

ARTICLE

## Safety Risk Assessment of Overturning Construction of Towering Structure Based on Cloud Matter–Element Coupled Model

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### ABSTRACT

Rapid urbanization has led to a surge in the number of towering structures, and overturning is widely used because it can better accommodate the construction of shaped structures such as variable sections. The complexity of the construction process makes the construction risk have certain randomness, so this paper proposes a cloud-based coupled matter-element model to address the ambiguity and randomness in the safety risk assessment of overturning construction of towering structures. In the pretended model, the digital eigenvalues of the cloud model are used to replace the eigenvalues in the matter–element basic element, and calculate the cloud correlation of the risk assessment metrics through the correlation algorithm of the cloud model to build the computational model. Meanwhile, the improved hierarchical analysis method based on the cloud model is used to determine the weight of the index. The comprehensive evaluation scores of the evaluation event are then obtained through the weighted average method, and the safety risk level is determined accordingly. Through empirical analysis, (1) the improved hierarchical analysis method based on the cloud model can incorporate the data of multiple decision-makers into the calculation formula to determine the weights, which makes the assessment results more credible; (2) the evaluation results of the cloud-based matter-element coupled model method are basically consistent with those of the other two commonly used methods, and the confidence factor is less than 0.05, indicating that the cloud-based physical element coupled model method is reasonable and practical for towering structure overturning; (3) the cloud-based coupled element model method, which confirms the reliability of risk level by performing Spearman correlation on comprehensive assessment scores, can provide more comprehensive information of instances compared with other methods, and more comprehensively reflects the fuzzy uncertainty relationship between assessment indexes, which makes the assessment results more realistic, scientific and reliable.

### KEYWORDS

Cloud matter-element model; clouded hierarchical analysis method; towering structure; overturning formwork construction; risk assessment



## 1 Introduction

The concept of risk and risk assessment has a long history; as early as more than 2400 years ago, the Athenians suggested that they could assess risk before making decisions [1]. In the 1930s, risk assessment emerged in the United States, and the U.S. Insurance Management Association first introduced risk assessment into education in the 1960s, which meant that risk assessment became a new discipline [2]. Since then, many scholars have conducted extensive and in-depth research on the types, characteristics, causes, and management of risks in various industries. In the construction process of large and complex structural systems such as towering buildings, various risk factors can interact and couple with each other, leading to safety accidents. The turnover formwork construction is a modern evolution of traditional climbing technology; its relatively unique process and technology can effectively adapt to the diversified variable section construction. This technology has been widely used in barrel, rectangular, and other engineering practices in recent years [3–5]. However, the complexity and uncertainty of the turnover formwork construction plan are still considerably due to the complexity of the structure system, the number of related processes involved, the high technical requirements, and the imperfect research on its safety [6]. Such difficulty leads to serious losses once the safety production accidents of the turnover formwork construction occur. Thus, in order to improve the effectiveness of safety risk assessment in the construction of towering buildings with overturned forms, it is important to combine traditional construction processes with information calculation models for systematic analysis of risks in the construction phase of projects.

In terms of risk assessment methods, existing studies have made important contributions to the safety risk assessment of overturning construction of towering structures. Literature [7] first proposed a new method of accident tree embedding artificial neural network based on the traditional accident tree analysis method through fuzzy set theory and hierarchical analysis. Literature [8] proposed a newly developed hierarchical distributed agent model for network risk assessment inspired by deep learning neural network theory. Literature [9,10] both incorporated the original analytic hierarchy process (AHP) and triangular fuzzy number-based AHP (TFN-AHP) into a geographic information system (GIS) to assess the risk of flood. The purpose of TFN-AHP is to effectively address complex situations where a significant number of experts have a high degree of preference variation. Literature [11] proposed a gray fuzzy material element theory construction risk assessment method based on Delphi method to improve the scientific aspect and effectiveness of construction risk source determination and increase the accuracy of construction risk factor analysis. Literature [12] introduced for the first time a new method of multi-indicator assessment of risk level based on entropy weight normal cloud model to address the situation that the assessment indexes will be affected by multiple factors when the construction risk assessment object itself has great complexity and ambiguity. Literature [13] first developed a stepwise process of fuzzy affiliation function based on the improved fuzzy set theory to determine risk variables, which provides a solution to the lack of historical data and reliance on expert opinions when assessing construction risk. Literature [14] identified the most important factors influencing the formwork systems by applying the relative importance index method through questionnaires and interviews with experts. Literature [15] addressed the problem of insufficient quantitative assessment of tall formwork construction risks based on the improved LEC method (operational conditions hazard rating method) for risk assessment indicators of tall formwork considering the influence of the formwork system itself and human factors.

Through the summary and analysis of the relevant risk assessment methods, it can be found that the commonly used assessment methods can be divided into subjective and objective evaluation methods. Subjective evaluation methods include Delphi method, hierarchical analysis, etc.; the objective ones include gray system evaluation, matter-element extension method [16,17], cloud model, artificial

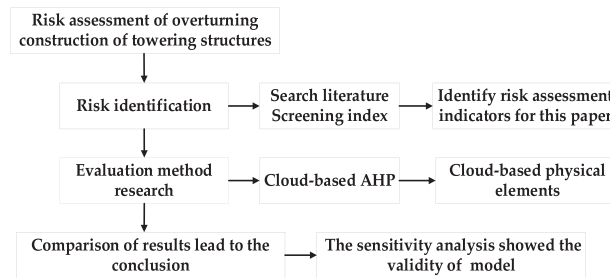
neural network, etc. In complex building construction safety risk assessment, analysis and decision-making are affected by various factors, making the information often present a certain degree of ambiguity and randomness [18,19]. Moreover, the building construction risk assessment is affected by various factors, and the relationship between them is difficult to express quantitatively. Therefore, considering these factors comprehensively is difficult for a single evaluation method. Hierarchical analysis is often used in the evaluation process to quantify the index weights [20]. However, this approach is also inclined to depend on expert opinions, and the fuzziness and incompatibility between evaluation indexes contribute to certain limitations of the above evaluation methods. Cloud model has attracted the attention of many scholars because of their ability to express the ambiguity and randomness of assessment information [21], the advantage of the widely used matter–element theory in recent years lies in its scientific solving of the incompatibility problem of evaluation objects. With the continuous development of theoretical methods, both cloud model and matter-element theory are gradually applied to practical problems, such as fuzzy evaluation, multi-attribute decision making and risk assessment [22]. In the field of uncertain information processing, seeking scientific and reasonable combination assessment methods to analyze and deal with various uncertainties in the process of security risk assessment is the direction and difficulty of scientific assessment.

Therefore, this paper proposed a novel risk assessment method for overturning construction of towering structures. The main contributions of this paper are as follows. (1) An index system for overturning construction of towering structures is established on the basis of overturning construction process to deal with the problem of complex formwork construction technology and numerous construction projects. (2) The cloud model is used to improve the hierarchical analysis method to determine the index weights and avoid the adverse impact of over-reliance on expert opinions. (3) The risk assessment of overturning construction of towering structures is conducted by the cloud matter–element coupled model to estimate the uncertainty of different evaluation indexes scientifically. (4) A specific project in Xinjiang Province, China, is selected for the case study, and the research results provide new ideas for the safety risk assessment of similar projects.

This paper consists of five sections. [Section 1](#) introduces the present situation of the study and highlights the need of the study. [Section 2](#) describes the establishment of the construction risk evaluation index system of overturning construction of towering structure, and presents the novel construction risk assessment method based on cloud matter-element coupled model. [Section 3](#) presents the case study of the overturning construction project of Turpan area in Xinjiang Province, China. [Section 4](#) presents the discussion of the study and the last section presents the conclusions made from this study.

## 2 Risk Assessment Model for Overturning Construction of Towering Structures

The purpose of this section is to establish a safety risk assessment model for turnover formwork construction of towering structures. This section contains five aspects: (1) Establishment of construction safety risk assessment index system by risk identification; (2) Grading of assessment indexes and safety risk metrics; (3) Determination of risk assessment index weights; (4) Introduction of the basic idea of cloud matter–element coupling model; (5) Creation of the construction safety risk assessment model based on the first four parts. The specific evaluation process is shown in [Fig. 1](#).



