

FUZZY ANALYTIC HIERARCHY BASED DBMS SELECTION IN TURKISH NATIONAL IDENTITY CARD MANAGEMENT PROJECT

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ABSTRACT

Database Management Systems (DBMS) play an important role to support enterprise application developments. Selection of the right DBMS is a crucial decision for software engineering process. This selection requires optimizing a number of criteria. Evaluation and selection of DBMS among several candidates tend to be very complex. It requires both quantitative and qualitative issues. Wrong selection of DBMS will have a negative effect on the development of enterprise application. It can turn out to be costly and adversely affect business process. The following study focuses on the evaluation of a multi criteria decision problem by the usage of fuzzy logic. We will demonstrate the methodological considerations regarding to group decision and fuzziness based on the DBMS selection problem. We developed a new Fuzzy AHP based decision model which is formulated and proposed to select a DBMS easily. In this decision model, first, main criteria and their sub criteria are determined for the evaluation. Then these criteria are weighted by pair-wise comparison, and then DBMS alternatives are evaluated by assigning a rating scale.

KEYWORDS

AHP, Fuzzy AHP, Fuzzy Logic, DBMS Selection, Multi Criteria Decision Making

1. INTRODUCTION

The selection of the Database Management System (DBMS) has been always been considered as a major source of uncertainty. The cost of purchasing and running DBMS efficiently is very high; therefore, DBMS selection is very crucial step for the software firms. DBMS software packages provide a large number of features that are customizable to meet specific needs of the organizations. Selecting a right DBMS to meet the organizational requirements is a difficult task. It needs a full examination of many factors. A questionnaire is used for determining the user preferences in software engineering process.

There has been a growth in the number of multiple-criteria decision-making (MCDM) methods to evaluate several alternatives to achieve a certain goal. The analytic hierarchy process (AHP) [1] is one of the most widely used methods. AHP method is a multi-criteria decision-making approach. The AHP method has attracted the interest of many decision makers to its ability to solve complex decision problems. It organizes the basic rationality by breaking down a problem into smaller constituents and then calls for only simple pairwise comparison judgments, to

develop priorities in each level [2]. AHP decomposes a decision problem into a hierarchy of more easily comprehended sub-problems. The element of the hierarchy can relate to any aspect of the decision problem. Once the hierarchy is built, the decision makers evaluate its elements systematically by comparing them to other one, using two elements at a time [3].

In this article, evaluating DBMS alternatives is considered independently, because each DBMS solution is relatively different and independent. Pair-wise comparisons used in Fuzzy Analytical Hierarchy Process (FAHP) may result in errors or inconsistency if they are used with the preferences of Industry users or in all evaluations of DBMS selection process.

