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Article in Journal of Research and Practice in Information Technology · August 2012

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Model-Driven Learning Design

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Learning Design is a framework of elements that are used for the formal specification of learning courses. Learning Design languages have been defined to facilitate the editing of online courses, usually including a number of technical formalisms that have proved to be scarcely comprehensible to non-technical staff. In this work we describe a model-driven development approach based on a Learning Design domain-specific language that makes it possible to author and generate learning design courses and to export them into languages and formats of common learning environments and course players.

Keywords: Learning design, model-driven development, domain-specific languages

ACM Classifications: D.2.13 [Reusable Software]: Domain engineering, D.3.2 [Language Classifications]: Design languages, K.3.1 [Computer Uses in Education]: Computer-assisted instruction

1. Introduction

Learning Design (LD) is an educational discipline that aims at converting a set of instructional or didactic principles into a learning environment (Smith and Ragan, 2005). From a technology-based viewpoint, the LD discipline aims at providing a framework of elements used for the formal specification of courses prepared to be played by a computer (Olivier and Tattersall, 2005). Such learning courses are normally hosted and run on software platforms known as Learning Management Systems (LMSs) or Virtual Learning Environments (VLEs), which often use a specific format to package and exchange the courses they store and manage. Because of this, Educational Modeling Languages (EMLs) and Learning Design Languages (LDLs) have been produced to express the software artefacts that result from the LD process.

Developing a generic LD language that makes it possible to author and deploy LD courses for the existing diversity of VLE platforms can be a challenge. The IMS LD specification (IMS, 2003) is a widely recognized LD language that has succeeded in doing so. Instructional designers and technology-enhanced learning researchers, however, have found it difficult to edit IMS LD courses (Burgos and Griffiths, 2005; Griffiths and Blat, 2005).

The authors of an LD may have a certain degree of contact with the conceptual structure of the LDL, depending on the authoring application and its usability. It seems reasonable to explore the
way in which learning designers engage with an LDL through an authoring application. However, the usability of the interfaces of the application (and hence users’ effectiveness with them) does not reveal much about the complexity of understanding the EML. Derntl et al (2010) propose to tackle the complexity issue by implementing improved authoring metaphors in the tools rather than by providing improved versions of the LDL. Nevertheless, shifting complexity from the LDL side to the developer side would significantly increase the effort required to improve the usability of the tools.

This work deals with a model-driven approach to reduce the complexity of editing LDs (France and Rumpe, 2007). Model-Driven Development (MDD) and Model-Driven Engineering (MDE) provide a set of principles and techniques to formalize and represent specific knowledge about a particular interest. MDD is intended to build software artefacts more flexibly and quickly, based on the development and transformation of models. The main idea is to produce executable code through successive transformations of models provided at a higher level of abstraction. The expected benefit of MDD is a reduction in the complexity of LD development and, more importantly, in terms of reusability of the learning model of a course for different platforms.

The objective of the paper is to define a Domain-Specific Language (DSL) that allows an LD to be edited without being committed to the technical formalisms of a given LDL. The domain of such a DSL is the technology-enhanced LD discipline, which aims at formalizing how a set of people in specific groups and roles are engaged in the structure of learning activities, resources and services that describe a learning experience (Olivier and Tattersall, 2005). LD shares many features with the business process management domain (Karampiperis and Sampson, 2007), such as workflows, activities, and roles. In the field of learning experiences, however, assessments are especially distinctive (Strijbos, 2011). Assessments are inevitable when designing the activities, resources, and roles involved in a learning experience. Assessments have also been the basis of proposed extensions to IMS LD (Sitthisak and Gilbert, 2010). In this vein, assessments and assessment-based adaptations of the learning flow are the foundation on which our DSL is based.

2. Review of Related Works

LD languages define a model of elements (namely learning resources, activities, flow connectors, and user roles, among others) that can be combined to design the structure of a course. LD languages are thought to deliver a runnable version of the designed learning experiences to a course player engine (Martens and Vogten, 2005) that can run them. Such course players can also be embedded in specific learning platforms or VLEs (Escobedo et al, 2007). On the other hand, there are relevant examples of VLE-specific playable courses, such as LAMS (Dalziel, 2003), which do not use a standard LD language for the specification of courses. Another example of a widespread VLE that does not implement any LDL is Moodle, which uses a non-standard package format instead to store and backup courses (Dougiamas and Taylor, 2003). LD languages and VLE-specific course formats are very diverse. The aim of having a generic LDL can hinder the deployment of a course, especially if it has to be rendered for a specific VLE platform. In the following, major LDL approaches and their issues are studied, beginning with IMS LD as the main source of research in the LD domain.

