# A multilevel fuzzy analysis model of higher education teaching quality

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#### Abstract

To improve the overall quality of teachers and enhance the teaching ability as well as the teaching quality of institutes of higher learning for the purpose of nurturing high-tech talents, this paper proposes a multilevel fuzzy analysis model of higher education teaching quality based on fuzzy system theory. It constructs a multilevel evaluation system and acquires the fuzzy evaluation set of teaching quality and fuzzy value of a quantity. Through calculation we can get the fuzzy membership between teaching quality and fuzzy evaluation set. Fuzzy membership is applied to standardization according to different types and scales of indicators to get the integrated weighted fuzzy membership. This will realize the evaluation on teaching quality of institutes of higher learning and helps to increase the overall quality of teachers. A case study is introduced to prove the efficacy of the model and the algorithm.

Keywords: higher education; teaching quality; fuzzy analysis; multiple attribute decision making; evaluation model

#### 1 Introduction

With the development of science and technology, modern society has yearning for high-tech talents. Teachers of institutes of higher learning play an important role in promoting the overall quality of high-tech talents. Management on teaching quality is given priority by the school as teaching becomes the most essential task. Therefore, it is significant to carry out a multilevel evaluation on teaching quality and standardize management on teaching in order to increase the efficiency and improve the quality of the school. Meanwhile, it also serves as guidance for teachers to improve their teaching ability [1-3].

Many scholars and experts have already studied the teaching quality of higher education and registered some achievements in terms of the evaluation system and evaluation method [4-8]. Based on previous researches and the fuzzy mathematic theory [9-12], this paper proposes an improved multilevel evaluation system of teaching quality

of higher education. It constructs a multilevel fuzzy analysis model through fuzzy evaluation set and fuzzy membership of different indicators.

## 2 Multilevel evaluation system of the teaching quality of higher education

Teaching quality of higher education is mainly presented in two perspectives. One is the teacher's ability to impart knowledge. The other is student's ability to learn and apply knowledge. This paper focuses on whether the evaluation is scientific, objective, with target and reasonable based on surveys and tailored to real situation. It constructs a three-layer evaluation system, namely, the system layer, the criterion layer and the indicator layer. The system layer consists of five perspectives: teaching quality, teaching ability, student's learning ability, student's research ability and student's other ability. The layers and the indicators are shown in Table 1.

TABLE 1 Multilevel evaluation system of the teaching quality of higher education

system layer	criterion layer	indicator layer	Type of indicator
Evaluation system of the teaching quality of higher education <i>C</i>	Teaching quality $C_{\scriptscriptstyle 1}$	Professional knowledge $c_{11}$	qualitative indicator
		Course planning $c_{12}$	qualitative indicator
		Classroom teaching effect $c_{13}$	qualitative indicator
		Advanced teaching method $c_{14}$	qualitative indicator
	Teaching ability $C_2$	Project number $c_{21}$	quantitative indicator

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	Teaching ability $C_2$	Number of papers $c_{22}$	quantitative indicator
	reaching ability $C_2$	Number of awards $c_{23}$	quantitative indicator
Evaluation system of		Number of researches $c_{24}$	quantitative indicator
the teaching quality of		Professional knowledge learning ability $c_{31}$	qualitative indicator
higher education	Student's learning	Creativity $c_{32}$	qualitative indicator
C	ability $C_3$	Practice ability $c_{33}$	qualitative indicator
		Proportion of award $c_{34}$	quantitative indicator
	Student's research ability $C_4$	Number of participations in scientific and technological design $c_{41}$	quantitative indicator
		Number of awards for scientific and technological design $c_{42}$	quantitative indicator
		Number of patent application and papers $c_{43}$	quantitative indicator
		Training on expanding ability $c_{51}$	qualitative indicator
	Student's other ability $C_5$	Training on environmental adaptability $c_{52}$	qualitative indicator
	- 3	Training on the ability to solve problems independently $c_{53}$	qualitative indicator

