Toward an Adaptive Educational Hypermedia System (AEHS-JS) based on the Overlay Modeling and Felder and Silverman’s Learning Styles Model for Job Seekers

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Abstract—In this paper we introduce a new design of an adaptive educational hypermedia system for job seekers, this proposal is based, for the part of learning objectives, on a job model which allows adapting the content and the path of education to the intended jobs, and, for the learner model construction, on a specific use of the learning styles of Felder and Silverman.

First, we present existing literature to give a general review on adaptive educational hypermedia systems, in that way; we have reported the related items to different notions in the adaptive educational Systems area as the differentiated pedagogy, the learning objects, and the learner profile. Then we argued our choice of the components of our model and we detailed the new ones.

As designed, the model can produce a suitable learning path for the user to match the job characteristics and the learning style of the person in order to help the user owning the job sought. With the possibility of linking the required competencies to the education skills, we aim to map business tasks to learning activities.

Based on this approach, we designed an Adaptive Educational Hypermedia System named AEHS-JS that will help to improve the efficiency and pragmatism of job search activities.

In plus of the social impact of this work as it help job seekers to complete their profiles and get the career they are looking for, this work will allow companies to find the candidates that match the job criteria sought.

Keywords—Adaptive educational hypermedia systems; learning object, learning styles, learning path, Overlay Modeling, differentiated instruction, Job model, Employability.

1 Introduction

Online learning is increasingly used in universities and schools; in this fast-growing world, most e-learning systems do not yet make the business requirements as
their priority. To meet this need Research has shifted to adaptive learning, which represents a domain of research in education in continuous development.

Traditional learning systems treats all users the same way, they deliver them the same content. When it comes to different people this way to do has shown its limits in terms of efficiency and effectiveness. The slogan “all things to all people” do not work.

There are several forms of adaptation, namely: adaptation based on user modeling, or those that use learning objectives to provide the learner with an education that fit with his profile [1]. So-called personalized Educational Hypermedia Systems are systems that use a methodology based on the objectives, preferences and knowledge of each, while adapting to their necessities [2]. In this way, a student in an adaptive educational model will be attribute a content that suit to his expertise on the theme.

Making personalized learning a big step towards improving online learning outcomes, but it is not enough because it does not take into account the needs of organizations and the learner's requirements. The purpose of learning is the knowledge of a learner who will contribute to the success of the learning process.

By working on this area, we are looking to suggest a new model that meets the business requirements to the learner part. The main goal is to set up an educational framework built on the work prerequisites and the user; this system will allow you to connect to the market.

We will begin the second section with a listing of related works. Then we will introduce the uses cases and the screenplay of the learning activity, thereafter in section four we will go through our model of Adaptive Educational Hypermedia System, we will zoom on the job model. To finish we will discuss the results, the conclusions, and draw some perspectives.

2 Literature Review

2.1 The adaptive learning systems

Recently, several studies have been interested in adaptive learning systems that provide the suitable material in the appropriate way to the student by analyzing its profile (Figure 1), these methodologies may be classified into 3 categories:

The preferences category is related to models that define the education experience based on the learner preferences as the personality style. In this sense, we find the works of DeCoux [3], Maseleno [4], PRITCHARD [5], Savage [6] and TRUONG & Huong [7].

These models uses and analyze the student behavior [8-9], observes it comport-ment [11-27], gives the user an online or mobile survey [12], [28], [14] and let him precise his choices and individual characters [15-16].

In our paper [10] we introduced a model based on the Felder and Silverman’s learner styles to define the learner model and give the student a learning experience that meet the job needs.
The second class, which is based in the user background and his knowledge, is considered as the most important feature for Adaptive Educational Hypermedia System [17].

This aspect can be represented in many ways. The first one is a rated System where we measure the degree of the learner knowledge in a domain by a rate (Exp.: a value between 0 and 5) or a grade (Exp: High, normal, lower), this method is simple to implement but it remains insufficient for some innovative and adaptive techniques.

Besides this model, Overlay illustrate the learner data [18], this model allocates a numerical number, 0 or 1, to every portion, showing the learner knowledge or not of an object.

The last category focus on the learner aims to select the suitable education experience for every single user.

![Profile](image)

Fig. 1. The learner dimensions

Today, the challenge is not to bring people the knowledge and help them to fully learn it, the principal aim of education systems is to help learners to get the knowledge they need to get into the enterprise and and being able to satisfy the job requirement.

All of these systems make the learner model at the center of their models to establish the user learning experience, this make possible to adjust the path of education to the learner but does not allow thinking through the requirement of the business or the work opportunity we are targeting.

