Is a Multi-Criteria Evaluation Tool Reserved for Experts?

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Abstract:
The objective of this investigation was to determine whether the analytical hierarchy process algorithm is suitable for the evaluation of software by evaluators with little Information Technology experience. The scope of the research was the evaluation of two free and open source e-learning systems at the Open University of Tanzania using 33 stakeholders with diverse levels of Information Technology experience. Both quantitative and qualitative research methods were used. The qualitative methods comprised participative observation and interviews. Questionnaires and the analytical hierarchy process, a multiple-criteria decision-making algorithm, represented the quantitative methods. The results showed that of the two e-learning systems evaluated, Moodle was preferred over ATutor. Furthermore it was found that the analytical hierarchy process algorithm is appropriate for the evaluation of software in a situation where Information Technology experience is limited. It is anticipated that the paper contributes to the theory and practice of decision making in developing countries such as Tanzania.

Keywords: free and open source software, e-learning systems, software quality, multi-criteria evaluation tool, analytical hierarchy process, novice user, developing country

1. Introduction

Finding suitable software for a specific purpose is a universal problem. Available software needs to be considered and some method found to determine which application is more suitable for the purpose of the application and the needs of the user. Evaluating available software for a specific purpose is in most cases left to software experts and the views of stakeholders are used only if the experts deem it necessary.

In the commercial world, software packages are developed for specific usages and the opinion of stakeholders’ are valued and incorporated throughout the software development cycle. This is not the case with free and open source software (FOSS). FOSS is often developed by software specialists interested in a specific application (Nichols and Twidale, 2003). They develop software for this application and then make it freely available for everyone who is interested in adapting it or changing it for their own needs. Experts are able to adapt FOSS for their needs, however it is difficult for end-users to evaluate, choose an appropriate software application (from the many that are freely available on the Internet) and adapt it for their specific needs.

In many organisations the selection of FOSS is done by trial and error: Software packages are introduced and if they do not “live up to expectations” another package is found and introduced until most of the needs of the users are satisfied. However there are methods for the evaluation of software, for example the qualitative weight and sum (QWS) method, the analytical hierarchy process algorithm (AHP) etc. These methods, each with its pros and cons, are employed by software specialists to evaluate software.

The Open University of Tanzania (OUT) is a distance learning institution that is in the process of considering using e-learning system to distribute their study material and manage their students. Since this is a new approach, OUT is still in the process of appointing more Information Technology (IT) personnel to implement and maintain an e-learning system. Currently IT experience at OUT is limited; therefore if a method could be found where any stakeholder could participate in the evaluation of the software it would benefit this environment. It is from this stance that we evaluated two e-learning systems at OUT using the analytical hierarchy process algorithm. The reason for choosing AHP is because it allows the incorporation of both subjective judgments and objective decision-making. For the subjective evaluation we made use of 33 evaluators: 30 students (so-called stakeholders) and three IT experts or staff. These were the only stakeholders that had experience with both the e-learning packages that were being considered.
Our assumption is that AHP is a sophisticated tool which is generally used by technical users when evaluating software. The question that needs to be asked is: Can any user of a computer system, deployed in a developing country, be asked to participate in the evaluation of FOSS with AHP?

Traditional methods of software evaluation involve expert judgments (Boehm, 1981a). Usually an expert evaluates software independently; the results from different experts are then merged to find the most appropriate software for a specific purpose. This approach is subjective and can thus be biased according to Somerville (2004). A more objective approach was suggested by Basili, Caldeira, and Rombach (1994) who used the Goal Question Metric (GQM) paradigm to evaluate software. They found that the advantage of GQM is that it is very flexible. They however indicated that a disadvantage of GQM is that it can only be used by experienced users who can define and answer questions relevant to the purpose of the evaluation (which differs depending on the environments). A further disadvantage is that GQM neither allows for the comparison of software systems nor does it provide a means of combining different attributes into a single value for the sake of comparison.

