

An Improved Interval AHP Method for Assessment of Cloud Platform-based Electrical Safety Monitoring System

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Abstract – Electrical safety monitoring System (ESMS) is a critical component in modern power systems, which is characterized by large-scale access points, massive users and versatile requirements. For convenience of the information integration and analysis, the software development, maintenance, and application in the system, the cloud platform based ESMS is established and assessed in this paper. Firstly the framework of the system is proposed, and then the assessment scheme with a set of evaluation indices are presented, by which the appropriate cloud product can be chosen to meet the requirements of a specific application. Moreover, to calculate the weights of the evaluation indices under uncertainty, an improved interval AHP method is adopted to take into consideration of the fuzziness of expert scoring, the qualitative consistency test, and the two normalizations in the process of eigenvectors. Case studies have been made to verify the feasibility of the assessment approach for ESMS.

Keywords: Cloud computing, Consistency test, Electrical safety monitoring system, Interval analytic hierarchy process

1. Introduction

In Smart Grid, electrical safety monitoring system (ESMS) has been drawn more and more attention and has developed a variety of applications, such as anti-tampering, VIP customer management, and power supply quality monitoring. There are many studies on ESMS from the aspects of uncertainty analysis [1], model building [2, 3], optimal control [4, 5], and software development [6]. In [7], it discusses a strategy aligned with leading edge developments in advanced safety management of hazards with high potential for fatality. However, there are numerous challenges need to be addressed such as the limited function, low utilization ratio, and low load ratio of monitoring equipment. Moreover, ESMS has the characteristics of enabling large-scale access points, supporting massive users and accommodating versatile requirements. Therefore, cloud computing is an ideal way to resolve those issues with the advantages of integration, flexibility, and resource on-demand.

Cloud computing is one of the most popular cyber solutions for many applications such as in education and industry. An excellent summary of main developments of cloud computing was given in [8]. The contemporaneous research involves in the definition [9], data storage [10, 11],

and load evaluation [12]. K. Zhou [13] proposes a new cloud-based power system operation model. In power systems, especially in the application of ESMS, it is critical to improve the performance and service level through cloud computing [14]. The difficulty to face is to choose the type of cloud platform (CP) from cloud providers among private cloud, public cloud and hybrid cloud platforms. Therefore, the criterion for the choice of the type of cloud platform (CP) from cloud providers among private, public and hybrid cloud platform need be studied. Also, an evaluation index set and a method of CP based ESMS are essential to guide the future construction, operation and maintenance of ESMS. The solution of the multi-attribute decision making problem in the evaluation considering the fuzziness of expert scoring results is also crucial. Currently, the AHP and fuzzy AHP algorithm keep attracting extensive attention, and many results have been achieved on the construction of the judgment [15, 16], uncertainty [17-20] and expert scoring [21, 22].

This paper proposes an improved interval AHP method for the assessment of CP-based ESMS. The main contributions of this paper are as follows: 1) To deal with the problems of massive different user requirements and large-scale decentralized access points that EMSM faces, a framework of CP based EMSM is proposed; 2) A serial of cloud evaluation indices is proposed to guide the construction of the CP based ESMS for different users, which could be adopted by customers to guide the choice of appropriate cloud products from a large number of cloud service providers; 3) In the process of solving the index weight, an improved interval AHP is adopted to solve the problems of fuzzy expert scoring, latency of consistency

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test and computational efficiency, which provides the reference for multi-criteria decision making in similar situations.

The rest of this paper is organized as follows. The next section gives the framework of a CP based ESMS. In section 3, we present the evaluation indices of a cloud computing application for ESMS. In section 4, the improved interval AHP is proposed to calculate the weights of the index. Section 5 presents case studies to verify the advantages of the proposed interval AHP method. Finally, the paper concludes with the summary of the main points.

2. The CP Based ESMS

Cloud computing, which has high scalability, is an integration of grid computing and virtualization technologies. It can also be easily upgraded with the integrated low computability parts, and is very convenient for information integration and analysis, software development, maintenance, and utilization. In this section, the details of the framework of CP based ESMS will be presented. The framework of cloud platform based ESMS is demonstrated in Fig. 1, which consists of infrastructure layer, platform layer, application layer, user layer and cloud safety.

2.1 Infrastructure layer

The infrastructure layer is the IaaS layer of the cloud computing architecture, which includes physical and management layers. Physical layer virtualizes two types of devices in ESMS. On one hand, it can manage some devices of ESMS by using virtual machine, which includes servers, disc arrays, and computers. Those servers consist of the front servers, web server and memory server. The disc arrays are mainly data centers. The computers include office PC, and monitoring client. On the other hand, the management of communication devices of ESMS is

operated by virtual network technology, which includes routers, switches and hubs. The management layer can dispatch and manage IT resources dynamically. It can allocate the computer resources dynamically via the load balance process and manage the resources and external request by resource management process. Therefore, the universal cloud-level management can be achieved by managing the idle computer hardware resources under monitoring.

Therefore, in order to meet the massive and heterogeneous hardware and software in end-user, the infrastructure layer of CP based ESMS not only needs to have a strong ability to virtual, and also has a fast respond to ensure the plug and play.

2.2 Platform layer

Platform layer is the Platform-as-a-Service (PaaS) layer of the cloud calculation framework. It mainly includes the Software Development Kit (SDK) exploitation environment, the sharing of the editing tools, and cross-platform exploitation environment in order to build the exploitation of the platform quickly. The SDK exploitation environment endorses different types of programming language, such as C, C++, C#, Java, and so on. Cross-platform exploitation environment supports Windows, Linux, Mac, etc.