**Figure 1:** Evaluation flow chart

## 2.1 Establishment of Security Risk Assessment Index System

The first step of risk assessment is to identify risk factors [23]. Based on the “Standard for Safety Inspection of Building Construction” (JGJ 59-2011) [24], this paper started from the perspective of the technique and process of overturning construction and referred to the engineering geological survey report, project design drawings, national standards and local codes. Combined with a large number of engineering practice and theoretical research of many experts, the risk assessment index system was constructed by selecting the indexes that play an important role in overturning construction.

### 2.1.1 Risk Indicators for the Formwork Design and Installation Stage

The factors affecting the installation of formwork include wind load calculation [25], pressure calculation of the main beam [26], pull-out test computation of buried parts, and permissible deviation of verticality of formwork installation [27]. The installation stage of the formwork should consider the load-bearing capacity requirements of the formwork in the design stage and the installation error of the formwork in the construction process on site.

### 2.1.2 Risk Indicators for Concrete Pouring Phase

For the construction of towering structures with turnover formwork on site, the risks involved in concrete placement operations cannot be ignored. During the pouring process, considering the influence of concrete slump, newly poured concrete lateral pressure [28], and concrete pouring speed under the influence of wind load [29] is important.

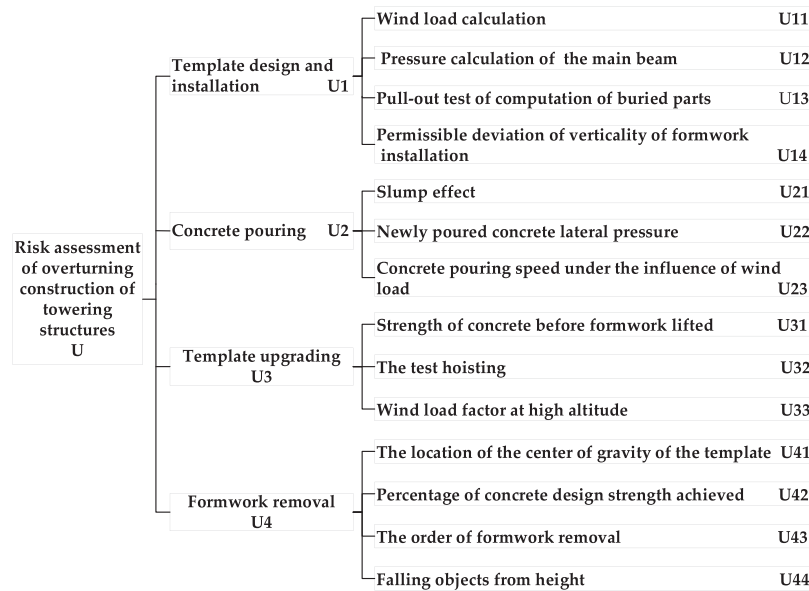
### 2.1.3 Risk Indicators for the Template Upgrading Phase

Factors, such as the strength of the concrete before the formwork is lifted, the test hoisting, and the wind load factors at high altitude, are all important influencing features that constitute the construction risk during the turnover formwork upgrading stage [30].

### 2.1.4 Formwork Removal Stage Risk Indicators

The risk factors influencing the removal of turnover formwork [31,32] include the location of the center of gravity of the formwork, the percentage of concrete design strength achieved, the order of formwork removal, and falling objects from height.

The resulting hierarchy of risk indicators for overturning the construction of towering structures is shown in Fig. 2.



**Figure 2:** The hierarchical model of risk causes in overturning construction of towering structures

**2.2 Grading of Assessment Indicators and Security Risk Metrics**

The selected evaluation indexes must first be judged for the safety status to evaluate the safety risk status of the overturning construction of towering structures based on the published and implemented “Technical Regulations for Concrete in Tall Buildings” (JGJ3-2010) [33], “Safety Technical Specification for Building Construction Formwork” (JGJ162-2008) [34], “The Code for Structural Loading of Buildings” (GB 50009-2012) [35], and “Pressure of freshly cast concrete on vertical formwork” (DIN18218-2010) [36]. By combining the research results of relevant literature and other relevant codes, the severity of safety risk of overturning construction of towering structures was divided into five levels: high risk, higher risk, medium risk, lower risk, and low risk, which were indicated by I–V as shown in Table 1.

**Table 1:** Risk accident safety level classification

Security risk level	Possibility description	Severity of consequences
Low risk (V)	Hardly ever ( $P < 0.01\%$ )	Negligible
Lower risk (IV)	Hardly ( $0.01\% \leq P < 0.1\%$ )	To be considered
Medium risk (III)	Occasionally ( $0.1\% \leq P < 1\%$ )	Serious
Higher risk (II)	Possibly ( $1\% \leq P < 10\%$ )	Very serious
High risk (I)	Frequently ( $P \geq 10\%$ )	Catastrophic

**2.3 Methods for Determining the Weight of Evaluation Indicators**

The commonly used weight determination methods are gray correlation, hierarchical analysis, and fuzzy evaluation. Among these methods, hierarchical analysis is widely used because of its simple operation and concise system. However, the hierarchical analysis method also has the following disadvantages: failure to solve the conflicts of multiple decision making during comprehensive evaluation

and incomprehensive and subjective constructed judgment matrix. Simultaneously considering the fuzziness and randomness of the problem is also difficult. Therefore, this paper aims to improve the traditional hierarchical analysis method by using the cloud model to overcome the shortcomings of excessive dependence on subjective experience to determine the index weights.

## 2.4 Theoretical Preparation

### 2.4.1 Cloud Model

Based on the concept of fuzzy set theory and probability statistics, the cloud model theory was developed by a Chinese scholar, Academician Li et al. [37]. This theory, which effectively combines the randomness and fuzziness of objective things by a unified mathematical expression, was first proposed in 1995; thus, the cloud model has been widely used in fuzzy evaluation, multi-attribute decision making, and risk assessment [38,39]. The detailed theoretical content of cloud models can be found in reference [37].

### 2.4.2 Matter–Element Extension Theory

The matter–element extension theory is a novel creation of Chinese scholar Cai in 1983 [40,41]. This theory incorporates the core ideas of the matter–element theory into topology by using the matter–element as the basic component to describe the thing. Thus, an ordered triad denoted as  $R = (N, C, V)$  is formed, where  $N$  is the name of the one described,  $C$  is the feature of the thing, and  $V$  is the characteristic value of the object. The topological evaluation method can determine the classical and node domains by changing the topology of the evaluated thing and calculating the correlation function and affiliation degree to perform the evaluation process combining quantitative and qualitative, which can effectively solve the uncertainty and contradiction problem.

## 2.5 Cloud-Based Coupled Matter–Element Model

### 2.5.1 Improved AHP Empowerment Method Based on Cloud Model

The cloud model was used in this study to transform the judgment matrix in AHP. The numerical judgment matrix was utilized to express the decision information quantitatively through the 1–9 scaling method. Assuming the existence of the theoretical domain  $U = \{i\}, i = 1, 2, 3, \dots, 9$ ,  $A_i$  is defined as the nine-cloud model with the structure  $A_i = (E_{x_i}, E_{n_i}, H_{e_i})$ , where  $E_{x_i}, E_{n_i}, H_{e_i}$  respectively denote the expectation, entropy, and hyperentropy. In reference [42], the golden section method was used to calculate [42] the above  $A_i$  parameters, as shown in Eq. (1).

$$\begin{cases} E_{x_i} = 1, 2, 3 \dots 9, i = 1, 2, 3 \dots 9, E_{x_i} \in Z \\ E_{n_1} = E_{n_3} = E_{n_5} = E_{n_7} = E_{n_9} = 0.382\alpha(x_{\max} - x_{\min})/6 = 0.437 \\ E_{n_2} = E_{n_1}/0.618, E_{n_4} = E_{n_3}/0.618, E_{n_6} = E_{n_5}/0.618, E_{n_8} = E_{n_7}/0.618 \\ H_{e_1} = H_{e_3} = H_{e_5} = H_{e_7} = H_{e_9} = 0.382\alpha(x_{\max} - x_{\min})/36 = 0.073 \\ H_{e_2} = H_{e_1}/0.618, H_{e_4} = H_{e_3}/0.618, H_{e_6} = H_{e_5}/0.618, H_{e_8} = H_{e_7}/0.618 \end{cases} \quad (1)$$

where  $x_{\max} = 9, x_{\min} = 1$  and  $\alpha$  are the adjustment factors, generally taking 0.858.