2.1 IMS Learning Design

The IMS LD specification (IMS, 2003; Olivier and Tattersall, 2005) is a formal framework of components described in a technical specification consisting of three levels. The information
Model of level A describes elements required to express a static LD, that is, a predefined structure of activities and roles in the course. Level B elements make it possible to describe the dynamic adaptations that can occur in the learning flow according to events and conditions that might happen during execution of the course (Koper and Burgos, 2004; Burgos et al, 2006). Finally, level C defines the kinds of notifications that can be issued to enable the running of course activities. All these components are described in a manifest file and packaged along with content resources in a Unit of Learning (UoL) for each learning course. The core concept around which IMS LD level A components are set is the activity, which can be of two different types, namely learning activities and support activities, depending on their relationship with the learning objectives of the course. Activities can be aggregated as activity-structures, which can be one of two kinds: sequence or selection. Someone playing a role (e.g. a learner or a member of staff) runs the activity. Activities and roles are associated by means of role-parts, which are instantiations of links between a role and an activity running within the method of the LD. The method is structured using the metaphor of a theatrical play, which includes plays and acts. Other IMS LD level A components are learning objects and services, which can be directly or indirectly – that is, through an environment – linked to an activity. The dynamic part of IMS LD is provided by level B elements, which are used to adapt the execution flow of the learning process (Koper and Burgos, 2004). Any event that might influence the learning flow is modeled by a number of properties that govern structured conditions to allow branching from one activity to another. An LD may include adaptations of the learning flow of activities through the use of properties that represent values on the branching decisions to be made, conditions to express combinations of conditional clauses on which the learning flow can branch, and showing or hiding elements to show or hide activities on each branching flow.

The IMS LD specification (IMS, 2003; Olivier and Tattersall, 2005) is a formal framework of components described in a technical specification consisting of three levels. The information model of level A describes elements required to express a static LD, that is, a predefined structure of activities and roles in the course. Level B elements make it possible to describe the dynamic adaptations that can occur in the learning flow according to events and conditions that might happen during execution of the course (Koper and Burgos, 2004; Burgos et al, 2006). Finally, level C defines the kinds of notifications that can be issued to enable the running of course activities. All these components are described in a manifest file and packaged along with content resources in a Unit of Learning (UoL) for each learning course. The core concept around which IMS LD level A components are set is the activity, which can be of two different types, namely learning activities and support activities, depending on their relationship with the learning objectives of the course. Activities can be aggregated as activity-structures, which can be one of two kinds: sequence or selection. Someone playing a role (e.g. a learner or a member of staff) runs the activity. Activities and roles are associated by means of role-parts, which are instantiations of links between a role and an activity running within the method of the LD. The method is structured using the metaphor of a theatrical play, which includes plays and acts. Other IMS LD level A components are learning objects and services, which can be directly or indirectly (that is, through an environment) linked to an activity. The dynamic part of IMS LD is provided by level B elements, which are used to adapt the execution flow of the learning process (Koper and Burgos, 2004). An LD may include adaptations of the learning flow of activities through the use of properties that represent values on the branching decisions to be made, conditions to express combinations of conditional clauses on which the learning flow can branch, and show or hide elements to select the visible activities of each branching flow.
Most significant editing approaches prevent users from authoring some IMS LD elements. For instance, environments, role-parts, properties, and conditions have been found to present significant challenges to teachers’ understanding (Derntl et al., 2010). Consequently instructional designers ought to learn and use technical concepts that are far from their pedagogical background (Griffiths and Blat, 2005). There are also findings on the low scalability of IMS LD course artefacts (Gutiérrez-Santos et al., 2008), because of the excessive number of properties and conditions required to express branching learning flows. Authoring approaches that aim at facilitating the editing of IMS LD UoLs also focus on these issues (Martínez-Ortiz et al., 2009).