2.3 Application layer

Application layer is the Software-as-a-Service (SaaS) layer of cloud computing framework for the related business application of ESMS. It includes user authentication, communication administration, electric tampering administration, key electricity consumers' administration, new customer' administration and other related applications.

The application layer not only faces a large number of real-time online users, but also should quickly and timely reply to the different requests of the end-users, and therefore, this layer must be able to guarantee the power quality of service.

2.4 User layer

User layer allows the flexible access from the users and administrative staff. It mainly comprises the users' communication mode and interaction methods. The communication mode refers to wireless internet, General Packet Radio Service (GPRS), Ethernet network and private optical fiber access, etc. The users' interaction methods includes client-side, wireless terminal, web explorer, etc.

Since there are so many end-users accessing to the cloud platform through a variety of different ways and different modes of communication, the cloud platform should support vast amounts of communication protocols.

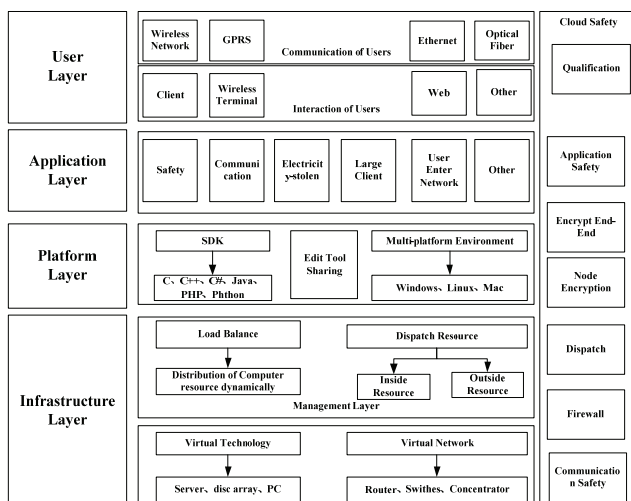


Fig. 1. Framework of cloud platform based ESMS

2.5 Cloud safety

Cloud safety is a significant factor which could directly impact the normal operation of the ESMS. It is mainly concerned with network communication safety, firewall of software and hardware, permissions of hierarchical administration, nodes' encryption, point to point encryption, the safety of applications, etc.

3. Evaluation Index of a CP Based ESMS

In the construction process of CP based ESMS, how to choose the appropriate CP for ESMS is a critical problem. In this section, a serial of evaluation indices for the cloud computing applications of ESMS is proposed from three aspects of switching capability of network architecture A_{11} , fast response capability of memory architecture A_{12} , and virtual ability A_{13} . The indices are summarized in Table 1. Some key indices are expounded as follows.

3.1 Switching capability of network architecture index set A_{11}

A_{11} is a capacity of the network for cloud platform in ESMS, mainly including routing throughput A_{111} , packet forwarding per second A_{112} , time-lag A_{113} , transmission speed A_{114} , total number of supporting router protocols and standards A_{115} and total number of checklist per second A_{116} . To measure this ability of communication requirement of massive end-user in cloud platform, we choose the A_{111} and A_{113} . The index A_{111} is calculated as.

$$A_{111} = R * \sum_{i=1}^n S_i \tag{1}$$

where R is the number of node for receiving data, and S_i is a packet received by the receiving node.

The index A_{113} has the form

$$A_{113} = \sum_{i=1}^n (Ts_i + Tc_i + Th_i + Tq_i) \tag{2}$$

where T_{si} , T_{ci} , T_{hi} and T_{qi} denote a sending time delay, a propagation delay, a time of processing delay, and a time of queuing delay, respectively.

3.2 Fast response capability of memory architecture index set A_{12}

A_{12} is a reflection of the overall storage capacity in CP based ESMS. It main includes total number of user connections per second A_{121} , total number of user service per second A_{122} , server utilization ratio A_{123} , memory reading and writing rate A_{124} , I/O ratio A_{125} and service response time A_{126} . We compute the quantitative data service of massive user according to (3)-(4).

$$A_{121} = \sum_{i=1}^n R_i, R_i = \begin{cases} 1, \text{on-line} \\ 0, \text{off-line} \end{cases} \tag{3}$$

where R_i belonging to $\{0,1\}$ is the network nodes of cloud computing. While $R_i=1$, the network node is online, otherwise, the node is offline. In addition, the service response time A_{126} is given by

$$A_{126} = \sum_{i=1}^k N_i + \sum_{j=1}^m A_j \tag{4}$$

where N_i and A_j are the network latency and waiting time for an application of cloud computing, respectively.

3.3 Virtual ability index set A_{13}

Virtual ability is a reflection of virtualization level in CP based ESMS. It can be evaluated by the total time of running A_{131} , total number of virtual CPU A_{132} , total number of virtual memory A_{133} , total number of virtual I/O A_{134} , deceleration ratio A_{135} , and acceleration ratio A_{136} . Among those indices, A_{132} is the most important one reflecting virtual ability, defined as

$$A_{132} = \sum_{i=1}^n V_i \tag{5}$$

where V_i is the number of virtual CPU in the i -th node of cloud computing.