The final numerical characteristics of the critical scale cloud model are shown in Table 2. The table shows that a large expected  $E_{x_i}$  value indicates its high importance.

**Table 2:** Important scale

Important scale	Definition
$A_1 = (E_{x_1}, E_{n_1}, H_{e_1}) = (1, 0.437, 0.073)$	$u_i$ and $u_j$ are equally important compared to each other
$A_3 = (E_{x_3}, E_{n_3}, H_{e_3}) = (3, 0.437, 0.073)$	$u_i$ is slightly more important than $u_j$
$A_5 = (E_{x_5}, E_{n_5}, H_{e_5}) = (5, 0.437, 0.073)$	$u_i$ is more important compared to $u_j$
$A_7 = (E_{x_7}, E_{n_7}, H_{e_7}) = (7, 0.437, 0.073)$	$u_i$ is very important compared to $u_j$
$A_9 = (E_{x_9}, E_{n_9}, H_{e_9}) = (9, 0.437, 0.073)$	$u_i$ is extremely important compared to $u_j$
$A_2 = (E_{x_2}, E_{n_2}, H_{e_2}) = (2, 0.707, 0.118)$	The degree of importance is between the two adjacent clouds mentioned above
$A_4 = (E_{x_4}, E_{n_4}, H_{e_4}) = (4, 0.707, 0.118)$	
$A_6 = (E_{x_6}, E_{n_6}, H_{e_6}) = (6, 0.707, 0.118)$	
$A_8 = (E_{x_8}, E_{n_8}, H_{e_8}) = (8, 0.707, 0.118)$	

According to the above decision method, the cloud model scalar judgment matrix for the clouded hierarchical analysis was established in the form of Eq. (2).

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} & \cdots & A_{1n} \\ A_{21} & A_{22} & \cdots & A_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \cdots & A_{nn} \end{bmatrix} \tag{2}$$

The entropy and hyperentropy of the elements on the diagonal in the above equation were 0 and the expectation was 1; n is the number of evaluation indicators. When the indicators were compared two by two, the  $a_{ji} = \frac{1}{a_{ij}}, A_{ji} = \frac{1}{A_{ij}} = \left( \frac{1}{E_x}, \frac{E_n}{(E_x)^2}, \frac{H_e}{(E_x)^2} \right)$ .

The obtained evaluation results of the experts were aggregated, and the average value was taken. The aggregation formula is shown in Eqs. (3)–(5) as follows:

$$\bar{E}_x = \frac{1}{n} \left( \sum_{i=1}^n E_{x_i} \right) \tag{3}$$

$$\bar{E}_n = \frac{1}{n} \left( \sum_{i=1}^n E_{n_i}^2 \right)^{1/2} \tag{4}$$

$$\bar{H}_e = \frac{1}{n} \left( \sum_{i=1}^n H_{e_i}^2 \right)^{1/2} \tag{5}$$

The matrix was normalized and the multiplication calculation was introduced in cloud computing, and the relative weights  $A_{\bar{w}_i} (E'_{x_i}, E'_{n_i}, H'_{e_i})$  of expectation, ambiguity, and randomness of the elements are calculated using the square root method. Expression of the parameters are shown in Eqs. (6)–(8):

$$E'_{x_i} = \frac{E_{x_i}}{\sum E_{x_i}} = \frac{\left( \prod_{j=1}^n E_{x_{ij}} \right)^{1/n}}{\sum_{i=1}^n \left( \prod_{j=1}^n E_{x_{ij}} \right)^{1/n}} \tag{6}$$



$$E'_{n_i} = \frac{E_{n_i}}{\sum E_{n_i}} = \frac{\left( \left( \prod_{j=1}^n E_{x_{ij}} \right) \sqrt{\sum_{j=1}^n \left( \frac{E_{n_{ij}}}{E_{x_{ij}}} \right)^2} \right)^{1/n}}{\sum_{i=1}^n \left( \left( \prod_{j=1}^n E_{x_{ij}} \right) \sqrt{\sum_{j=1}^n \left( \frac{E_{n_{ij}}}{E_{x_{ij}}} \right)^2} \right)^{1/n}} \tag{7}$$

$$H'_{e_i} = \frac{H_{e_i}}{\sum H_{e_i}} = \frac{\left( \left( \prod_{j=1}^n E_{x_{ij}} \right) \sqrt{\sum_{j=1}^n \left( \frac{H_{e_{ij}}}{E_{x_{ij}}} \right)^2} \right)^{1/n}}{\sum_{i=1}^n \left( \left( \prod_{j=1}^n E_{x_{ij}} \right) \sqrt{\sum_{j=1}^n \left( \frac{H_{e_{ij}}}{E_{x_{ij}}} \right)^2} \right)^{1/n}} \tag{8}$$

The expected consistency test was performed by consistency metrics C and R. This test required

satisfying Eq. (9), where  $C = \frac{\lambda_{\max} - n}{n - 1}$ ,  $\lambda_{\max} \approx \frac{1}{n} \sum_{i=1}^n \left( \frac{\sum_{j=1}^n E_{x_{ij}} A_{w_i}}{A_{w_i}} \right)$ , and R are the average of the

consistency metrics of the same order random judgment matrix. The consistency ratio I is calculated as follows:

$$I = \frac{C}{R} < 0.1 \tag{9}$$

### 2.5.2 Improved Cloud-Based Model for Matter–Element Risk Assessment

The cloud matter–element was used in this paper to utilize the cloud model to redefine and construct the matter–element extension theory and to employ the general steps of extension evaluation to describe and evaluate things. The cloud matter–element [43] replaced the specific values  $V$  in the ordered triple  $R = (N, C, V)$  in the matter–element theory with  $(E_x, E_n, H_e)$  in the normal cloud model to realize the description of the randomness and fuzziness in the evaluation process. The cloud matter–element model is expressed as:

$$R = \begin{pmatrix} N & C_1 & (E_{x_1}, E_{n_1}, H_{e_1}) \\ N & C_2 & (E_{x_1}, E_{n_1}, H_{e_1}) \\ \vdots & \vdots & \vdots \\ N & C_n & (E_{x_1}, E_{n_1}, H_{e_1}) \end{pmatrix} \tag{10}$$

where expectation  $E_{x_i}$  indicates the center of the cloud distribution, which is the point value that best reflects the attribute concept; entropy  $E_n$  indicates the degree of uncertainty: a large entropy value leads to high uncertainty of the indicator; hyperentropy  $H_e$  is a measure of the degree of entropy uncertainty: a large hyperentropy leads to large ambiguity and uncertainty of entropy  $E_n$  and large thickness of cloud drops on the cloud map. The cloud model is an overall reflection of the quantitative characteristics of the qualitative concept by the three numerical characteristics.

### 2.5.3 Calculation of Cloud Model Parameters

In the cloud matter–element model, when dividing the interval of safety risk level for the overturning construction of towering structures, the fuzzy and random nature of the cloud model was used to fuzzy the interval boundary values of the traditional matter–element model. That is to say, the hierarchical boundary of the classification level boundary of each evaluation index for the



safety risk of turnover formwork construction of towering structures can help obtain the expectation  $E_x$ , entropy  $E_n$ , and hyperentropy  $H_e$ .

$$E_x = \frac{(C_{\min} + C_{\max})}{2} \tag{11}$$

$$E_n^{(1)} = \frac{(C_{\max} - C_{\min})}{6} \tag{12}$$

$$E_n^{(2)} = \frac{(C_{\max} - C_{\min})}{2.3548} \tag{13}$$

$$H_e = s \tag{14}$$

At present, the evaluation of indicators has two main methods of cloud entropy calculation: [formula \(12\)](#) is used for the classification of levels with clear boundaries, such as low-risk level, and [formula \(13\)](#) is used to classify various hazard levels other than low-risk level.  $C_{\min}$ ,  $C_{\max}$  are the minimum and maximum values of the classification level interval of the evaluation indicators, respectively;  $s$  is a constant, which can be adjusted in combination with the ambiguity and randomness of the risk assessment indicators of turnover formwork construction. This paper takes  $s = 0.08$  [44].

#### 2.5.4 Correlation Calculation of Cloud Matter–Element Model

The introduction of clouds has changed the ambiguity and randomness of the traditional matter–element theory in determining the grade boundaries. Thus, the calculation of the correlation degree of the matter–element model combined with the cloud model has also changed. If each assessed value of the turnover formwork construction to be evaluated  $x$  is regarded as a cloud droplet, then the correlation degree for calculating the cloud droplet  $x$  belonging to a specific grade is calculated by the following equation:

$$k = \exp\left(-\frac{(x - E_x)^2}{2(E_n')^2}\right) \tag{15}$$

where  $E_n'$  is the normal random number generated from the entropy  $E_n$  and the hyperentropy  $H_e$ .