2.2 Exchange LD Languages

There is a conceptual distinction between exchange LDLs and authoring LDLs (Martínez-Ortiz et al., 2008). On the one hand, the features of an exchange LDL are intentionally defined to run the LD course on a player, but they are not of special interest to the learning designer. IMS LD is the major example of an exchange LDL that can deliver ready-to-run UoLs. It is difficult for non-technical people to understand and use the IMS LD level B, however (Griffiths and Blat, 2005). On the other hand, authoring LDLs are closer to the learning designer’s common needs and practices, but are not readily playable. Some instances of authoring LDLs are E2ML (Botturi, 2006) and CompendiumLD (Conole and Weller, 2008). Most authoring LDLs define visual languages for both instructional designers and educational software developers (Caeiro-Rodríguez et al., 2010). In this section, we mainly focus on LDL issues as a means of delivering ready-to-run LD courses which can guide learners and instructors while running the LD on a player. The source LD abstractions required to edit such courses are sometimes excessively dependent on the underlying formalisms of the platform on which the course must be run. A lot of approaches use the IMS LD model as both the source and the target abstractions. Others restrict the source model to a subset of the IMS LD specification that does not present considerable difficulties for the engineering (i.e. source-to-target) or re-engineering (i.e. target-to-source) of LD courses based on model mappings (Martínez-Ortiz et al., 2009).

A deeper analysis of existing approaches to LDLs can be found in Dodero et al. (2007), Caeiro-Rodríguez et al. (2010), and Dodero et al. (2010). In the following, this analysis is revisited and updated from the perspective of exchange LDL issues.

The e-LD approach (Martínez-Ortiz et al., 2007, 2009) provides a visual workflow-like source model that must be transformed to the IMS LD course. The visual source language supports the level B conditions by adding flowchart-like branching representations, which are more attainable for the pedagogical designer.

CompendiumLD (Conole and Weller, 2008) is a visual tool used to describe LDs that are delivered as a set of HTML pages in a way that is easily interpretable by both instructor and students. It is useful to sketch a sequence of learning activities, but it does not deliver any executable artefact that complies with any runnable LD specification. Eventually, the instructional designers should use a different tool to deploy and run the course as an IMS LD UoL.

Template-based approaches have also been used to generate the basic structure of IMS LD-based courses (Boticario and Santos, 2008; Hernández-Leo et al., 2006). The template abstractions are parametrized simplifications of the design of some course activities, and have to be instantiated to deliver the final UoL. For example, the Collage tool (Hernández-Leo et al,
2006) provides a visual abstraction based on design patterns to edit an LD. These approaches are usually restricted to a predefined set of patterns that come with the tool.

Authors can also use rich sequencing graphs (Gutiérrez-Santos et al., 2008) to define learning flow strategies, which can then be converted into runnable UoLs. This approach can solve the limitations of the IMS LD level B specification in defining a learning flow. Nevertheless, sequencing graphs are not still a friendly language for instructional designers and pedagogues.

UML4LD (Laforcade et al., 2008) introduces meta-modeling and model-driven principles as part of the LD authoring. A UML profile is provided in a meta-CASE tool to reversely engineer a UML diagram from an IMS LD level A-compliant UoL. Understanding the delivered models, however, requires certain technical knowledge of UML, which is a barrier for instructional designers. More recent works (Laforcade, 2010) use domain-specific modeling tools to build visual instructional design languages, but the approach still requires software engineering skills.

The main drawback of existing LDLs has to do with using the same LD language as a means of both authoring and exchange. All these solutions share the idea that the authored LD model is to be run on an IMS LD-specific player, which is not necessarily the case. Besides, if the potential of a LDL is reduced to the IMS LD metamodel, the resulting LDL can be rather limited, because VLE-specific features cannot be envisaged. That is the rationale of the model-driven approach that is described next.

3. Model-driven Learning Design

The principles of MDD/MDE are applied in this work by the provision of a DSL and a tool that facilitates the transformation between an authoring LD model (that is known by domain experts) and the exchange LDL model that eventually has to be delivered. According to MDD terminology, the exchange LDL is the Platform-Specific Model (PSM), while the authoring LD provides the Platform-Independent Model (PIM). In the LD domain, common exchange LDLs (such as IMS LD or Moodle packages) can be considered as platform-specific languages for the implementation of learning environments. The result of the MDE process has to be the delivery of executable PSM-compliant course artefacts, which can be accomplished in two ways. The first way is to edit models with domain-specific software engineering modeling tools (Laforcade, 2010). The second is to define our own Learning Design Domain-Specific Language (LDDSL) in which LD-specific PIM models can be written. We have opted for this second option, which does not require authors to learn software engineering tools before providing them with an acceptable level of abstraction, flexibility, and control to edit an LD. By using simple annotations, our framework automatically generates PSM-compliant course artefacts from PIM models, allowing non-expert users to specify and generate new learning environments.

