

Research Article

Exploration of Normal Stochastic Multicriteria Decision-Making and Legal Dilemma in Project Investment

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The purpose is to help enterprise managers make more accurate, scientific, and objective decisions on the optimal supplier in project investment, improve investment return and overall efficiency, and reduce investment risk. Following an introduction of normal stochastic multicriteria decision-making, a normal stochastic multicriteria decision-making method is proposed based on Bayesian Network- (BN-) Technique for Order Preference by Similarity to an Ideal Solution (TOPSIS) to help investors select suppliers. First, product cost, product quality, and product service are selected as evaluation indexes, and then, a supplier evaluation model is built based on the BN concept for expert evaluation. Second, the evaluation information is integrated. Based on the TOPSIS normal stochastic multicriteria decision-making method, the optimal supplier is selected under the comparison of the fit degree between each supplier and the best and worst solution. Finally, the optimal solution is determined and verified by a simulation example. The results show that in the problem of supplier selection in the retail industry, the BN model can truly provide feedback the product quality, cost, and service situation of each supplier and fully consider their behavior diversity. The normal stochastic multicriteria decision-making method based on TOPSIS can select the suppliers that are most in line with the investors' development goals, especially in the face of good and bad suppliers. BN-based TOPSIS normal stochastic multicriteria decision-making method has strong logic and efficiency, is easy to operate, and has high practical value. Furthermore, the relevant laws of project investment decision-making are discussed through consultation on the relevant literature, thus providing help for the construction of relevant laws in the future.

1. Introduction

Simply, the project investment decision-making is an investment plan made for investment and investment return [1]. In practice, to maximize the investment return, investors often adopt scientific and reasonable ways to conduct a full investigation, analysis, and demonstration on such factors as investment projects, investment scale, and investment direction, and make the judgment and formulate the investment plan accurately and perfect [2, 3]. The most important business of investment enterprises is project investment, while investment decision-making is the most important factor to the success of the enterprise, whose scientificity and reasonableness determine the continuous increase of enterprise benefits [4].

Under the rapid economic development and fierce market competition, many problems have occurred in the investment industry in China, and investors are often blindly confident. Therefore, most of the investment decisions lack scientific rationality, which is believed to be caused by two reasons: first, the investment heat is driven by the Unicorn project; some investors have a strong momentum of development depending on their unique advantages, become unicorns in this field to obtain huge profits, and attract the favor of other investors who will eventually join them; investors' such reckless behavior, which often lacks the consideration of self-factors, undoubtedly increases the risk of investment decision-making. Second, blind confidence is brought by the emerging industries to investors. In the face

of emerging industry investment, many investors in traditional industries are overconfident. Once there is financing demand in an emerging industry, investors will rush in and make decisions without scientific analysis [5].

With the continuous development and improvement of the management field, decision-making is being affected by some single factors (criterion), as well as many interrelated and restrictive factors (criterion). In particular, decision makers often have to consider all these factors (criterion) as much as possible to make accurate and reliable decisions, namely the multicriteria decision-making problem [6]. In practice, many factors can affect decision-making in project investments, such as suppliers' cost control ability, overall product quality, and product-related services. These multicriteria decision-making problems often involve random variables due to the complexity and uncertainty of the situation. Studies have found that the random variables obey or approximately obey the normal distribution, and the criterion is the multicriteria decision-making of random variables that obey the normal distribution, namely, the normal random multicriteria decision-making [7]. To date, enterprises still cannot solve many multicriteria decision-making problems in actual project investments. It is imperative for enterprises to make accurate and scientific decisions under uncertainty [8]. Bayesian Network (BN) has a strong anti-uncertainty ability using conditional probability to express the correlation between various information elements; it can learn and reason under limited, incomplete, and uncertain information conditions. BN can be used to comprehensively evaluate the reliability of suppliers [9]. Technique for Order Preference by Similarity to an Ideal Solution (TOPSIS) is a sort method by approximating the ideal solution; its basic principle reads: first, the initial data matrix is normalized to find out the best scheme and the worst scheme, then the distance is calculated between each alternative scheme and the best scheme and the worst scheme and is sorted by the proximity of each alternative scheme to the best scheme, and the best scheme is found out among the alternatives [10]. BN TOPSIS method can be used for enterprise project investment to help managers make accurate, scientific, and objective decisions on the optimal supplier, and it plays an important role in improving investment return, overall efficiency, and reducing investment risk.