According to [Eq. \(15\)](#), the cloud correlation between each index to be evaluated and the standard normal cloud of security risk level was calculated, and the cloud correlation matrix  $K$  was obtained, which was in the form of [Eq. \(16\)](#).

$$K = \begin{bmatrix} k_{11} & k_{12} & k_{13} & k_{14} & k_{15} \\ k_{21} & k_{22} & k_{23} & k_{24} & k_{25} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ k_{n1} & k_{n2} & k_{n3} & k_{n4} & k_{n5} \end{bmatrix} \tag{16}$$

where  $k_{ij}$  is the cloud correlation between the evaluation index  $i$  and the security risk level  $j$ ;  $i$  ( $i = 1, 2, \dots, n$ ) is the number of evaluation index (in this paper,  $n$  is 14);  $j$  is the security risk level number (in this paper,  $j$  is the integer of  $1 \sim 4$ ).

### 2.5.5 Determination of the Security Risk Level

The integrated evaluation vector  $B$  was obtained by multiplying the weight coefficient  $A_{\bar{w}_i}$  with the cloud correlation matrix  $K$ , and the formula is shown in (17).

$$B = A_{\bar{w}_i} \cdot K \quad (17)$$

where the evaluation index weight vector  $A_{\bar{w}_i}$  comprises the weights of each evaluation index.

A weighted average method was applied to the composite assessment vector  $B$  to derive the composite assessment score  $r$  [45] as in Eq. (18).

$$r = \frac{\sum_{i=1}^5 b_i f_i}{\sum_{i=1}^5 b_i} \quad (18)$$

where  $b_i$  is the component of vector  $B$ ;  $f_i$  is the score of the  $i$  level, which corresponds to the scores of 5, 4, 3, 2, and 1 for the judgment levels I to V.

When calculating the correlation degree, the presence of normal random numbers produced common calculation results. Therefore, the combined evaluation score of expectation  $E_{rx}$  and entropy  $E_{rn}$  after multiple simulations must be calculated to reduce the influence of random factors. The calculation equations are shown in Eqs. (19) and (20).

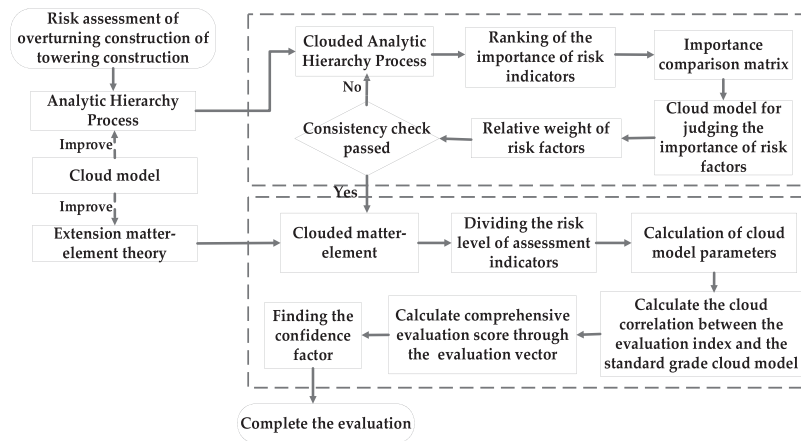
$$E_{rx} = \frac{r_1(x) + r_2(x) + \dots + r_m(x)}{m} \quad (19)$$

$$E_{rn} = \sqrt{\frac{1}{m} \sum_{h=1}^m (r_h(x) - E_{rx})^2} \quad (20)$$

where  $m$  is the number of simulation calculations, which was taken as 100 in this paper;  $r_h(x)$  is the comprehensive evaluation score of the  $h$ -th simulation calculation. The final expected value is the evaluation score that best represents the level of risk in the construction of overturned forms of towering structures; when the score is close to the value, the safety risk level is judged as the level corresponding to that value. The entropy is the measure of the variance of the evaluation result, and as the value becomes larger, the evaluation result is distributed. Meanwhile, the confidence factor  $\theta$  [46] is defined as Eq. (21).

$$\theta = \frac{E_{rn}}{E_{rx}} \quad (21)$$

A large  $\theta$  value in the formula indicates the large variance and the low credibility of the evaluation results; otherwise, the credibility of the evaluation results is high. The process of the specific evaluation method is shown in Fig. 3.



**Figure 3:** Flow chart of evaluation method

### 3 Empirical Analysis

#### 3.1 Project Background

The project is located in Turpan City, Xinjiang, China. The climatic conditions of the project location are remarkably extreme, with frequent windy weather in summer belonging to the continental monsoon climate and four indistinguishable seasons. Thus, the impact of the natural environment on construction cannot be ignored. The structure of the project is a reinforced concrete cylindrical tower structure with a building height of 149.85 m, 28 floors above ground, and 2 floors underground. The radius of the structure is 10.5 m, and the span of each floor beam support is more than 18 m. The concrete wall outer mold adopts cantilevered overturning mold. Among them, the formwork panel adopts an imported Vespa board, wherein the size of the formwork is 5.6 m × 4.131 m (height × width) and its thickness is 18 mm; meanwhile, the vertical flute of formwork adopts a woodworking word beam, and the main back flute is double 14# channel steel. The support system comprises a triangle frame, a picket frame, a hanging platform, and an embedded load-bearing device. The formwork structure is shown in Figs. 4 and 5 below.

#### 3.2 Determination of Index Weights

The importance of risk indicators affecting safety was ranked by familiarizing with relevant project information and consulting with relevant experts. In this paper, 20 experts with a considerable amount of successful experience in the construction and risk control of overturning sites of towering structures were invited to conduct a questionnaire survey. Detailed information is shown in Table 3. Their questionnaire results were randomly selected to compare and analyze the risk factors of overturning construction sequentially, and the flowchart is shown in Fig. 6. The importance ranking of the first-level indicators was obtained as follows. Expert 1 thought the relative importance ranking of four risk factors was  $U_2 > U_4 > U_3 > U_1$  and the risk comparison matrix was  $Q_1$ ; Expert 2 believed that the relative importance of the four risk factors was  $U_2 > U_3 > U_4 > U_1$  and the risk comparison matrix was  $Q_2$ ; Expert 3 thought that the relative importance of the four risk factors was  $U_2 > U_4 > U_3 > U_1$  and the risk comparison matrix was  $Q_3$ . The summary results of the risk comparison matrix of the first-level indicators are shown in Table 4.

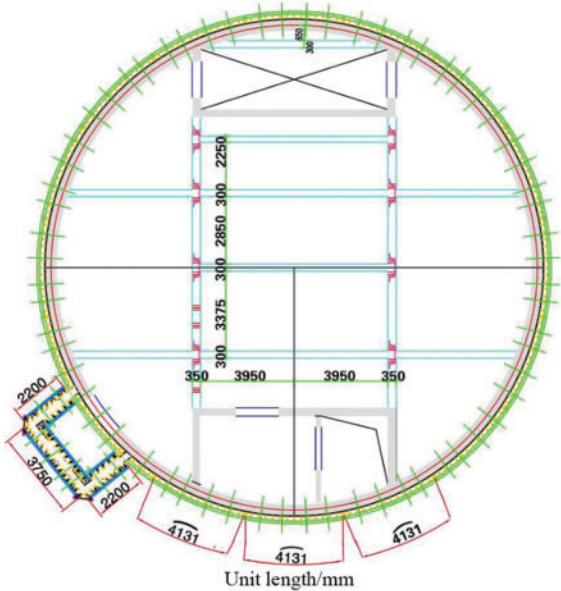


Figure 4: Formwork plan layout

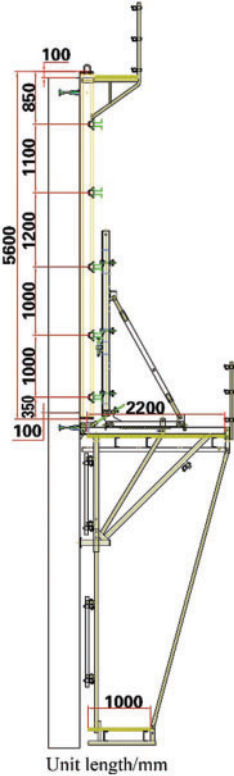
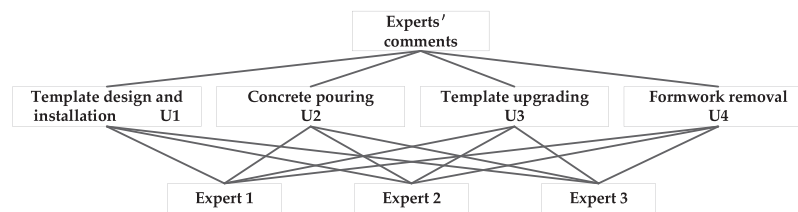