3.1 Learning Design DSL

The original requirements of the project included the development of an editing tool with a visual interface and a text-based interface, since the former might pose accessibility issues. In the following, the LDDSL is defined in terms of its abstract syntax or metamodel and its concrete syntax. For the concrete notation, we have defined a text-based and a visual LD syntax; however, here we will focus on the text-based notation.
3.1.1 Abstract syntax

The metamodel of LDDSL is depicted in Figure 1. We restrict our analysis to the core elements, which are used to describe learning activities, learning flows, and learning assessments, as summarized in Table 1.

The core concept is the Activity class. Activities are associated to a set of Resources, which represent both learning resources and services that are required to deploy them in a Web environment. Both simple and composite activities can be defined. The Flow class represents conditional expressions used to alter the learning flow, which can be branched to other activities, depending on the result of a conditional Expression. Since assessments are the usual way to adapt the learning flow, the

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
<th>Icon</th>
</tr>
</thead>
<tbody>
<tr>
<td>SimpleActivity</td>
<td>A simple activity associated to a resource and an assessment, to be included within an assessment activity or composite activity</td>
<td><img src="SimpleActivity.png" alt="SimpleActivity Icon" /></td>
</tr>
<tr>
<td>CompositeActivity</td>
<td>An activity consisting of other activities, to be included within an assessment activity or another composite activity</td>
<td><img src="CompositeActivity.png" alt="CompositeActivity Icon" /></td>
</tr>
<tr>
<td>AssessmentActivity</td>
<td>An entity consisting of an activity plus an assessment</td>
<td><img src="AssessmentActivity.png" alt="AssessmentActivity Icon" /></td>
</tr>
<tr>
<td>Assessment</td>
<td>An entity composed of a set of assessment parameters, which will be the output from an assessment activity or the input to a conditional flow</td>
<td><img src="Assessment.png" alt="Assessment Icon" /></td>
</tr>
<tr>
<td>SimpleFlow</td>
<td>A directed connection between two activity entities; the target entity will be run after the source entity</td>
<td><img src="SimpleFlow.png" alt="SimpleFlow Icon" /></td>
</tr>
<tr>
<td>BranchingFlow</td>
<td>A directed connection between two or more activities, with one source activity and multiple target activities; it may contain a conditional expression that determines which target activity will be run after the source activity and when this will occur</td>
<td><img src="BranchingFlow.png" alt="BranchingFlow Icon" /></td>
</tr>
</tbody>
</table>

Table 1: Summary of LDDSL elements that represent learning activities, learning flows, and learning assessments.
AssessmentActivity links an activity with an Assessment required for conditional adaptations of the flow. The Assessment class holds the Parameters required for the assessment. Finally, a set of Roles can be defined and mapped to activities and assessment activities, in order to provide advanced models of assessment (e.g. self-assessment and peer assessment).

3.1.2 Concrete syntax
The concrete syntax of the text-based version of LDDSL is defined using ANTLR (Parr and Quong, 1995). Listing 1 depicts a condensed version of the LDDSL grammar that covers assessments, activities, flows, and roles. For brevity, basic data types, expressions, and custom tokens (e.g. ID, STR, NATURAL) are not described here.

```
course : courseTag 'course' ID '{' roles courseBody '}';
courseBody : assessmentActivity * ;

// Assessment activities
assessmentActivity :  
  ( activityTags ) ('begin') 'assessment_activity' ID  
  '(' ( compositeActivity assessment | simpleActivity assessment ) | learningFlow ')' ;
compositeActivity :  
  ( activityTags ) 'composite_activity' ID  
  '(' activityRole 'title' STR activity * assessmentActivity * ')' ;
simpleActivity :  
  ( activityTags ) 'simple_activity' ID
```