In the process of project investment decision-making, there are more and more problems caused by the lack of laws. The main reasons are as follows: first, there is no legal reference for decision-making due to the lack of relevant laws [11, 12]. At present, the laws for project investment decision-making in China are only written in the government's policy documents, without forming a system of rules and regulations. As a result, the legal base for investment decisions is lacking, so investors often make decisions only from the income without the corresponding approval process [13, 14]. Meanwhile, the existing policy provisions are not implemented effectively [15]. Even under strict requirements, policy loopholes can be found by some speculators.

Thereupon, the purpose is to solve the investment problem and ensure the feasibility, scientificity, and

rationality of project investment decision-making. Here, the BN-based TOPSIS normal stochastic multicriteria decision-making method is used to evaluate the retail industry suppliers and analyze their reliability, thereby help investors select suppliers and make more scientific investment decisions in the complex economic environment. Shortly, the results have some theoretical and reference significance for future project investment decision-making and law-making research and have certain application prospects and economic value. This research aims to provide an important reference for the comprehensive evaluation of optimal suppliers in the process of enterprise project investment, the decision-making, and the construction of relevant laws in the future.

2. Theory and Method

2.1. Normal Stochastic Multicriteria Decision Theory. Gaussian distribution is a very important continuous probability distribution with the characteristics of random variables in mathematics and is also called normal distribution, which literally means the most obvious probability distribution. In practical problems, the probability distribution of many random variables is found to be in the form of normal distribution, because of which normal distribution is widely used in the field of science and engineering. The normal distribution probability curve is called the bell curve because of its symmetry of being low on both sides and high in the middle. In practical applications, Sun believed that the outbreak of plague and its change with time approximately followed the normal distribution and used this function for research as a reference for actual epidemic prevention work [16]. Liu et al. thought that the amount of fertilizer used in rice was normally distributed and used the theory to study the implementation effect of fertilizer in rice [17]. Ye et al. applied normal distribution to study the ornamental characters of different species of cut chrysanthemum [18]. In daily life, there are also phenomena, such as product quality, company sales volume, income, and investment, in which the random variables involved follow or roughly follow the normal distribution [19–21].

With the development of project investment decision-making management, the influence of a single factor should be considered, along with the influence of interrelated and mutual restricting factors for project investment decision-making. To improve the reliability of decision-making, investors should often consider all factors (factors are also called criteria), the process of which is usually called multicriteria decision-making [22, 23]. The criterion of multicriteria decision-making approximately conforms to the normal distribution in reality. The decision information obtained by multicriteria decision-making is aggregated with specific methods, and the schemes are sorted according to the corresponding rules, from which the best scheme is selected according to the results. Decision information generally includes criteria, criteria weight, and subjective preference. More scholars are focusing on the practical problems of normal random variables and have made implementable achievements [24, 25].

Generally, in terms of the analysis on the practical problems of normal stochastic multicriteria decision-making, the scheme set of a multicriteria decision-making problem is assumed to be $A = \{A_1, A_2, \dots, A_m\}$, and the criterion set is $C = \{c_1, c_2, \dots, c_n\}$, based on which the multicriteria decision-making model is established, as shown in Table 1. b_{ij} ($i = 1, 2, \dots, m; j = 1, 2, \dots, n$) is the value of the scheme A_i under the criterion c_j .

The problem-solving process of the multicriteria decision-making model is shown in Figure 1.

2.2. Supplier Evaluation Model Based on BN. BN is a probability network, also known as causal network, which is a widely used mathematical model based on probability theory and can digitally present strong logical reasoning problems through calculation [26]. In short, Bayesian is a vector acyclic graph, which can show the relationship between variables intuitively and systematically and reduce the difficulty of information decision processing [27]. The nodes in the vector acyclic graph can indicate random variables, and the relationship between random variables is represented by directed line segments. The expression reads as follows:

$$BN = |G(I, E), P|. \quad (1)$$

In equation (1), $G(I, E)$ represents a vector acyclic graph, in which I denotes the set of nodes, and E indicates the vector set between nodes; P stands for the joint probability distribution. If $X = (X_i)$, $i \in I$ is a random variable node i , then equation (2) is obtained.

$$P(X) = \sum_{i \in I} P(X_i | X_{pa(i)}). \quad (2)$$

In equation (2), $pa(i)$ represents the factor of the random variable i . If the number of random variables is large, the expression reads as follows:

$$P(X_1, X_2, \dots, X_n) = P(X_n | X_1, X_2, \dots, X_{n-1}) \cdots P(X_2 | X_1) P(X_1). \quad (3)$$

The simple BN structure is shown in Figure 2, where a_4 depends on a_1 , a_2 , and a_3 and a_1 , a_2 , and a_3 have no parent nodes, then $P(a_3) = P(a_4 | a_3, a_2, a_1) P(a_3) P(a_2) P(a_1)$.