Figure 5: Formwork elevation

**Table 3:** The background information of the interviewed experts

Numbers	Job position	Age	Education	Field	Engaged in engineering construction industry
1	Project-level leadership	43	Bachelor	Project supervision	11–20 years
2	Middle manager	45	Bachelor	Safety management	11–20 years
3	Middle manager	38	Bachelor	Project supervision and consultation	11–20 years
4	Top manager	48	Master	Project management	21–30 years
5	Top manager	52	Bachelor	Construction design	21–30 years
6	Project-level leadership	43	Master	Project consultation	6–10 years
7	Middle manager	47	Master	Construction management	21–30 years
8	Professor, PhD supervisor	45	PhD.	Risk management	11–20 years
9	Professor, PhD supervisor	42	PhD.	Project management	11–20 years
10	Professor, PhD supervisor	53	PhD.	Risk management	21–30 years
11	Project-level leadership	40	Bachelor	Construction design	11–20 years
12	Middle manager	36	Master	Project management	6–10 years
13	Professor, PhD supervisor	45	PhD.	Structural Engineering	11–15 years
14	Project-level leadership	42	Bachelor	Construction design	11–20 years
15	Project-level leadership	39	Master	Construction management	11–15 years
16	Professor, PhD supervisor	54	PhD.	Safety management	21–30 years
17	Middle manager	42	Bachelor	Construction management	11–20 years
18	Project-level leadership	41	Bachelor	Project supervision	11–15 years
19	Project-level leadership	44	Bachelor	Project consultation	11–20 years
20	Middle manager	39	Master	Construction management	6–10 years



**Figure 6:** Flow chart of risk evaluation

**Table 4:** Comparative matrix of relative importance of risk factors for the first-level indicators of overturning formwork construction

	$Q_1$				$Q_2$				$Q_3$			
	$U_1$	$U_2$	$U_3$	$U_4$	$U_1$	$U_2$	$U_3$	$U_4$	$U_1$	$U_2$	$U_3$	$U_4$
$U_1$	1	5	3	4	1	7	3	5	1	7	4	3
$U_2$	1/5	1	1/5	1/4	1/7	1	1/4	1/4	1/7	1	1/5	1/4
$U_3$	1/3	5	1	1/5	1/3	4	1	1/3	1/4	5	1	1/3
$U_4$	1/4	4	5	1	1/5	4	3	1	1/3	4	3	1

The cloud model-based linguistic judgment scales  $A_{12}$  given by three experts for the importance of two elements  $U_1$  and  $U_2$  are as follows:  $Q_{1(A_{12})} = (5, 0.437, 0.073)$ ,  $Q_{2(A_{12})} = (7, 0.437, 0.073)$ , and  $Q_{3(A_{12})} = (7, 0.437, 0.073)$ . After the aggregation of formulas (3)–(5), the cloud model for the importance judgment of  $U_1$  and  $U_2$  can be obtained as (3.333, 0.313, 0.052). In detail,  $(\overline{E_x})_{A_{12}} = (5 + 7 + 7)^{1/3} = 6.333$ ,  $(\overline{E_n})_{A_{12}} = \frac{1}{3}\sqrt{(0.437^2 + 0.437^2 + 0.437^2)} = 0.252$ ,  $(\overline{H_e})_{A_{12}} = \frac{1}{3}\sqrt{(0.073^2 + 0.073^2 + 0.073^2)} = 0.042$ . Similarly, the cloud model of importance judgment of other risk factors obtained by the same method is summarized in Table 5.

**Table 5:** The importance of risk factor cloud model matrix

$i$	$j$			
	1	2	3	4
1	(1, 0, 0)	(6.333, 0.252, 0.042)	(3.333, 0.313, 0.052)	(4.000, 0.313, 0.052)
2	(0.162, 0.007, 0.001)	(1, 0, 0)	(0.217, 0.015, 0.002)	(0.250, 0.026, 0.003)
3	(0.306, 0.029, 0.005)	(4.667, 0.313, 0.052)	(1, 0, 0)	(0.289, 0.021, 0.004)
4	(0.261, 0.021, 0.004)	(4.000, 0.408, 0.052)	(3.667, 0.252, 0.042)	(1, 0, 0)

The relative weights calculated by Eqs. (6)–(8) are shown in Table 6. In detail,  $E_{x_1} = (1 \times 6.333 \times 3.333 \times 4)^{\frac{1}{4}} = 3.031$ , similarly,  $E_{x_2} = 0.306$ ,  $E_{x_3} = 0.801$ ,  $E_{x_4} = 1.399$ .

**Table 6:** Relative weights of risk factors

$A_{w_i}$	$A_{\overline{w}_i}$
(3.031, 1.815, 1.160)	(1.399, 0.867, 0.539)
(0.306, 0.183, 0.113)	(0.055, 0.055, 0.053)
(0.801, 0.488, 0.312)	(0.145, 0.146, 0.147)
(1.399, 0.867, 0.539)	(0.253, 0.259, 0.254)

$E'_{x_1} = 3.031 / (3.031 + 0.306 + 0.801 + 1.399) = 0.547$ , similarly,  $E'_{x_2} = 0.055$ ,  $E'_{x_3} = 0.145$ ,  $E'_{x_4} = 0.253$ .

$$E_{n_1} = \left\{ (1 \times 6.333 \times 3.333 \times 4) \left[ 0 + \left( \frac{0.252}{6.333} \right)^2 + \left( \frac{0.313}{3.333} \right)^2 + \left( \frac{0.313}{4.000} \right)^2 \right]^{1/2} \right\}^{1/4} = 1.185,$$

similarly,  $E_{n_2} = 0.183, E_{n_3} = 0.488,$

$$E_{n_4} = 0.867. E'_{n_1} = 1.185 / (1.185 + 0.183 + 0.488 + 0.867) = 0.541, \text{ similarly, } E'_{n_2} = 0.055,$$

$$E'_{n_3} = 0.146, E'_{n_4} = 0.259. H_{e_1} = \left\{ (1 \times 6.333 \times 3.333 \times 4) \left[ 0 + \left( \frac{0.042}{6.333} \right)^2 + \left( \frac{0.052}{3.333} \right)^2 + \left( \frac{0.052}{4.000} \right)^2 \right]^{1/2} \right\}^{1/4} = 1.160, \text{ similarly,}$$

$$H_{e_2} = 0.113, H_{e_3} = 0.312, H_{e_4} = 0.539. H'_{e_1} = 1.16 / (1.16 + 0.113 + 0.312 + 0.539) = 0.546,$$

$$H'_{e_2} = 0.053, H'_{e_3} = 0.147, H'_{e_4} = 0.254.$$

The consistency test was performed by Eq. (10) to obtain  $I = 0.052 < 0.1$ . In detail,

$$\begin{bmatrix} 1 & 6.333 & 3.333 & 4 \\ 0.162 & 1 & 0.217 & 0.25 \\ 0.306 & 4.667 & 1 & 0.289 \\ 0.261 & 4 & 3.667 & 1 \end{bmatrix} \begin{bmatrix} 3.031 \\ 0.306 \\ 0.801 \\ 1.399 \end{bmatrix} = \begin{bmatrix} 13.236 \\ 1.320 \\ 3.560 \\ 6.352 \end{bmatrix}, \lambda_{\max} \approx \frac{1}{4} \left( \frac{13.236}{3.031} + \frac{1.32}{0.306} + \frac{3.56}{0.801} + \frac{6.352}{1.399} \right) = 4.175, C = \frac{\lambda_{\max} - n}{n - 1} = \frac{4.175 - 4}{4 - 1} = 0.058, I = \frac{C}{R} = \frac{0.058}{0.89} = 0.065. \text{ Therefore,}$$

the first-level indicators of template design and installation  $U_1$ , concrete pouring  $U_2$ , template upgrading  $U_3$ , and formwork removal  $U_4$  had a weight vector of  $w = (0.547, 0.055, 0.145, 0.253)$ , respectively.