2.2 The learning style model of silver man

Several models in the literature have treated The Learning Style ([3] DeCoux [3], Maseleno [4], Pritchard [5] and Savage [6]).

In our case, we used the learning styles of Felder and Silverman and adapt it to our new model in order to have a system that bring the whole adaptation power. Indeed, we are leaning towards this choice for its easiness and its weight in the education researches [7] The model of Felder & Silverman Learning Style is a questionnaire of 44-item that identifies the learner's learning style. In order to spare the heaviness of the survey in an education system intended to be accessible online, we have restricted the list of questions to 20 relevant ones that represent the main questions [23].
2.3 The differentiation in pedagogy

In her research, Carol Ann explained the importance of differentiation in pedagogy, she based her conclusions on the fact that we cannot treat a group of students in the same way. The objective is to engage the learner in a challenging learning activities that enhance the the learning experience and build a system that take the student to higher level of compliance, This can be done over four ways: through content, the style of learning, product/process, and the background [32], [24], in this work we focused on differencing the learning style and content, at the end, the result and environment are already known in an adaptive learning system.

The content: Adopting a strategy of differentiation at the content level consists on defining the learning objects list and organizing them in a number of ranks of knowledge and intellectual performance starting from lower-level intellectual expertise to higher-level. In our model we chosen the Taxonomy of Bloom (a grouping of 6 ranks comportment of intellectual starting by bottom-level intellectual abilities to higher-level). [25]

Therefore, we can build six different forms from one learning object; those new LO versions will be attributed to the student according to his score. By way of illustration, a student who has no experience in a domain will be given LOs with an inferior degree: memorizing and understanding, and a different one with some experience will receive learning objects for a higher level as applying and analyzing.

Learning styles: Every student has its favorite learning style, an important way of differencing is based on the provided support of education: it can be visual, auditory or written. Our choice in the model we want to apply is based on The Felder and Silverman's Learning Styles Model (see section III)

3 Research Methodology

In order to validate our model and reach our goals we based our research methodology on following instruments:

- Interviews with the human resources responsible for the job offers: this part of work will allow us to compare the results of different applicants and evaluate the value added of the learning experience on the process of selecting and recruiting candidates. It will bring us the result of each candidate and allow us to calculate the success rate. As the positions opened were only two in numbers, this part will bring us more information about each candidate independently of his results.
- A survey for the job seeker who have benefited from the learning experience, to collect quantitative data and their feedback. The survey is made up of 3 parts, the first one contain general instructions, the second specify the personal information (Gender, Age, years of experience…etc.) , the third part is the body of the questionnaire where we gather candidate's opinion on the experience and the level of usefulness of the skills acquired by questions like: "what was the content that help you the most in the interview?" or a collect of opinion like: "The following items describe statements about the learning experience, please indicate your degree of..."
agreement or disagreement about them: 1- The skills learned in the training helped me to convince the recruiter about my profile, 2-…etc.” with 6 choices (Strongly disagree, Disagree, Disagree somewhat, Agree somewhat, Agree, and Strongly Agree)

The participants were chosen from a list of 89 resumes received by the human resources responsible for the job offer. This task was done using a matching engine [35] to select resumes that match the most to the job need. The objective is to keep only the profiles that do not represent a big difference with the need.

In this study, we have limited our scope to a single job offer "Banking Business Analyst", the same process can be applied iteratively to a list of job offers.

A group of 23 candidates was selected to take the interviews with the job offer responsible. From this group, we have contacted some candidates to benefit from the learning experience based on our designed model.

In order to compare the result of the experience to a reference, we divided our group of 23 preselected candidate into two subgroups, the first one with 13 candidate to take the learning experience and the second with 10 candidates who will follow the normal interview process without any training experience.

In our study, the learning experience consists of the following main steps:

- **Step I:** The Job seeker authenticate in the system and fill in his account profile (individual data, competencies, CV...)
- **Step II:** The Job seeker response to a survey that determine his learner style
- **Step III:** He select the job he is looking for from the opened positions. (the model works for different job offers, but in this case we will focus on one job offer)
- **Step IV:** The system build a learning path that take into consideration both the job selected and the Job Seeker analyzed profile.

This learning path will lets the Job Seeker expand the competencies and befitted as the perfect applicant for the associated job.

Regarding the inductive part of our research, we established a new model of an adaptive learning system (cf. next section for more detail) that will be applied to the selected candidates.

