



# An Integrated Neutrosophic SWARA and VIKOR Method for Ranking Risks of Green Supply Chain

Abdullah Ali Salamai<sup>1</sup>

<sup>1</sup>Community college, Jazan University, Jazan, Kingdom of Saudi Arabia, Email: abSalamai@jazanu.edu.sa

**Abstract**: The green supply chain (GSC) plays a vital role for companies and organizations. Though there are several risks thread GSC Hence, these risks need to be ranked for companies. So, the goal of this study ranking these risks under a neutrosophic environment due to this problem contains uncertain information. So, this study proposed a multi-criteria decision making (MCDM) for dealing with conflicting criteria and used MCDM methods. This study introduces an integrated model with Stepwise Weight Assessment Ratio Analysis (SWARA) and visekriterijumsko kompromisno rangiranje (VIKOR). The SWARA method is used to calculate the weights of criteria and the VIKOR method is used to rank the risks of GSC based on six main criteria and twenty sub-criteria with ten risks (alternatives). Then the proposed model was evaluated by a numerical example. Finally, the sensitivity analysis is conducted.

Keywords: Green Supply Chain (GSC), Neutrosophic, SWARA, VIKOR, Risks, SVNSs

## 1. Introduction

Green Supply Chain (GSC) introduces several benefits and advantages to companies like increasing the financial power and enable companies to share their market strongly by improving the capacity of the environment and reduce the negative impact of environmental[1]. The gaining advantage competitive and keep it is a vital role for the company in performing the creativities green in GSC[2, 3]. The success of establishments in the supply chain becomes more difficult[4].

There are several risks when performing the initiatives green in GSC[5]. Reduce cost and increase customer satisfaction are the goals for improving performance in the supply chain[6]. The risks of GSC make many problems in operations and reduce GSC performance[7]. There are many problems that may result from risks of GSC like negative impact of environmental, issues of quality, failure in operations, reduce performance, and disarray of supply materials[8]. So, these risks are necessary to analysis and ranking for companies for adoption the initiatives green in GSC.

neutrosophic sets. It work with three value (T,I,F).

The analysis and rank risks in GSC contain vague, inconsistent, and uncertain information[7]. To overcome this uncertainty and vague information some studies proposed fuzzy sets. They used linguistic terms for their assessment. But the fuzzy set cannot deal with indeterminacy [9]. So, this study proposed the neutrosophic set to overcome this uncertainty information. The neutrosophic set is generalized to fuzzy sets. It is contained with truth, indeterminacy, and false (T.I,F). The neutrosophic sets are proposed in several fields like manufacturing, healthcare and others [10-12]. In this paper proposed he single valued neutrosophic sets (SVNSs). The SVNSs is a subset of

The GSC contains many conflict criteria. So, use multi-criteria decision-making (MCDM) to deal with this problem. This study proposed two MCDM methods. First, the SWARA method to calculate the weights of criteria. The SWARA has two advantages first, the criteria are compensatory. Second, the criteria are independent of each other. Then the VIKOR method is applied to rank the risks of GSC [13]. The VIKOR method is used to solve problems with conflicting criteria [14]. This paper used the VIKOR method for ranking the risks of GSC.

The rest of this paper is organized as follows: the literature review is presented in section 2. Section 3 introduces the methodology of this paper. Section 4 introduces the numerical example to validate the methodology. The sensitivity analysis is presented in section 5. Finally, section 6 introduces the conclusion of this paper.

## 2. Literature Review

There are several works to evaluate and analyze the risks of GSC[15, 16]. For instance Allen H.Hu et al.[17] used the analysis of effects and failures mode to rank and analysis the risks of the green component to with the European Union in compliance. They used the fuzzy AHP to calculate the weights of four criteria. Then the risks are ranked for each green component. Zhen-kun Yang and Jian Li [18] are ranked the risks of GSC and describe the operations of GSC. They used the fuzzy AHP to calculate the weights of the criteria and then rank the risks of GSC. The aim of their study to introduce the risk control of organization and reliability for selection of supply chain.

Dan-li Du et al. [19] used the gray theory for assessing the risks manufacturing of GSC. The aim of their study that provides stability of running the GSC and evade risks appearing. Li Qianlei [20] used the systematic analysis to recognize the risks of products of agriculture GSC and introduce measures risks management for agriculture products GSC. Xiaojun Wang et al. [7] proposed two phase fuzzy AHP for evaluation risks of GSC.

This problem contains uncertain information[21]. Wei Wang et al. [22] discuss the demand uncertainty in the GSC. Kuo-Jui Wu et al. [23] discuss uncertainty for exploring decisive factors in GSC practice. They used the fuzzy DEMATEL to overcome this uncertainty. To overcome this uncertainty information proposed single-valued neutrosophic sets. M.Abdel-Baset et al. [24] proposed single-valued neutrosophic sets to assess the GSC management practices. The risks of GSC contain different conflicting criteria so, the MCDM is proposed to deal with these criteria. [25] Morteza Yazdani proposed an integrated MCDM for GSC. Hsiu Mei Wong Chen et al. [26] proposed the fuzzy MCDM methods for GS selection.

This study proposed the SWARA method for calculating the weights of criteria. Serap Akcan and Mehmet Ali Taş [27] proposed the SWARA method for green supplier assessment to decrease environmental risk factors. Selçuk Perçin [28] proposed a fuzzy SWARA method for outsourcing provider selection. After calculating the weights of criteria then needs to rank the risks of GSC. The VIKOR method is used to rank the risks of GSC. Reza Rostamzadeh et al. [29] proposed the fuzzy VIKOR method for assessment GSC management practices. Xiaolu Zhang and Xiaoming Xing [30] introduce the VIKOR method for assessing the GSC initiatives.