Figure 1: LDDSL abstract syntax to describe learning activities, learning flows, and assessments.
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```
'(' activityRole 'title' STR activityResource ')';
activity:
  simpleActivity | compositeActivity;
activityResource:
  'resource' STR;
assessment:
  assessmentProcType '(' assessmentRoles 'title' STR assessmentParameters ')';
assessmentProcType:
  'assessment' | 'self_assessment' | 'peer_assessment';
assessmentParameters:
  assessmentParameter *;
assessmentParameter:
  'collection' ID collection
  | 'natural' ID rank
  | 'boolean' ID ('true' | 'false');
  | 'string' ID (STR);
// Learning flow
learningFlow:
  branchingFlowTag 'flow': [condition]
  | branchingFlowTag 'flow': ID |
branchingFlowTag:
  ( layoutTag | flowTag )+ |
// Roles
roles:
  | 'roles' '{( role )* }';
assessmentRoles:
  assessorRole ( activityRole ) |
assessorRole:
  'evaluator_role' ID (',', ID)* |
activityRole:
  'teacher_role' ID (',', ID)* | 'learner_role' ID (',', ID)* |
```

Listing 1: ANTLR grammar of the text-based LDDSL.

To implement exchange LDL-specific extensions to LDDSL, attribute-oriented programming techniques have been applied (Pawlak, 2005). Such techniques use tagged annotations (usually having an @ prefix) that can be included in the source model before certain source elements. For instance, an @imsld tag before an activity description can be used to describe the IMS LD-specific type of activity to be generated. An excerpt of the ANTLR grammar rules that define extension tags is shown in Listing 2.

