Modeling of Indicators using UTL A Study Case with Hop3x System

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Abstract: In this paper, we present a novel approach for obtaining pedagogical indicators from tracks logged by Technology Enhanced Learning (TEL) systems. These indicators can help teachers to improve their pedagogical scenarios in order to making them relevant to students as well as to regulate learners' activities. The Usage Tracking Language (UTL) and its extension named the Data Combination Language for UTL (DCL4UTL) were proposed to formally specify the calculation method of these indicators. A typical use of UTL and DCL4UTL is also presented to illustrate the calculation of indicators, which are used to regulate a learning session, from tracks generated by Hop3x system.

1 INTRODUCTION

The work presented in this paper is related to the calculation modeling of pedagogical indicators from raw data collected during a learning session. Our research team has defined UTL (Choquet and Iksal, 2007) language that allows modeling indicators in form of design patterns. Since these patterns were designed for knowledge capitalization and sharing and not for being computed, we have proposed the UTL's extension named the Data Combination Language for UTL (DCL4UTL) (Pham-Thi et al., 2009). This language is used to formally describe indicators and to facilitate their automation. We present a study case that illustrates the modeling and the calculation of indicators from tracks generated by Hop3x system using UTL. These indicators are calculated and addressed to teachers in real time to help them in the regulation learners' activities during the ongoing session.

The remainder of this paper is structured as follows. The next section elaborates the UTL language and its extension DCL4UTL. The third section focused on the study case about UTL language's use. Finally, the last section concludes this paper with an outlook on future works.

2 UTL

The Usage Tracking Language (UTL) is designed to

make easier the capitalization of data analysis techniques and teachers' know-how in the analysis of a learning session. It is a generic language to describe tracks and their semantics, including the definition of the observation needs and the means required for data acquisition. Furthermore, it can be used to structure tracks from raw data, which are acquired from and provided by the learning system during the learning session, to indicators, which are significant for their users (teachers, tutors, learners, curriculum managers, etc.). These data are capitalized independently from any format of tracks generated by learning systems.

2.1 Conceptual Model

UTL allows describing two high-level types of data: the primary datum (PD) and the derived datum (DD). The primary data are not calculated or elaborated with the help of other data or knowledge. They consist of raw-data (RD), content-data (CD) and additional-data (AD). The derived data are calculated or inferred from primary data and/or other derived data. They consist of indicator and intermediate-data (ID). Each UTL type of data is defined according to three facets : the Defining (D) facet defines the observation needs; the Getting (G) facet describes the observation means to implement for producing data and the Using (U) facet defines the data's uses once they are calculated.

Models of different UTL data types have been published in (Choquet and Iksal, 2007).

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2.2 DCL4UTL

Although UTL allows describing the necessary data for formalizing the indicator (the Getting facet), its description of the data acquisition method is informal. Therefore it is difficult to generate the automatic analysis tools to compute indicators. For these reasons, we have proposed an extended part for UTL called Data Combination Language for UTL (DCL4UTL). The main aim of this add-on language is to create a method allowing to combine UTL data for producing a new one (indicator or intermediate datum). It is a declarative language and is able to be processed automatically by the DCL4UTL interpreter (Pham-Thi et al., 2009).

Another research goal of this language concerns the capitalization of these combination methods for the re-usability of derived data. Two characteristics of DCL4UTL facilitating the realization of this goal are namely the possibility to integrate external functions (Pham-Thi et al., 2010) and to create parameterized intermediate data (Pham-Thi, 2011).

2.3 UTL Tool

This section presents tools allowing the modeling and the calculation of indicators.

UTL Editor. To facilitate the definition and modeling of indicators, we developed an UTL editor based on Web. This editor helps analysts in their tasks. It is used to describe any UTL data according to three facets of the DGU model, especially the calculation method of indicators using DCL4UTL. It integrates a DCL4UTL parser verifying the DCL4UTL syntax in the formalization and expression of calculation methods of indicators. This tool also allows creating a new UTL data.

