Modeling and evaluating dialog success in the LAST MINUTE corpus

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Abstract
The LAST MINUTE corpus comprises records and transcripts of naturalistic problem solving dialogs between N = 130 subjects and a companion system simulated in a Wizard of Oz experiment. Our goal is to detect dialog situations where subjects might break up the dialog with the system which might happen when the subject is unsuccessful. We present a dialog act based representation of the dialog courses in the problem solving phase of the experiment and propose and evaluate measures for dialog success or failure derived from this representation. This dialog act representation refines our previous coarse measure as it enables the correct classification of many dialog sequences that were ambiguous before. The dialog act representation is useful for the identification of different subject groups and the exploration of interesting dialog courses in the corpus. We find young females to be most successful in the challenging last part of the problem solving phase and young subjects to have the initiative in the dialog more often than the elderly.

Keywords: Human-Companion-Interaction, Dialog Success, Wizard-Of-Oz

1. Introduction
The LAST MINUTE corpus comprises records and transcripts of naturalistic problem solving dialogs between N = 130 subjects and a companion system simulated in a Wizard of Oz (WoZ) experiment (Rösner et al., 2012b). The WoZ scenario is designed in such a way that many aspects of user companion interaction (UCI) that are relevant in mundane situations of planning, re-planning and strategy change (e.g. conflicting goals, time pressure, ...) will be experienced by the subjects (Rösner et al., 2012a).

With 56 hours of recorded and transcribed1 interactions the LAST MINUTE corpus is an invaluable resource for collaborating groups that employ e.g. the recorded audio or video streams to train classifiers for their analyses. This includes emotion detection from facial expressions or from prosodic aspects of speech signals (Frommer et al., 2012a). Other WoZ experiments in the companion paradigm (Wilks, 2010) include (Legát et al., 2008) and (Webb et al., 2010).

2. LAST MINUTE dialogs
The overall structure of an experiment is divided into a personalisation module, followed by the LAST MINUTE module. These modules serve quite different purposes and are further substructured in a different manner (for more details cf. (Rösner et al., 2012b)).

In the bulk of LAST MINUTE – the problem solving phase – the subject is expected to pack a suitcase for a two week holiday trip by choosing items from an online catalogue with twelve different categories that are presented in a fixed order.

Barriers The normal course of a sequence of repetitive subdialogs is modified for all subjects at specific time points. These modifications or barriers are:

• after the sixth category, the current contents of the suitcase are listed verbally (listing barrier),
• during the eighth category, the system for the first time refuses to pack selected items because the airline’s weight limit for the suitcase is reached (weight limit barrier),
• at the end of the tenth category, the system informs the user that now more detailed information about the target location Waiuku is available (Waiuku barrier).

Additional difficulties for the subjects may occur depending on the course of the dialog. These are typically caused by user errors or limitations of the system or a combination of both.

3. Previous work
A first global measure In order to compare different dialogs we started with the following coarse global measure for the course of interaction of the LAST MINUTE problem solving dialogs: We distinguish turns that are – based on the logged system response – judged as successful from those that are judged as unsuccessful or faulty. We then use the ratio of unsuccessful turns in relation to all turns as a measure of the relative faultiness of the dialog as a whole. (Rösner et al., 2012a)

In detail:

• successful turns are those with an explicit wizard confirmation of success,
• the turns that are counted as failed or unsuccessful include those with the following system responses: un-processable input, item not in suitcase, weight limit reached again, system enforced category change.
• system responses that are neutral or ambiguous are ignored.

1 The audio recordings were transcribed using with FOLKER. (Schmidt and Schütte, 2010) The transcription followed the the GAT-2 minimal standard. (Selting et al., 2009)
For a cohort of N = 130 subjects the values for this global measure range between 9% and 73% (unsuccessful turns) with a mean of approximately 26% and a variance of 10%. The only information source for this measure were the log files of the system utterances, because the full transcription of all experiments was not available at that time. The log files allowed to assume which subject request precedes a system utterance but many situations remained ambiguous and all assumptions were unproven. Now, with the full transcripts and annotation of the packing phase of all subjects, better insight and finer analyses are possible.

Refined measure In this work we refine the segmentation from turns to dialog acts which allows finer analysis of the subjects’ intentions in the dialog. We also refine the operationalization of successful dialog in two ways which help to answer the two questions:

- Did the system react as the subject expected it? (cf. section 5.1.)
- Did the user expect what the system did? (cf. section 5.2.)