The proposed index system takes into considerations the requirement of massive participants, scattered points, the difficulty of the subdividing service, and the high demand to service level. It allows us to solve the following

Table 1. Evaluation indices of CP based ESMS

First index	second-level index	third-level index
Evaluation index of cloud platform based ESMS A_1	Switching capability of network architecture A_{11}	Routing throughput A_{111}
		Packet forwarding per second A_{112}
		Time-lag A_{113}
		Transmission speed A_{114}
		Total number of support routing protocols and standards A_{115}
		Total number of checklist per second A_{116}
	Fast response capability of memory architecture A_{12}	Total number of user connections per second A_{121}
		Total number of user service per second A_{122}
		Service utilization ratio A_{123}
		Memory read and write speed A_{124}
		Input/output ratio A_{125}
		Service response time A_{126}
	Virtual Ability A_{13}	Total time of running A_{131}
		Total number of virtual CPU A_{132}
		Total number of virtual memory A_{133}
		Total number of virtual I/O A_{134}
		Deceleration ratio A_{135}
		Acceleration ratio A_{136}

problem, that is, different suppliers have different product performance among private cloud, public cloud and hybrid cloud, but the evaluation criteria recognizing cloud computing technology is insufficient. And it can also be used to guide the selection of CP for different user groups such as power grid enterprises, industrial users, commercial users and resident users, which have relatively large discrepancy in demand.

4. Interval AHP Method

For the index system built previously, a key issue is to determine the weight for each index. This is a multi-group decision making problem. Currently, there exist the subjective or objective methods to solve this problem. In this section, we propose an improved analytic hierarchy process (IAHP). The basic process of IAHP is illustrated in Fig. 2, and we explain it in detail as follows.

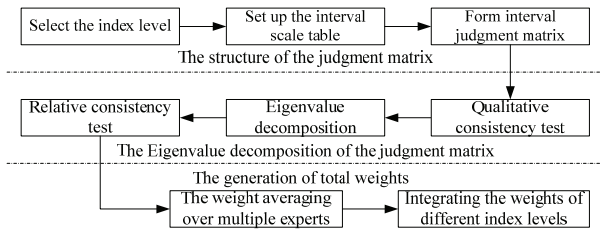


Fig. 2. The basic process of IAHP

4.1 Main idea of IAHP

IAHP can be divided into three phases: the construction of the judgment matrix, the eigenvalue decomposition of the judgment matrix, and the generation of total weights. In the first phase, we use intervals to represent the weights of indices, considering the weights of index are relative and uncertain to some extent from the perspective of cognition. Specifically, the interval scales from 1 to 5 are adopted to simplify the complexity of the expert scoring. The IAHP is used to solve the problem of fuzzy insufficiency of experts scoring in AHP.

A difficulty we face is the consistency test of the judgment matrix, so in the second phase, the qualitative consistency test is firstly performed so as to improve the effectiveness and rationality of the expert scoring. Then, in order to pass through the consistency test for the judgment matrix, the eigenvalue decomposition is conducted and in the decomposition two normalizations are used to improve the convergence speed. Last, in the third phase, we provide a strategy to obtain the total weight of the selected level index by the combination of the eigenvalues corresponding to multiple scoring experts.

4.2 The judgment matrix construction

The judgment matrix is constructed according to the

following steps.

Step 1: Select the index level

As described in the previous section, the analytical index system for CP based ESMS is formed by the first indices, the second-level indices, and the third-level indices, corresponding to the target layer, index layer, and the scheme layer in AHP. To construct the judgment matrix, we first select the second-level indexes, and let selected N experts score according to the next step. Once we finish computing the weights for the second level indexes, the third-level indexes will be selected, for which we continue to calculate their weights.

Step 2: Set up the interval scale table

Interval scale consists of the interval midpoint and the width. The interval midpoint is a cardinal number of the expert scoring, which is denoted by a_{ij} while comparing the i -th index to j -th index in the considered index level. The width is a fuzzy number according to the uncertainty or ambiguity of the expert's experience, which is denoted by μ . In general, the experts determine firstly the midpoint a_{ij} , then, determine the width μ based on the uncertainty and fuzziness. The interval proportional scale of the expert scoring is listed in Table 2.

Table 2. Interval proportional scale

Importance (compare i to factor j)	Interval mid-point values a_{ij}	Interval widths μ
The same important	1	$0 < \mu < 1$
Slightly important	2	
More important	3	
Strongly important	4	
Extremely important	5	

Step 3: Form interval judgment matrix A_m

The interval judgment matrix A_m is of size $n \times n$, where n denotes the total number of the indicators in the selected index level, as shown in (6). Each element of the matrix A_m , denoted by $a'_{ij} = [a_{ij}, \mu] = [a'_{ij-}, a'_{ij+}]$ at (i, j) , is a vector composed of the upper and lower bounds of the interval determined by the interval midpoint a_{ij} and the width μ . Letting $a'_{ij} = [a_{ij}, \mu] = [a'_{ij-}, a'_{ij+}]$, the element can be obtained according to the following rules.

$$A_m = \begin{bmatrix} a'_{11} & a'_{12} & \cdots & a'_{1n} \\ a'_{21} & a'_{22} & \cdots & a'_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a'_{n1} & a'_{n2} & \cdots & a'_{nn} \end{bmatrix} \quad (6)$$

When the i -th index is thought to be more important than the j -th one in the selected index level, resulting in $a_{ij} \geq 1, i \neq j$, the element is determined by

$$a'_{ij} = [a'_{ij-}, a'_{ij+}] = \begin{cases} a'_{ij-} = a_{ij} - \mu \\ a'_{ij+} = a_{ij} + \mu \end{cases} \quad (7)$$

Oppositely, if the j -th index is thought to be more important than the i -th one, which gives rise to $1/a_{ij} \geq 1, j \neq i$, the element is computed as

$$a'_{ij} = [a'_{ij-}, a'_{ij+}] = \begin{cases} a'_{ij-} = 1/(a_{ij} + \mu) \\ a'_{ij+} = 1/(a_{ij} - \mu) \end{cases} \quad (8)$$

4.3 The eigenvalue decomposition of the judgment matrix

In this phase, two operators should be carried out on the constructed matrix A_m , the consistency test and the eigenvalue decomposition.