For the second-level indicators under the risk of template design and installation, Expert 1 believed that the relative importance of the four risk factors was ranked as  $U_{1-3} > U_{1-4} > U_{1-2} > U_{1-1}$  and the risk comparison matrix was  $Q_{1-1}$ ; Expert 2 believed that the relative importance of the four risk factors was ranked as  $U_{1-3} > U_{1-4} = U_{1-2} > U_{1-1}$  and the risk comparison matrix was  $Q_{1-2}$ ; Expert 3 believed that the relative importance of the four risk factors was ranked as  $U_{1-3} > U_{1-4} > U_{1-2} > U_{1-1}$  and the risk comparison matrix was  $Q_{1-3}$ .

For the second-level indicators under the risk of concrete pouring, Expert 1 believed that the relative importance of the three risk factors was ranked as  $U_{2-2} > U_{2-3} > U_{2-1}$  and the risk comparison matrix was  $Q_{2-1}$ ; Expert 2 believed that the relative importance of the three risk factors was ranked as  $U_{2-2} > U_{2-3} > U_{2-1}$  and the risk comparison matrix was  $Q_{2-2}$ ; Expert 3 believed that the relative importance of the three risk factors was ranked as  $U_{2-2} > U_{2-3} > U_{2-1}$  and the risk comparison matrix was  $Q_{2-3}$ .

For the second-level indicators under the risk of template upgrading, Expert 1 believed that the relative importance of the three risk factors was ranked as  $U_{3-2} > U_{3-3} > U_{3-1}$  and the risk comparison matrix was  $Q_{3-1}$ ; Expert 2 believed that the relative importance of the three risk factors was ranked as  $U_{3-2} > U_{3-3} > U_{3-1}$  and the risk comparison matrix was  $Q_{3-2}$ ; Expert 3 believed that the relative importance of the three risk factors was ranked as  $U_{3-2} > U_{3-3} > U_{3-1}$  and the risk comparison matrix was  $Q_{3-3}$ .

For the second-level indicators under the risk of formwork removal, Expert 1 believed that the relative importance of the four risk factors was ranked as  $U_{4-2} > U_{4-4} > U_{4-1} = U_{4-3}$  and the risk



comparison matrix was  $Q_{4-1}$ ; Expert 2 believed that the relative importance of the four risk factors was ranked as  $U_{4-2} > U_{4-4} > U_{4-3} > U_{4-1}$  and the risk comparison matrix was  $Q_{4-2}$ ; Expert 3 believed that the relative importance of the four risk factors was  $U_{4-2} > U_{4-4} > U_{4-3} > U_{4-1}$  and the risk comparison matrix was  $Q_{4-3}$ . The summary comparison matrix of all levels of indicators is shown in Tables 7–10.

**Table 7:** Comparative matrix of relative importance of risk factors of secondary indicators under the template design and installation

	$Q_{1-1}$				$Q_{1-2}$				$Q_{1-3}$			
	$U_{1-1}$	$U_{1-2}$	$U_{1-3}$	$U_{1-4}$	$U_{1-1}$	$U_{1-2}$	$U_{1-3}$	$U_{1-4}$	$U_{1-1}$	$U_{1-2}$	$U_{1-3}$	$U_{1-4}$
$U_{1-1}$	1	3	7	5	1	4	6	4	1	2	6	5
$U_{1-2}$	1/3	1	4	2	1/4	1	3	1/2	1/2	1	4	3
$U_{1-3}$	1/7	1/4	1	1/2	1/6	1/3	1	1/4	1/6	1/4	1	1/2
$U_{1-4}$	1/5	1/2	2	1	1/4	2	4	1	1/5	1/3	2	1

**Table 8:** Comparative matrix of relative importance of risk factors of secondary indicators under the concrete pouring

	$Q_{2-1}$			$Q_{2-2}$			$Q_{2-3}$		
	$U_{2-1}$	$U_{2-2}$	$U_{2-3}$	$U_{2-1}$	$U_{2-2}$	$U_{2-3}$	$U_{2-1}$	$U_{2-2}$	$U_{2-3}$
$U_{2-1}$	1	5	3	1	7	3	1	7	4
$U_{2-2}$	1/5	1	1/5	1/7	1	1/4	1/7	1	1/5
$U_{2-3}$	1/3	5	1	1/3	4	1	1/4	5	1

**Table 9:** Comparative matrix of relative importance of risk factors of secondary indicators under the template upgrading

	$Q_{3-1}$			$Q_{3-2}$			$Q_{3-3}$		
	$U_{3-1}$	$U_{3-2}$	$U_{3-3}$	$U_{3-1}$	$U_{3-2}$	$U_{3-3}$	$U_{3-1}$	$U_{3-2}$	$U_{3-3}$
$U_{3-1}$	1	3	2	1	6	3	1	4	3
$U_{3-2}$	1/3	1	1/3	1/6	1	1/4	1/4	1	1/5
$U_{3-3}$	1/2	3	1	1/3	4	1	1/3	5	1

**Table 10:** Comparative matrix of relative importance of risk factors of secondary indicators under the formwork removal

	$Q_{4-1}$				$Q_{4-2}$				$Q_{4-3}$			
	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$
$U_{4-1}$	1	5	1	4	1	7	2	5	1	6	3	5

(Continued)

**Table 10 (continued)**

	$Q_{4-1}$				$Q_{4-2}$				$Q_{4-3}$			
	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$
$U_{4-2}$	1/5	1	1/3	1/4	1/7	1	1/4	1/3	1/6	1	1/5	1/2
$U_{4-3}$	1	3	1	2	1/2	4	1	2	1/3	5	1	3
$U_{4-4}$	1/4	4	1/2	1	1/5	3	1/2	1	1/5	2	1/3	1

The same method was used to calculate the weights under different indicators in turn. With the inclusion of the second-level index, the weight vectors of the template design and installation, concrete pouring, template upgrading, and formwork removal were {0.553, 0.225, 0.067, 0.155}, {0.655, 0.078, 0.267}, {0.592, 0.105, 0.303}, and {0.509, 0.066, 0.288, 0.137}, respectively. The consistency test met the requirements, and the final comprehensive weight is shown in Table 11.

**Table 11:** Weights of risk indicators for overturning formwork construction

First level indicator	Weights	Secondary indicator	Weights	Combined weights
Template design and installation	0.547	Wind load calculation (KN/m)	0.553	0.303
		Pressure calculation of the main beam (KN)	0.225	0.123
		Pull-out test of computation of buried parts (KN)	0.067	0.036
		Permissible deviation of verticality of formwork installation (mm)	0.155	0.085
Concrete pouring	0.055	Slump effect (mm)	0.655	0.036
		Newly poured concrete lateral pressure (KN)	0.078	0.004
		Concrete pouring speed under the influence of wind load (m/h)	0.267	0.015
Template upgrading	0.145	Strength of concrete before formwork lifted (MPa)	0.592	0.086
		The test hoisting	0.105	0.015
		Wind load factor at high altitude (KN)	0.303	0.044
Formwork removal	0.253	The location of the center of gravity of the template	0.509	0.129
		Percentage of concrete design strength achieved (%)	0.066	0.017
		The order of formwork removal	0.288	0.073
		Falling objects from height	0.137	0.034

### 3.3 Cloud-Based Risk Assessment

The index system contained 4 qualitative indicators and 10 quantitative indicators. For quantitative indicators, the data from public sources, such as statistical yearbooks and design specifications, are used for the interval division criteria; for qualitative indicators, levels I, II, III, IV, and V were respectively assigned 2, 4, 6, 8, and 10, as shown in Table 12.