To evaluate FOSS e-learning systems, Graf and List applied a qualitative weight and sum (QWS) method (Graf and List, 2005:164). QWS has the advantage of being simple and novice users can be used in the evaluation. However, its disadvantage is that it is subjective and it is complex to compare the evaluation of the different software packages. This is because symbols are used (rather than numerical values) when assigning software quality attributes.

Saaty (1980:78) showed that the AHP model can be used to solve any multi-criteria decision making problem. The advantage of using AHP is that it is simple and caters for a step-wise evaluation of software which makes it suitable for evaluating FOSS.

In developing countries the success of FOSS depends on the following factors:

- user acceptance (keeping in mind that users are from varied cultural and educational backgrounds with differing computer expertise).

- resources (since resources are scarce) and

- the quality of the software

These factors (mentioned above) impact on the evaluation of e-learning systems according to Wesson (2003:52). Other factors that may influence users’ impression of the software may be the quality of the content/course materials available for the e-learning system being evaluated.

Researchers have categorized the quality of software as either technological or pedagogical (Collace et al., 2003). The focus of this paper is not on the pedagogical, but rather on the technical factors, that determine software quality. But what determines the quality of software? The quality of software is determined by: (i) the quality of the development process; (ii) the quality of the product as well as; (iii) the quality of the product-in-use (Fenton and Pfleeger, 1997:100). Manieri, Begin, Meglio, and Rippa (2008) argue that most FOSS applications are poorly documented in terms of requirement specification, system architecture use and system prototypes tested. Consequently it is difficult to determine the quality of the FOSS development process and thus the best way to evaluate FOSS software, is to consider the software product-in-use.

Nichols and Twidale (2003) argue that the rationale for involving novice users in the evaluation of software is that software developers of FOSS do not necessarily design and develop software with features that take the typical user into account. It is therefore important to involve the ordinary user in the evaluation process. Technical and non-technical users perceive software differently and thus we advocate the participation of all stakeholders to bridge the gap between the novice and the experienced user.

OUT uses traditional distance learning methods for teaching and learning (McHarazo and Olden, 2004:205). Currently there is an initiative to improve learning and teaching at OUT through the implementation of e-learning. Many FOSS e-learning systems are available on the Internet. From these, OUT chose ATutor and Moodle as the e-learning systems to test in a pilot project. The idea is to select one of these two for full implementation in the near future. This will enable OUT to have access to an inexpensive but effective e-learning system to support learning and teaching. However the cost of a FOSS e-learning system lies not only in the creation of content for the system, but also in the customization and the maintenance of the system.
E-learning is becoming an attractive alternative to traditional distance learning in Tanzania because of the recent growth of Internet usage (and coverage) (Isamuyo, 2006:10). Students scattered throughout the country will be able to access e-learning systems at work, Internet cafés, community centres, telecentres or at home. The use of blended learning, traditional learning and teaching complemented by e-learning, is thus an appropriate option for OUT. Furthermore the university as well as the government has enacted information communication and technology (ICT) policy to guide the incorporation of ICTs into teaching and learning, creating opportunities for e-learning.

We argue that, in order for the uptake and implementation of FOSS to be successful in Tanzania or another developing country, certain criteria of software quality must be considered. In our case study, we considered three software quality criteria, namely: usability, maintainability, and deployability, which we consider to be the most important criteria to consider in a developing country. These criteria depend on the environment and context where the system is implemented. They are not universal in all developing countries due to the difference in availability of resources, experience of users to maintain the systems and the environment where the software systems must be deployed.

The quality of software depends on the user’s ability to interact with the software; the knowledge to maintain the software and the resources available to deploy the system in particular environment. These three criteria are quite challenging in Tanzania because of: (i) the computer illiteracy of the population; (ii) the low bandwidth of networks and; (iii) the fact that some user interfaces will be in a language which is foreign to the Swahili user (Swahili is the language spoken in Tanzania).

Both qualitative and quantitative methods were used in this study. The qualitative method comprised collecting data on site using participative observation and focused group interviews. Quantitative methods consisted of data collection using questionnaires and data analysis using AHP.