The qualitative consistency test of A_m is passed only when it satisfies the equation (10) in the condition of the equation (9). If the test fails, the contradiction information will be transmitted to the experts. Then, the procedure returns to the step 2, where the expert will decide the score again according to the feedback information.

$$\begin{cases} a'_{ij} > a'_{ij} & (i \neq l, k \neq l) \\ a'_{ij} > a'_{kl} \end{cases} \quad (9)$$

$$a'_{ij} > a'_{kl} \quad (i \neq l, k \neq l) \quad (10)$$

Next, the eigenvalues of the matrix A_m passing through the above test is derived by an improved power method. For convenience, we use k to denote the iteration number. The initial vector $X^{(0)}$ is a randomly chosen non-zero vector, and its updated version for the k -th iteration is represented by $X^{(k)}$. The iterative vector is denoted by $Y^{(k)}$ for the k -th iteration. With the introduced notations, the improved power method is described as follows.

4.3.1. Initialization

The vectors $X^{(0)}$ and $Y^{(0)}$ are initialized by

$$\begin{cases} X^{(0)} = (x_1^{(0)}, x_2^{(0)}, \dots, x_n^{(0)}) \\ \quad = ([x_{1-}^{(0)}, x_{1+}^{(0)}], [x_{2-}^{(0)}, x_{2+}^{(0)}], \dots, [x_{n-}^{(0)}, x_{n+}^{(0)}]) \\ k = 0 \\ m_0 = \|X^{(0)}\| = \max_i |x_i^{(0)}| = \max_i |x_i^{-(0)}| \\ Y^{(0)} = X^{(0)}/m_0 \end{cases} \quad (11)$$

where m_0 is a temporary parameter, and $\|\cdot\|_q$ stands for l_q norm, and $|\cdot|$ is an absolute value operator.

4.3.2. Iteration

The vector $X^{(k+1)}$ for the $k+1$ -th iteration and the corresponding convergence factor m_1 can be obtained by the iterative rule

$$\begin{cases} X^{k+1} = AY^{(k)} = \sum_{r=1}^n A_{ir} y_r^{(k)} = \sum_{r=1}^n [a_{ir-}, a_{ir+}] [y_{r-}^{(k)}, y_{r+}^{(k)}] \\ \quad = ([\sum_{r=1}^n a_{ir-} y_{r-}^{(k)}, \sum_{r=1}^n a_{ir+} y_{r+}^{(k)}]) \\ m_1 = \|X^{(k+1)}\| = \max_i \{x_i^{-(k+1)}\} \end{cases} \quad (12)$$

4.3.3. The first normalization

In this step, the iterative vector $Y^{(k+1)}$ for the $k+1$ -th iteration and the temporary parameters m_2 and m_3 are determined by

$$\begin{cases} Y^{(k+1)} = X^{(k+1)}/m_1 \\ m_2 = \|Y^{(k+1)}\| = \max_i \{y_i^{-(k+1)}\} \\ m_3 = (m_1 + m_2)/2 \end{cases} \quad (13)$$

4.3.4. The second normalization

Here, the iterative vector $Y^{(k+1)}$ is updated again and the temporary parameters m_4 and m_5 are computed by

$$\begin{cases} Y^{(k+1)} = X^{(k+1)}/m_2 \\ m_4 = \|Y^{(k+1)}\| = \max_i \{y_i^{-(k+1)}\} \\ m_5 = m_1 - \frac{(m_2 - m_1) * (m_3 - m_1)}{m_4 + m_1 - m_2 - m_3} \end{cases} \quad (14)$$

4.3.5. Convergence

$$\text{If } |m_5 - m_0| \leq \varepsilon, \quad (15)$$

Where ε is 1×10^{-6} , let $k=k+1$, and $m_0=m_5$, and go to b. Otherwise, continue.

4.3.6. Normalization

The largest eigenvalue and eigenvector, denoted by λ_{\max} and w respectively, are obtained by the normalization operator

$$\begin{cases} w = Y^{(k+1)} / \frac{1}{2} \sum_{i=1}^n (y_{i-}^{(k+1)} + y_{i+}^{(k+1)}) \\ \lambda_{\max} = m_5 \end{cases} \quad (16)$$

Last, we carried out the relative consistency test on the matrix A_m . We use CR to denote the value of the relative consistency test, which is calculated as (17), where the freedom index RI takes values from Table 3, and $CI=..$

Table 3. Freedom index RI

Dimension (n)	1	2	3	4	5	6	7	8	9
RI	0	0	0.58	0.96	1.12	1.24	1.32	1.41	1.45

Generally, the smaller CR , the better the consistency of A_m is. If CR of A_m is less than 0.1, A_m is feasible and passes the consistency test. Otherwise, A_m fails to pass the test, then the procedure goes back to the step (2), and reconstructs the qualified judgment matrix A_m .

$$\begin{cases} CR = CI/RI \\ CI = (\lambda_{\max} - n)/(n-1) \end{cases} \quad (17)$$

4.4 The generation of total weights

For N experts, by repeating the above operations, we can obtain the N eigenvectors $w_i, i=1, \dots, N$. Then, we calculate the average value of these eigenvectors as

$$\bar{w}_2 = \frac{1}{N} \sum_{i=1}^n w_i \quad (18)$$

and use it as the comprehensive weight of the selected level index.