## 3 A multilevel fuzzy analysis model of higher education teaching quality

#### 3.1 MULTILEVEL EVALUATION SET OF HIGHER EDUCATION TEACHING QUALITY

The analysis of teaching quality of higher education is a complicated decision-making process. It requires a reasonable description and effective measurement of all evaluation results, which indicates the necessity to have a multilevel evaluation set. Therefore, we divide the teaching quality into five grades and the evaluation set is expressed

$$U = \{U_1, U_2, U_3, U_4, U_5\}$$
 (1)

In the expression, refers to excellence, refers to good, refers to medium, refers to pass and refers to fail.

There are two types of evaluation, the qualitative and the quantitative ones. This paper discusses them respecttively. For qualitative indicators, we can get the membership degree through fuzzy membership function.

$$\varphi_{i}^{j} = f_{U_{i}^{j}}\left(v\left(x_{i}\right) \otimes \left[v_{min}\left(U_{i}^{j}\right), v_{max}\left(U_{i}^{j}\right)\right]\right) \tag{2}$$

In the expression, the membership degree of qualitative indicator is shown as Table 2.

TABLE 2 Value of fuzzy membership of qualitative indicator

membership $oldsymbol{arphi}_i$	Explanation of qualitative indicator
0	No membership
0.2	Poor membership
0.4	Membership
0.6	Good membership
0.8	Fair membership
1.0	Absolute membership
0.1, 0.3, 0.5, 0.7, 0.9	In between

If it is a quantitative evaluation indicator, the calculation of fuzzy membership refers to 3.3 based on fuzzy distance

## 3.2 WEIGHT ALLOCATION OF MULTILEVEL EVALUATION INDICATORS OF TEACHING QUALITY

Many indicators need to be taken into account in the evaluation as they have different influence on teaching quality. Traditional AHP, through capable of acquiring the weight of indicators [13], is subjective in the course of producing weight. Therefore, this paper adopts an indicator negotiation weighing method based on AHP to be more objective.

Suppose an expert uses 1-9 ratio scale to evaluate indicator. The acquired weight of indicator according to AHP is and fits and. The negotiated weight of indicator is:

$$w_{i}^{\Theta} = \frac{1}{M} \sum_{n=1}^{M} w_{i}^{p} + \rho_{i} * \left| \max_{1 \le p \le M} \left( w_{i}^{p} \right) - \min_{1 \le p \le M} \left( w_{i}^{p} \right) \right|$$
(3)

In the expression, refers to the number of experts; refers to negotiation coefficient; is the number of evaluation indicators.

Then, the allocated model of weight of indicator is::

$$\overline{w}_{i} = \frac{w_{i}^{\Theta}}{\sum_{i=1}^{n_{c}} w_{i}^{\Theta}}$$

$$(4)$$

3.3 FUZZY MEMBERSHIP OF TEACHING QUALITY Suppose the indicator  $c_i$  is a quantitative indicator, its value of a quantity is  $V(c_i) = \left[v_{left}(c_i), v_{right}(c_i)\right]$ .

If the ideal fuzzy value of a quantity field for teaching quality level j is  $V^0\left(c_i\right) = \left[v_{left}^0\left(c_i\right), v_{right}^0\left(c_i\right)\right]$ , then the fuzzy distance  $D_i^j\left(P\right)$  between indicator  $c_i$  and teaching quality level j is

$$D_{i}^{j}(P) = \left( \left| v_{left}^{0}(c_{i}) - v_{left}(c_{i}) \right|^{P} + \left| v_{right}^{0}(c_{i}) - v_{right}(c_{i}) \right|^{P} \right)^{\frac{1}{P}} / 2^{\frac{1}{P}}$$
 (5)

In particular, if P = 1, the fuzzy distance is the Hamming distance.  $D_i^j(1)$  Is expressed as:

$$D_{i}^{j}(1) = \left( \left| v_{left}^{0}(c_{i}) - v_{left}(c_{i}) \right| + \left| v_{right}^{0}(c_{i}) - v_{right}(c_{i}) \right| \right) / 2$$
(6)

