Creating and Annotating a Corpus of Health Coaching Dialogue

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Abstract

In this paper, we present our data collection process, annotation schema and agreement results for extracting health goals from SMS conversations between a health coach and the patients. This is our first step towards building an autonomous virtual assistant health coach that learns from expert demonstration to interact with patients via SMS.

1 Introduction

Many health problems faced by individuals can be mitigated with changes in health behavior. However, successfully implementing healthy behaviors in ones daily life requires significant motivation that most people, individually, find difficult to initiate and maintain. Health coaching has been identified as a successful method for motivating and maintaining health behavior changes by having a peer or professional convey relevant medical information, help to set realizable, yet challenging goals tied to health behavior change, and provide encouragement in adhering to those goals (Kivelä et al., 2014; Wolever et al., 2010). But unfortunately, personal health coaching is time-intensive, uneconomical for the low-income patients, and has limited accessibility.

Therefore, we aim to create a dialogue-based virtual assistant health coach that will converse with the patients via text messages and will help them increase physical activity by setting Specific, Measurable, Attainable, Realistic and Time-bound (SMART) goals (Doran, 1981). Even though an influential strand of work on conversational agents has been conducted by Bickmore and his group, their systems rely on predefined set of utterances from the patients (Bickmore et al., 2015, 2011; Schulman and Bickmore, 2009). We plan to build an autonomous system that learns from observed communications between human health coach and

participants to increasingly automate the generation of dialogue (Rizzolatti and Craighero, 2004). The ability to recognize proposed health goals in dialogues is a key initial capability needed for this type of learning from demonstration.

In this paper we talk about our data, annotation schema, agreement results and future work on extracting goals from patient-coach dialogue.

2 Annotation Process

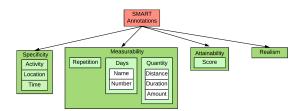


Figure 1: SMART goal annotation schema

In this section we discuss the data collection process and annotation schema for SMART goals.

2.1 Corpus

We recruited 28 patients between the age of 21 to 65 years who were interested in increasing their physical activity. A health coach, trained in SMART goal setting, conversed with these patients to set goals on weekly basis for a month via SMS. The patients were given Fitbit Alta to monitor their progress. The coach also monitored patients' progress using Fitbit application. The conversation involved setting a specific, measurable, and realistic goal and establishing any barriers to goal attainment. The coach also sent reminders based on patient's preference and provided motivational feedback on their progress. Only one patient did not complete the study.

We have a corpus of 2858 messages in total, where approximately 54% of messages were sent

Coach: Also, what would you like to set as your SMART goal this week? [Message 2, 3 omitted for brevity] **Patient:** My goal for this week is $2 \frac{\text{miles}(M_quantity_distance)}{\text{quantity_distance}}$ a $\frac{\text{day}(M_repetition)}{\text{day}}$ [Message 5, 6 omitted for brevity] Coach: Okay so you want to stick with your same goal as last week? FYI 2 miles(M_quantity_distance) is $4000 \text{ steps}(M_quantity_amount)$ **Patient:** Let's make it <u>3 miles</u>(*M_quantity_distance*) Coach: Base on your steps(S_activity) the first week I think 3 miles(M_quantity_distance) a day(M_repetition) is doable... R_intent Coach: How many days do you want to go for? M_days_number_intent **Patient:** Five(*M_days_number*)? Patient: I think I can do it all my days **Patient:** Let's go for a week(*M*_days) Coach: ok so all $7 \frac{\text{day}(M_days_number)}{?}$ Patient: Yes **Coach:** What do you think will make it easy to accomplish/achieve your goal? **Coach:** Me walking(S_activity) everyday(M_days) A_intent

Figure 2: SMART goal annotation for a conversation between the health coach and the patient.

Level	Specificity	Measurability	Attainability	Realism
Message	0.967	0.965	0.907	0.694
Word	0.878	0.895	0.515	0.549

Table 1: Results for inter-annotator agreement.

by the coach and 46% by the patients. This tells us that patients were equally involved in setting the goal. We plan to collect more data with different health coaches in the next phase of the project. Unfortunately, the dataset cannot be released because of human subject protection restrictions.

2.2 SMART Annotation Schema

15 patient-coach conversations were used to design the annotation schema (Bovend'Eerdt et al., 2009). After four iterations, we finalized the schema for annotating SMART goals as shown in Figure 1. We didn't annotate Timeliness as a new goal is set every week, and hence by default it is one week. Each of the annotations can either be categorized as a *slot value* or an *intention*. A slot value is a word or group of words that captures a particular piece of information such as 'walk' is a slot value for *specific activity*. Where as intention is an utterance that tries to gain information about a slot. Each of the SMART annotation category can have other optional tags such as *previous* for annotating anything related to previous

week, *accomplished* or *remaining* is used to annotate the progress of the patient, *update* to add another slot value to an existing one and *other* for anything which doesn't belong to previous or current week. Figure 2 shows the use of SMART annotation schema for a small sample of conversation from our dataset.

3 Annotation Results

Two annotators annotated four previously unseen patients' data for SMART goals. Inter-Annotator Agreement (IAA) was measured using kappa statistics on individual SMART categories. (Cohen, 1960)

We measured kappa on two levels: message and word. In message level, we consider an agreement if both the annotators labeled at least one word in the message with the given tag (not necessarily the same word). In word level, we consider an agreement if both the annotators labeled the same word with the given tag.

In total, 447 messages were annotated. There were 128 messages with *Specificity* tag, 120 with

	Specificity	Measurability	Attainability	Realism
Slot	653	608	60	0
Intent	31	87	260	65
Total	684	695	320	65
Total (%)	23.93	24.32	11.20	2.27

Table 2: Messages in each SMART category.

Measurability tag, 45 with Attainability tag and 13 with Realism tag. We observe approximately 90% reliability for Specificity and Measurability and only 50% reliability for Attainability and Realism on word level. This is because {S, M} tags are easy to annotate and have high number of occurrences in the data as compared to {A, R} which are hard to distinguish from each other and have very few occurrences. It should also be noted that for {S, M} word level annotation is more important where as for {A, R} message level annotation makes more sense.

Table 2 shows the statistics for SMART categories over the entire dataset of 2858 messages. One can observe that the percentage of $\{R\}$ in the entire dataset is fairly small when compared to the $\{S, M, A\}$ tags. It is not surprising as the coach only questions the realism of the goal if he thinks the goal is either too difficult or too easy based on the patient's past performances.

4 Discussion and Future Work

In this paper we discussed our data collection process and SMART goal annotation schema. We plan to use these annotations to train a classifier that can extract and summarize the SMART goal set by the patient. Using Stanford Temporal Tagger (SUTime) and rules such as Part-Of-Speech followed by a given keyword, we are able to obtain a minimum F1 score of 70% on all the SMART categories. We think that detecting different goal setting stages such as goal identification, goal negotiation, anticipate barriers, and follow-up can help us to improve the results further. Therefore, we are working on annotating the data for stages and will build a classifier for the same. Our aim is to combine these two classifiers to present the summary of the goal set by the patient. This will be the first step towards building a health coaching dialogue system for increasing physical activity.

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