```
courseTag:
  '@moodle' | '@imsld' | '@lams' | '@lpcel';
activityTags:
  ( layoutTag | activityTag )+ |
activityTag:
  '@imsld' ':' imsldActivityAttr
  | '@moodle' ':' moodleActivityAttr
```

Listing 1: ANTLR grammar of the text-based LDDSL.
Attribute-oriented programming techniques reduce the redundancy between the abstract and concrete syntax of the DSL (Krahn et al., 2007). Tagged annotations can be detached from the source specification, which can then be reused for designs having another LDL or VLE as the PSM. In the visual environment, such attribute-oriented extensions are deployed as forms that make it possible to give values to specific properties.

3.1.3 Transformations

From an LDDSL source specification, specific code of an exchange LDL can be generated. For that aim, first we defined a number of transformations between the LDDSL source model and the IMS LD model to enable the LD course to be run on regular IMS LD engines. Second, another set of transformations were defined to consider specific VLEs as the PSM. In this case, the target of such transformations can be Moodle or other VLEs that include their own specific types of activities, role mappings, and learning flows, among other details. PIM-to-PSM transformations are driven by the tagged attributes that are specific to the target LDL or VLE. The details of IMS LD UoLs and Moodle exchange packages are defined by the permitted attribute values of \texttt{@imsld} and \texttt{@moodle} tags, respectively. The LDL PSM of the editing tool can be extended to include other exchange LDLS. For instance, we have easily extended the concrete notation (i.e. the last rules of Listing 2) in order to configure a tool variant prepared for the LPCEL language (Torres et al., 2006). LPCEL is an LDL that uses improved synchronization mechanisms for learning processes (e.g. concurrent split-join branching flow) not considered in other LDLS.

3.2 LDDSL Editing Example

The LDDSL editing environment is available through a stand-alone tool implemented using ActionScript. It can be used to design and export playable LD courses. The editing interface provides common LDDSL abstractions (such as activities, roles, learning flows, and assessments)
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to edit a source LD specification. For that aim, it provides a text-based and a visual interface. The LDDSL source of a course can be visually edited in the tool or directly edited in a text view. The current version of the editing environment can generate IMS LD UoLs, LPCEL learning processes, and Moodle packages. For the tool serialization notation we took advantage of the text-based concrete syntax, and extended the language with additional @layout tags that hold layout information (e.g. the position of forms and connectors in the canvas). That provides a rudimentary form of reverse engineering from LDDSL specifications, which can be modified with external text editors and imported again in the tool.

Listing 3 depicts the LDDSL representation of a course, designed to cover the main elements of a regular LD situation. The first sentence is the specification of roles, which defines a new peer role that is defined as an extension of learner and evaluator predefined roles and is intended for the final assessment activity. After the definition of roles, four assessment_activities are defined. They are surrogates of simple activities (ae1, ae2 and ae4) and one composite activity (ae3). Activity ae1 is marked with the begin flag and is connected by a conditional flow with ae2 and ae3. Assessment activities are wrappers of regular activities that encompass all assessment-related information. This is a common design technique for extending a class interface without changing it. In this way, each activity is provided with an assessment specification which describes the type of assessment (i.e. self-assessment, peer assessment, or regular assessment), the assessment parameters, and the roles that must develop it. The ae1 activity renders as its output a natural parameter that can be read later to branch the learning flow. For example, after ae1, the flow will branch to ae2 activity, looking for some learning support if the grade value is below 5. Assessments may include a resource that defines the application or service to be used for assessment. In ae2 and ae4, two assessment instruments hosted in the http://evalcomix.uca.es/service are used for this purpose.

course sample_course {
roles { role peer extends learner, evaluator }
begin assessment_activity ae1 {
  simple_activity as11 {
    learner_role learner
    teacher_role teacher
    title “Beginning”
  }
  assessment {
    evaluator_role evaluator
    teacher_role teacher
    natural grade [0,10]
  }
  flow: if: grade < 5
    then: ae2
    else: flow: ae3
}
assessment_activity ae2 {
  @imsld:support_activity
  simple_activity as21 {
    learner_role learner
    teacher_role teacher
  }
}
title “Support”
}
assessmet {  
evaluator_role evaluator  
resource "http://evalcomix.uca.es/instrument/view?id=85cf15"  
natural grade [0,10]  
boolean finished  
collection scale ["good", "regular", "poor"]
}
flow: ae3

assessment_activity ae3 {
  @imsld:learning_activity  
  @imsld:activity_structure sequence  
  composite_activity ac1 {
    learner_role learner  
    teacher_role teacher  
    title “Content”  
    simple_activity as31 {
      title “Activity 1”  
      resource "http://docs.google.com/document/d/1CA7e47dSi"  
    }
    simple_activity as32 {
      title “Activity 2”  
      resource "http://docs.google.com/document/d/jD6kd273DS"  
    }
  }

assessment {
  evaluator_role evaluator  
  teacher_role teacher
}
flow: ae4

assessment_activity ae4 {
  simple_activity as43 {
    learner_role learner  
    teacher_role teacher  
    title “Final Activity”
  }

assessment {
  evaluator_role peer  
  teacher_role teacher  
  resource "http://evalcomix.uca.es/instrument/view?id=74fd04"  
  collection final_grade ["fine","medium","poor"]
}
}

Listing 3: Text-based representation of a learning course using LDDSL.
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Listing 3 also shows an example of PSM annotations including tagged values to represent two IMS LD activities: one is an IMS LD support-activity configured with a teacher-based assessment service, and the other is a sequence-type activity-structure for an IMS LD learning-activity that provides access to a pair of Google Docs resources.

4. Evaluation