In the same way, the comprehensive weight vector of the third index level can be obtained. Last, the weight vector for the first index level, i.e., the total weight, is written as,

$$\bar{w}_1 = \bar{w}_2 * \bar{w}_3 \quad (19)$$

where \bar{w}_1 indicates that each dimension of \bar{w}_2 is multiplied by the corresponding dimension of \bar{w}_3 .

5. Case Study

5.1 Case 1

In order to verify the effectiveness of the qualitative

consistency test in the improved interval AHP, a case study has been presented through the second level index of the CP based ESMS. A power grid expert has been selected to score for the index. The results are shown in Table 4 and Table 5.

In Table 4, S1 represents a no-qualitative consistency test, and S2 represents a qualitative consistency test.

We can infer from Table 4 and 5 as follows.

- 1) In Table 4, considering the expert scoring with no-qualitative consistency test, the midpoint of the important interval of A_{13} comparing to A_{11} is 5, that can be depicted as $A_{13}-A_{11}=5$. And the midpoint of the importance interval of A_{12} comparing to A_{11} is 4, that is depicted as $A_{12}-A_{11}=4$. According to the basic mathematical rule, $A_{13}-A_{12}=1$, that means the importance of A_{13} comparing to A_{12} is around 1. However, from Table 4, the midpoint of the importance of A_{13} comparing to A_{12} is 4, which is contradicted with the result from the basic mathematical rule. So, the inconsistent problem arises.
- 2) In case of a qualitative consistency test, the inconsistent problems are found and solved e.g. in table 4, the expert scoring in case of S2 has been modified through the feedback information, the importance interval of A_{12} comparing to A_{11} is changed from [3.5, 4.5] to [2.5, 3.5], and the importance interval of A_{13} comparing to A_{12} is changed from [3.6, 4.4] to [1.6, 2.4]. The midpoint of the importance interval of comparison among A_{11}, A_{12} and A_{13} , that can be depicted as $A_{13}-A_{11}=5, A_{12}-A_{11}=3$, and $A_{13}-A_{11}=2$. Now it completely matches the basic theory of mathematics.
- 3) In Table 5, there are less difference between the largest eigenvalue and the weight. In partial, it has the same trend of the weight, that is $W_{A_{11}} < W_{A_{12}} < W_{A_{13}}$. Therefore, it is obvious that the interval AHP has strongly fuzzy,

Table 4. Scores of power grid expert

A_1	S1			S2		
	A_{11}	A_{12}	A_{13}	A_{11}	A_{12}	A_{13}
A_{11}	[1.0, 1.0]	[0.222, 0.286]	[0.175, 0.233]	[1.0, 1.0]	[0.285, 0.4]	[0.175, 0.233]
A_{12}	[3.5, 4.5]	[1.0, 1.0]	[0.227, 0.278]	[2.5, 3.5]	[1.0, 1.0]	[0.417, 0.625]
A_{13}	[4.3, 5.7]	[3.6, 4.4]	[1.0, 1.0]	[4.3, 5.7]	[1.6, 2.4]	[1.0, 1.0]

Table 5. Weight of the second level index

A_1	S1		S2	
	λ_{\max}	W	λ_{\max}	W
A_{11}		[0.0838, 0.0948]		[0.1018, 0.1208]
A_{12}	[0.2363, 0.2530]	[0.3256, 0.3356]	[0.2600, 0.2750]	[0.3201, 0.3269]
A_{13}		[0.5796, 0.5806]		[0.5590, 0.5713]

Table 6. Scores of AHP and improved interval AHP

A_1	AHP			Improved Interval AHP		
	A_{11}	A_{12}	A_{13}	A_{11}	A_{12}	A_{13}
A_{11}	1.0	0.25	0.2	[1.0, 1.0]	[0.285, 0.4]	[0.175, 0.233]
A_{12}	4.0	1.0	0.25	[2.5, 3.5]	[1.0, 1.0]	[0.417, 0.625]
A_{13}	5.0	4.0	1.0	[4.3, 5.7]	[1.6, 2.4]	[1.0, 1.0]

and S2 is convenient to the expert scoring.

5.2 Case 2

In order to verify the convergence speed of the eigenvector in the improved interval AHP, a case is presented to compare between AHP and improved interval AHP. The input data of the improved interval AHP is the power grid expert scoring, and the AHP of which is the interval midpoint of the power grid expert scoring. The results are shown in Table 6-7, where S3 is a AHP and S4 is an improved interval AHP.

We can find the following rules.

In Table 6, the input data of AHP is the interval midpoint of the improved interval AHP. The results of the AHP belong to the interval range of the improved interval AHP in Table 7, includes CR, the largest eigenvalue and the weights. So, there is a good consistent between AHP and the improved interval AHP.

The convergence of the improved interval AHP is better than that of AHP, for it need less iterations of calculation. And the calculation speed of the improved interval AHP is faster too.

5.3 Case 3

A CP based ESMS for a large city in the Southern China is selected for a case study. Some experts are invited to guide the construction of this ESMS, who possesses knowledge in the three different fields of the grid enterprises, industrial users and commercial users. For the application of evaluation index of CP based ESMS, the expert scorings are listed in Table 8-11.