### 4 An Adaptation System For Job Seekers

#### 4.1 Business and uses cases

Before presenting the adaptation system architecture, we will start with a learning scenario, which represent the theoretical basis for the design of our system architecture; this scenario may take into consideration the job model, which is the principal component of our SHEA.

A job seeker who logs on to the system for the first time must complete a registration form including his resume and answer a survey that help us to determine his learning style and competencies.
Thereafter, the job seeker selects one or more job offers that interest him / her, at this step our model is completed by the learning goals of the candidate, which is an important information for our system.

The learning experience is done through the pursuit of the learning path generated. Accordingly, it is possible to clearly distinguish three essential phases in the process of learning for a job:

- Registration form including the upload of resume
- Survey to determine the learning style
- The choice of a list of interesting job by the job seeker
- The system build content and learning path (matching the learner profile, the job wanted and the learning elements)

4.2 The adaptation model

The ALS (Adaptive Learning System) we designed is built with the following models:

- The Job Model
- The Domain model
- The Learner Model
- The Instruction Model
- and the Adaption Model (Fig. 2).
This global model is detailed by its components as below:

**The Domain Model:** This model is made of a universe of LOs (learning objects) based on the SCORM Principles, various forms of an identical LOs and the metadata content as established by [21]

![Learning Objects Diagram](image)

**Fig. 3.** Figure 1: Learning objects

As described by Robert BECK [28], the LO is “a collection of content items, practice items, and assessment items that are combined based on a single learning objective”, in our case the learning objective is the job skills need.

We choose the learning objects for our domain model to prevent the “reinventing the wheel”, it’s a way to save time by reusing and sharing content to focus on specialized tasks and learning objects.

Our choice of SCORM standard will make the object reused in any learning management system or browser and easily manageable. In this standard, the learning object must comply with the Metadata of Learning objects LOM IEEE. The figure 1 in [34] describe the UML Class Diagram of IEEE LOM Metadata Standard.

In order to give an example of our learning objects, first we must specify our educational problem, in our case it is related to job we are looking for. Secondly, we will define the metadata to make our LO searchable in several search engines. Finally, we have to specify the requirement needed to be able to pursue the related LO.

Once the Learning Object’s defined, our system have to choose between all indexed LOs in our Domain Model to propose to the learner the most suitable LO to his profile. Therefore we must build our Learner model that represent the profile of the learner.

**The Learner Model:** The learner model is set to define the Job seeker description, the learner model, as Brusilovsky [20] presented it, take 2 proportions, the major concern the data, and the other is based on the style of learner.
In terms of data line, we used the model of overlay for the ease of setup and for the reason that it is a model that use a numerical value to qualify the degree of knowledge of an item. The principal notion of overlay modeling consist on considering the learner model a subgroup of domain model, and for all items in the model of learner (related to every item in the domain model) has a particular rate that quantify the learner’s knowledge of this item. This rate is named the mastery of the item.

For each learner we define the level of mastery for a unit of course. It’s can progress with the experience of learning and be evaluated based on the quizzes established. As shown in the table below: learners who obtain a result percentage less than 50 are classed as “Novice” and the level of mastery $M=1$. The other class $M=2$ is intermediate with percentage between 50 and 70. The advanced one $M=3$ is reserved for learners who obtain a percentage greater than 70. Only learners with a mastery of $M=3$ degree can move to next level of topic.

The information about the learner model is warehoused in the database and will be used by the adaptation model to generate the learning path adequate to the learner.

| Level     | Score             | Mastery |
|-----------|-------------------|---------|
| Novice    | $M(X) < 50$       | $M=1$   |
| Intermediate | $50 \leq M(X) < 70$ | $M=2$   |
| Advanced  | $70 \leq M(X)$   | $M=3$   |

The learner relational schema can be present as follow:
In practice, the proficiency of knowledge will be evaluated regarding the CV analysis of the student.

For the learning style level, as we chosen the FSLS (Felder-Silverman learning style) model [22] (cf. sect. III) our relational model will contain the below classes,
where the feeding of the learning style class linked to a learned is done by a resume data treatment and the result of the Felder and Silverman’s Learning Styles test.

![Learning styles tables](http://www.i-jet.org)

**Job Model:** In this part we will define the job needs based on the list of skills and requirements that the person looking for this job must satisfy to get the job done.

This model organize and describe the business needs (in terms of jobs and offers). It connects between the components of business need (The Enterprise characteristics, the offer description, the looked-for profile, the wanted competencies, a competencies assessment...)
The advantage of using an identical system to characterize the knowledge notion in job and learner models will help for the correspondence processing in the adaptation model. In the same way established in learner model, we use the overlay model to characterize the required skills by a Job offer. At the level of the job model we define for every unit of course the level of mastery needed. It’s can be liked to a job description to categorize a group of job offers (cf. table 2 below).