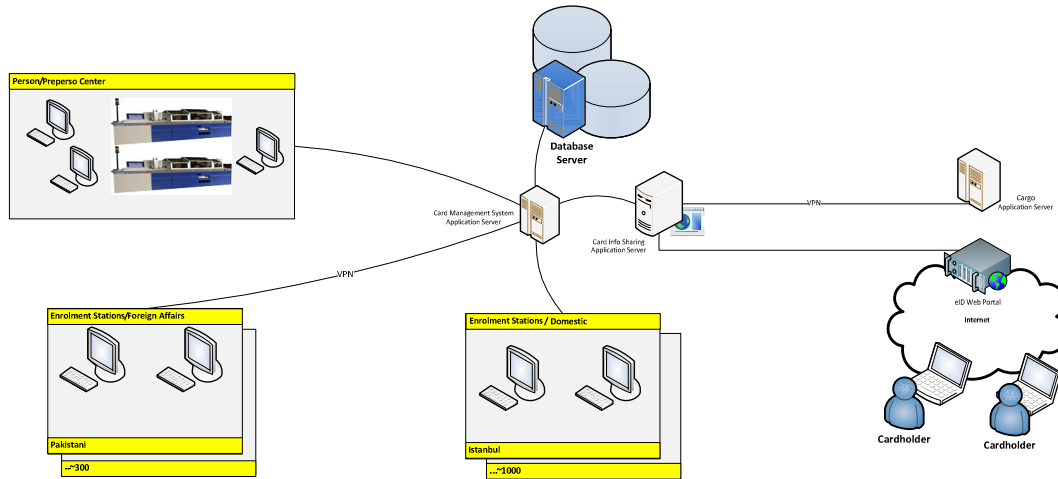


Figure 1. Turkish eID Management System Overview

The population of the Turkey is around 75 million people according to last census. The identification paper in Turkey is mandatory and every citizen required having an eID (electronic identity card). All signed operations related to card life cycle and data contained in the card occupy 40KB. The whole process for distributing eIDs to all citizens will be completed in three years. The selected DB for storing citizens' data must be able to manage/govern massive data block. At least three terminals in every district will have connections to the main server where services related to eID system are provided. Also, there will be total 3000 enrolment terminals connected and getting service from the main server. Moreover, every citizen who applied for an eID will be able to query the status of the eID by using a device called KIOSK. Based on all this information mentioned in the above sections, the selection of the database plays a crucial role in this research. The selection criterions are also a vital issue for the customer as well.

Proposed model is implemented for Turkish National Identity Card Management Project DBMS selection process of the Scientific and Technological Research Council of Turkey (TUBITAK), the leading agency for management, funding and conduct of research in Turkey. TUBITAK is responsible for promoting, developing, organizing, conducting and coordinating research and development in line with national targets and priorities. TUBITAK acts as an advisory agency to the Turkish Government on science and research issues, and is the secretariat of the Supreme Council for Science and Technology (SCST), the highest S&T policy making body in Turkey. Analytic Hierarchy Process (AHP) and Fuzzy Analytic Hierarchy Process (Fuzzy AHP) are used to determine the relative weights of the attributes to evaluate DBMS selection in the National ID Card project.

2. FUZZY ANALYTIC HIERARCHY PROCESS (FAHP)

The FAHP method is an advanced analytical method which is developed from the AHP. In spite of the popularity of AHP, this method is often criticized for its inability to adequately handle the inherent uncertainty and imprecision associated with the mapping of the decision-maker's perception to exact numbers [4]. In FAHP method, the fuzzy comparison ratios are used to be able to tolerate vagueness. In literature, there are several methods to obtain the priorities. Xu's fuzzy least square method [5] is concerned with the question of estimation of weights of factors by least squares from a fuzzy judgment matrix. First, the expression form of least squares under fuzziness is given. Then the associated system of equation is developed. Buckley's geometric mean method [6], Chang's synthetic extend analysis [7], Mikhailov's fuzzy preference programming [8], Wang's two stage logarithmic programming [9]. One of the best known of these different methods is Fuzzy Extend Analysis proposed by Chang. Chang's fuzzy extend analysis is used in this research to evaluate the DBMS alternatives.

3. A SIMPLIFIED DECISION MODEL FOR THE EVALUATION OF DBMSs

In order to evaluate different Database Management Systems (DBMS), we developed a simplified decision model. Obviously, the simplification reduces the decision hierarchy. In this study, critical factors are grouped into eight main categories which are OS support, cluster capabilities, data size, table and view properties, database features, partitioning, backup and recovery and security. 27 sub criteria are taken to indicate the DBMS selection process. These are the independent variables of the decision model. It may also vary according to the business requirements of the enterprise information system. The critical factors used in our decision model are shown in Table 1.

Table 1. Critical factors used in DBMS Selection by Fuzzy AHP

Criteria	Sub criteria	Description
OS Support (OS)	Windows	Operating systems supported
	Linux	
	BSD	
	Mac OS X	
Cluster Support (CLU)	Active-Active Cluster	Supported cluster types
	Active-Passive Cluster	
	Standby DB	
Data Size (DS)	Blob Data Size	Max data size
	Row Length	
	Max. Table Size	
Table and Views (TV)	Temporary Tables	Supported table and view types
	Materialized Views	
Database Features (FEAT)	Parallel Query	Supported database features
	Windowing	
	Automatic optimization SQL Statements	
Partitioning (PART)	Range	Supported partitioning types
	Hash	
	List	
	Shadow	