Here, when the supplier evaluation model is built with BN, cost, service, and revenue are selected as evaluation indexes. The corresponding network structure is shown in Figures 3–5, respectively.

The selection basis of each evaluation index is as follows.

2.2.1. Product Cost. Cost refers to the total expenditure of an enterprise for manufacturing and selling a specific number of products within a specific period of time, which can directly affect the profit of the enterprise and is one of the important factors in supplier evaluation. The cost item includes purchase, distribution, and waste costs. Specifically, the purchase cost of a supplier includes the total price of goods, procurement, and other related expenses, which are related to the customer's demand and the unit price of

products. Distribution is an important operation link of suppliers. Distribution cost refers to all the inevitable expenses of suppliers in product distribution. According to the actual situation, the distribution cost is determined by customers' personalized needs, distribution distance, and distribution modes. Waste costs may occur during enterprise operation due to improper management, such as redundant personnel, excess inventory, and large material consumption in production, which will increase the total expenditure of the enterprise and reduce enterprise profit. Thus, waste cost directly affects the enterprise management quality.

2.2.2. Product Quality. Product quality refers to the collection of the product features and performance to meet the specified needs and potential needs. Any product is manufactured to meet the needs of users. Product quality is an important index to measure the competitiveness of suppliers. Product quality affects the market price of products and the reputation of enterprises. It is the lifeline of enterprises. The supplier's product manufacturing is mainly divided into new product production, product recycling and storage, and remanufacturing of wasted and expired products. Therefore, the quality item includes three parts: new product quality, inventory product quality, and recycled product quality. The quality of new products is the basis for enterprise operation, profit acquisition, and sound reputation; generally, inventory products can induce losses if not handled properly. Thus, there is a need to pay attention to the quality monitoring of inventory products. Remanufacturing wasted and expired products can save material costs, which is also another form of new product manufacturing that needs considerable quality monitoring. In short, these three product quality indexes are important assessment bases for supplier reliability.

2.2.3. Product Service. The purpose of product services is to enhance user experience and improve customer satisfaction. Suppliers with strong service awareness, good service attitude, and high service quality help improve the utilization value of goods, and good service, in turn, can also strengthen the partnership between retailers and suppliers. The service items are mainly divided into three parts: presales service, in-sales service, and after-sales service. Presales service refers to a series of services, such as providing information, online and offline consultation, and market research and prediction before customers contact the products; it plays an important role in collecting customer data, retaining customers, and expanding the enterprise market. In-sales service is provided to customers in the process of product sales, which is an activity to deeply understand customers' needs and assist them in purchasing the most appropriate products; the quality of in-sales service is a crucial factor of customer's purchase intention. After-sales service refers to various service activities provided following the sold goods or products, the more thoughtful the after-sales service is, the higher the service level of the supplier is; enterprises can

TABLE 1: Multicriteria decision matrix model.

Programs	Criteria				
	c_1		c_2	...	c_n
A_1	b_{11}		b_{12}	...	b_{1n}
A_2	b_{21}		b_{22}	...	b_{2n}
...
A_m	b_{m1}		b_{m2}	...	b_{mn}

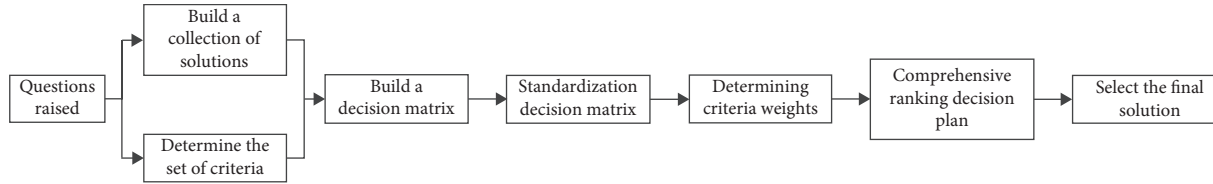


FIGURE 1: Multicriteria decision-making process.

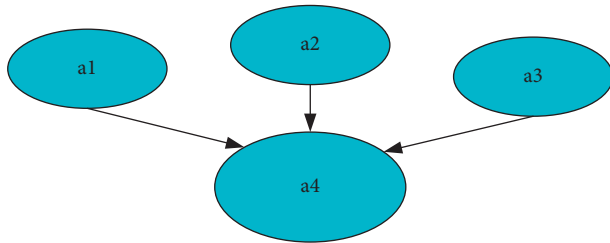


FIGURE 2: Simple BN structure.

