Abstract
This paper presents approaches and solutions for eLearning personalization, which are implemented in some contemporary Learning Management Systems (LMSs). The importance of the appropriate user model construction and its relation towards support of the different types of personalization are briefly described in section 1. In the section 2 of the paper is described a solution for role- and competency-based learning customization that uses Web Services approach. A method for development and design of adaptive learning content and on the processes of customization in eLearning in terms of learning strategy system support is represented in section 3. A key requirement of LMSs is personalization of learners’ access to learning objects by providing results tailored to the individual or group of learners as the response to search queries. Some aspects of personalization can also take place even before a query is submitted for evaluation. These issues are included in section 4. The last part presents how personalization techniques are implemented in Learning Grid-driven applications.

Keywords

1. INTRODUCTION
Contemporary eLearning systems support the creation, storage and presentation (often via a web browser) of learning materials in a structured way. eLearning materials are developed as appropriately indexed learning objects (LOs) at the ‘course’ level of granularity. LOs are generally available and searchable via the Web and are collected in learning object repository (LOR). The ‘courses’ deal with by LMSs are an aggregation of the different kinds of objects found in LOR. The possibilities for adaptation, personalization, modality, record-keeping on student’s performance, and usage statistics for the system as a whole are key components of the LMS functionality [1].

Learning personalization includes how to find and filter the learning information that fits the user’s preferences and needs, how to represent it and how to give the user tools to reconfiguration the systems, in consequence, reconfiguration system could be part of personalized environment in some systems.

For instance, WebCT can provide personalised learning paths for users, as access to objects can be conditioned on a wide range of personal data. There would seem to be a lot more scope
for individual personalization than this, as the WebCT system records quite a lot of information on the behaviour of the users (which documents they visit, how long they spend viewing them, test results and grades. At present this information seems to be used solely for the production of reports, but could be used towards providing a truly personalised educational experience without the need for large investments of course designers’ time.

Xtensis claims to be “a revolution in the management and delivery of e-learning” as it is specifically designed to handle LOs and their (IMS and SCORM-compliant) metadata. It is usable ‘out of the box’ as a learning management system, but can be configured to reflect the structure of an organisation and is more an architecture than a single product. It is used as the content management system for several UK-based LOR projects, including the National Learning Network (http://www.nln.ac.uk/), the Seeveaz Key Skills repository and Iconex (http://www.iconex.hull.ac.uk/).

Moodle (Modular Object-Oriented Dynamic Learning Environment) is an open source software package for producing internet-based courses and web sites. Depending on the difficulty level of the learning content and the general level of knowledge and skills of the learners, learning process can be realized dynamically. The exemplary scenario is the following. The capacity for personalization of the environment is subject to be improved further in the next main release of the product.

The user modelling is the process of constructing (often computer-based) users models, while the user model means all the information collected about a user that logs to a web site, in order to take into account her needs, wishes, and interests. Every LMS has its techniques to modelling his users so as to construct the user model or profile. The process of user modelling covers user model as an output, as well as background knowledge, and the user behaviour. Construction of an effective user model and tracking of its continuous changes are a real challenge in contemporary LMS. The quality and expressive power of the user model is crucial in respect to the implementation of intelligent support for different adaptive teaching strategies and their switching during the personalization, of [2] [3] [4]:

- the learning content, based on learner’s preferences, educational background and experience, learning content tailored to individual learning style of the user;
- the representation manner and the form of the learning content (for example, learning content in the form of the adaptive learning sequences of learning objects).
- a combination of the previous two types.

Personalization in current Learning Management Systems tends to be concerned with remembering which courses the user is allowed to view and how they like their pages to be presented. In some cases users (learners, teacher and administrators) are able to edit their own profile; to maintain their personal calendar (monthly and weekly) which keeps track of their event transactions; to subscribe to forums, etc. Observing the educational process as a whole, learners are very rarely allowed to get access to learning objects which are conditioned on a wide range of personal data including achievement, date/time and class code.