Backup and Recovery (BR)	Full Backup	Supported Backup and Recovery types
	Incremental Backup	
	Partial Backup	
	Online Backup	
Security (SEC)	Audit	Security features of DBMS
	Brute Force Protection	
	Network Security	
	Password Complexity Rule	

3.1. Extend Analysis Method on Fuzzy AHP

For the estimation of the importance of these criteria, we used the FAHP. Let \tilde{A} represent a fuzzified reciprocal $n \times n$ judgment matrix containing all pairwise comparisons \tilde{a}_{ij} between elements i and j for all $i, j \in \{1, 2, \dots, n\}$.

$$\tilde{A} = \begin{bmatrix} (1,1,1) & \tilde{a}_{12} & \dots & \tilde{a}_{1n} \\ \tilde{a}_{21} & (1,1,1) & \dots & \tilde{a}_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \tilde{a}_{n1} & \tilde{a}_{n2} & \dots & (1,1,1) \end{bmatrix}$$

where $\tilde{a}_{ji} = \tilde{a}_{ij}^{-1}$ and all are triangular fuzzy numbers (TFN) $\tilde{a}_{ij} = (l_{ij}, m_{ij}, u_{ij})$ with l_{ij} the lower and u_{ij} the upper limit and m_{ij} is the point where the membership function $\mu(x) = 1$.

In the following, first the outlines of the extend analysis method on Fuzzy AHP are given and then the method is applied to a DBMS selection problem. Let $X = \{x_1, x_2, \dots, x_n\}$ be object set, and $U = \{u_1, u_2, \dots, u_m\}$ be a goal set. According to the method of Chang [7], each object is taken and analysis for each goal, g_i , is performed, respectively. Therefore, m extent analysis values for each object can be obtained, with the following signs:

$$M_{g_i}^1, M_{g_i}^2, \dots, M_{g_i}^m \quad i = 1, 2, \dots, n$$

where all the $(i = 1, 2, \dots, m)$ are triangular fuzzy numbers whose parameters are represented as (a, b, c) , describing least, most and largest possible values respectively.

Step 1: The value of fuzzy synthetic extent with respect to the i_{th} object is defined as:

$$S_i = \sum_{j=1}^m M_{g_i}^j \otimes \left[\sum_{i=1}^n \sum_{j=1}^m M_{g_i}^j \right]^{-1} \tag{1}$$

to calculate $\sum_{j=1}^m M_{g_i}^j$ we perform the ‘‘fuzzy addition operation’’ of m extent analysis values for a particular matrix given in equation 2 below, at the last step of calculation, new (l, m, u) set is obtained and used for the next:

$$\sum_{j=1}^m M_{g_i}^j = \left(\sum_{j=1}^m l_j, \sum_{j=1}^m m_j, \sum_{j=1}^m u_j \right) \tag{2}$$

where l is the lower limit value, m is the most promising value and u is the upper limit value and to obtain the reciprocal in equation 1, we perform the “fuzzy addition operation”

$M_{g_i}^j$ ($j = 1, 2, \dots, m$) values as given in equation 3:

$$\sum_{i=1}^n \sum_{j=1}^m M_{g_i}^j = \left(\sum_{j=1}^m l_j, \sum_{j=1}^m m_j, \sum_{j=1}^m u_j \right) \quad (3)$$

and then compute the inverse of the vector in the equation (3) to obtain equation (4) :

$$\left[\sum_{i=1}^n \sum_{j=1}^m M_{g_i}^j \right]^{-1} = \left(\frac{1}{\sum_{i=1}^n u_i}, \frac{1}{\sum_{i=1}^n m_i}, \frac{1}{\sum_{i=1}^n l_i} \right) \quad (4)$$