**Table 12:** Weights of risk indicators for overturning formwork construction

First level indicator	Secondary indicator	Risk level				
		I	II	III	IV	V
Template design and installation	Wind load calculation (KN/m)	(4, 6.2)	(6.2, 7.4)	(7.4, 8.6)	(8.6, 9.8)	(9.8, 11.0)
	Pressure calculation of the main beam (KN)	(30.76, 42.76)	(42.76, 44.76)	(44.76, 46.76)	(46.76, 48.76)	(48.76, 50.76)
	Pull-out test of computation of buried parts (KN)	(50.46, 61.46)	(61.46, 63.46)	(63.46, 65.46)	(65.46, 67.46)	(67.46, 69.46)
	Permissible deviation of verticality of formwork installation (mm)	(4, 5)	(3, 4)	(2, 3)	(1, 2)	(0, 1)
Concrete pouring	Slump effect (mm)	(180, 200)	(160, 180)	(130, 160)	(90, 130)	(50, 90)
	Newly poured concrete lateral pressure (KN)	(70.36, 76.36)	(76.36, 78.36)	(78.36, 80.36)	(80.36, 82.36)	(82.36, 84.36)
	Concrete pouring speed under the influence of wind load (m/h)	(8, 10)	(6, 8)	(4, 6)	(2, 4)	(0, 2)
Template upgrading	Strength of concrete before formwork lifted (MPa)	(0, 15)	(15, 20)	(20, 25)	(25, 40)	(40, 50)
	The test hoisting	(0, 2)	(2, 4)	(4, 6)	(6, 8)	(8, 10)
	Wind load factor at high altitude (KN)	(9, 10)	(7, 9)	(5, 7)	(3, 5)	(1, 3)
Formwork removal	The location of the center of gravity of the template	(0, 2)	(2, 4)	(4, 6)	(6, 8)	(8, 10)
	Percentage of concrete design strength achieved (%)	(0, 50)	(50, 60)	(60, 75)	(75, 90)	(90, 100)
	The order of formwork removal	(0, 2)	(2, 4)	(4, 6)	(6, 8)	(8, 10)
	Falling objects from height	(0, 2)	(2, 4)	(4, 6)	(6, 8)	(8, 10)

After obtaining the grade division index, the standard cloud model of each secondary index was calculated by the method of determining the cloud model parameters. The specific numerical calculation results are shown in [Table 13](#).

**Table 13:** Numerical characteristics of the cloud model of secondary indicators in turnover model construction

Safety risk level	Boundary cloud model parameter values for each assessment index classification level ( $E_x, E_n, H_e$ )				
	$U_{1-1}$	$U_{1-2}$	$U_{1-3}$	$U_{1-4}$	$U_{2-1}$
High risk (I)	(5.1, 0.934, 0.08)	(36.76, 5.096, 0.08)	(55.96, 4.671, 0.08)	(10, 1.699, 0.08)	(190, 8.493, 0.08)
Higher risk (II)	(6.8, 0.51, 0.08)	(43.76, 0.849, 0.08)	(62.46, 0.849, 0.08)	(7, 0.849, 0.08)	(170, 8.493, 0.08)
Medium risk (III)	(8, 0.51, 0.08)	(45.76, 0.849, 0.08)	(64.46, 0.849, 0.08)	(4.5, 1.274, 0.08)	(145, 12.74, 0.08)
Lower risk (IV)	(9.2, 0.51, 0.08)	(47.76, 0.849, 0.08)	(66.46, 0.849, 0.08)	(2, 0.849, 0.08)	(110, 16.987, 0.08)
Low risk (V)	(10.4, 0.2, 0.08)	(49.76, 0.333, 0.08)	(68.46, 0.333, 0.08)	(0.5, 0.167, 0.08)	(70, 6.667, 0.08)
	$U_{2-2}$	$U_{2-3}$	$U_{3-1}$	$U_{3-2}$	$U_{3-3}$
High risk (I)	(73.36, 2.548, 0.08)	(9, 0.849, 0.08)	(7.5, 6.37, 0.08)	(1, 0.849, 0.08)	(9.5, 0.425, 0.08)
Higher risk (II)	(77.36, 0.849, 0.08)	(7, 0.849, 0.08)	17.5, 2.123, 0.08)	(3, 0.849, 0.08)	(8, 0.849, 0.08)
Medium risk (III)	(79.36, 0.849, 0.08)	(5, 0.849, 0.08)	(22.5, 2.123, 0.08)	(5, 0.849, 0.08)	(6, 0.849, 0.08)
Lower risk (IV)	(81.36, 0.849, 0.08)	(3, 0.849, 0.08)	(32.5, 6.37, 0.08)	(7, 0.849, 0.08)	(4, 0.849, 0.08)
Low risk (V)	(83.36, 0.333, 0.08)	(1, 0.333, 0.08)	(45, 1.667, 0.08)	(9, 0.333, 0.08)	(2, 0.333, 0.08)
	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$	
High risk (I)	(1, 0.849, 0.08)	(25, 21.233, 0.08)	(1, 0.849, 0.08)	(1, 0.849, 0.08)	
Higher risk (II)	(3, 0.849, 0.08)	(55, 4.247, 0.08)	(3, 0.849, 0.08)	(3, 0.849, 0.08)	
Medium risk (III)	(5, 0.849, 0.08)	(67.5, 6.37, 0.08)	(5, 0.849, 0.08)	(5, 0.849, 0.08)	
Lower risk (IV)	(7, 0.849, 0.08)	(82.5, 6.37, 0.08)	(82.5, 6.37, 0.08)	(7, 0.849, 0.08)	
Low risk (V)	(9, 0.333, 0.08)	(95, 1.667, 0.08)	(95, 1.667, 0.08)	(9, 0.333, 0.08)	

The MATLAB program was prepared in accordance with the three numerical characteristics of the cloud model of the classification level boundaries of each assessment index in [Table 13](#). Therefore, each safety risk level corresponded to a cloud, and five normal clouds were generated using the normal

cloud generator. Among them, the standard cloud diagram of the risk factor of wind load is shown in Fig. 7. The standard cloud diagrams of other risk factors were not comprehensively described in the paper due to space limitation.

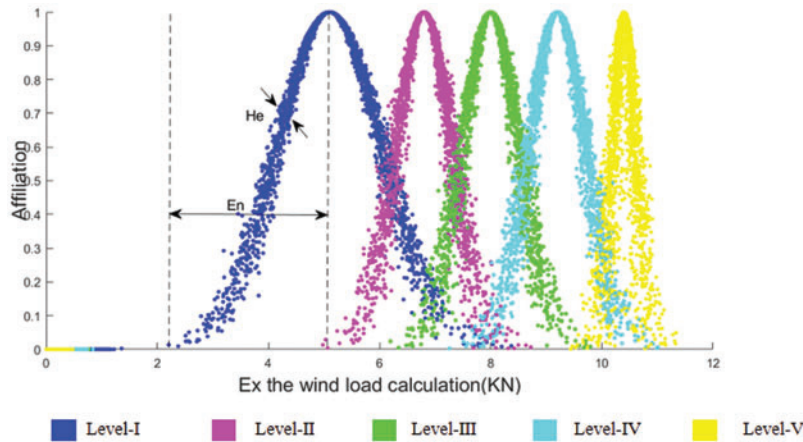


Figure 7: Standard cloud diagram of risk factors

By investigating the field situation and schedule of the construction site, three groups of working conditions with different orientations of the same construction height were selected in this study. The actual measured or score values of the 13 evaluation indicators in were inputted in Table 14, and the cloud correlation degree between each evaluation indicator and the normal cloud of the security risk level standard was calculated in accordance with Eq. (15). Taking working condition one as an example, the cloud correlation matrix  $K$  was calculated as follows:

$$K = \begin{bmatrix} 0.0001 & 0.0001 & 0.0017 & 0.4989 & 0.0108 \\ 0.9998 & 0.0001 & 0.0001 & 0.0001 & 0.0001 \\ 0.0479 & 0.0001 & 0.0019 & 0.4995 & 0.0109 \\ 0.4988 & 0.4995 & 0.0229 & 0.0001 & 0.0001 \\ 0.0001 & 0.0001 & 0.1458 & 0.8408 & 0.0001 \\ 0.0019 & 0.0001 & 0.0019 & 0.4986 & 0.0109 \\ 0.0001 & 0.0001 & 0.0128 & 0.8386 & 0.0001 \\ 0.0020 & 0.0001 & 0.0019 & 0.9258 & 0.0001 \\ 0.0617 & 1.0000 & 0.0608 & 0.0001 & 0.0001 \\ 0.0001 & 0.0001 & 0.2085 & 0.8386 & 0.0001 \\ 0.0605 & 1.0000 & 0.0599 & 0.0001 & 0.0001 \\ 0.0092 & 0.0001 & 0.0019 & 0.4995 & 0.0111 \\ 0.0001 & 0.0599 & 1.0000 & 0.0609 & 0.0001 \\ 0.0018 & 0.4985 & 0.4987 & 0.0018 & 0.0001 \end{bmatrix}$$

The sample evaluation vector  $B = \{0.1765, 0.2079, 0.1159, 0.3432, 0.0040\}$  can be calculated in accordance with Eq. (15), and the weighted average evaluation score  $r$  was 3.2475. In detail,  $r = \frac{0.1765 \times 5 + 0.2079 \times 4 + 0.1159 \times 3 + 0.3432 \times 2 + 0.004 \times 1}{0.1765 + 0.2079 + 0.1159 + 0.3423 + 0.004} = 3.2475$ . The above steps are repeated 100 times to reduce the effect of random factors. Eqs. (19) and (20) indicate that the mean and standard deviation of the comprehensive evaluation scores were 3.246 and 0.0134, respectively.