AHP, a multi-criteria decision-making (MCDM) algorithm, uses pairwise comparisons to derive weights (of importance) for the attributes identified (Saaty, 1980). AHP was used to evaluate two FOSS e-learning systems: Moodle and ATutor.

The results showed that the evaluation was consistent for the attributes that were identified for the criteria: deployability; maintainability and usability. It confirmed that ordinary users can be employed in the evaluation process and indicated that the e-learning package Moodle is preferred over ATutor.

2. Method

The basic steps for software product evaluation, as identified by Comella-Dorda et al. (2002:87), were followed. They involve planning the evaluation process; identifying criteria for evaluation; collecting data; and finally analyzing the data.

2.1 Planning the evaluation process

The participants were selected by means of purposive sampling (Van Vuuren and Maree, 1999:281). Purposive sampling was used because there were only a limited number of users, with the necessary experience, who could participate in the study. Thirty three participants from OUT (30 second year BSc. Information Communication Technology students and three IT lecturers) with experience in both ATutor and Moodle were selected to participate in the study. They were given a consent form to complete before taking part in the research project.

2.2 Identification of attributes that were used in the data collection

For the three main criteria (usability, maintainability, and deploy-ability), sub-criteria were identified as proposed in the literature (Boehm et al., 1976b:595; Cavano and McCall, 1978:136; IEEE, 1998; ISO, 2001). For each of these sub-criteria, the authors and participants decided on attributes that best describe the selected sub-criteria. These attributes were combined with attributes identified in the literature (Coleman et al., 1994; ISO, 2001; Bertoa and Vallecillo, 2002:65).

<table>
<thead>
<tr>
<th>LEVEL 1 criteria</th>
<th>LEVEL 2 sub criteria</th>
<th>LEVEL 3 attributes</th>
</tr>
</thead>
</table>

Table 1: The hierarchical structure of the quality attributes for the proposed evaluation
The identified criteria, sub-criteria, and attributes were combined and used in the formulation of a questionnaire (some of the questions were adopted from a Software Usability Measurement Inventory (SUMI) questionnaire (Ryu, 2005: 192)).

Attributes which the authors felt were duplicated, were removed from the questionnaire. The aim was to come up with a list of representative attributes agreed to by all the participants (see Table 1).

### 2.3 Data collection

The questionnaire was refined by means of a pilot study. In the actual study participants were asked to complete the questionnaire and some of the participants were interviewed using open ended questions or probes. In addition, the authors took field notes (i.e. participative observation) during the process.
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data collection process. This provided rich data which informed the research process. Focus groups helped to obtain a wish-list of weights for the criteria and sub-criteria (weights allocated to level 1 and level 2 of Table 1).

It was decided to use output from the questionnaire as input to the AHP method. This was done instead of allowing the users to directly use AHP (Turban, 1993:221). Some researchers might argue that this is not ideal but we did this to facilitate non technical users’ involvement in the evaluation process.

2.4 Data analysis

All the questions in the questionnaire required categorical responses. Numerical values were arbitrarily allocated to these categorical responses (Stevens, 1946:679) (see Table 2). Mapping software quality phenomena to a numerical value is accepted in measurement theory and software measurement (Fenton and Pfleeger, 1997).

According to Turban (1993:221) defining intensities (measures) the way we did in this research, is the best approach to avoid comparison ambiguity associated with large number of criteria and alternatives.

Table 2: An example of the interpretation of the attribute symbols

<table>
<thead>
<tr>
<th>Symbol</th>
<th>A</th>
<th>B</th>
<th>C</th>
<th>D</th>
<th>E</th>
</tr>
</thead>
<tbody>
<tr>
<td>Nominal</td>
<td>Strongly agree</td>
<td>Agree</td>
<td>Disagree</td>
<td>Strongly disagree</td>
<td>I don't have enough information to answer</td>
</tr>
<tr>
<td>Scale</td>
<td>9</td>
<td>7</td>
<td>5</td>
<td>3</td>
<td>1</td>
</tr>
</tbody>
</table>

There are two methods for computing the combined group decisions in AHP: (i) first method by using either the arithmetic mean or geometric mean of individual respondents’ comparison judgments (i.e. opinions) and (ii) second method by using either the arithmetic mean or geometric mean of the individually calculated priorities (Forman and Peniwati, 1998). The first method is used if the group of the respondents wants to act as a unit while the second method is used if the group wants to act as a combination of individuals. In our case study we used the first method.

