## Adapting a Virtual Agent to Users' Vocabulary and Needs

Ana Cristina Mendes, Rui Prada, and Luísa Coheur

Instituto Superior Técnico, Technical University of Lisbon/INESC-ID ana.mendes@12f.inesc-id.pt

## 1 Introduction

Duarte Digital is an agent that engages in inquiry-oriented conversations about an art artifact. Since it was build for a Museum, interactions are supposed to be directed to different types of audience: an interaction with an art expert should be carried out in a different way than an interaction with a child; likewise, interactions with users interested in learning should be distinct from interactions with users having only entertainment goals. Being so, an agent needs to undergo two tasks: it must understand the user's knowledge about the topic, and his/her learning goals; it should adapt its vocabulary and dialogue strategy to cope with the user's characteristics and expectations.

This paper presents a simple and straighforward model of interaction that allows a virtual agent to understand its interlocutors based on their vocabulary and to adapt to their expertise and needs.

## 2 Interaction Model

The model of interaction is described in terms of a two dimensions graph (Fig. 1): the X axis corresponds to the level of the user's expertise on the topic; the Y axis relates with the degree of the interaction's orientation towards a sub-topic.

Duarte Digital holds a knowledge base (KB) with the possible users' questions and the answers the agent can provide to users. The agent's lexicon is weighted based on its difficulty level and clustered according to sub-topics. Every answer in Duarte's KB is tagged with: easy, neutral or complex, based on the difficulty level of its compounding words; and, as concise, neutral or detailed, depending on their informative content.

The user's expertise is calculated by:  $E_m = E_{(m-1)} + D_q$ , where:  $E_m$  is the perceived expertise in a moment m given a question q, and  $D_{(q)}$  is the difficulty level of the uttered question q. Answers are provided according to ( $\alpha$  is a threshold): if  $E_{(m)} > \alpha$  the user is an expert and give a complex answer; if  $E_{(m)} < -\alpha$  the user is unacquainted and give an easy answer; otherwise give a neutral answer (X axis). The orientation of an interaction on a moment m towards a topic t is defined as  $I_{t(m)}$  and calculated based on the history of questions the user has formulated and their proximity with the sub-topics. Answers are given according to ( $\beta$  is a threshold): if  $I_{t(m)} > \beta$  the interaction is directed, therefore

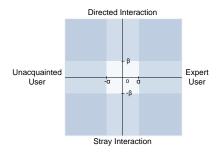


Fig. 1. The interaction model.

utter a detailed answer; if  $I_{t(m)} < -\beta$  the interaction is *stray*, therefore utter a concise answer and guide the dialogue; otherwise utter a neutral answer (Y axis).

## 3 Evaluation

Two experiments were carried out to evaluate the model: a) simulated interactions; and, b) interactions with human subjects in a controlled environment.

Virtual users were created by picking 10 random questions from KB. These questions were presented to Duarte, emulating real interations. To evaluate the X axis, virtual users were classified according to their expertise (as *expert*, *neutral* or *unacquainted*). When Duarte used the model of interaction, in 87% of the questions it gave an answer adequated to the virtual user's expertise (against 53%). To evaluate the Y axis, interactions were classified as being oriented or not. When the model was used, 9 questions were answered by Duarte that a real user would not need to ask if he would be interested in a particular sub-topic

Human subjects were also manually classified as expert, neutral or unacquainted, depending on their age, academic background and familiarity with the artifact. After the interaction with the agent, most of the users (11/12) reported having understood every word. We believe this situation occurred because no complex answers were given. One user revealed a non-understanding about the words employed: Duarte classified him as neutral (instead of unacquainted), suggesting the applicability of a model that differenciates users based on their expertise. Some users did not formulate questions as expected: the manually classified expert users did not fully explore the agent's capabilities by employing complex terms. No user was oriented towards a sub-topic; like expected, users did not know what to ask about the artifact. The agent's guidance was here of great use: in nearly 60% of the situations, the user's next question was based on a hint provided by the agent. They found it interesting and useful that the agent provide them with directions. Nearly 88% of the users were satisfied or very satisfied with the interaction, and they get more frustrated when Duarte provides a wrong answer than when it gives an answer which they might not understand.