If P = 2, the fuzzy distance is the Euclidean distance.  $D_i^j(2)$  Is expressed as:

$$D_{i}^{j}(2) = \left( \left| v_{left}^{0}(c_{i}) - v_{left}(c_{i}) \right|^{2} + \left| v_{right}^{0}(c_{i}) - v_{right}(c_{i}) \right|^{2} \right)^{\frac{1}{2}} / \sqrt{2}$$
(7)

As quantitative indicator can be divided into positive and adverse indicators, they are equipped with different fuzzy distance. To unify the evaluation standard, we need to apply  $D_i^j(P)$  to standardization.

If the quantitative indicator has positive distance, then the standardized fuzzy membership  $\varphi_i^j(P)$  is:

$$\varphi_{i}^{j}(P) = \begin{cases}
\frac{D_{i}^{j}(1) - \min_{1 \leq j \leq N} \left(D_{i}^{j}(1)\right)}{\max_{1 \leq j \leq N} \left(D_{i}^{j}(1)\right) - \min_{1 \leq j \leq N} \left(D_{i}^{j}(1)\right)} & P = 1 \\
\frac{D_{i}^{j}(2) - \min_{1 \leq j \leq N} \left(D_{i}^{j}(2)\right)}{\max_{1 \leq j \leq N} \left(D_{i}^{j}(2)\right) - \min_{1 \leq j \leq N} \left(D_{i}^{j}(2)\right)} & P = 2
\end{cases}$$
(8)

N refers to the number of teaching quality level.

If the quantitative indicator has adverse distance, then the standardized fuzzy membership  $\varphi_i^j(P)$  is:

$$\varphi_{i}^{j}(P) = \begin{cases}
\frac{\max_{1 \leq j \leq N} (D_{i}^{j}(1)) - D_{i}^{j}(1)}{\max_{1 \leq j \leq N} (D_{i}^{j}(1)) - \min_{1 \leq j \leq N} (D_{i}^{j}(1))} & P = 1 \\
\frac{\max_{1 \leq j \leq N} (D_{i}^{j}(2)) - D_{i}^{j}(2)}{\max_{1 \leq j \leq N} (D_{i}^{j}(2)) - \min_{1 \leq j \leq N} (D_{i}^{j}(2))} & P = 2
\end{cases}$$
(9)

According to the physical significance of the fuzzy distance, when indicator  $C_i$  of the evaluation object is closer to teaching quality level j, the fuzzy distance is smaller. When indicator  $C_i$  of the evaluation object is less close to teaching quality level j, the fuzzy distance is bigger.

Given that different evaluation indicator has different weight  $W_i$ , we can get the weighted fuzzy membership

 $\varphi_i^j(DL)$  between quantitative indicator  $C_i$  and teaching quality level j:

$$\varphi_{i}^{j}(DL) = \sum_{i=1}^{n_{c1}} (\overline{w}_{i} * (1 - \varphi_{i}^{j}(P))).$$
 (10)

In the expression,  $n_{c1}$  refers to the number of quantitative indicators in the same criterion layer.

Similarly, the weighted fuzzy membership  $\varphi_i^{^J}ig(DXig)$ between qualitative indicator  $c_i$  and teaching quality level

$$\varphi_i^j(DX) = \sum_{i=1}^{n_{c2}} (\overline{w}_i * \varphi_i^j), \tag{11}$$

n the expression,  $n_{c2}$  refers to the number of qualitative indicator in the same criterion layer.

And there is 
$$n_{c1} + n_{c2} = n_c$$
.