According to the calculation process in the section 4, the results of those selected expert scoring are shown in table 12 and figure.3, including the weight of the second level index W_{10} and the total index W_1 .

We can infer the following results.

Table 7. Weight of S3 and S4

	S3			S4		
	A_{11}	A_{12}	A_{13}	A_{11}	A_{12}	A_{13}
CR		-2.6577			[-2.6607, -2.6461]	
λ_{max}		0.2425			[0.2600, 0.2750]	
W	0.1097	0.3258	0.5645	[0.1018,0.1208]	[0.3201,0.3269]	[0.5590, 0.5713]
Time(s)		1.87			1.5	
Iterations		9			5	

Table 8. The expert scoring of the first level index A_1

	A_1	A_{11}	A_{12}	A_{13}
	Grid Enterprise	A_{11}	[1.0, 1.0]	[0.285, 0.4]
A_{12}		[2.5, 3.5]	[1.0, 1.0]	[0.417, 0.625]
A_{13}		[3.4, 4.6]	[1.6, 2.4]	[1.0, 1.0]
Industrial User	A_{11}	[1.0, 1.0]	[0.4, 0.667]	[1.8, 2.2]
	A_{12}	[1.5, 2.5]	[1.0, 1.0]	[3.5, 4.5]
	A_{13}	[0.455, 0.556]	[0.222, 0.285]	[1.0, 1.0]
Commercial User	A_{11}	[1.0, 1.0]	[4.5, 5.5]	[2.4, 3.6]
	A_{12}	[0.182, 0.222]	[1.0, 1.0]	[0.455, 0.556]
	A_{13}	[0.278, 0.416]	[1.8, 2.2]	[1.0, 1.0]

Table 9. The expert scoring of the second level index A_{11}

	A_{11}	A_{111}	A_{112}	A_{113}	A_{114}	A_{115}	A_{116}
	Grid Enter prise	A_{111}	[1.0, 1.0]	[0.909, 1.111]	[0.196, 0.204]	[0.189, 0.213]	[0.278, 0.417]
A_{112}		[0.9, 1.1]	[1.0, 1.0]	[0.238, 0.263]	[0.244,0.256]	[0.455, 0.556]	[0.833, 1.25]
A_{113}		[4.9, 5.1]	[3.8, 4.2]	[1.0, 1.0]	[0.833, 1.25]	[1.8, 2.2]	[2.9, 3.1]
A_{114}		[4.7, 5.3]	[3.9, 4.1]	[0.8, 1.2]	[1.0, 1.0]	[1.7, 2.3]	[2.8, 3.2]
A_{115}		[2.4, 3.6]	[1.8, 2.2]	[0.455, 0.556]	[0.435, 0.588]	[1.0, 1.0]	[0.9, 1.1]
A_{116}		[1.7, 2.3]	[0.8, 1.2]	[0.323, 0.345]	[[0.313, 0.357]	[[0.909, 1.111]	[1.0, 1.0]
Industrial User	A_{111}	[1.0, 1.0]	[1.5, 2.5]	[1.2, 2.8]	[0.6, 1.4]	[4.2, 5.8]	[3.6, 4.4]
	A_{112}	[0.4, 0.667]	[1.0, 1.0]	[0.5, 1.5]	[0.833, 1.25]	[3.2, 4.8]	[2.6, 3.4]
	A_{113}	[0.357, 0.833]	[0.667, 2.0]	[1.0, 1.0]	[0.417, 0.625]	[2.1, 3.9]	[1.5, 2.5]
	A_{114}	[0.714, 1.667]	[0.8, 1.2]	[1.6, 2.4]	[1.0, 1.0]	[3.3, 4.7]	[2.6, 3.4]
	A_{115}	[0.172, 0.238]	[0.208, 0.313]	[0.256, 0.476]	[0.213, 0.303]	[1.0, 1.0]	[0.455, 0.556]
	A_{116}	[0.227, 0.278]	[0.294, 0.385]	[0.4, 0.667]	[0.294, 0.385]	[1.8, 2.2]	[1.0, 1.0]
Commercia l User	A_{111}	[1.0, 1.0]	[0.4, 0.667]	[0.455, 0.556]	[1.3, 2.7]	[2.4, 3.6]	[3.7, 4.3]
	A_{112}	[1.5, 2.5]	[1.0, 1.0]	[0.833, 1.25]	[2.8, 3.2]	[2.9, 3.1]	[3.1, 4.9]
	A_{113}	[1.8, 2.2]	[0.8, 1.2]	[1.0, 1.0]	[2.6, 3.4]	[3.7, 4.3]	[4.7, 5.3]
	A_{114}	[0.370, 0.769]	[0.313, 0.357]	[0.294, 0.385]	[1.0, 1.0]	[0.8, 1.2]	[1.4, 2.6]
	A_{115}	[0.278, 0.417]	[0.323, 0.345]	[0.233, 0.270]	[0.833, 1.25]	[1.0, 1.0]	[0.9, 1.1]
	A_{116}	[0.233, 0.270]	[0.204, 0.323]	[0.189, 0.213]	[0.385, 0.714]	[0.909, 1.111]	[1.0, 1.0]