The following approaches (based on user model) can be used to apply the learning personalization:

- Personalization, controlled by the learner – It requires direct input of the learner's needs and preferences by filling question forms or by choosing options and alternatives.
- Personalization, based upon an existing user profile and meta-descriptions of the information content - In this case, the learners' preferences are stored in their profile.
- Personalization via searching for a correlation between the learners - Correlation is through the values of the attributes, describing the learner's profile. If there is a strong correlation, there is a possibility that the content for a given profile is suitable for applying to its close (adjacent) profiles.
User modelling is a complex and sophisticated one. Application of the semantic web approach to representing student model based on multiple student data with respect to the most important and well-developed learner model standards will help and decrease the difficulty of the process. Ontology is a key Semantic Web technology for marking up electronic resources. Ontologies typically consist of definitions of concepts relevant for the domain, their relations, and axioms about these concepts and relationships.

2. CUSTOMIZED LEARNING AND WEB SERVICES APPROACH

Customized learning, presenting just the right material to the learner on demand, can be described using data representations from learning technology standards (learner profiles, competency definitions, sequencing rules, learning objects). William Blackmon and Daniel Rehak offer a web services-based methodology for customization by profile, specifically one of eliminating LOs from a course [7] because either:

a. Learner's current role does not require the learning objective taught by the LO, or
b. Learner's profile indicates the learner has already achieved the objective taught by a LO.

The learning content and data used in customization are represented in a set of standards-based data models. These are used in a content authoring and delivery process that customizes the activities delivered to the learner based on the learner's role and competencies [9] [10].

Content and learning activity customization uses six sets of data elements (with data representations taken from current learning technology standards):

- **Learning Objects** -- the collection of content and learning resources maintained in a content repository.
- **Content Structure** -- the organization of learning objects in a tree or hierarchical structure.
- **Roles** -- definitions of the job roles of a learner.
- **Competency Definitions** -- definitions of the skills and knowledge acquired by a learner.
- **Learner Information Package** -- the collection of stored profile information about a learner.
- **Sequencing** -- rules used to select content and sequence the learner through a content structure.

The major steps for a customized course preparation and delivering are:\footnote{Assuming there is a globally defined set of learner job roles and competency definitions}:

- **Create Course and Content Description** -- describe the course (content structure and set of LOs) and behaviour rules used to express the progression of the learner through the content:
  - Associate role and competency definitions with each learning object by mapping a sequencing objective id (used to label the objective) to a competency definition id or to a role id.
  - Specify the conditional rules used to customize the course by eliminating learning objects from the activity sequence.
- **Establish Learner Profiles** -- specify the role of the learner (which in turn may yield a set of competencies required to perform the role), and contain data on the learner's record relative to each of the specified competencies.
- **Register Learners** -- register the learner for the course.
- **Deliver Course** -- deliver the course, matching the course description to the learner's profile to select content. As the learner completes instruction, the profile may be
updated to include mastery of subject matter. Delivery and customization continues until all required activities have been completed.

The customization process has been implemented through a set of web services. Rather than building large, closed systems, the focus is on flexible architectures that provide interoperability of components and learning content, and that rely on open standards for information exchange and component integration. The overall web services architecture for learning is divided into layered services. The layers from top to bottom in this services stack are:

- **User Agents** -- provide interfaces between users (both end user applications and program agents) and the learning services. Agents provide the major elements of learning technology systems: authoring of content, management of learning, and actual delivery of instruction to learners.

- **Learning Services** -- collection of (many small, simple) data models and independent behaviours. Service components are characterized as providing a single function that implements a particular behaviour. Each service is identifiable, discoverable, (de)referenceable, and interoperable. They include built-in security and rights management, and assume an unreliable underlying network. Services are grouped into logical collections, where upper-level services rely on the support from the lower-level services:
  - **Tool Layer** - Tools provide high-level, integrated server applications. Accessed via known, published interfaces, they provide the public interface to the learning tools (tutors, simulators, assessment engines, collaboration tools, registration tools, etc.). User agents and end user applications are built using collections of tool services.
  - **Common Applications Layer** - These are services that provide the commonly used learning functions and application support behaviours used by tools and agents (sequencing, managing learner profiles, learner tracking, content management, competency management, etc.).
  - **Basic Services Layer** - Basic services provide core features and functionality that are not necessarily specific to learning, but which may need to be adapted for learning (storage management, workflow, rights management, authentication, query/data interfaces, etc.).