Step 2: The degree of possibility of $M_1 \geq M_2$ is defined as:

$$V(M_1 \geq M_2) = \text{SUP}_{x \geq y} [\min(\mu_{M_1}(x), \mu_{M_2}(y))] \quad (5)$$

where SUP represents supremum and when a pair (x, y) exists such that $y \geq x$ and $\mu_{M_1}(x) = \mu_{M_2}(y)$ then we have $V(M_2 \geq M_1) = 1$. Given that M_1 and M_2 are convex fuzzy numbers, so $V(N_1 \geq N_2) = 1$ if $n_{11} \geq n_{21}$. And

$$V(M_2 \geq M_1) = \mu_{M_2}(d) = \begin{cases} 1 & m_2 \geq m_1 \\ 0 & l_1 \geq u_2 \\ \frac{l_1 - u_2}{(m_2 - u_2) - (m_1 - l_1)} & \text{otherwise} \end{cases} \quad (6)$$

where d is the ordinate of the highest intersection point D between μ_{M_1} and μ_{M_2} . (See Figure 2)

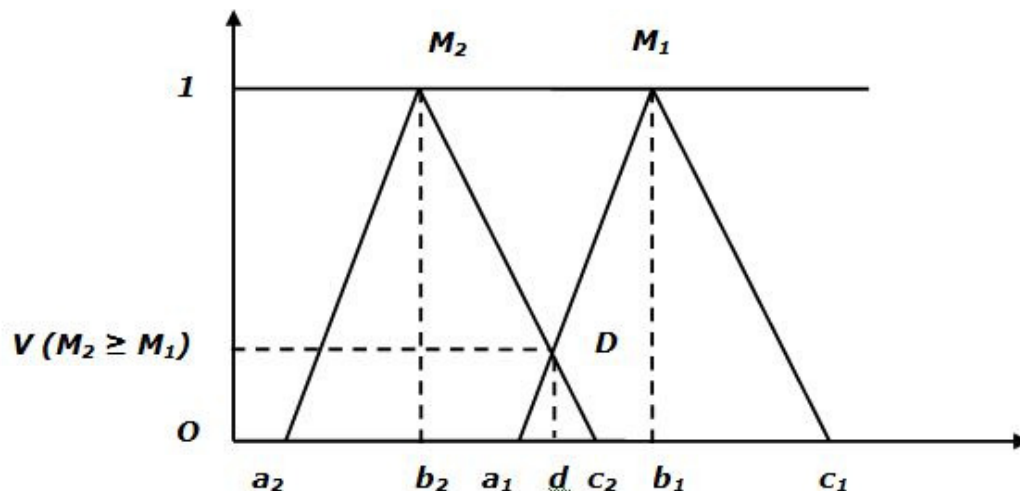


Figure 2. Intersection between M_1 and M_2

Step 3: The degree of possibility for a convex fuzzy number to be greater than k convex fuzzy numbers M_i ($i = 1, 2, \dots, k$) can be defined by

$$V(M \geq M_1, M_2, \dots, M_k) = V[(M \geq M_1) \text{ and } (M \geq M_2) \text{ and } \dots \text{ and } (M \geq M_k)] \\ = \min V(M \geq M_i), \quad i = 1, 2, \dots, k$$

assume that equation 7 is

$$d'(A_i) = \min V(S_i \geq S_k) \tag{7}$$

For $k = 1, 2, \dots, m; k \neq i$. Then the weight vector is obtained as follows

$$W' = (d'(A_1), d'(A_2), \dots, d'(A_n))^T \tag{8}$$

where $A_i (i = 1, 2, \dots, n)$ are n elements.

Step 4: After normalization, the normalized weight vectors are given in the equation 11

$$W = (d(A_1), d(A_2), \dots, d(A_n))^T \tag{9}$$

where W is not a fuzzy number.

In this methodology, we have determined the criteria as given in Table 1. The fuzzy conversion scale is shown in Table 2. Different scales can be found in the literature as in Abdel-Kader and Dugdale's study.