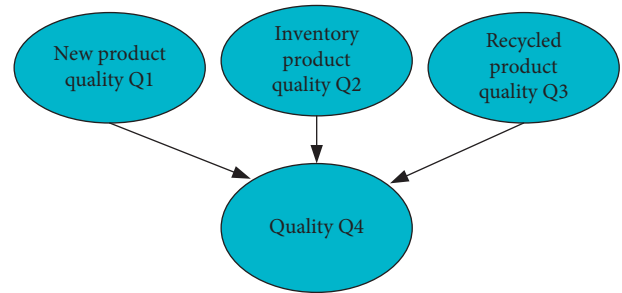


FIGURE 4: BN quality structure.

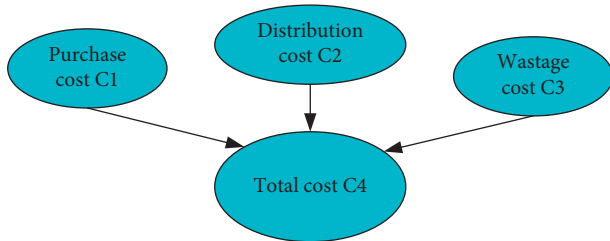


FIGURE 3: BN cost structure.

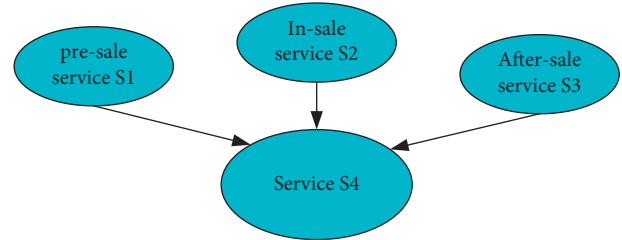


FIGURE 5: BN service structure.

improve their reputation and expand their market share through excellent after-sales service.

After the construction of the network structure, the parameters are set. First, cost parameters setting: The cost approximately obeys the normal distribution, and the cost probability $P(C_4)$ reads as follows:

$$P(C_4) = P(C_4 | C_3, C_2, C_1)P(C_3)P(C_2)P(C_1). \quad (4)$$

Similarly, according to the requirements of investors, the quality probability $P(Q_4)$ reads as follows:

$$P(Q_4) = P(Q_4 | Q_3, Q_2, Q_1)P(Q_3)P(Q_2)P(Q_1). \quad (5)$$

According to the consumer satisfaction survey, the service probability $P(S_4)$ reads as follows:

$$P(S_4) = P(S_4 | S_3, S_2, S_1)P(S_3)P(S_2)P(S_1). \quad (6)$$

2.3. Normal Stochastic Multicriteria Decision-Making Based on TOPSIS. TOPSIS is a ranking method proposed by C. L. Hwang and K. Yoon in 1981, which is based on the degree of fit between the evaluation object and the ideal target. Its essence is to sort the evaluation object through distance calculation and analysis between the evaluation object and the optimal solution and between the evaluation object and the worst solution. If the evaluation object is closest to the

optimal solution and farthest away from the worst solution, the result is the best; otherwise, the result is not optimal. Each index of the optimal solution reaches the optimal value of each evaluation index. Each index value of the worst solution reaches the worst value of each evaluation index. Since its proposal, the TOPSIS method has been used in the field of multicriteria decision-making, has also been widely expanded, and is distinguished according to different classification standards. For example, according to the number of decision-making experts, the TOPSIS method can be divided into individual decision-making and group decision-making; according to the attribute and weight of criterion, it can be divided into interval type, fuzzy type, and mixed type. In practice, TOPSIS is applied to specific issues, such as brand valuation, risk assessment, and importance evaluation [28–31].

The decision-making steps are as follows:

- (1) If there are m alternative suppliers and n evaluation indexes (also called factor or criterion), then the initial judgment matrix reads as follows:

$$Y = [y_{ij}]_{m \times n} \quad (i = 1, 2, \dots, m; j = 1, 2, \dots, n). \quad (7)$$

- (2) The initial judgment matrix is normalized, and the normalized matrix reads as follows:

$$Z = [z_{ij}]_{m \times n} \quad (i = 1, 2, \dots, m; j = 1, 2, \dots, n). \quad (8)$$

- (3) According to the normalized matrix $Z = [z_{ij}]_{m \times n}$, the entropy of the j th evaluation index reads as follows:

$$E_j = -k \sum_{i=1}^m z_{ij} * \ln z_{ij}. \quad (9)$$

In equation (9), $k = 1/\ln m > 0$, $E_j > 0$. If the j th evaluation index of the alternative suppliers is the same, then $y_{1j} = y_{2j} = \dots = y_{mj}$ and $y_{ij} = 1/m$, and at this time, the entropy E_j has the maximum value $E_j = 1$.