All services are built on and use a common infrastructure model. The infrastructure layer relies on basic Internet technologies (e.g., HTTP, TCP/IP) to connect service components over the network. The services themselves are implemented using web services bindings. Messaging is done with SOAP; service descriptions are catalogued with UDDI, and described in WSDL - all are XML representations [6]. Overall service coordination is expressed in a workflow or choreography language. These standard technologies permit the upper-level services to be implemented in a platform-neutral manner, and provide interoperability across different implementations of the actual learning services.

3. METHODOLOGY FOR DEVELOPMENT AND DESIGN OF ADAPTIVE LEARNING CONTENT

The term “adaptive learning” means the capability to modify any individual student’s learning experience as a function of information obtained through their performance on situated tasks or assessments. With the integration of the IMS Simple Sequencing (IMS SS) Specification [10], SCORM [11] allows the learning strategies to be translated into sequencing rules and actions, which are associated with the activities a learning experience consists of. The sequencing rules are based on learner’s progress and performance and affect the availability of the learner is allowed to experience.
All learning activities can be associated with sequencing information defined by the content author. In run time, each activity experienced by the learner is associated with tracking status data, which may affect the overall sequencing process. This means that learners with difficulties in satisfying the learning objective should be able to experience additional activities (or repeat some of the activities) to improve their knowledge level and skills. Some restrictions concerning number of attempts and/or period of time for any activity could be set by the content author.

The process of defining a specific sequence of learning activities begins with the creation of a learning strategy for the achievement of the determined pedagogical aim/s. Learning strategy specifies types of learning activities and their logical organization (the activity tree) as well as the prerequisites and expected results for each activity. The rules for managing the instructional flow are the other important part of the strategy. Describing the rules by means of IMS SS elements and attributes the content author transforms the sequencing strategy into strategy for the activity tree traversal management. The author establishes an aggregation of learning objects associating leafs of the activity tree with appropriate Sharable Content Objects (SCOs). The outcome of this process is a content package. The imsmanifest.xml file of the package describes SCOs organization and their sequencing. The implementation of adaptive learning in an eLearning environment could be promoted and facilitated by providing of sequencing templates for the development and design of instructional flows. The following figure describes the main stages of the sequencing loop.

![Figure 1: Main stages of the sequencing loop](image)

The sequencing template describes the conceptual organization of the learning content as a sequence of template pages and provides the learning strategy implementation translating it into sequencing strategy. Such sequencing template can be used in different knowledge domains from different instructors who want to follow the described in the package content organization and the implemented learning strategy. In this case, instructor is responsible only to identify (or create) and then to incorporate the relevant multimedia content in each of the template pages accordingly the subject matter of the course taking into consideration the concrete learning objectives and context [12].
Simple Sequencing recognizes only the role of the learner and does not define sequencing capabilities that utilize or are dependent on other actors. The specification includes a limited number of widely used sequencing behaviours. In particular, IMS SS does not address, but does not necessarily preclude, artificial intelligence-based sequencing, schedule-based sequencing, collaborative learning, or sequencing requiring data from multiple parallel learning activities.

The main advantage of the Simple Sequencing approach is that the sequencing rules are described outside the learning objects’ content. In this way, the instructional designer can change the rules (i.e. the learning strategy) without any changes in the content or its organization. Nesting manifests of the developed sample packages the content author can developed more complex strategies and content structures.

4. TAILORING TO INDIVIDUAL LEARNING STYLES

In the contemporary learning environments personalization techniques of learners’ access to learning objects have to provide results tailored to the individual or group of learners and their learning styles as the response to search queries. When users search for LOs the results returned to them will depend on who they are as well as their query, since different LOs may be more appropriate for different learners. Personalization will have an effect on search results returned from a keyword-based query at three different levels [8]:

- **Filtering** of the returned LOs - excluding those LOs deemed unsuitable for the learner, even though they satisfied the original query;
- **Ranking** of the returned LOs - the ‘best’ LO for one user may be different from the ‘best’ LO for another, but personalized ranking means that they can both have the most suitable LO for them returned at the top of their search results;
- **Presentation** of results - users will have different preferences for the display of their search results (e.g. display results as trails or as a simple list, display 10 results per page or 50 results per page).

Some aspects of personalization can also take place even before a query is submitted for evaluation: personalized queries can be constructed using information stored in the profile, by re-formulating or annotating the user’s original query to reflect elements of their profile. The user profile has to contain information about preferences, aims, and educational history that can be used by the system. This is the first stage of filtering.