Table 2 Triangular Fuzzy Conversion Scale

Linguistic Scale	Triangular Fuzzy Scale	Traingular Fuzzy reciprocal Scale
Just Equal	(1,1,1)	(1,1,1)
Equally Important	(1/2,1,3/2)	(2/3,1,2)
Weakly Important	(1,3/2,2)	(1/2,2/3,1)
Strongly More Important	(3/2,2,5/2)	(2/5,1/2,2/3)
Very strong more important	(2,5/2,3)	(1/3,2/5,1/2)
Absolutely more important	(5/2,3,7/2)	(2/7,1/3,2/5)

4. APPLICATION OF THE MODEL

The proposed model is implemented for Turkish National Identity Card project's DBMS selection process of the Scientific and Technological Research Council of Turkey (TUBITAK), the leading agency for management, funding and conduct of research in Turkey. TUBITAK is responsible for promoting, developing, organizing, conducting and coordinating research and development in line with national targets and priorities. TUBITAK acts as an advisory agency to the Turkish Government on science and research issues, and is the secretariat of the Supreme Council for Science and Technology (SCST), the highest S&T policy making body in Turkey. Fuzzy Analytic Hierarchy Process (Fuzzy AHP) is used to determine the relative weights of the attributes to evaluate DBMS selection in the Turkish National ID Card project.

Experts who are participating in the implementation of this model have regarded eight important criteria for DBMS selection. Table 3 shows the aggregated fuzzy pairwise comparisons of the four experts.

Table 3 Aggregated Fuzzy Comparison Matrix of the Attributes for Fuzzy AHP Model

	OS	CLU	DS	TV	FEAT	PART	B&R	SEC
OS	1,1,1	0.29,0.3,0.4	0.67,1,1.5	0.67,1,1.5	0.67,1,1.5	0.7,1,1.5	0.67,1,1.5	1,1.33,2
CLU	2.5,3,3.5	1,1,1	1.5,2,2.5	1.5,2,2.5	1.5,2,2.5	1.5,2,2.5	1.5,2,2.5	1.5,2,2.5
DS	0.67,1,1.5	0.4,0.5,0.67	1,1,1	1,1,1	1,1,1	0.4,0.5,0.7	0.67,1,1.5	0.67,1,1.5
TV	0.67,1,1.5	0.4,0.5,0.67	1,1,1	1,1,1	0.4,0.5,0.7	0.67,1,1.5	1,1,1	1,1,1
FEAT	0.67,1,1.5	0.4,0.5,0.67	1,1,1	1.5,2,2.5	1,1,1	1,1.33,2	0.67,1,1.5	1,1,1
PART	0.67,1,1.5	0.4,0.5,0.67	1.5,2,2.5	0.67,1,1.5	0.5,0.75,1	1,1,1	0.67,1,1.5	1.5,2,2.5
BR	0.67,1,1.5	0.4,0.5,0.67	0.67,1,1.5	1,1,1	0.67,1,1.5	0.67,1,1.5	1,1,1	0.67,1,1.5
SEC	0.5,0.75,1	0.4,0.5,0.67	0.67,1,1.5	1,1,1	1,1,1	0.4,0.5,0.7	0.67,1,1.5	1,1,1

Table 4 Aggregated Fuzzy Comparison Matrix of the Attributes for Fuzzy AHP Model

$\sum_{i=1}^8 M_{gt}^1 = (5.62,7.67,10.9)$	$\sum_{i=1}^8 M_{gt}^5 = (7.23,8.83,11.167)$
$\sum_{i=1}^8 M_{gt}^2 = (12.5,16,19.5)$	$\sum_{i=1}^8 M_{gt}^6 = (6.9,9.25,12.167)$
$\sum_{i=1}^8 M_{gt}^3 = (5.8,7,8.83)$	$\sum_{i=1}^8 M_{gt}^7 = (5.73,7.5,10.167)$
$\sum_{i=1}^8 M_{gt}^4 = (6.13,7,8.3)$	$\sum_{i=1}^8 M_{gt}^8 = (5.63,6.75,8.3)$

Aggregated fuzzy comparison matrix of the attributes for our fuzzy decision model is show in Table 4.

The degree of possibility of superiority is calculated and is denoted by $V(M_i \geq M_j)$. The calculated values are show in Table 5.