Finally, the confidence factor was found to be 0.0041 according to formula (21). A small factor leads to a small dispersion of the evaluation results but produces reliable evaluation results. The correlation of the ranks is determined using the Spearman’s correlation coefficient  $r_s$ .

**Table 14:** Measured value of each risk indicator for overturning formwork construction

Item	$U_{1-1}$	$U_{1-2}$	$U_{1-3}$	$U_{1-4}$	$U_{2-1}$	$U_{2-2}$	$U_{2-3}$	$U_{3-1}$	$U_{3-2}$	$U_{3-3}$	$U_{4-1}$	$U_{4-2}$	$U_{4-3}$	$U_{4-4}$
1	9.8	36.67	67.46	8	120	82.36	2.5	30	3	4.5	3	90	5	4
2	9.6	34.20	65.40	8	125	80.30	3	25	3	4.5	3	80	4	4
3	9.8	35.00	65.80	7	120	82.00	3	30	2	5	2	90	5	4

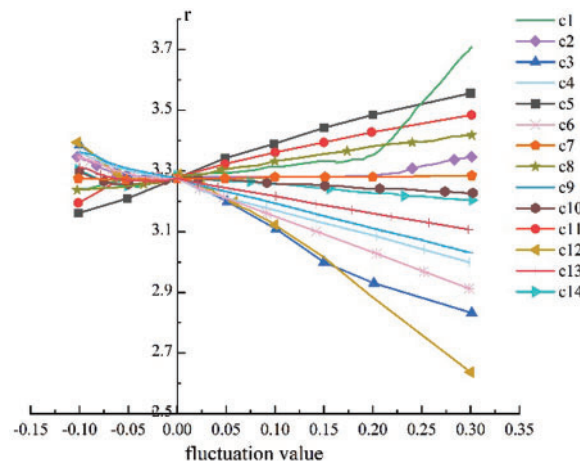
Meanwhile, the evaluation results were compared with the conventional matter–element method and cloud model to verify the effectiveness of the method in this paper. The results of the final evaluation are shown in Table 15.

**Table 15:** Risk assessment results of different evaluation methods

Work condition	Method of this article		Evaluation results of the matter-element method	Evaluation results of cloud model method
	Evaluation results	Confidence factors		
1	III ( $E_{s,r}$ ) = 3.275	0.0041	III	III
2	III ( $E_{s,r}$ ) = 3.165	0.0036	III	IV
3	III ( $E_{s,r}$ ) = 3.421	0.0052	III	III

**3.4 Cloud-Based Risk Assessment**

Sensitivity Analysis is performed based on the risks index system of overturning construction of towering structure. When the weights of risk index in Table 11 changed by  $\pm 0.1$ ,  $\pm 0.05$ ,  $+0.2$ ,  $+0.3$ , we obtained the result shown in Fig. 8.



**Figure 8:** Sensitivity analysis of the weights of risk index

As we can see from Fig. 8, with the weight of the risk indexes  $c_1, c_5, c_{11}$  increasing, the whole risk level deviates from the medium level to larger level gradually and the weight of  $c_1$  is the most sensitive. With the weight of the risk indexes  $c_{12}, c_3, c_6$  increasing, the whole risk level deviates from the medium level to smaller level gradually and the weight of  $c_{12}$  is the most sensitive. The weights change of risk index  $c_7, c_{10}, c_{14}$  have little effect on the risk level, so their sensitivities are weak.

From above sensitivity analysis, we can make a conclusion that  $c_1, c_{12}$  are sensitive factors in the safety risk assessment of overturning construction of towering structure. They are the wind load calculation and the percentage of concrete design strength achieved. In the construction risk management process, these indexes should be analyzed mainly in order to make a better risk aversion measure and reduce the risk of overturning construction of towering structure.

#### 4 Discussion

Table 15 shows that the evaluation results obtained by the evaluation methods in this paper were consistent and the confidence factors were all less than 0.05, indicating that the cloud matter–element coupled model evaluation method for the safety risk assessment of overturning construction of towering structures had a high confidence level. The Spearman's rank test for correlation between the weighted average evaluation score  $r$  and level III, the mean value of  $r_s$  was 0.979, which indicates an extremely large correlation of level III. In addition, the weight calculation in this paper combined the decision results of all experts. Thus, the proposed evaluation model considered the randomness and fuzziness among the risk factors in construction risk assessment. The final result of the risk assessment for the construction of towering structure overturning was level III (medium risk level), which is consistent with the actual situation of the project.

The evaluation results of the cloud-based matter–element coupled model in this paper had high similarity individually compared with the two other commonly used methods. Among them, the assessment results of the proposed method in this paper were consistent with 100% compared with the traditional matter–element method, and two-thirds of the assessment results were consistent compared with the cloud model method. In the two other methods, the proportion of level III was 83.3%. Thus, the assessment results of the cloud matter–element coupled model were representative.

In addition, the cloud matter-coupled model defined the safety risk level by using a comprehensive assessment score mean  $E_{x,r}$ , which can provide more integrated information on the assessment cases compared with other methods. For example, the assessment results of cases 1 and 2 were III. However, the comprehensive assessment score mean of case 1 was 3.275 higher than that of case 2, which indicated that the risk of case 1 was higher than that of case 2. The evaluation method in this paper retains the advantages of traditional matter–element theory in dealing with the incompatibility problem among evaluation indicators. Thus, the introduction of the cloud model can quantitatively convert the ambiguity and randomness in the process of risk assessment of overturning construction to reflect its uncertainty characteristics accurately. Therefore, compared with other evaluation methods, the proposed evaluation method in this paper was comprehensive to evaluate the safety risk level of the overturning construction of towering structures.

#### 5 Conclusion

This paper first adopted the cloud matter–element coupled model to solve the complex uncertainty problem of safety risk assessment of overturning construction of towering structures effectively. This model maximized the advantages of matter–element theory and cloud model to deal with the fuzziness and randomness in the safety risk assessment of the overturning the construction of



towering structures. The improved hierarchical analysis method was then employed on the basis of the cloud model through the aggregation algorithm of the cloud model to bring all the assignments of multi-person decision making into the calculation formula. The determination of the weights of various construction risk indicators was more objective and reliable. Finally, a reinforced concrete cylindrical tower structure overturning construction safety risk assessment was taken as a case to describe the application steps of the proposed method. The evaluation results of the proposed method are also compared with those of the traditional matter–element and cloud model methods to test the application effect of the proposed method. The findings showed that the assessment results of the cloud matter–element coupled model method were consistent with those of two other common methods. Moreover, the confidence factor of each assessment case was less than 0.05, which indicated that the proposed method in this paper was effective in the safety risk assessment of overturning the construction of towering structures. The proposed model in this paper used a comprehensive assessment score mean to determine the safety risk level, which can provide more integrated information regarding the assessment cases compared with other assessment methods, and reflect the uncertain relationship among evaluation indexes more comprehensively compared with other evaluation methods, thus increasing the reliability of the evaluation results.

However, in the process of safety risk assessment of towering structure overturning construction, the classification criteria of the safety risk level of each assessment index are not constant, and the scoring value of the assessment index is susceptible to subjective factors, both of which have an impact on the reasonableness of the assessment results. Therefore, further supplementing and improving the safety risk assessment index system of towering structure overturning construction is necessary for the subsequent research. Moreover, the issues, such as the scoring value of the assessment index and its safety risk classification criteria, should be investigated further to improve the scientificity of the assessment index system and enhance the comprehensiveness and objectivity of the assessment results.

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