2.4.1 Detailed explanation of data analysis

The collected data was analyzed empirically using AHP (Saaty, 1980) since AHP simplifies the quantification of software quality attributes and allows for a more objective evaluation process. In AHP a problem is decomposed into a number of hierarchical levels (similar to Table 1), these are then compared pairwise and prioritization is determined.

The pairwise comparison of elements in each level was done using Saaty’s comparison technique (Appendix A and Appendix C show the technique used to obtain a pairwise comparison matrix from the numerical values of criteria, sub-criteria and attributes). After computing the pairwise comparison matrices for each of the levels of the hierarchy in Table 1, normalization matrices (Saaty, 1980) were calculated. From the normalization matrix, a priority vector was calculated. The priority vector (eigenvector) determined the relative ranking of the alternatives for each criterion. To check for consistency the Saaty’s procedure was followed (see Appendix B). Checking for consistency of measurement is the basis of objective measurement for the subjective evaluations (Moses and Farrow, 2008:276).

3. Results

The AHP algorithm involves linear modelling during the analysis of data. The advantage of this algorithm is that it does not need many historical datasets (Cavano and McCall, 1978:137).

The linear equation used is as follows:

$$\begin{bmatrix} a_{c_1} & b_{c_1} \\ a_{c_2} & b_{c_2} \\ a_{c_3} & b_{c_3} \end{bmatrix} = \begin{bmatrix} P_a \\ P_b \end{bmatrix}$$  (1)
where: vectors $a$ and $b$ are alternatives with respect to a specific objective, and $c_1$, $c_2$ and $c_3$ are the weights assigned to criterion 1, criterion 2 and criterion 3 respectively, and $(a_{1c}, b_{1c}), (a_{2c}, b_{2c}), (a_{3c}, b_{3c})$, are the priority of the alternatives $a$ and $b$ with respect to the three criteria. In our case ATutor and Moodle were represented by $a$ and $b$; and $c_1$, $c_2$ and $c_3$ were used to depict the criteria: usability, deploy-ability, and maintainability.

The result for equation (1) gives the total priority for each e-learning system (see Appendix D for the detailed calculation). Table 3 shows the ranked results. Moodle ranks above ATutor for each of the scales used.

### Table 3: Ranking

<table>
<thead>
<tr>
<th></th>
<th>Total Priority</th>
</tr>
</thead>
<tbody>
<tr>
<td>ATutor</td>
<td>0.468554988</td>
</tr>
<tr>
<td>Moodle</td>
<td>0.531445012</td>
</tr>
</tbody>
</table>

Finally the consistency indices for each pairwise comparison were computed and were equal to approximately zero (see Appendix B and Appendix C which means that all the pairwise comparison judgments were consistent (Saaty, 1980). It concurs with Moses and Farrow (2008:286) who argue that the best approach to validate results for subjective judgments is to check consistency.

### 4. Discussion and conclusion

The objective of this study was to determine the suitability of the analytical hierarchy process algorithm in the evaluation of FOSS when the evaluators are mostly users with little technical IT experience. In the evaluation, qualitative methods (using participative observation and focus group interviews) were complimented by quantitative methods (questionnaires and AHP).

We have shown that AHP is suitable for evaluating software in a developing country where Information Technology experience is limited because of its simplicity, accuracy and flexibility in making a logical, consistent and informed decision.