- (4) The mutation coefficient and weight of the j th evaluation index calculate as follows:

$$G_j = 1 - E_j \quad (j = 1, 2, \dots, n), \quad (10)$$

$$w_j = \frac{G_j}{\sum_{j=1}^n G_j} \quad (j = 1, 2, \dots, n). \quad (11)$$

In equation (10) and (11), G_j and w_j represent the value of mutation coefficient and weight of the j th evaluation index, respectively.

- (5) According to the weight of each index calculated by equations (8) and (9), the weighted normalized matrix $P = (p_{ij})_{m \times n}$ calculates as follows:

$$p_{ij} = w_j z_{ij} \quad (i \in m, j \in n). \quad (12)$$

- (6) The calculation of positive and negative ideal solutions P^+ and P^- reads as follows:

$$P^+ = \{p_1^+, p_2^+, \dots, p_n^+\} = \left\{ \max_i p_{ij} \mid j \in T_1, \min_i p_{ij} \mid j \in T_2 \right\}, \quad (13)$$

$$P^- = \{p_1^-, p_2^-, \dots, p_n^-\} = \left\{ \max_i p_{ij} \mid j \in T_2, \min_i p_{ij} \mid j \in T_1 \right\}. \quad (14)$$

In equation (13) and (14), T_1 represents the benefit evaluation index and T_2 denotes the cost evaluation index, which corresponds to the proposed cost, quality, and service. Specifically, the cost evaluation index corresponds to the cost, while the benefit evaluation index corresponds to the quality and service.

- (7) The distances between the suppliers and the positive ideal solution P^+ and between the suppliers and the negative ideal solution P^- calculate as follows:

$$d_i^+ = \sqrt{\sum_j^n (p_{ij} - P^+)^2} \quad (i = 1, 2, \dots, m), \quad (15)$$

$$d_i^- = \sqrt{\sum_j^n (p_{ij} - P^-)^2} \quad (i = 1, 2, \dots, m). \quad (16)$$

- (8) The degree of fit between the evaluation suppliers and the positive ideal solution is calculated, and the best supplier is selected, as shown in the following:

$$C_i^+ = \frac{d_i^-}{d_i^+ + d_i^-}. \quad (17)$$

The greater the degree of fit is, the more reliable the supplier is, and the more worthy of selection is. On the contrary, the smaller the degree of it is, the less reliable the supplier is, and the less worthy of selection is. Finally, the best supplier is selected according to the sorting size.

2.4. Case Simulation. Due to the fierce and dynamic market competition, the form of competition among investors has also changed and is gradually inclined to the supply chain. Lv et al. proposed that compared with joint venture suppliers, self-operated retailers had higher retail prices and lower unit operating costs, thus, lower operating costs [32]. When investors make decisions, they should consider supply chains management, such as evaluation and selection, commodity design, commodity quality, commodity sales, and commodity transportation, of which the evaluation and selection of suppliers are the most important, and it is also the key link of the whole decision-making. A proper supplier selection can save costs and benefits the follow-up development. Therefore, accurate, objective, comprehensive, and reliable evaluation and decision-making of suppliers will promote the healthy development of the investment project and obtain the maximum benefits.

Given various practical problems related to multicriteria decision-making, here, it is assumed that an investment

enterprise will invest in a retail project, and there are three suppliers A, B, and C to choose from. BN-based TOPSIS normal stochastic multicriteria decision-making method can analyze and evaluate these suppliers, thereby finding the best supplier. The experimental results can evaluate the proposed method's reliability and effectiveness in practical application. Firstly, according to the conditional probability distribution determined by the expert knowledge and the prior distribution of various evaluation indexes of each supplier, the BN simulation analysis is carried out using Netica. Through probability inference, the probability of True cost, quality, and service of each supplier is obtained, and the percentage probability is taken as the reliability score of the supplier. Secondly, these suppliers are analyzed and evaluated based on TOPSIS normal stochastic multicriteria decision-making method. Finally, the accuracy of the results is verified under the comparative analysis of references.