Keyword-based query is not the only way that users can locate LOs – the schema of the LO descriptions can also be browsed to find relevant LOs, providing facilities such as ‘browse by author’ and ‘browse by subject’. Personalization of the browsing process can occur at two levels:

- Allowing users to restrict the information they see to only those attributes of interest to them, organised in their preferred manner.
- LMS can use knowledge of a user’s preferences (either those explicitly supplied by the user or those learned by the system itself) to recommend individual LOs or categories of LOs to the user as they are browsing.

**Filtering and ranking search results**

The query service will return a set of LO descriptions - all those LOs that satisfy the user’s query. The user wants to be able to find exactly the right LO quickly, without having to browse too many of the results, so rather than present the results exactly as they are returned by the query service some processing is done first.
If a profile of the user is not available (or the user has personalization turned off) then all that can be done at this stage is some rudimentary ranking of the result set, possibly using standard ranking techniques from information retrieval and web search.

However, we anticipate that usually some minimal profile will be available to the system, as users should supply at least some minimum information into their profile when first registering. In this case the ranking of LOs will involve personalization. This means that the system can attempt to show the user only those results likely to be most relevant to them personally, as well as relevant to the query in general.

The first step in this processing is to filter the results - remove all those LOs that we are certain will be of no use to the user. At this stage, for example, any LOs in languages that the user does not understand can be eliminated, as can those not meeting accessibility requirements, those at a far too high or low level of difficulty and possibly those covering only material that the learner is already completely familiar with.

Next, the remaining set of LO descriptions must be ranked in order of relevance to the user. Whereas filtering can be done with just the user profile, ranking a set of results should take the original query into consideration too (i.e. relevance must be judged against the combination of user profile and query, not just the profile).

The best algorithm to use for this ranking is still an open question, but it will take into consideration:

- Relevance of the LO to the query;
- How well the LO caters for the user’s accessibility requirements;
- Whether the user has the prerequisite knowledge and experience;
- Matching between the user’s goals and the learning objectives of the LO;
- If the user’s learning styles are catered for by the LO;
- If the user is likely to prefer it for other reasons (it is by a preferred author, say);
- The user’s most recent activity.

The clear individual semantics of each section of the user profile allows focussed matching against relevant sections of the LO descriptions. For a LO to be a “good” LO for the user, the greatest possible number of different elements will match to some degree. Clearly, though, some factors are more important than others to the user and a good algorithm for combining them will reflect this. For example:

- If LO X caters for one of the user’s learning styles but is not very relevant to the original query then other, more relevant LOs should be ranked higher even if their descriptions don’t list one of the user’s learning styles;
- If LO Y has a learning outcome that matches one of the user’s goals but is far too difficult for the user to tackle (they have none of the prerequisite knowledge, say) then again other LOs (closer to the user’s level) should be ranked higher.

With so many factors to take into consideration, discovery of which algorithms work better or worse for which groups of users requires much further work and testing, and is beyond the scope of this project. It may be that the ranking algorithm itself needs to adapt to the individual, and will differ from user to user (an additional section could be added to the user profile to store information about parameters used by the ranking algorithm).

**Support for browsing as a trail**

As the user is browsing LOs the trails and adaptation service can actively recommend the next LO to look at, effectively generating trails of length two (i.e. a trail consisting of the current LO and a suggestion for the next one) at every stage of the user’s browsing, based on the user profile.

The recommendations can be derived in several ways:
• from the semantic relationships between the current LO and other LOs in the LMS repository;
• from the user’s profile plus LO metadata - perhaps suggesting LOs that cover more advanced material on the same topic, and also suit the user’s preferences (learning style, accessibility, etc.);
• through a process of collaborative filtering, suggesting as the next step a LO that other similar users browsed after seeing the current LO (where similar users can be identified by having similar preferences or similar histories of LO access).