The questions are quite similar but lead to different perspectives on the corpus and to supplementary results.

4. Annotation

We analyze the task-related success, so the annotation is also focused on the task. In the previous work and in following we only consider the problem solving phase of LAST MINUTE.

4.1. Dialog acts of the subject

The utterances of the subjects show a large variance with respect to many linguistic features: lexical choice, syntactic patterns, well- vs. ill-formedness, etc. (Rösner et al., 2012c) Fully automatically correct classification of all dialog acts is not possible up to now, so all dialog acts from the packing phase of the subjects were annotated by a human rater.

During the annotation an initially tiny tagset was adjusted to distinguish many domain specific dialog acts but to contain mostly frequent tags. The annotation process was computer assisted: The annotations of ca. 1/2 of the segments which are easy to detect correctly (e.g. “two jeans”, “next category”) were retrieved automatically (rule-based) – the rater verified the automatic annotations and annotated the remaining segments. The annotator normally annotated the written transcript, but whenever the transcript was ambiguous or the annotator needed more information he was free to listen to the audio recordings.

During transcription very long subject contributions were split into smaller segments, so sometimes a single dialog act spreads over several segments. Such segments were marked as associated and were glued automatically for the analyses presented here. Some transcript segments contain more than one dialog act, so they could be annotated with several annotations (e.g. a segment with the utterance “ein pullover (.) weiter” - engl. “a pullover (.) go on” contains a packing request and a request for category change and was annotated with both – preserving the temporal order). Thus the segmentation of the transcribers does not interfere with the dialog act annotation.

4.2. Dialog acts of the wizard

Most of the Wizard contributions were preformulated and could be annotated automatically by using regular expressions.

4.3. Dialog act representation (DAR)

The first measure – based on unambiguously classifiable wizard utterances only – obviously is quite coarse. We now work with measures that are based on sequences of successive subject and wizard utterances.

The dialog act representation is hierarchical (x stands for ‘unknown’): The first capital letter indicates the speaker (S for subject, W for wizard). The second capital letter stands for the dialog act:

- Sx: R = request, O = offtalk, Np = non-phonological and pauses, A = answer
- Wx: A = accept, Rj = reject, I = information, Q = question/request

A third capital letter may refine a dialog act subtype:

- SRx: P = packing, U = unpacking, E = exchange, F = finalization, L = listing, C = category change
- SOx: T = offtalk, Q = question
- WAx: P = packing, U = unpacking, F = finalization
- WRjx: P = packing, U = unpacking, Np = non-processable
- WIX: C = category change, W = weather, F = finalization, L = listing
- WQx: F = finalization, I = intervention, C = comment, E = elaboration

The lowercase letters are inserted for readability. The hierarchical organization of the DARs is shown in figure 1.

![Figure 1: Hierarchy of the DAR for wizard and subject actions](image)

The distribution of the different DARs can be seen in figure 2. Several subsequent non-phonological events might be annotated with a single SNp – if they were transcribed within one segment. Sometimes – e.g. when there was a pause between two non-phonological utterances – they might both be transcribed separately. In this case the pause transcription was then filled in by the transcription software and all three segments were later annotated as SNp, leading to the
sequence: SNp SNp SNp. So the length of a SNp and the number of SNp annotations in a row may vary for similar cases.

Subject requests (SRx) may either be accepted (and performed, W Ax) or rejected (WRjx = subject exceeded the weight limit, WRjNp = utterance could not be processed) by the system/wizard. Some subjects requested ‘exchange X for Y’ (SRE) – the system did not provide this functionality, but could either unpack (WAU) or pack (WAP), which was half of the task and considered as success here. Accepted requests are further described in section 5.

We are interested in finding common and uncommon interaction patterns by means of exploring dialog act sequences. The DAR allows the quantitative analysis of many dialog act combinations – e.g. the system’s reaction to offtalk and questions, the subjects’ reaction to system questions (SQx) and information messages (SIx) – and even longer subsequences, but this is not in the focus of this paper and will be addressed in other publications.