Table 10. The expert scoring of the second level index A_{12}

A_{12}		A_{121}	A_{122}	A_{123}	A_{124}	A_{125}	A_{126}
Grid Enterprise	A_{121}	[1.0, 1.0]	[0.9, 1.1]	[0.227, 0.278]	[0.233, 0.270]	[0.238, 0.263]	[0.244, 0.256]
	A_{122}	[0.909, 1.111]	[1.0, 1.0]	[0.192, 0.208]	[0.196, 0.204]	[0.313, 0.357]	[0.323, 0.345]
	A_{123}	[3.6, 4.4]	[4.8, 5.2]	[1.0, 1.0]	[0.8, 1.2]	[1.5, 2.5]	[2.6, 3.4]
	A_{124}	[3.7, 4.3]	[4.9, 5.1]	[0.833, 1.25]	[1.0, 1.0]	[1.4, 2.6]	[1.2, 2.8]
	A_{125}	[3.8, 4.2]	[2.8, 3.2]	[0.4, 0.667]	[0.385, 0.714]	[1.0, 1.0]	[1.3, 2.7]
	A_{126}	[3.9, 4.1]	[2.9, 3.1]	[0.294, 0.385]	[0.357, 0.833]	[0.370, 0.769]	[1.0, 1.0]
Industrial User	A_{121}	[1.0, 1.0]	[4.6, 5.4]	[0.9, 1.1]	[1.5, 2.5]	[2.8, 3.2]	[0.833, 1.25]
	A_{122}	[0.185, 0.217]	[1.0, 1.0]	[0.208, 0.313]	[0.286, 0.4]	[0.294, 0.385]	[0.185, 0.217]
	A_{123}	[0.909, 1.111]	[3.2, 4.8]	[1.0, 1.0]	[1.3, 2.7]	[1.5, 2.5]	[0.455, 0.556]
	A_{124}	[0.4, 0.667]	[2.5, 3.5]	[0.370, 0.769]	[1.0, 1.0]	[0.9, 1.1]	[0.270, 0.435]
	A_{125}	[0.313, 0.357]	[2.6, 3.4]	[0.4, 0.667]	[0.909, 1.111]	[1.0, 1.0]	[0.222, 0.286]
	A_{126}	[0.8, 1.2]	[4.6, 5.4]	[1.8, 2.2]	[2.3, 3.7]	[3.5, 4.5]	[1.0, 1.0]
Commercial User	A_{121}	[1.0, 1.0]	[0.833, 1.25]	[3.5, 4.5]	[2.1, 3.9]	[3.2, 4.8]	[1.4, 2.6]
	A_{122}	[0.8, 1.2]	[1.0, 1.0]	[3.2, 4.8]	[2.6, 3.4]	[4.3, 5.7]	[1.3, 2.7]
	A_{123}	[0.222, 0.286]	[0.208, 0.313]	[1.0, 1.0]	[0.909, 1.111]	[1.8, 2.2]	[0.313, 0.357]
	A_{124}	[0.256, 0.476]	[0.294, 0.385]	[0.9, 1.1]	[1.0, 1.0]	[1.5, 2.5]	[0.357, 0.833]
	A_{125}	[0.208, 0.313]	[0.175, 0.233]	[0.455, 0.556]	[0.4, 0.667]	[1.0, 1.0]	[0.323, 0.345]
	A_{126}	[0.385, 0.714]	[0.370, 0.769]	[2.8, 3.2]	[1.2, 2.8]	[2.9, 3.1]	[1.0, 1.0]

Table 11. The expert scoring of the second level index A_{13}

A_{13}		A_{131}	A_{132}	A_{133}	A_{134}	A_{135}	A_{136}
Grid Enterprise	A_{131}	[1.0, 1.0]	[0.8, 1.2]	[1.6, 2.4]	[2.5, 3.5]	[3.6, 4.4]	[4.8, 5.2]
	A_{132}	[0.833, 1.25]	[1.0, 1.0]	[0.9, 1.1]	[1.6, 2.4]	[2.6, 3.4]	[3.4, 4.6]
	A_{133}	[0.417, 0.625]	[0.909, 1.111]	[1.0, 1.0]	[0.9, 1.1]	[1.8, 2.2]	[2.4, 3.6]
	A_{134}	[0.286, 0.4]	[0.417, 0.625]	[0.909, 1.111]	[1.0, 1.0]	[0.8, 1.2]	[1.6, 2.4]
	A_{135}	[0.227, 0.278]	[0.294, 0.385]	[0.455, 0.556]	[0.833, 1.25]	[1.0, 1.0]	[0.8, 1.2]
	A_{136}	[0.192, 0.208]	[0.217, 0.294]	[0.278, 0.417]	[0.417, 0.625]	[0.833, 1.25]	[1.0, 1.0]
Industrial User	A_{131}	[1.0, 1.0]	[1.4, 2.6]	[1.5, 2.5]	[2.6, 3.4]	[0.4, 0.667]	[0.455, 0.556]
	A_{132}	[0.385, 0.714]	[1.0, 1.0]	[0.9, 1.1]	[1.8, 2.2]	[0.294, 0.385]	[0.303, 0.370]
	A_{133}	[0.4, 0.667]	[0.909, 1.111]	[1.0, 1.0]	[1.2, 2.8]	[0.208, 0.313]	[0.217, 0.294]
	A_{134}	[0.294, 0.385]	[0.455, 0.556]	[0.357, 0.833]	[1.0, 1.0]	[0.192, 0.208]	[0.196, 0.204]
	A_{135}	[1.5, 2.5]	[2.6, 3.4]	[3.2, 4.8]	[4.8, 5.2]	[1.0, 1.0]	[0.6, 1.4]
	A_{136}	[1.8, 2.2]	[2.7, 3.3]	[3.4, 4.6]	[4.9, 5.1]	[0.714, 1.667]	[1.0, 1.0]
Commercial User	A_{131}	[1.0, 1.0]	[1.5, 2.5]	[2.6, 3.4]	[3.8, 4.2]	[0.8, 1.2]	[0.909, 1.111]
	A_{132}	[0.5, 0.667]	[1.0, 1.0]	[1.5, 2.5]	[2.6, 3.4]	[0.455, 0.556]	[0.294, 0.385]
	A_{133}	[0.294, 0.385]	[0.4, 0.667]	[1.0, 1.0]	[1.2, 2.8]	[0.476, 0.526]	[0.313, 0.357]
	A_{134}	[0.238, 0.263]	[0.294, 0.385]	[0.357, 0.833]	[1.0, 1.0]	[0.227, 0.278]	[0.196, 0.204]
	A_{135}	[0.833, 1.25]	[1.8, 2.2]	[1.9, 2.1]	[3.6, 4.4]	[1.0, 1.0]	[0.370, 0.769]
	A_{136}	[0.9, 1.1]	[2.6, 3.4]	[2.8, 3.2]	[4.9, 5.1]	[1.3, 2.7]	[1.0, 1.0]