3. Results and Analysis

3.1. Supplier Evaluation Results Based on BN. It is assumed that investors have the following requirements when choosing suppliers: (1) *Cost*. The purchase cost is controlled within 20,000 RMB, the distribution cost is controlled within 2,000 RMB, and the wastage cost is controlled within 1,000 RMB. (2) *Quality*. The damage rate of new products is less than 3%, the damage rate of inventory products is less than 3%, and the damage rate of recycled products returned to the factory is less than 4%. (3) *Service*. The satisfaction feedback of presales service shall be at least 95%, the satisfaction feedback of the in-sale service shall not be less than 85%, and the satisfaction feedback of after-sales service shall not be less than 70%. The evaluation indexes of each supplier are determined by expert interviews and field investigation, as shown in Table 2.

The Netica is employed to analyze the evaluation index of each supplier, and the obtained statistical data are used for the reliability evaluation of the proposed method. The details are shown in Figure 6-8 (only the score of supplier A is shown).

The cost, quality, and service evaluation scores of the remaining two suppliers B and C are obtained according to the same operation and all scores are summarized in Table 3:

3.2. TOPSIS Normal Stochastic Multicriteria Decision-Making Method. Given the above-mentioned supplier selection decision problem, TOPSIS normal stochastic multicriteria decision-making method is used for sorting:

- (1) Table 3 reveals that a judgment matrix is constructed to make a preliminary judgment on the evaluation of each supplier. The expression of the preliminary judgment matrix reads as follows:

$$Y = \begin{bmatrix} 82.3 & 89.8 & 85.7 \\ 83.4 & 85.2 & 87.8 \\ 76.9 & 90.3 & 81.5 \end{bmatrix}. \quad (18)$$

- (2) According to equation (8), the initial judgment matrix Y is normalized by column, and the expression of the normalized matrix Z reads as follows:

$$Z = \begin{bmatrix} 0.86 & 0.90 & 0.67 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}. \quad (19)$$

- (3) According to equations (9) to (17), TOPSIS calculation is performed using MATLAB to calculate the fit degree of alternative suppliers to positive ideal solution p^+ and negative ideal solution p^- , as shown in Table 4.

Table 4 demonstrates the fitting degree of participating suppliers A, B, and C with the positive ideal solution. According to their size, the order is $B > C > A$. Thus, supplier B has the best fitting degree and is the best choice for investment.

Considering the verification of the reliability and accuracy of the proposed method, other methods are chosen based on the literature review [33] and used for comparative analysis. According to the five optional suppliers and nine evaluation indexes mentioned in the literature, three suppliers are selected respectively, and the three indexes (buying price, inventory level, and delivery reliability) are specifically analyzed. Generally, the buying price plays a decisive role in reducing the investment cost and improving the competitiveness of investors; the inventory level of the supplier can reflect the hardware strength of the supplier and the production efficiency of the market demand and can judge whether it can meet the demand of investors quickly; finally, the reliability of delivery shows the supplier's ability to deliver timely. When making decisions, investors have to consider carefully. The stronger the reliability is, the more reassuring the investor is, and the more secure the development of the investment project is.

Subsequently, the selected data are sorted out. The three candidate suppliers are set as $Q = \{Q_1, Q_2, Q_3\}$, and the three evaluation criteria are set as $P = \{P_1, P_2, P_3\}$. Here, P_1, P_2, P_3 represent buying price, inventory level, and delivery reliability, respectively. Three decision-making experts with equal decision-making authority are invited for evaluation, and the evaluation results of each evaluation index of each supplier follow the normal random distribution. Table 5 is the decision matrix, and the best supplier is selected.

Regarding the decision-making problem, the calculation process is as follows:

- (1) According to the decision matrix of normal random variables given in Table 5, the optimal solution and the worst solution read as follows:

$$\text{Optimal solution: } C^+ : \{N(462, 112^2)N(171, 32^2)N(15, 6^2)\}.$$

$$\text{Worst solution: } C^- : \{N(449, 131^2)N(168, 37^2)N(5, 7^2)\}.$$

- (2) According to equations (15) and (16), the distances are calculated between the alternative solution and

TABLE 2: Prior distribution of evaluation indexes of each participating supplier.

