual objects scattered over the entire scene. It follows that alignment is consistently used as a strategy but there are other factors that trigger changes in FoR.

In this work-in-progress we look at Swedish dialogues between 4 pairs of Swedish native speakers that amount to 794 turns overall. We continue to collect and annotate data both in English and Swedish which we plan to release to the research community. Here are two interesting excerpts:

**Excerpt 1**

17 P5.P2: Ska vi börja från din ända av bordet?
18 P5.P1: ja
19 P5.P1: jag har en rak linje med fyra muggar med öron
20 P5.P1: från mitt vänster till höger: röd, blå, vit och röd
22 P5.P2: Från din vänster till höger.
23 P5.P1: okej
24 P5.P1: då fattas det nog en röd mugg längst till mitt höger
25 P5.P1: den är lite längre ut på kanten än de andra
26 P5.P2: Nej. Ser ingen. :-)  
27 P5.P1: okej, då har vi hittat en där :)  

17 P6.P2: vem perspektiv ska vi utgå ifrån?
18 P6.P1: ta Katies
19 P6.P2: oj okej
20 P6.P1: eller det blir konstigt kanske
21 P6.P2: på Katies högra nedre sida står en gul mugg
22 P6.P2: nej vi testar!
23 P6.P1: japp
24 P6.P2: lite till vänster om den står en vit
25 P6.P1: ja som har en annan form och till vänster i hörnet en röd
26 P6.P2: jag har ingen röd! och den vita har samma form
27 P6.P2: jag ritar in en röd först
28 P6.P1: humm okej
29 P6.P2: yani har du ingen mugg eller...
30 P6.P1: den röda har öra och sitter i Ks högra hörne
31 P6.P2: vänta, nyss sa du vänster?
32 P6.P1: framför K finns det en blå mugg
33 P6.P1: oj jag menade vänster
34 P6.P1: sorry'
We investigate if the findings from the English study hold cross-linguistically, when resources for resolving misunderstandings may not be the same across languages. We also examine whether a selection/change of the FoR could be predicted from the (textual) dialogue data. We hypothesise that dialogue turns contain sufficient information about the dialogue games that conversational participants are engaged in and to which the FoR assignment appears to be linked.

Through quantitative data analysis we attempt to identify features that are predictive of FoR changes and which would be useful for annotating and extending our corpus described above. The overall goal is to provide a training dataset for machine learning that would allow us to build a model of FoR assignment. Finally, we also investigate the suitability of different machine learning models for the task.

References