For the evaluation of our approach we will follow first an analytical method and then a case study evaluation, as proposed by Hevner et al (2006). The former will evaluate the quality of LDDSL as a design artefact. It serves to examine the structure of an LDDSL specification for static qualities such as computational complexity. We assume that IMS LD is expressive enough to fulfil many LD situations (van Es and Koper, 2006) and it can be used to represent most learning scenarios. Since LD languages are Turing-complete, it is pointless to compare them only from the viewpoint of expressiveness (van der Aalst and van Hee, 2002). Kiepuszewski et al (2003) evaluate workflow languages based on suitability, which responds to whether the language provides direct support for patterns which appear frequently in domain-specific designs. With that aim, a case study evaluation has also been designed, consisting of a set of common LD situations. The aim is to find out the main source of complexity in the final courses that are delivered in a target language. We have selected the IMS LD specification as the target language in order to test if IMS LD is limited in expressing varied learning flows (Gutiérrez-Santos et al, 2008).

4.1 Analysis of Complexity

The complexity of an LD specification can be expressed as a function of the evaluable attributes in the graph of the learning activities. Graph nodes represent activities; graph edges represent the flow between activities. The complexity of a course can be described as an expression of the following variables:

- $a_1 = \text{Number of source activities, that is, activities in which a flow starts}$
- $a_2 = \text{Number of sink activities, that is, activities in which a flow ends}$
- $a_3 = \text{Number of clauses in conditional branching flows}$

The type and number of IMS LD elements required to represent each element of LDDSL is computed as follows. Table 2 shows the types of IMS LD elements required to implement each adaptation feature of the conceptual model, the number of elements, and the formulas used to compute the overall number of elements. Simple, composite, and assessment activities require a finite number of IMS LD level A activity elements. Conditional branching flows require a number of IMS LD level B conditional elements that is a function of $\sum_{i=1}^{3} (c_i a_i)$ which has $O(n)$ complexity, with $n = max(a_i)$ and $c_i$ constants. If conditional flows are chained, the complexity increases by a multiplicative factor of $n$ due to the additional nested branching that is required. So the complexity of IMS LD specifications is $O(n^k)$, with $k$ being the maximum nesting level of conditional flows. In contrast, an equivalent LDDSL specification will only require a constant number of additional elements for each conditional branch, so the complexity remains linear.
This evaluation could be complemented with an analysis of the performance of the UoL in an IMS LD player. Nevertheless, the purpose of our research is not to compare the performance of different LD specifications, but to evidence the lesser complexity of editing LD adaptations with LDDSL.

4.2 Learning Design Cases

For the second evaluation, three common LD cases have been designed. These situations are characterized by the kinds of activity elements included (simple and composite), the types of assessments, and the kinds of flow primitives included. The flow of activities can be sequential or conditionally branched. The learning flow can go forward or backward, so that loops can be defined for remediation activities. Either the teacher or the learner role can carry out assessments. The cases have been designed to cover all the learning flow possibilities of the LDL.

- **Case 1** (see Figure 2a and Listing 4) describes a set of activities and a conditional branching that can flow forward and backward until the *Last* activity is reached. If remediation is needed, the flow is directed to the *Support* activity, which has a backward connection to the *First* activity. Assessments that determine the need for remediation are carried out by the teacher role.

- **Case 2** (see Figure 2b and Listing 5) describes a set of activities and a non-conditional branching flow so that the learner can select between two ways of action (i.e. *Activity 1* or *Activity 2*). Assessments are carried out by the learner roles on a peer-assessment basis.

<table>
<thead>
<tr>
<th>LDDSL element</th>
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<th>Count</th>
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<td>locpers-property</td>
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</tr>
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<td>property-ref</td>
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<td>property-value</td>
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</tr>
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<td></td>
<td>complete</td>
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<td></td>
</tr>
</tbody>
</table>

Table 2: Type and number of IMS LD elements required to represent the elements of the LDDSL metamodel.
Figure 2: View of three LDDSL courses that consist of assessment activities and branching flows.
Case 3 (see Figure 2c and Listing 6) describes a set of activities and a conditional branching that flows forward. After the beginning activity, the flow can branch forward to the Content activity or to the Support activity. Assessments that determine the need for remediation are carried out by each learner role as self-assessment.

course case_1 {
    @layout:(50,63)
    begin assessment_activity ae1 {
        simple_activity as11 {
            learner_role learner
            teacher_role teacher
            title "First Activity"
        } assessment {
            evaluator_role evaluator
            teacher_role teacher
            boolean ok false
        }
    }
    @layout:(273,104)
    flow: if: ok = true
        then: ae2
        else: flow: ae3
    }
    @layout:(285,229)
    assessment_activity ae2 {
        simple_activity as21 {
            learner_role learner
            teacher_role teacher
            title "Last Activity"
        } assessment {
            evaluator_role evaluator
            teacher_role teacher
        }
    }
    @layout:(53,231)
    assessment_activity ae3 {
        simple_activity as31 {
            learner_role learner
            teacher_role teacher
            title "Support Activity"
        } assessment {
            evaluator_role evaluator
            teacher_role teacher
        } flow: ae1
    }
}

Listing 4: Text LDDSL specification of the course of Figure 2a.
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course case_2 {
    begin assessment_activity ae1 {
        composite_activity ac11 {
            title "Composite Activity"
            simple_activity as111 {
                title "Simple Activity"
            }
        }
        assessment ( ... )
        @imsld: optional
        flow: ae2, ae3
    }
    assessment_activity ae2 {
        simple_activity as21 {
            title "Activity 2"
        }
        assessment ( ... )
        flow: ae4
    }
    assessment_activity ae3 {
        simple_activity as31 {
            title "Activity 3"
        }
        assessment ( ... )
        flow: ae4
    }
    assessment_activity ae4 {
        composite_activity ac41 {
            title "Final Assessment"
            simple_activity as411 {
                title "Simple Activity"
            }
        }
        assessment ( ... )