Therefore, the unified weighed fuzzy membership in the criterion layer is:

$$\varphi_{i}^{j} = \varphi_{i}^{j}(DL) + \varphi_{i}^{j}(DX) = \sum_{i=1}^{n_{c1}} (\overline{w}_{i} * (1 - \varphi_{i}^{j}(P))) + \sum_{i=1}^{n_{c2}} (\overline{w}_{i} * \varphi_{i}^{j})$$
(12)

$$\varphi^{j} = \sum_{i=1}^{n_z} \left( w_i * \varphi_i^{j} \right) \tag{13}$$

In the expression,  $n_z$  refers to the number of criterion layer in the evaluation system.

According to optimization principle of membership, when

$$\varphi^{j} = max(\varphi^{1}, \dots, \varphi^{k}, \dots, \varphi^{N}) = \varphi^{t}$$
(14)

It indicates that the evaluation object is the closest to teaching quality level t.

#### 3.4 MODEL AND ALGORITHM

The multilevel fuzzy analysis model has the following steps:

- Step 1: Construct the multilevel evaluation system for teaching quality of higher education according to Section 2. And get different evaluation sets according to Section 3.1.
- Use expression (3) and (4) to get different Step 2: evaluation rules and weight of indicators according to Section 3.2.
- Step 3: For qualitative indicators, get its fuzzy membership from Table 2;

- Step 5: Get the fuzzy membership between qualitative indicator and teaching quality level according to expression (10) to (12):
- Use expression (13) to get the integrated weighted Step 6: fuzzy membership between evaluation object and teaching quality level. Use expression (14) to acquire the membership degree of the evaluation object based on integrated weighted fuzzy membership.

#### 4 Case study and test

This paper takes teachers of an institute of higher learning as examples to prove the efficacy of the model and the algorithm. Through data collection, statistics and analysis and based on school policies and institutions, we have original data for teaching quality evaluation at hand. Teacher A is chosen as the example. Relevant information about qualitative indicator is shown in Table 3 and about quantitative indicator, Table 4.

TABLE 3 Qualitative indicator of teaching quality

criterion					f	uzzy membershi	D	
layer	layer weight	indicator layer	weight	Excellence	Good	Medium	Pass	Fail
		$c_{11}$	0.25	0.70	0.85	0.30	0.10	0
C	0.314	$c_{12}$	0.30	0.95	0.70	0.20	0.10	0
$C_1$	0.314	$c_{13}$	0.30	0.85	0.80	0.40	0.20	0
		$c_{_{14}}$	0.15	0.60	0.85	0.60	0.30	0
	0.241	$c_{31}^{}$	0.25	0.70	0.85	0.85	0.60	0
$C_3$		$c_{32}$	0.20	0.40	0.60	0.80	0.60	0.30
		$c_{33}$	0.25	0.55	0.75	0.80	0.40	0.10
$C_5$	0.076	$c_{51}$	0.30	0.60	0.80	0.80	0.40	0
		$c_{52}$	0.30	0.60	0.85	0.75	0.30	0
		$c_{53}$	0.40	0.60	0.95	0.85	0.10	0

TABLE 4 Quantitative indicator of teaching quality

criterion wei	:-1.4	indicator	vyoi oht	Ideal value of a quantity field of quantitative indicator					Value of a quantity
	weight	layer	layer weight	Excellence	Good	Medium	Pass	Fail	Evaluation object
		$c_{21}^{}$	0.30	3-5	3	2	1	0	2
C	0.270	$c_{22}^{}$	0.25	5-10	3-5	2-3	1-2	0-1	4
$C_2$	0.270	$c_{23}^{}$	0.30	2-5	1-2	0-1	0-1	0	1
		$c_{24}^{}$	0.15	6-10	4-6	2-4	0-2	0	3
$C_3$	0.241	$c_{34}$	0.30	0.45-1.00	0.30-0.45	0.15-0.30	0.05-0.15	0-0.05	0.35-0.40
		$c_{41}^{}$	0.30	5-10	3-5	1-3	0-1	0	2
$C_4$	0.099	$c_{42}$	0.35	3-5	2-3	1-2	0-1	0	2
		$c_{43}$	0.35	8-15	6-8	4-6	2-4	0-2	10

## Calculate the fuzzy distance of the quantitative indicator in Table 4 according to Section 3.3. The result is shown in Table 5.