UTL Indicators Calculation Tool. As we have presented, UTL and its extension DCL4UTL are used to describe indicators and their calculation method in a form independent from any format of tracks generated by learning systems and from the architecture of databases. To validate our propositions, we have developed a tool executing the DCLUTL language, named UTL Indicators Calculation Tool (UICT). The Figure 1 presents the proposed architecture of this tool. It is composed of six components:

• The *Connection service* component communicates and connects to a TEL system to receive tracks and provide them to *Transformation data* component.

• The *Data transformation* component transforms all tracks collected by TEL systems into UTL primary data (RD, CD or AD) stored into a database.

• The DCL4UTL *interpreter* reads the DCL4UTL source code and will execute the code if there is no syntax error in the code. This interpreter is specially designed to allow executing the pre-defined functions and/or the external operators. It uses UTL data in the database and combines them to produce a new one formatted by the *Data format* component.

• The *Data format* component whose the role is to format the result according to the derived data definition.

• To calculate indicators in real-time, the tool needs a component to automatically trigger the calculation. The *Event management* component is added for this purpose. It manages events that trigger the calculation of indicators in real time. In our context, this calculation can be started after a given time interval, at the time of receiving a particular data or a user's request. It connects to the *Data transformation* component to verify if a special event is generated in the track.

• The *Data service* component provides necessary functionalities that allow sending calculation requests to the *Interpreter* and receiving the results as well as adding new derived data modeled in the database. It also includes services that allow interrogating the UTL database.



Figure 1: UTL indicators calculation tool.

3 STUDY CASE WITH Hop3x

This section presents a case study using UTL to model indicators from tracks generated by Hop3x system. For this case, the *Connection service* component was adapted for exchanging data with Hop3x. The UICT receives tracks from Hop3x and sends it back indicators. These indicators are calculated in real time and used to help teachers/tutors in their tutoring actions.

3.1 Description of Hop3x

The Hop3x (Lekira et al., 2011) system is a platform helping graduate students to learn and practice Java object-oriented programming techniques at the Computer Science faculty, University of Maine. Hop3x consists of three parts. Hop3x-Student is used by learners to create, modify, compile and execute programs. Hop3x-Server logs all activities of students during the session learning. These activities are classified into types of event: text insertion, text deletion, question selection, manual compilation, program execution, etc. Hop3x-Teacher is intervention tool for teachers. It allows them to manage and regulate in real-time learners in distance learning sessions.

3.2 Description of the Experiment

We have made several experiments with Hop3x to verify our prototype. In these experiments, we wanted to calculate a set of indicators concerning learners' activities in a practice learning session of Java programming course. Learners were beginners in Java object-oriented programming. One of these experiments was carried out between May and June 2010 with six groups of first-year students, five groups of 15 students and one group of 11 students. Each session held in three hours. Learners had to answer a set of 12 questions concerning to concepts of the object-oriented programming such as overriding, scope, data encapsulation using accessors and mutators, etc. A set of indicators related to these concepts were calculated and addressed to tutors in real time to help them in the regulation learners' activities during the ongoing session.

Two teachers worked remotely (in their office) as tutors. They observed, regulated the activities of learners and made interventions via Hop3x-Teacher.

3.3 UTL Data

Hop3x generates about 23 events. Every event corresponds to a UTL raw data (RD). These RD were modeled before the learning session (the *Using* facet's *Data* field of these data is empty). During session, the *Data transformation* component transformed all events collected by Hop3x into UTL RD (the *Using* facet's *Data* field is filled).

The content of Java files created by learners was extracted by a component of the Hop3x-Server and transformed into UTL primary data in form of content data (CD). All UTL data were modeled before the learning session according to three facets of DGU model and stored in the database. We modeled about 80 indicators. These indicators were defined by a teacher in Computer Science faculty, University of Maine. They were calculated from two types of UTL primary data: raw data and content data.