4.4. LAST MINUTE workbench
The result of dialog act tagging of the corpus is available as an HTML file with the packing phase sequences for all subjects. The DAR is part of the workbench – a collection of tools for the exploration of the LAST MINUTE corpus. There is a one-lined version to search for sequences and a two-lined version for better readability where the first letter is omitted and each speaker is represented in a separate line. A part of the two-lined version can be seen in figure 3. Wizard dialog acts that are a clear rejection of a subject’s action are colored red, confirming utterances are green, neutral and unclear ones are black. This ensures a quick overview of a subject’s task success. The three barriers of the packing phase are marked by | symbols (cf. 2.). All dialog act tags feature a mouseover text with the time-point and text of the underlying transcript segment and the long form annotation. With this information every dialog act can be found in the original transcript. An example can be seen in figure 4.

Some interaction patterns are ‘interrupted’ (depending on what is looked for) by nonphonological segments (SNp). For all exploration tasks that are disturbed by SNp segments both representations, the one-lined and the two-lined, were also computed without the nonphonological segments. An example: There may be sequences like SRC SNp WIC in the original version. If a successful category change is searched (SRC WIC) it can only be found in the version without SNp.

In the following we consider the 5294 SNp as non-functional and work with a version without SNp. This sequence visualisation allows a quick overview and the search for abstract interaction patterns. The HTML document also comprises a script for the calculation of dialog act bigram statistics. The bigram statistic can be calculated for specific user groups and for specific phases of the experiment with the script.

4.5. Bigram statistic
In figure 2 we can see, that there are many packing requests (SRP) but less packing confirmations (W AP). What happens after subject requests can be seen in a bigram statistic. In table 1 we show the bigram statistic for the packing phase. This is useful to find common bigrams or to analyze what other tag followed a certain tag (lines) or what happened before a certain tag (columns). The bottom left and the top right of the four large boxes show the numbers for bigrams of one speaker. The top right large box shows bigram counts for subject dialog acts followed by a system response.

4.6. Ambiguous interpretation
For the proper differentiation of some dialog acts the context is needed, e.g. Wizard utterances announcing category
changes. Here the same wording is used for any category change, be it system initiated or requested by the subject.

4.7. An example

The following DAR example is taken from a dialog segment where a subject (20110401adh) tries to pack a (winter) coat but the packing attempt is rejected several times (SRP WRjP pairs) and therefore the subject has to unpack several other items (SWAU) in order to create sufficient space. SNPs stand for nonphonological utterances of the subject and can be interpreted as expressions of the experienced efforts.

Below is an excerpt of the corresponding transcript. Following the GAT-2 minimal standard (Selting et al., 2009) short pauses are noted as (.), longer pauses with their duration in brackets, e.g. (1.77). English glosses are added for convenience. Please note the emotional expression of relief (‘gott sei dank’, engl. ‘thank god’) when the subject finally succeeds.

Below is an excerpt of the corresponding transcript. Following the GAT-2 minimal standard (Selting et al., 2009) short pauses are noted as (.) and (-), longer pauses with their duration in brackets, e.g. (1.77). English glosses are added for convenience. Please note the emotional expression of relief (‘gott sei dank’, engl. ‘thank god’) when the subject finally succeeds.

Table 1: Bigram statistic for the tags of the packing phase. The first tag of a bigram is on the left, the second on the top. As # stands for beginning or end of the packing phase.

| Tag | SA | SOQ | SOT | SRC | SRF | SRL | SRU | SWAU | WRJNp | WRJP | WRJU | Sum |
|-----|----|-----|-----|-----|-----|-----|-----|------|-------|------|------|-----|
| #   | 1  | 2   |     |     |     |     |     |      |       |      |      | 17  |
| W   | 1  |     |     |     |     |     |     | 56   | 64    | 22   | 10  | 132 |
| WAP | 1  |     |     |     |     |     |     | 64   | 22    | 12   | 2   | 67  |
| WIC | 1  |     |     |     |     |     |     | 22   | 12    | 2    | 1   | 6   |
| WIF | 1  |     |     |     |     |     |     | 12   | 2     | 1    | 1   | 1   |
| WIL | 1  |     |     |     |     |     |     | 1    | 1     | 1    | 1   | 1   |
| WIR | 1  |     |     |     |     |     |     | 1    | 1     | 1    | 1   | 1   |
| WJW | 168 | 2   | 14  |     |     |     |     | 2    | 1     | 1    | 1   | 2   |
| WQI | 168 | 2   | 14  |     |     |     |     | 2    | 1     | 1    | 1   | 2   |
| WQC | 44  | 1   | 1   |     |     |     |     | 1    | 1     | 1    | 1   | 1   |
| WQF | 168 | 2   | 14  |     |     |     |     | 2    | 1     | 1    | 1   | 2   |
| WRJNp|168 | 2   | 14  |     |     |     |     | 2    | 1     | 1    | 1   | 2   |
| WRJP|168 | 2   | 14  |     |     |     |     | 2    | 1     | 1    | 1   | 2   |
| WRJU|168 | 2   | 14  |     |     |     |     | 2    | 1     | 1    | 1   | 2   |
| Sum | 262 | 2   | 14  |     |     |     |     | 2    | 1     | 1    | 1   | 2   |