Evaluation index	Participating supplier		
	A	B	C
Purchase cost C_1	(19000, 200^2)	(19500, 200^2)	(18500, 200^2)
Distribution cost C_2	(1800, 100^2)	(1900, 100^2)	(1800, 100^2)
Wastage cost C_3	(800, 10^2)	(900, 10^2)	(850, 10^2)
New product quality Q_1	(150, 4)	(200, 5)	(180, 5)
Inventory product quality Q_2	(100, 2)	(300, 8)	(150, 4)
Recycled products quality Q_3	(100, 3)	(120, 4)	(200, 7)
Presale service S_1	(0.95, 0.05)	(0.98, 0.02)	(0.95, 0.05)
In-sale service S_2	(0.85, 0.15)	(0.90, 0.10)	(0.85, 0.15)
After-sale service S_3	(0.70, 0.30)	(0.75, 0.25)	(0.70, 0.30)

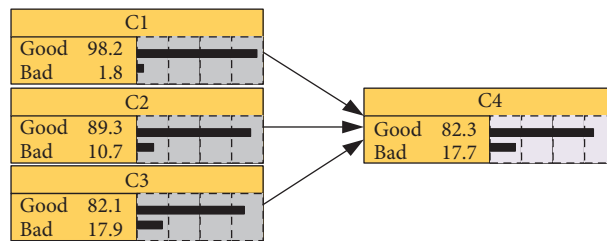


FIGURE 6: Cost evaluation score of supplier A.

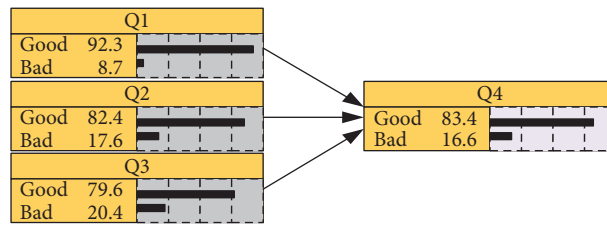


FIGURE 7: Quality evaluation score of supplier B.

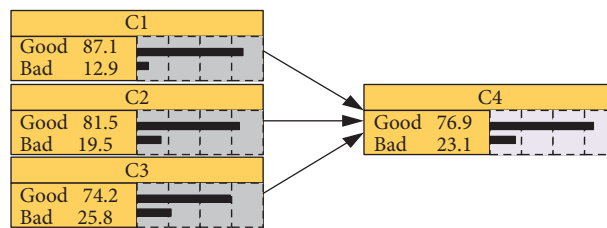


FIGURE 8: Service evaluation score of supplier A.

TABLE 3: Summary of reliability score of each supplier evaluation index.

Evaluation index	Participating supplier		
	A	B	C
Cost	82.3	89.8	85.7
Quality	83.4	85.2	87.8
Service	76.9	90.3	81.5

the criterion of the optimal solution and between the alternative solution and the criterion of the worst solution, as shown in Figures 9 and 10.

(3) The distance D_i^+ from each solution to the optimal solution, and the distance D_i^- from each solution to the worst solution are calculated.

TABLE 4: The best fit degree of all participating suppliers.

	Participating supplier		
	A	B	C
Positive ideal solution	0.175	0	0.027
Negative ideal solution	0	0.132	0.126
Fitting degree	0	1	0.831

TABLE 5: Normal random multicriteria decision matrix.

Supplier	Factors		
	P_1	P_2	P_3
Q_1	$N(449, 131^2)$	$N(171, 37^2)$	$N(5, 7^2)$
Q_2	$N(454, 112^2)$	$N(168, 32^2)$	$N(10, 7^2)$
Q_3	$N(462, 131^2)$	$N(170, 37^2)$	$N(15, 6^2)$

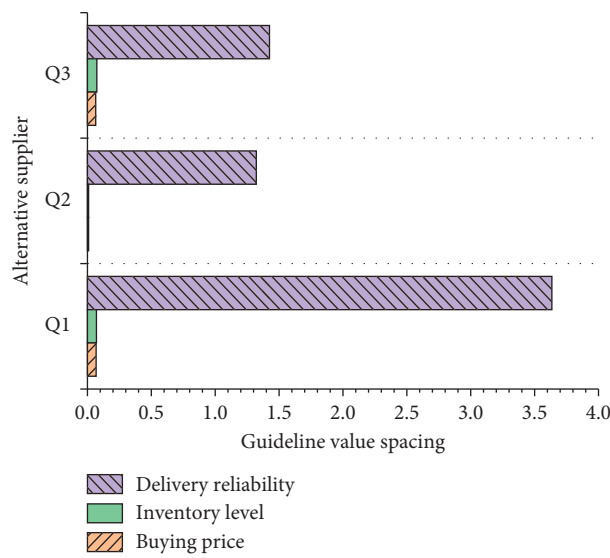


FIGURE 9: Distance between evaluation indexes of alternative solutions and the optimal solution.