5. PERSONALISATION TECHNIQUES IN LEARNING GRID-DRIVEN APPLICATIONS

The philosophy and the approach behind Grid technologies [9] show the right characteristics for achieving an effective learning. Indeed, they allow to access and integrate the different technologies, resources and contents that are required in order to realise new paradigms in eLearning. They are the most promising approach to realise an infrastructure that will allow learning process actors to collaborate, to take part in realistic simulations, to use and share personaliselly high quality learning data and to innovate solutions of learning and training. Grid will be able to support learning processes allowing each learner to use, in a transparent and collaborative manner, the resources already existing on-line, by facilitating and managing dynamic conversations with other human and artificial actors available on the grid, etc.

The SeLeNe (Self eLearning Networks, http://www.dsc.bbk.ac.uk/selene/) project was funded as an EU FP5 Accompanying Measure (IST-2001-39045) running from 1st November 2002 to 31st January 2004. This project was part of action line V.1.9 CPA9 of the IST 2002 Work Programme, contributing to the objectives of Information and Knowledge Grids by allowing access to widespread information and knowledge, with eLearning as the test-bed application. The developers conducted a feasibility study into using Semantic Web technology for syndicating knowledge-intensive resources (such as learning objects) and for creating personalized views over such a Knowledge Grid. A self e-learning network consists of web-based learning LOs that have been made available to the network by its users, along with metadata descriptions of these learning objects and of the network’s users. The architecture of the network is distributed and service-oriented. The personalization facilities include: querying learning object descriptions to return results tailored towards users’ individual goals and preferences; the ability to define views over the learning object metadata; facilities for defining new composite learning objects; and facilities for subscribing to personalised event and change notification services.

ELeGI (European Learning Grid Infrastructure, http://www.elegi.org/) is an EU-funded Integrated Project that aims at facilitating the emergence of a European GRID infrastructure for eLearning and stimulating research of technologies to enhance and promote effective human learning. The project is supported by the European Community under the Innovation Society Technologies (IST) programme of the 6th Framework Programme for RTD - project ELeGI, contract IST-002205.

ELeGI promotes and supports a learning paradigm shift focused on knowledge construction using experientials based and collaborative learning approaches in a contextualized, personalized and ubiquitous way. This new paradigm is based on a learner centred approach, to replace the classical, content centred approach to learning. The learner plays an active and central role in the learning process. Rather than stressing the memorization of information, learning activities are aimed at aiding the learner in the construction of an autonomous, functional base of knowledge and skills. In keeping the learner at the centre of the learning process, personalisation/individualisation (creating and adapting learning paths according to
learner’s previous knowledge, preferences, skills, preferred learning style, and collaboration (with other students, teachers, tutors, or experts) become relevant aspects to be supported by technologies through the creation of the appropriate context. Considering humans at the centre, learning is clearly a social, constructive phenomenon. It occurs as a side effect of realistic simulations, interactions, conversations, collaborations and enhanced presence in dynamic Virtual Communities.

Project DILIGENT (Digital Library Infrastructure on Grid Enable Technology, http://www.diligentproject.org/) [5] is an integrated project funded in part by the European Commission FP6 IST Programme. DILIGENT is aimed at the creation of virtual digital libraries on the basis of grid-based infrastructure so that the integration of metadata, personalization services, semantic annotation, and on-demand availability of information collection and extraction to be supported. Such new decentralized and service-oriented architecture for digital library assures a better and adaptive tailoring of the content and service offer to the needs of the relevant community as well as to the current service and content offer, and a more systematic exploration of existing resources.

![Grid-based digital library infrastructure](image)

Figure 3: Grid-based digital library infrastructure

6. CONCLUSIONS

Important requirement for each modern LMS is the learning adaptation to be assured for each learner in respect to her/his necessities, preferences, needs, performance, and progress. The achievement of interoperability and content reusability in the existing diversity of software and hardware platforms is a real challenge. One big limitation of the web-based interaction is the smaller communication bandwidth than traditional face-to-face interaction. The term bandwidth represents the amount of information that can be transferred in a unit of time through any means possible. In the face-to-face communication mode, if a verbal instruction is not understood, the clue can be available to the counterpart through gestures, group dynamics and other such means, but the clues in the web-based mode are not always so clear and in many cases not available at all. Therefore, tailoring the information to the right-level for the receiver to understand and integration of different appropriate methods for learning adaptation are crucial factors for the success of any LMS.
REFERENCES


12. Zheleva M. (2005), Methods, models and algorithms for development of adaptive learning content in interactive educational environments, PhD thesis.