Table 1: Bigram statistic for the tags of the packing phase. The first tag of a bigram is on the left, the second on the top. As # stands for beginning or end of the packing phase.

4.8. Relation to other tagsets

Our analyses are focused on the task specific semantic contents of a user’s or wizard’s utterance and do not take other information like length and number of pauses, breathing, non-task specific offtalk, questions etc. into account. Therefore most of the features of larger, standardized tagsets DAMSL (Core and Allen, 1997) or DiAML (Bunt et al., 2012) are not needed and we chose to construct our own DAR tailored specifically to the problem domain.

This DAR can easily be mapped to DiAML or DAMSL annotations though. As an example DARs of the form SRX would be mapped to a Directive annotation in DAMSL while WAX/WRjx would be mapped to Agreement Accept/Reject, WRjNp to Signal-Non-Understanding, WX to
 Assert and so on.
We plan to investigate measures utilizing other tagsets than the DAR. These measures will incorporate information from e.g. the analysis of questions and answers, offtalk and the non-task-solving phase etc.

4.9. Errors and inconsistencies
In spite of intensive training and a detailed manual (Frommer et al., 2012b) the wizards did not always operate consistent and accurate. This is not surprising given the large number of subjects and the time span of nearly a year for the completion of all N = 130 experiments.
We found inconsistent wizard behavior by analyzing the subject initiated category changes. It turned out that some rejected wordings would have been accepted by different wizards or even by the same wizard in other situations.
We also found some wizard errors, meaning thereby situations, where a wizard did not operate according to the guidelines of the manual. One type of such wizard error is the rejection of a subject request with ‘your input could not be processed’ (WRjNp) when indeed the intention of the subject was clearly recognizable and the intended action was performable.
DAR analysis uncovers candidates for such wizard errors and inconsistencies: at least 55 candidates for an erroneously rejected packing request, 31 for category change requests and 23 for unpacking, resp. These numbers show that the wizards did not work perfect but made only few mistakes which do not influence the following calculations significantly.

5. DAR based measures
The dialog act based representation (DAR) characterizes dialog courses with sequences of dialog act labels. This allows to define a variety of measures for dialog success or failure.
A first approach is to calculate the relative frequency of successful subject commands (SRx). Here we count the following bigrams as successful requests (numbers in brackets: number of successful action / number of action. Calculated for all subjects, cf. table 1):
- packing: SRP WAP (3551/5057)
- category change: SRC WIC (1342/1971)
- unpacking: SRU WAU (615/868)
- exchange: SRE WAU/WAP (13/44)
- listing: SRL WIL (89/101)
- finalization: SRF WQF (22/56)
For single subjects, the number of successful actions ranges from 19 to 74, the relative frequency of successful actions ranges from 39.6% to 87.5% with a mean of 69.6% and a standard deviation of 9.5%.
Table 2 shows the correlation of absolute and relative frequency of successful actions with age and gender calculated using Kendall’s tau. Missing values where not significant on a $p < 0.05$ level.

| age/sex | male | female |
|----------|------|--------|
| young    | 5    | 13     |
| elderly  | 5    | 1      |

Table 3: Top performers after weight limit.