The information about the Job model is stored in the database and will be used by the adaptation model to generate the learning path adequate to the job wanted.

| Unit of course | Job description | Mastery required |
|----------------|-----------------|------------------|
| Unit 1         | Project Manager | M=3              |
| Unit 2         | Software Engineer | M=1             |
| Unit 3         | English Professor | M=2            |

The job offer relational schema can be present as follow:

**Instructional model:** The model of instruction define the education strategy adopted to allocate the LOs existing and organized in the model of domain. In this global model we chosen the differentiated pedagogy for the possibilities its offer in terms of adjusting of the instruction at the different levels as the content, the process or the learning materiel (Cf. Sect. IV)

**Adaptation model:** The adaptation model represent our engine that to define the list of LOs based on the analyze of learner profile (Mastery, LS) and the details of the chosen employment (necessary competencies...) , the main work consist on assigning LOs that tie with the required skill by the business (job offer) and the learner characteristic. (Cf. Fig. 2).
In order to determine the list of learning objects and to define the learning path for the learner, we eliminate the LOs list assimilated by the candidate from the required by the job offer. As the LOs are designed into the same model (domain model) this subtraction can be modeled in its standard form under the following mathematical formula:

\[ \sum_{k=0}^{n} \text{Final}(LOs) = \sum_{k=0}^{n} \text{Required}(LOs) - \sum_{k=0}^{n} \text{Acquired}(LOs) \]

Where,
- \( \text{LOs} \): The Learning Objects list. This list must contain the metadata related to LS of the candidate.
- \( \text{Final (LOs)} \): The final list that will be used in the learning experience.
- \( \text{Required (LOs)} \): The LOs list necessary in the job offer
- \( \text{Acquired (LOs)} \): The Learning objects list already learned by the Student (in our case: the job seeker).

Fig. 10. The Adaptation relationnel schema
Algorithms principal steps:

BEGIN

Step 1: Filling the learner data from CV and skills tests: $M(X) = \sum ML_i$ where $i=$ competency acquired by the learner and $ML_i =$ level of mastery of the competency
Skills $M(X) = \sum$Skills/mastery level $= \sum ML_i$
Step 2: take survey to define the learning style
Learning style $LS(X) = LS_i$
Step 3: filling the job model data based on the job description chosen
Job description $= \sum ML_j$ where $j =$ competency required by the job
Step 4: generating the LP(learning path) established on the model of job and selecting the learning objects from the domain model
Initiate the unit number $u = 1$
Initiate content level $l = 1$
Initiate learner level $= ML_i$ (based on the model of learner)
Step 5: take the learning experience (courses and assessments of unit $u$)
Present the unit $u$
If the skill test related to the unit $u$ $M(Xu) > ML_u$
go to next unit $u = u+1$
Else present unit $u$ with level $= l+1$
Go to step 5
Step 6: update the learner model
Step 7: Evaluate if the learner model skills $\sum ML_i$ include the job model skills required $\sum ML_j$

$\sum ML_i \subseteq \sum ML_j$: present the CV of the job seeker to the company
Else: Go to step 4
END

5 Results

Survey Results: To validate our model of learning designed for job seekers, we explored the results of the survey taken by the candidates.

Regarding the impact of the training on the interview process, we received the response detailed in the table 3

| The skills learned in the training helped you to convince the recruiter about your profile? |  |
|---|---|
| Strongly disagree | 7% |
| Disagree | 13% |
| Disagree somewhat | 5% |
| Agree somewhat | 7% |
| Agree | 40% |
| Strongly agree | 28% |

Furthermore, 69% of candidates who agreed or strongly agreed on the usefulness of skills learned in the training have a number of years of experience less than 4 years.

The table 4 shows the results related to the most useful skill acquired for the interview process.

| The most useful Skill learned by the job seeker for the interview |  |
|---|---|
| Banking functional | 43% |
| Project management fundamentals | 28% |
| Business Analyst CCBA | 17% |
| Communication | 8% |
| Others | 4% |

It was found that candidates who qualified themselves as adapted to the position were more satisfied about the experience and had less difficulty reported in the interview process.

The result shows us that more than 70% of candidates were in favor of repeating the experience or recommending the learning to other people.