Table 5 Degree of Possibility of Superiority

$V(M_1 \succcurlyeq M_2) = 0.32$	$V(M_3 \succcurlyeq M_4) = 0.91$	$V(M_5 \succcurlyeq M_6) = 1$	$V(M_7 \succcurlyeq M_8) = 0.90$
$V(M_1 \succcurlyeq M_3) = 1$	$V(M_3 \succcurlyeq M_5) = 0.13$	$V(M_5 \succcurlyeq M_7) = 0.374$	$V(M_7 \succcurlyeq M_9) = 0.262$
$V(M_1 \succcurlyeq M_4) = 1$	$V(M_3 \succcurlyeq M_6) = 1$	$V(M_5 \succcurlyeq M_8) = 1$	$V(M_7 \succcurlyeq M_2) = 1$
$V(M_1 \succcurlyeq M_5) = 0.87$	$V(M_3 \succcurlyeq M_7) = 0.75$	$V(M_5 \succcurlyeq M_9) = 1$	$V(M_7 \succcurlyeq M_4) = 1$
$V(M_1 \succcurlyeq M_6) = 0.84$	$V(M_3 \succcurlyeq M_8) = 0.72$	$V(M_5 \succcurlyeq M_2) = 0.954$	$V(M_7 \succcurlyeq M_5) = 0.843$
$V(M_1 \succcurlyeq M_7) = 1$	$V(M_3 \succcurlyeq M_9) = 0.93$	$V(M_5 \succcurlyeq M_3) = 1$	$V(M_7 \succcurlyeq M_6) = 0.801$
$V(M_1 \succcurlyeq M_8) = 1$	$V(M_3 \succcurlyeq M_2) = 1$	$V(M_5 \succcurlyeq M_4) = 1$	$V(M_7 \succcurlyeq M_7) = 1$
$V(M_1 \succcurlyeq M_9) = 1$	$V(M_4 \succcurlyeq M_1) = 0.91$	$V(M_6 \succcurlyeq M_1) = 1$	$V(M_8 \succcurlyeq M_1) = 0.865$
$V(M_2 \succcurlyeq M_3) = 1$	$V(M_4 \succcurlyeq M_2) = 0.073$	$V(M_6 \succcurlyeq M_2) = 0.451$	$V(M_8 \succcurlyeq M_2) = 0.071$
$V(M_2 \succcurlyeq M_4) = 1$	$V(M_4 \succcurlyeq M_3) = 1$	$V(M_6 \succcurlyeq M_3) = 1$	$V(M_8 \succcurlyeq M_3) = 0.959$
$V(M_2 \succcurlyeq M_5) = 1$	$V(M_4 \succcurlyeq M_5) = 0.72$	$V(M_6 \succcurlyeq M_4) = 1$	$V(M_8 \succcurlyeq M_4) = 0.958$
$V(M_2 \succcurlyeq M_6) = 1$	$V(M_4 \succcurlyeq M_6) = 0.693$	$V(M_6 \succcurlyeq M_5) = 1$	$V(M_8 \succcurlyeq M_5) = 0.699$
$V(M_2 \succcurlyeq M_7) = 1$	$V(M_4 \succcurlyeq M_7) = 0.923$	$V(M_6 \succcurlyeq M_6) = 1$	$V(M_8 \succcurlyeq M_6) = 0.671$
$V(M_2 \succcurlyeq M_8) = 1$	$V(M_4 \succcurlyeq M_8) = 1$	$V(M_6 \succcurlyeq M_7) = 1$	$V(M_8 \succcurlyeq M_7) = 0.889$

The minimum degree of possibility of superiority for each criterion over another is calculated. The weight vector is given as

$$W = (0.3214, 1, 0.1299, 0.0734, 0.3741, 0.4509, 0.2624, 0.0716).$$

The normalized weight vector is calculated as

$$W = (0.1198, 0.3726, 0.0484, 0.0274, 0.1394, 0.1680, 0.098, 0.0267).$$

The normalized weight of each criterion is shown in Figure 3.