Since the AHP algorithm involves linear modelling, contributing metrics values can in certain cases create a non-compensatory effect. This means that a system under evaluation might be ranked higher than another, even though one or more of its constituent criteria, sub-criteria, and attributes have lower weights than the system it is compared to. The use of the analytical network process algorithm (ANP) may remedy this effect as it takes the dependence of criteria, sub-criteria, and attributes at the same level of the hierarchy, into account.

AHP deals with crisp (real) values of evaluation judgments, but human reasoning is imprecise, uncertain and fuzzy (Mikhailov and Tsvetinov, 2004:23). Furthermore when the number of criteria considered increases, the number of pair-wise comparisons will increase geometrically. This can lead to inconsistencies or even that the AHP algorithm fails completely. Fuzzy AHP could address this problem and is proposed as a possible alternative method for imprecise problems or problems with more criteria. In a future study it will be investigated how ANP and Fuzzy AHP can complement the decision making of AHP.

The results of this study agree with the results from Graf and List (2005:165) who found in their study (using QWS) that Moodle is preferred above 36 FOSS e-learning systems. The stakeholders at OUT regarded usability as the most important criterion required for the successful implementation of an e-learning system (see Table 4).

In our study it has been shown that — with a well planned evaluation process, good data collection and analysis methods — both novice and technical users can be successfully involved in the evaluation of software. The results contribute to the theory and methodology of the evaluation of FOSS and can be used to design and develop a framework for the evaluation of FOSS for developing countries.
We feel that using AHP in a developing country is suitable because it simplifies a complex problem by breaking it up into smaller steps that can be understood by all. The steps of AHP includes: defining the goals and outcomes of the problem; decomposing the problem into a hierarchical structure of criteria, sub-criteria, attributes, and alternatives; computing pairwise comparisons; employing the Eigenvalue method to estimate relative weights; checking consistency and finally combining the relative weight to obtain the overall rating for the alternatives.

Deployability, maintainability and usability were the only criteria considered in this evaluation as they were assumed to be the most important criteria needed for the implementation of software in a developing country. Computer competency of technical users affects maintainability, bandwidth of networks affects deployability and the fact that the user interface is in a language foreign to the OUT user, affects its usability.

The subjective evaluation was consistent for the attributes identified for the criteria deployability; maintainability and usability. The evaluation was shown to be consistent since the computed consistency indices, for all pairwise comparisons, were equal to zero thus confirming that all stakeholders can participate in the evaluation process. Involving all stakeholders in the selection of software has the added advantage of greater acceptance of the system.

5. Appendix A: AHP computations

5.1 Pairwise comparison matrix

The first pairwise comparison matrix is obtained by comparison of criteria with respect to the objective.

<table>
<thead>
<tr>
<th>Objective</th>
<th>criterion 1</th>
<th>criterion 2</th>
<th>criterion 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>criterion 1</td>
<td>( \frac{c_1}{c_1} = a_{11} )</td>
<td>( \frac{c_1}{c_2} = a_{12} )</td>
<td>( \frac{c_1}{c_3} = a_{13} )</td>
</tr>
<tr>
<td>criterion 2</td>
<td>( \frac{c_2}{c_1} = a_{21} )</td>
<td>( \frac{c_2}{c_2} = a_{22} )</td>
<td>( \frac{c_2}{c_3} = a_{23} )</td>
</tr>
<tr>
<td>criterion 3</td>
<td>( \frac{c_3}{c_1} = a_{31} )</td>
<td>( \frac{c_3}{c_2} = a_{32} )</td>
<td>( \frac{c_3}{c_3} = a_{33} )</td>
</tr>
</tbody>
</table>

Then three pairwise comparison matrices are computed with respect to the three criteria. In our case A can stand for ATutor and B for Moodle.