Interviews Results: Through the 23 candidates preselected for the job interview process, two candidates were admitted to the position; both of them had completed the learning experience.
Furthermore, the interview results show that appreciations of eight candidates qualified them to be accepted if there were more opened positions. Six of these candidates were in the list who completed the learning experience.

The 13 candidates who passed the learning experience were better rated in all major required skills except “communication” where the rest was slightly better appreciated; these candidates demonstrated greater enthusiasm for the position than others were.

The job offers responsible mentioned that some candidates declared that they had followed a training for the interview and he appreciated this.

Table 5. Interviewers’ statements revealed some remarkable feedbacks on candidates (table 5).

| Candidate | Statement on the candidate |
|-----------|---------------------------|
| M. B      | He is certified in project management, he has good communication skills but it is clear that he is not motivated by the position. |
| G. L      | He has demonstrated knowledge of the field, he is certainly not an expert but he will be able to satisfy the job needs |
| M. A      | The candidate M.A is very motivated for the position, he has some communication difficulties, he has a lot of potential and masters project management and business analysis approach |
| Y. B      | The profile is interesting, he can support us in the training of new recruits, but he shows that he has never practiced project management in operating conditions; it will take him a little time to master the project approach. |

Generally, results show that the interviewers were more interested by the profiles who did the training. In deed, they qualified them like more “interesting”, “fitting to the job”, “motivated”, and “initiated in the domain”.

6 Discussions

The main goal of this work was to set up a system that help job seekers to boost their chances to get the position they are looking for.

Based on the results, the adaptive learning system designed for the job seekers, improved the employability related to the job offer in question by 56%, indeed, The effectiveness of the system was also confirmed by the candidates feedbacks who reported the utility of the education content as it helped them to gain greater profile and showcase their competencies.

These findings confirm the works of Ng and Feldman, [36] and Krueger & Rouse [37], on how positively education is influencing core task performance and employability.

Seeing that the majority of candidates who were satisfied by the experience are junior is interesting, this can be explained by the fact that this category of candidates is still in touch with education, and regarding their experience, they are the ones who needs the most such as learning experience.

It is normal that the impact of the system was more visible on some skills than others, indeed, some skills like communication and relationship management needs more
time and practice to see significant results, the system shows more efficiency on knowledge skills like functional banking or management. This finding shows the limitation of our model on some particular skills, but it will not impede us to include these type of elements in our content, because for some candidate who have a minimum of level on this type of skill, the result can be more interesting. These results support the studies of [36] where they classified the knowledge into two form, the declarative knowledge that can be learned based on suitable content and the procedural one that need more practice.

The system is more efficient for persons who are initiated in the fields, this can be explained by the reason that qualified people are more able to understand concept that are related to what they master.

The results of the interviews with the persons in charge of the recruitment brought out that they were aware that some candidates have followed a training, and they valued it positively. This statement drives us to explore the opportunity of applying this system within the companies during the integration of new recruits.

To sum up, this work showed us the importance of education in improving the employability of job seekers, and how can a job-oriented learning lead us to better results, these comply with different studies carried out in this field [38, 39, 40].

7 Limitations

As with all studies, there are limitations. In our case, we applied our model on one job offer and collect data related to the job seekers looking for this job, this can be done for other jobs and different populations to make our work more global and confirm the generalizability of the finding.

8 Conclusion and Future Works

Online learning provide people a new way to improve their skills in different area. In our case we focused on it use to enhance job seekers chances to satisfy the jobs requirements. Study have revealed the role of education in employment and job performance. In this work, we designed an Adaptive Educational Hypermedia System for Job Seekers “AEHS-JS” to evaluate the utility of job oriented education on increasing the employability and job performance.

We used for this work the Overlay Modeling and Felder and Silverman’s Learning Styles Model for Job Seekers. Our model is based on two principals element, the first one is the model of learner and the second one is the job model, the main goal is to establish a learning experience able to make the student fits more to the profession he is looking for. By this approach, we are ensuring the link between learning systems and business requirements. The challenge is to build a system based on this model that works for businesses and job seekers both.

Like many studies, the results of our work confirm the importance of education in improving employment opportunities, a job oriented education can in enhance chances of succeeding for a job seeker by more than a half (in a similar context of the one
used in our test case). Moreover, even job seekers who did not succeed the opened position were interested in the experience for other job offers in future.

The same model can be used by companies to build a training program that help employees evolve from one position to another by pursuing an individual training plan. The next step will consist on working on a probabilistic approach of adaptation based on Bayesian Network with a new step added to the learning path “revaluation of the learning path”. This will be useful for the implementation and will allow trying out another adaptation algorithm and more data collection on different populations.

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