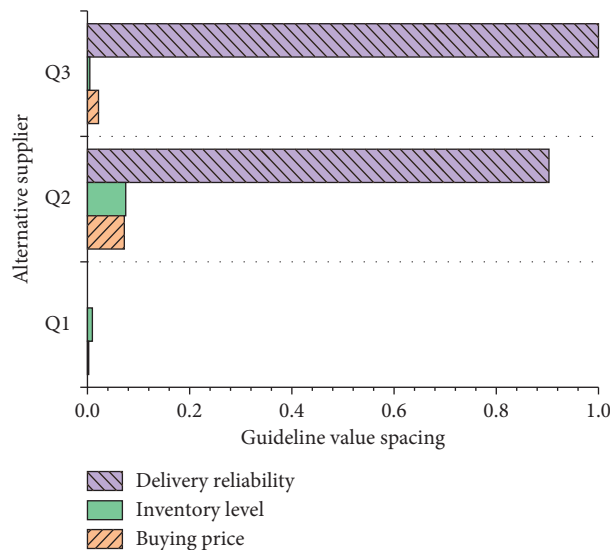


FIGURE 10: Distance between evaluation indexes of alternative solutions and the worst solution.

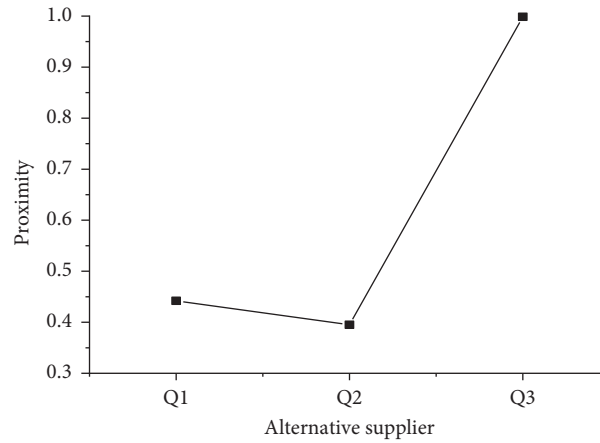


FIGURE 11: Closeness of the alternatives to the optimal one \tilde{C}_i .

$$D_1^+ = 0.516528, D_2^+ = 0.146973, \text{ and } D_3^+ = 0.173309.$$

$$D_1^- = 0.001499, D_2^- = 0.116417, \text{ and } D_3^- = 0.113148.$$

- (4) The closeness of each solution to the optimal solution is calculated by equation (17), and the solutions are sorted, as shown in Figure 11.

According to the closeness of each solution, the ranking result of alternative solutions in Figure 11 is $Q2 < Q1 < Q3$; that is, Q3 is the best supplier. The calculation results are consistent with the literature results, which shows that the method is more reliable and effective in helping investors make decisions.

4. Conclusion

Here, the normal random variables are chosen to deal with the stochastic multicriteria decision-making problem in which the criteria are a normal random variable and the weight of the criteria is determined, and the distances between the criteria of the alternatives and the optimal and the worst alternatives are found out. Meanwhile, product cost, product quality, and product service are selected as evaluation indexes, and a supplier evaluation model is built based on the BN concept. After expert evaluation, the obtained evaluation information is integrated. Then, based on TOPSIS normal stochastic multicriteria decision-making method, the most suitable supplier is selected under the comparison of the fit degree between each supplier and the optimal and the worst solution. Here, case analysis on the retail industry is conducted through the simulation experiment. The results suggest that in terms of decision-making for optimal supplier selection in the retail industry, the BN model can truly feed back the product quality, product cost, and product service of each supplier and fully consider the diversity of their behavior; the normal stochastic multicriteria decision-making method based on TOPSIS can select the optimal suppliers for investors; under complex supplier market, BN TOPSIS normal stochastic multicriteria decision-making method has strong logic and efficiency and is easy to operate; it has a high value in practical applications. Regarding the

legislative proposal, currently, project investment decision-making can rely on nothing but mainly based on the judgment of policy management and other practical problems. Here, it is argued that the legislation should be sped up for project investment decision-making; the accountability system should be implemented at an individual level. The shortcomings and proposed suggestions are summarized: (1) Only three supplier evaluation indexes (product cost, product quality, and product service) are selected, which is not comprehensive enough. In the follow-up study, the supplier's technology, environmental protection, and flexibility will be added to perfect the evaluation system; (2) The proposed stochastic multicriteria decision-making method is about selective ranking. In the follow-up study, the classification problem in stochastic multicriteria decision-making can be further explored; these stochastic multicriteria decision-making methods can be extended to more fields to enhance their practical applications.

Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

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