Pairwise matrix for alternative A and B with respect to criterion 1

<table>
<thead>
<tr>
<th>criterion 1</th>
<th>A</th>
<th>B</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>( \frac{a_{c_1}}{a_{c_1}} )</td>
<td>( \frac{a_{c_1}}{b_{c_1}} )</td>
</tr>
<tr>
<td>B</td>
<td>( \frac{b_{c_1}}{a_{c_1}} )</td>
<td>( \frac{b_{c_1}}{b_{c_1}} )</td>
</tr>
</tbody>
</table>

Pairwise matrix for alternative A and B with respect to criterion 2

<table>
<thead>
<tr>
<th>criterion 2</th>
<th>A</th>
<th>B</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>( \frac{a_{c_2}}{a_{c_2}} )</td>
<td>( \frac{a_{c_2}}{b_{c_2}} )</td>
</tr>
<tr>
<td>B</td>
<td>( \frac{b_{c_2}}{a_{c_2}} )</td>
<td>( \frac{b_{c_2}}{b_{c_2}} )</td>
</tr>
</tbody>
</table>

Pairwise matrix for alternative A and B with respect to criterion 2

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5.2 Normalization matrix

After obtaining the pairwise comparison matrices, the next step is to normalise the matrices with respect to the weights, and then obtain the principal Eigenvector – which is the priority vector. We did this process for level 1, and then for the level 2. Then we computed the overall composite weight for the alternatives A and B, which is a normalization of weighted priority vectors.

Normalization is obtained first by computing the total of each column \( j \) in pairwise comparison matrix (Saaty, 1980) and then each row in a column \( j \) is divided by the column total( \( \sum_{i=1}^{n} a_{ij} \)).

Normalization is given by the following expression:

\[
a_{ij}^* = \frac{a_{ij}}{\sum_{i=1}^{n} a_{ij}} \quad , i,j=1,2,\ldots,n
\]

Objective criterion 1 criterion 2 criterion 3

<table>
<thead>
<tr>
<th>criterion 1</th>
<th>criterion 2</th>
<th>criterion 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>3</td>
<td>3</td>
</tr>
<tr>
<td>( \sum_{i=1,j=1}^{3} a_{ij} )</td>
<td>( \sum_{i=1,j=2}^{3} a_{ij} )</td>
<td>( \sum_{i=1,j=3}^{3} a_{ij} )</td>
</tr>
<tr>
<td>2</td>
<td>2</td>
<td>3</td>
</tr>
<tr>
<td>( \sum_{i=1,j=1}^{3} a_{ij} )</td>
<td>( \sum_{i=1,j=2}^{3} a_{ij} )</td>
<td>( \sum_{i=1,j=3}^{3} a_{ij} )</td>
</tr>
<tr>
<td>3</td>
<td>3</td>
<td>3</td>
</tr>
<tr>
<td>( \sum_{i=1,j=1}^{3} a_{ij} )</td>
<td>( \sum_{i=1,j=2}^{3} a_{ij} )</td>
<td>( \sum_{i=1,j=3}^{3} a_{ij} )</td>
</tr>
</tbody>
</table>

After normalization, the sum of each column \( j \) of the matrix is equivalent to 1.

\[
\sum_{i=1}^{n} a_{ij} = 1
\]

In order to get the weight vector \( W_i \) for the criterion 1, criterion 2 and criterion 3, we take the average of each row of the normalization matrix (assume the normalized matrix is \( a_{ij}^* \)). Therefore

\[
\frac{1}{n} \cdot \sum_{i}^{n} a_{ij}^* = w_i
\]

This process is done for the other matrices in order to obtain the weight vector for each element in the hierarchy.
It gave the weight vector for each pairwise matrix which represents the relative weights of each alternative $p_j$ with respect to criterion $i$ $(i=1, 2, 3)$.

### 5.3 Computing global priorities

This is the step where the relative importance of each element within the level (local priorities) is merged/multiplied with the relative importance of each element in the parent level. This gives the global priorities for each alternative.

The computation for each alternative $j$ $(p_j)$ is done in order to obtain the overall composite weight vector, which is given by $\sum_{i=1}^{n} z_{ij} \cdot w_i$. (ix)

where $w$ is the weight vector or priority vector of the alternative software products and $z$ is the weight vector of the criteria.