Table 12. The weight of the second level index W_{10}

W_{10}	Power Grid	Industrial User	Commercial User
A_{11}	[0.1129, 0.1343]	[0.2746, 0.3058]	[0.6323, 0.6493]
A_{12}	[0.3484, 0.3541]	[0.5419, 0.5889]	[0.1233, 0.1285]
A_{13}	[0.5173, 0.5330]	[0.1365, 0.1532]	[0.2221, 0.2444]

1) In W_{10} , it is obvious that the cognition to the importance of the cloud platforms index is different for different experts since they come from different fields. An expert in the field of power grid enterprises believes that the most important index is virtual ability due to the maximum weight of A_{13} in table 12 ($W_{A_{13}} > W_{A_{12}} > W_{A_{11}}$). An expert from industrial users focuses on the fast response capability of memory architecture because of the maximum weight of A_{12} in table 12 ($W_{A_{12}} > W_{A_{11}} > W_{A_{13}}$). An expert from commercial users thinks that the switching capacity of network architecture is the most important according to the maximum weight of A_{11} in

table 12 ($W_{A_{11}} > W_{A_{13}} > W_{A_{12}}$). Therefore, we can infer that the CP based ESMS need to be promoted in future, and different users have had different service of the product.

- In W_1 , it is obvious that different users have different performance requirements for cloud platform, when the three biggest weights of W_1 are selected. The most important index of W_1 to power grid enterprise is A_{111} due to the maximum of $W_{A_{111}}$ in Fig. 3 ($W_{A_{111}} > W_{A_{122}} > W_{A_{133}}$). The industrial user of which is A_{126} because of the weight of $W_{A_{126}}$ in Fig. 3 ($W_{A_{126}} > W_{A_{123}} > W_{A_{121}}$). The commercial user is A_{113} according to the weight of $W_{A_{113}}$ in Fig. 3 ($W_{A_{113}} > W_{A_{112}} > W_{A_{111}}$). Therefore, we can infer that the product performance of the cloud computing needs the strong pertinence to different customer.
- According to the characteristics of cloud computing, and considering the type of cloud computing, we infer that

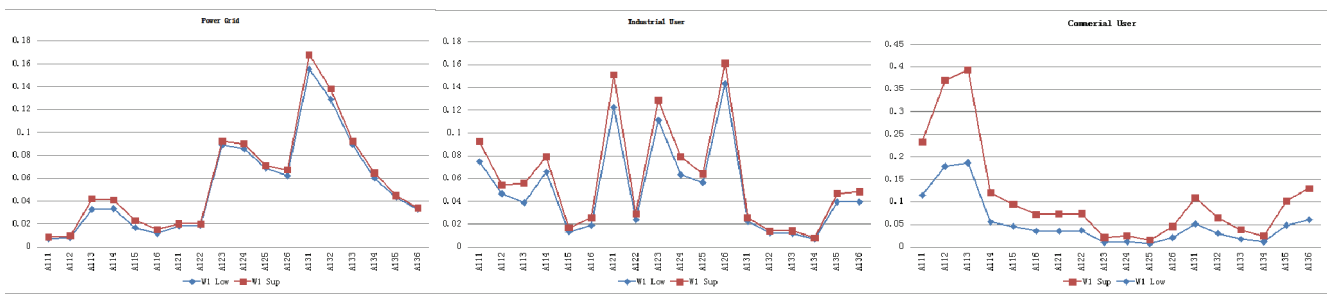


Fig. 3. Weight of the total index W_1

the commercial users are suitable for public clouds, the power grid enterprise users are suitable for private cloud, and the industrial users are suitable for hybrid cloud.

6. Conclusions

An improved interval AHP method is not only effective to solve the problem of the multi-level attribute and multi-decision making, but also provides a quantitative method for the assessment of CP based ESMS, and provides a strong basis for the construction of this platform.

The electric safety issue becomes more and more important and complicated with the integration of large-scale new devices into distribution system, such as electric vehicles (EVs), distributed generations (DGs) and energy storage devices, micro-grids, smart home,. A high standard for the management level and quality of service of ESMS are required, the challenges and opportunities will be coming in many aspects, including technology, personnel and management. Those related topics are worthy of further study.

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