The main objective of these indicators was to verify if students respected instructions given in questions and know how to create a class in Java. Some of these indicators are: Detecting the existence of a class is called "Point", verifying if a student often repeats the same errors or he/she assimilated errors that he/she fixed before, etc.

In this experiment, the calculation of indicators in real-time was started at the time of receiving a particular event: question selection, manual compilation or program execution. Each event corresponds to some indicators. For example, if the learner compiles his program, some indicators as "the frequency of manually compilation", "the rate of errors correction at compilation process", etc. are calculated. Moreover, each question also corresponds to some indicators. Every time the student carries out one of three events, all indicators concerning with this event and all indicators from the first question to the current question are calculated. So the more questions are carried out, the more indicators are recalculated. The indicators are recalculated because when the learner answers a question, he can modify existent codes. He can then make new mistakes. This recalculation aims to verify whether new errors can be produced.

3.4 Results

From the experiment, we obtained in total 218773 instances of raw data (RD), 3943 instances of the content-data (CD), 135581 instances of the indicators calculated and stored into the database. Indicators' instances were addressed to teachers and displayed on the interface of Hop3x-Teacher. Based on these results, tutors involved in the experiment were able to evaluate the learners' activities and regulate them in order to improve the learning session through interactions with students (for example, 148 audio interventions and 154 textual interventions were carried out).

3.5 Discussion

According to result above, on the average about 3.37 RD were produced (i.e. each student had achieved a maximum of $3.37/11 \sim 0.3$ event per second, one

group with 11 students) and 2.09 indicators are calculated per second. During the process of the experimentation, the tool had responded well to the calculation of indicators in real-time with a group of 15 students. In practice, there were moments where there were no indicators calculated, but occasionally, several indicators were calculated continuously. In the case where several indicators are calculated continuously, the maximum delay time between an event generated and an indicator's instance produced corresponding to this event could be ten seconds. This can be explained as follow. As we have presented in the section 3.3, the calculation of indicators is triggered when an event is generated and the number of indicators to be calculated is increased according to questions carried out by a learner. Therefore, when this learner is in question 10, with each event generated, there are about 70 indicators to be recalculated and it then takes time if many special events are generated continuously.

Moreover, to estimate the computational capacity of the UICT, we performed a simulation on raw data obtained in others experiments. This simulation was done on the same system (software and hardware) that the experimentation above was carried out. With a simulation time of 4187 seconds, the UICT got 65819 RD and produced 48102 instances of indicators. On the average, about 15.72 events were generated and 11.49 indicators were calculated per second. We can conclude that the UICT can meet up with 52 students (15.72/0.3 = 52.4). However, this capacity may be lower than this value. That depends on several factors, for example, the computer hardware, the learners' knowledge, the number of indicators to be calculated, etc. In addition, groups with few students are recommended because a tutor cannot observe many students at the same time.

We have proposed and used DCL4UTL because it is independent from the architecture of databases, from the format of any tracks and from programming languages. It then allows reusability of the calculation method of indicators. For example, we now use JavaCC and eXist XML database to implement the interpreter and DCL4UTL to describe the acquired method of indicators. In the future, we can modify the interpreter using other parser generator and other database, but the indicators modeled will be reused without modifying.

4 CONCLUSIONS

This paper introduces a case study illustrating an UTL's use in which UTL is used as a modeling

language to structure indicators and DCL4UTL is employed to specify how to establish indicators from UTL raw data and other data. Based on values of these indicators, tutors could observe activities of learners, therefore detect their problems and regulate their activities.

As a perspective, we consider the indicator visualization, for example in a graphical form. Actually, each UTL indicator capitalizes the teacher's observation need and the acquired method of data. In the context of tutoring actions, based on the values of indicators, tutors can detect problems, mistakes, misunderstanding, etc. of learners and therefore make interventions to propose solutions. We think that these (mistakes, solutions, etc.) are important and necessary for tutors and learners. However they are currently not capitalized in UTL. We will then work on the capitalization and the reuse of this knowledge. We will also improve our prototype through more experiments in other contexts.

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