### 6. Appendix B: consistency checking

We implemented consistency checking as outlined by Saaty (1980). Checking consistency is done using the following mathematical formulae:

**First procedure**

Let $A$ be a pairwise comparison matrix for $n$ criteria. Let $w$ be the weight vector computed for $A$. Compute

$$ Aw^T $$

(i)

**Second procedure**

Let $\lambda_{\text{max}}$ be a maximum or principal Eigenvalue. Let $n$ be the number of criteria. Compute the $\lambda_{\text{max}}$ given that the number of criteria is $n$.

$$ \lambda_{\text{max}} = \frac{1}{n} \sum_{i=1}^{n} (Aw^T)_i (w^T)_i $$

(ii)

**Third procedure**

Let $CI$ be the consistency index. Compute $CI$

$$ CI = \frac{\lambda_{\text{max}} - n}{n - 1} $$

(iii)

**Fourth procedure**

Let $CR$ be the consistency ratio. Let $RI$ be the random index.

If $CI = 0$ then $A$ is consistent; otherwise if the $CR = \frac{CI}{RI} \leq 0.10$ then $A$ is consistent enough. If $CR = \frac{CI}{RI} \geq 0.10$ then $A$ is seriously inconsistent; the Random index $RI$ for any square matrix is given as a constant value by Saaty (1980).

| Table 4: Random Index (adapted from Saaty, 1980) |
|---|---|---|---|---|---|---|---|---|---|
| n | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| RI | 0 | 0 | 0.58 | 0.9 | 1.12 | 1.24 | 1.32 | 1.41 | 1.45 | 1.49 |
7. Appendix C: example how AHP computations were done

To obtain priority/weight vector for level 1: respondents were asked to indicate the level of importance of three criteria usability, deployability and maintainability \((c_1, c_2, c_3)\) respectively. Then the mean of all the respondents were computed to obtain the input values for AHP. We got \((c_1 = 9, c_2 = 7\) and \(c_3 = 7\)) using scale shown in Table 2. After that the equation (i) from Appendix A was applied and we obtained the following pairwise matrix:

\[
\begin{pmatrix}
1 & 1.285714286 & 1.285714286 \\
0.777778 & 1 & 1 \\
0.777778 & 1 & 1
\end{pmatrix}
\]

Then we computed normalization matrix using equation (vi) from Appendix A and we obtained the following matrix:

\[
\begin{pmatrix}
0.391304 & 0.391304348 & 0.391304348 \\
0.304348 & 0.304347826 & 0.304347826 \\
0.304348 & 0.304347826 & 0.304347826
\end{pmatrix}
\]

After that the priority vector was computed using equation (ix) from Appendix A and we obtained the following weights for usability, deployability and maintainability:

\[
\begin{pmatrix}
0.391304348 \\
0.304347826 \\
0.304347826
\end{pmatrix}
\]

This answer is reflected in level 1 of appendix D.

Then we checked the consistency using the formulae shown in Appendix B.

Equation (i) in Appendix B gave the following result:

\[
\begin{pmatrix}
1.173913043 \\
0.913043478 \\
0.913043478
\end{pmatrix}
\]

The equation (ii) in appendix B was applied and we got

\[
= 1/3*(1.173913043/0.391304348+0.913043478/0.304347826+0.913043478/0.304347826) = 3
\]

After this computation, we calculated the CI using equation (iii) in Appendix B and the answer was

\[
= (3-3)/2 = 0 \text{ (the number of criteria \((n)\) was 3)}
\]

This indicated that the consistency index was equal to 0 (means consistent judgments).

These computations were done for all hierarchy levels and the results are shown in Appendix D.

8. Appendix D: priority weight obtained after using AHP

<table>
<thead>
<tr>
<th>LEVEL 1</th>
<th>LEVEL 2</th>
<th>LEVEL 3</th>
<th>ATutor</th>
<th>Moodle</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Weight</td>
<td>Priority</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>0.5000</td>
<td>0.5000</td>
</tr>
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**Acknowledgement**

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**References**