    }
}

Listing 5: Text LDDSL specification of the course of Figure 2b.

course case_3 {
    roles {
        role peer extends learner, evaluator
    }
    begin assessment_activity ae1 {
        simple_activity as11 {
            title "First Activity"
        }
        assessment {
            evaluator_role teacher
            boolean ok false
        } flow: if: ok = false
        then: ae2
        else: flow: ae3
    }
    assessment_activity ae2 {

simple_activity as21 {
    title “Support Activity”
} peer_assessment {
    evaluator_role peer
} flow: ae3

assessment_activity ae3 {
    composite_activity ac31 {
        title “Content Activity”
        simple_activity as311 {...}
        simple_activity as312 {...}
    } self_assessment {
        evaluator_role learner
    } flow: ae4

assessment_activity ae4 {
    simple_activity as41 {
        title “Final Activity”
    } assessment {
        evaluator_role teacher
    }
}

Listing 6: Text LDDSL specification of the course of Figure 2c.

The measurements of complexity of the IMS LD-based UoL generated in each evaluation case study are shown in Table 3. Some conclusions can be drawn from these results:

- Level A structural elements introduced by simple and composite activities are not the source of much complexity, as observed in cases 1 and 3. An exception is the need to include as many references to learning activities as the number of simple activities defined within each composite activity. The latter is observed in case 2, which has two composite activities including other simple activities.
- The main source of complexity for assessment activities is the need to include many activity references and properties, as obliged by the level B primitives that are generated.
- Non-conditional branching flows do not add much complexity, as shown in case 2. This is because they can be represented with simple references and attributes belonging to level A.
- Conditional branching flows are the major source of complexity, as shown in cases 1 and 3. The required number of level B elements is greater than in any other situation.

The main source of complexity added to the delivered IMS LD UoL artefact is due to the adaptation details required by level B of the IMS LD specification. Such details are automatically generated from an LDDSL specification. For instance, a single entity of LDDSL, such as a conditional flow, requires a number of elements that is O(1), versus the O(n) required if the level B properties and conditions of IMS LD are edited directly.
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<table>
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<tr>
<th>LDDSL element</th>
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<th>Case 3</th>
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<td>is, is-not</td>
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<td><strong>38</strong></td>
<td><strong>28</strong></td>
<td><strong>32</strong></td>
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</tbody>
</table>

Table 3: Complexity of the IMS LD elements that represent the evaluation cases.

5. Conclusion and Future Work

This work describes a model-driven approach to edit LDs and presents LDDSL, an LD-specific language that relieves the instructional designer of needing to know software development techniques, course exchange formats, and technical details of LD specifications. With the model-driven approach, the complexity of using an LD language is reduced so that the reusability of the learning models for different learning platforms can be improved. Both standard and specific LD models, such as IMS LD and Moodle, are generated from an LDDSL model simply by annotating it with LDL-specific tags that complement it. A different annotation on the same model would be enough to generate the same learning environment on another platform.

The LDDSL provides learning designers with a set of editing elements that can be more suitable than exchange LDL elements. LDDSL is not more expressive than IMS LD, but it contains a set of frequently recurring structures (such as assessments and learning flows) that, for a learning
designer, can become more convenient than IMS LD level B elements. Although such structures can be directly expressed with IMS LD technical details, the LDDSL tool implementation generates an equivalent IMS LD specification representing such patterns, which can be more easily understood by the learning designer. The elements of LDDSL have been analysed and compared with the corresponding ones of IMS LD, with the result that: (1) IMS LD is expressive enough to represent common LD situations described in LDDSL; and (2) LDDSL has less complexity for expressing learning flows and learning assessments. An authoring tool is provided with a visual and a text-based editing interface to design an LDDSL specification from which an exchange LDL-based course can be generated. An attribute-oriented tagging approach has been followed to extend LDDSL courses with details of a specific exchange LDL or learning platform. The LDDSL approach has been previously analysed with respect to the generation of IMS LD-compliant units of learning and Moodle 1.x courses (Dodero et al., 2010). Tagging has also been used to implement a rudimentary form of reverse engineering from an LDDSL text-based specification. Nevertheless, reverse engineering from fully fledged IMS LD UoLs or Moodle packages is still an issue that needs to be explored. In this vein, the Moodle 2.x platform is expected to provide a more widespread user environment test bed. Particularly interesting is the improved model of conditional activities and service integration of the latest Moodle versions, which were not available in former versions of the platform. In addition, the way in which users can engage with the LDDSL language through the authoring tool must be further investigated. Users’ effectiveness with the tool interface may confuse the complexity of understanding the DSL model with the usability of the tool, so a careful co-design experience must be carried out.

Acknowledgements

This work has been sponsored by a grant from the ASCETA project (ref. P09-TIC-5230) of the Andalusian Government Research and Development Programme and a research stay grant (ref. EST2011-075) from the University of Cádiz.

References


Model-Driven Learning Design


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Manuel Palomo-Duarte is a lecturer of the department of informatics of the University of Cádiz, in which he teaches courses on videogame development and operating systems using open source software. He obtained a CS degree from the University of Seville and a CS PhD from the University of Cádiz. He is currently coordinator of the CS degrees at the school of engineering, member of the software process improvement research group, and has been the former director of the Libre Software and Open Knowledge Office of the University of Cádiz.
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