TABLE 5 Fuzzy distance of quantitative indicator of teaching quality

Sandi and an Innove	Fuzzy distance of quantitative indicator						
indicator layer	Excellence	Good	Medium	Pass	Fail		
$c_{21}$	1.000	1.000	0	1.000	2.000		
$c_{22}$	1.000	0	1.581	2.549	2.236		
$c_{23}$	2.915	0	0.707	0.707	1.000		
$c_{24}^{}$	3.000	2.236	0	2.236	3.000		
c <sub>34</sub>	0.430	0.050	0.158	0.276	0.350		
$c_{41}$	3.000	2.236	0	1.581	2.000		
$c_{42}$	1.000	0	1.000	1.581	2.000		
C <sub>43</sub>	0	3.162	5.099	7.071	9.055		

## Standardize the fuzzy distance in Table 5 and the result is shown in Table 6

TABLE 6 Standardized fuzzy distance of quantitative indicator

in diagton lavon	Standardized fuzzy distance of quantitative indicator						
indicator layer	Excellence	Good	Medium	Pass	Fail		
$c_{21}^{}$	0.500	0.500	1.000	0.500	0		
$c_{22}$	0.608	1.000	0.380	0	0.123		
$c_{23}$	0	1.000	0.757	0.757	0.657		
$c_{24}^{}$	0	0.255	1.000	0.255	0		
$c_{34}$	0	1.000	0.716	0.405	0.211		
$c_{41}$	0	0.255	1.000	0.473	0.333		
$c_{42}$	0.500	1.000	0.500	0.210	0		
$c_{43}$	1.000	0.651	0.437	0.219	0		

Give weight to qualitative and quantitative indicator and get the fuzzy membership of indicators, as is shown in Table 7

TABLE 7 Fuzzy memberships of indicators

Indiantons	fuzzy members	hip			
Indicators	Excellence	Good	Medium	Pass	Fail
$c_{11}$	0.175	0.213	0.075	0.025	0
$c_{12}$	0.285	0.210	0.060	0.030	0
$c_{13}$	0.255	0.240	0.120	0.060	0
$c_{14}^{}$	0.009	0.128	0.009	0.045	0
$c_{21}$	0.150	0.150	0.300	0.150	0
$c_{22}$	0.152	0.250	0.095	0	0.031
$c_{23}$	0	0.300	0.227	0.227	0.197
$c_{24}$	0	0.038	0.150	0.038	0
c <sub>31</sub>	0.175	0.213	0.213	0.150	0
$c_{32}$	0.080	0.120	0.160	0.120	0.006
$c_{33}$	0.138	0.188	0.200	0.100	0.025
$c_{34}$	0	0.300	0.215	0.122	0.063
$c_{41}$	0	0.077	0.300	0.142	0.100
$c_{42}$	0.175	0.350	0.175	0.074	0
$c_{43}$	0.350	0.228	0.153	0.077	0
c <sub>51</sub>	0.180	0.240	0.240	0.120	0
$c_{52}$	0.180	0.255	0.225	0.009	0
C <sub>53</sub>	0.240	0.380	0.340	0.040	0

According to the fuzzy membership calculation model, we can get the comprehensive fuzzy membership sequence between teacher A and teaching quality level.

There is  $\varphi = (0.501, 0.778, 0.607, 0.323, 0.094)$ . Teacher A reached a good level in terms of teaching quality.

#### **5 Conclusions**

This paper proposes a multilevel fuzzy analysis model of

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higher education teaching quality based on fuzzy system theory. It constructs a multilevel evaluation system and acquires the weight of indicators and analyzes the fuzzy membership of qualitative and quantitative indicators. The integrated weighted fuzzy membership calculation model is given and the membership degree of teaching quality is confirmed based on fuzzy membership. It actually realizes the performance management of teaching quality of higher education on the computer.

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