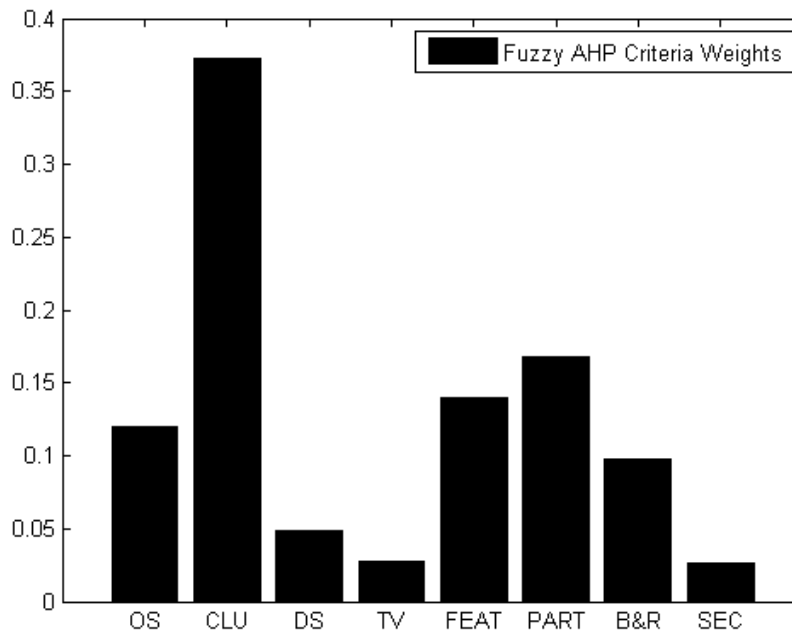


Figure 3. Weights of criteria used in DBMS Selection Model

Figure 4 shows the graphical representation of fuzzy priority numbers calculated from Fuzzy AHP pair wise comparison.

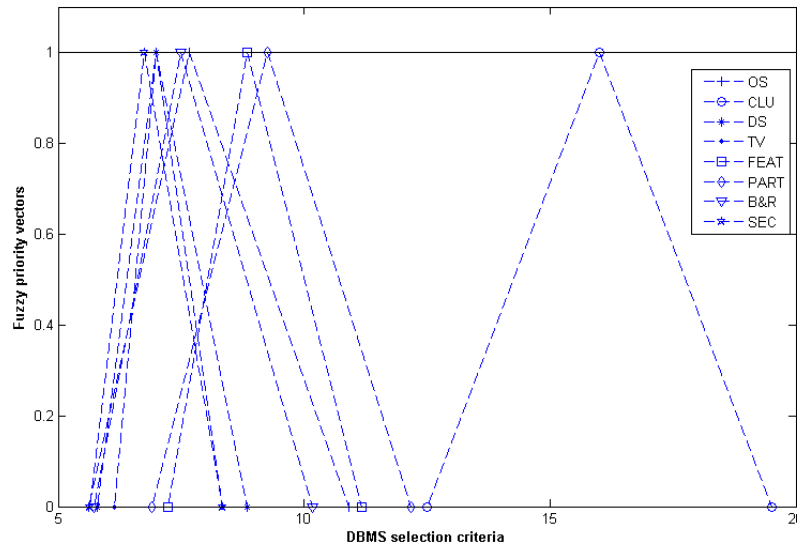


Figure 4. Fuzzy Membership Functions for DBMS Selection Features

6. CONCLUSIONS

The aim of this study is to investigate the application of the Fuzzy Analytic Hierarchy Process (FAHP) method of multi-criteria decision-making within a DBMS selection problem. The AHP provides an effective approach for solving multiple criteria decision making problems in software development. This case study shows, that although the selection of the database management system selection - is one of the most important activities of an IT project in a company in an IT project-, it does not mean it has to be complicated. An algorithm using Fuzzy AHP is emphasized and proposed as an original study of DBMS. Fuzzy AHP is utilized in order to eliminate or reduce cognitive biases in decision making. The developed model eases the decision maker's mission of choosing the quantitative weights and making further calculations.

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