5.2. Initiative in LAST MINUTE dialogs
An important aspect of dialogs is the initiative: Which participant has initiative and is thus driving the dialog? Losing the initiative leads to losing time to solve the task. The subjects solve the task under time pressure so we assume losing initiative and time ‘feels’ less successful and driving the dialog ‘feels’ more successful.
While the system generally has the initiative during the personalization phase, the problem solving phase is primarily characterized by user initiative: the subject expresses a request (R) for a system action and, as a response, the system either confirms (A) or rejects (Rj) the request. An action may be rejected based on aspects of the user’s utterance (‘your request could not be processed’) or – although the utterance was ‘understood’ and accepted – the action could not be performed for task related reasons. In sum: the pairs of successive dialog acts are SRx WAx, SRx WRjx or SRx WRjNp, but there are exceptions to this general rule.
One example is category change. If the allocated processing time of approximately one minute for a category is over the system takes initiative and performs an announced change of category (WIC).
Such system initiated category changes come in different structures:
Either the latest user request remains unprocessed (or ignored) resulting in a pair SRx WIC, or the request is performed and the change of category is announced immediately following the acceptance info resulting in a dialog act triple SRx W Ax WIC.
There are also situations where a subject does not give further commands (e.g. subject makes a thinking pause or does not know how to proceed) and the system changes the category. The subjects did not initiate such category change (even when they might approve it).
Wizard induced category changes come in variety of patterns:
- SRP W AC
- SRU W AC
- SRP WAP WAC
- SRU WAU WAC
- ...
A user initiated category change follows the dialog act pattern SRC WIC.

![Figure 5: Number of wizard induced category changes.](image)

We measure the wizard induced category changes for each subject. Figure 5 shows the distribution of this measure. The bulk (88%) of the subjects have between three and zero wizard induced category changes. The mean for this measure for all N = 130 subjects is 1.97, the standard deviation 1.96. After any category change the initiative is always again on the user side.
Self initiated changes of the category for selection are an indicator for the degree of control of the dialog flow and for the taken initiative that a subject exhibits. An in-depth analysis of self vs. system initiated category changes reveals a highly significant correlation between age and number of successful self initiated category changes after the weight limit barrier: calculating Kendall’s tau for these two quantities reveals that, with a tau statistic of \(-0.23\) and a p-value smaller than \(10^{-4}\), younger subjects have on average a higher number of such self initiated category changes than the elderly.

6. Summary
Both measures – successful actions and initiative – indicate how successful subjects were in the whole packing dialog or in parts of the interaction. This findings can be used for the correlation with personality and linguistic parameters to find indicators for critical dialog courses.
We have presented a dialog act based representation (DAR) for the problem solving dialogs in the LAST MINUTE corpus.
This representation is on the one hand abstract enough to generalize over the large variety of linguistic expressions in the user contributions. It is on the other hand specific enough to capture essentials of these contributions, i.e. the core dialog act and its underlying intention.
We have presented measures for success and failure of dialogs based on this representation. These measures allow to cluster transcripts into meaningful groups that deserve further qualitative analysis. In addition the relevant segments in the dialogs are identified. In other words: quantitative investigations guide subsequent qualitative in-depth interpretation of transcripts. Findings from such scrutiny are expected to inform the design of elaborated interaction strategies for future companion systems.
In addition the identification of those segments of the LAST MINUTE dialogs where problems accumulate or even escalate is an invaluable support for those collaborating groups that exploit audio or video records to train classifiers for their analyzes, be it emotion detection from facial expressions or from prosodic aspects of speech signals (Frommer et al., 2012a).
The identification of successful and unsuccessful bigrams can be used to calculate a task success value as in the PARADISE framework (Walker et al., 1997). We focus on explorative analyses and sequential properties here but plan to compare our measures to the PARADISE framework.

6.1. Future work
Ongoing and future work with the LAST MINUTE corpus includes:
- automate dialog act classification of user utterances,
- in depth analysis and classification of recorded offtalk,
- automatic detection of negative dialog situations,
- fine grained analyzes of failed dialogs in order to develop design guidelines for companion systems,
- investigate role of breathing, sighing and other non-phonological user contributions,
- finer grained analysis of pauses and hesitations in dialog course,
- compare to measures from the PARADISE framework (Walker et al., 1997),
- annotation of the dialog acts with more raters.
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Availability

The LAST MINUTE corpus is available for research purposes upon written request from the authors. For the reviewers a sample from the corpus with anonymized data is available from the following URL http://iws.cs.uni-magdeburg.de/a3/conf/index.htm with loginname reviewer and password review.

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