From the literature review, no research takes into consideration the indeterminacy value. So, in this study introduce the SVNSs to overcome this uncertain information. Then the SWARA and VIKOR methods are not used in previous research with this problem. So, the SWARA is used to calcite the weights of criteria and the VIKOR method to rank the risks of GSC.

#### 3. Methodology

The methodology of this study is proposed for ranking the risks of GSC, which it contains from two main stages. The first stage collects criteria and risks dimension and proposed the SWARA method to calculate the weights of criteria. The second stage proposed the VIKOR method to rank the risks of GSC.

#### 3.1. SWARA Method

This method is used to calculate the weights of criteria. It is a relatively simple use. Fig 1. Show the SWARA steps. The steps of SWARA is organized as follow [13]:





**Step 1.** Start with the opinions of three experts and decision-makers with the linguistic terms in Table 1.

**Step 2.** Convert the value of single-valued neutrosophic numbers (SVNNs) into crisp value by using the following score function

$$s(P_m^D) = \frac{2 + T_m^D - I_m^D - F_m^D}{3}$$
(1)

 $T^D_m, I^D_m, F^D_m\,$  Presents truth, indeterminacy, and falsity of the SVNNs and D refers to decision-makers

Step 3. Aggregate the crisp value to obtain one value by using the following equation

$$P_m = \frac{\sum_{D=1}^{D} P_m}{D} \tag{2}$$

Step 4. Calculate the coefficient (C) by using the following equation

$$C_m = \begin{cases} 1, \ m = 1 \\ P_{m+1}, \ m > 1 \end{cases} m = 1,2,3, \dots n \text{ number of criteria}$$
(3)

Step 5. Calculate the initial weight by using the following equation

$$A_m = \begin{cases} 1, \ m = 1\\ \frac{A_{m-1}}{c_m}, \ m > 1 \end{cases} m = 1, 2, 3, \dots n \text{ number of criteria}$$
(4)

Step 6. Compute the relative weight of criteria by using the following equation

$$W_m = \frac{A_m}{\sum_{m=1}^n A_m} \tag{5}$$

# 3.2 VIKOR Method

The VIKOR method is used to rank the risks of GSC Fig 2. Show the steps of the VIKOR method. The steps of the VIKOR method is organized as follow [31]:



Fig 2. The steps of the VIKOR method

**Step 7.** Start with building the decision matrix between the criteria and alternatives (risks) by opinions of experts with the linguistic term in Table 1 by using the following equation. Then convert the SVNNs to the crisp value by Eq. (1). Then combine the decision matrix into one matrix by using Eq. (2).

$$P^{D} = \begin{bmatrix} P_{11}^{D} & \cdots & P_{1y}^{D} \\ \vdots & \ddots & \vdots \\ P_{m1}^{D} & \cdots & P_{my}^{D} \end{bmatrix} m = 1,2,3, \dots n ; y = 1,2,3, \dots x$$
(6)

**Step 8.** Calculate the best and worst solution for positive and negative criteria Best solution  $P_m^+ = (P_{my})_{max}$  for positive criteria  $P_m^+ = (P_{my})_{min}$  for negative criteria (7) Worst solution  $P_m^- = (P_{my})_{min}$  for positive criteria  $P_m^- = (P_{my})_{max}$  for negative criteria (8)

**Step 9.** Calculate the value of  $g_m$ ,  $h_m$  by using the following equation

$$g_m = \sum_{y=1}^{x} (W_y * \frac{P_m^+ - P_{my}}{P_m^+ - P_m^-})$$
(9)

$$h_m = \max_{y} (W_y * \frac{P_m^+ - P_{my}}{P_m^+ - P_m^-})$$
(10)

**Step 10.** Calculate the value of  $Z_m$  by using the following equation

$$Z_m = f * \frac{g_m - g^*}{g^- - g^*} + (1 - f) * \frac{h_m - h^*}{h^- - h^*}$$
(11)

Where  $g^* = \min g_m$ ,  $g^- = \max g_m$ ,  $h^* = \min h_m$ ,  $h^- = \max h_m$  and f is recognized as a weight for the strategy of maximum group utility, whereas (1 - f) is the weight of the separate remorse. Usually, the value of f is set as 0.5. Though, f can set any value from 0 to 1.

Step 11. Rank the risks according to ascending value of  $Z_m$ 



Fig 3. The criteria and risks (alternatives) of GSC

## 4. Numerical Example and discussion

The criteria and risks of GSC are extracted from the literature review. Fig 3. shows the criteria and alternatives of this problem. Firstly, the weights of criteria are obtained from section 3.1 by the SWARA method. This problem introduces the three decision-makers and the value of SVNNs is presented in Table 1. The SVNNs contain from (T,I,F). After taking the opinions of experts the three value (T,I,F) is converted to one value by score function by Eq. (1). Then aggregate the three values of three decision-makers into one value by using Eq. (2). Then the coefficient value is obtained by using Eq. (3). Then the initial weight is obtained by using Eq. (4). Then the weights of main and subcriteria are obtained by using Eq. (5) in Table 2.

The weights of main criteria found that the operational risks are the highest value with 0.369 and demand risks is the lowest weight with value 0.04. The weights of sub-criteria found that the green

technology level is the height weight with value 0.1617 and failures of getting keeping design risks is the lowest weight value 0.00655. Fig 4. Show the weights of the main criteria. Fig 5. Show the weights of sub-criteria.



Fig 4. The weights of main criteria.





| Table 1. SVNSs scale. |  |  |  |  |
|-----------------------|--|--|--|--|
| Linguistic Term       | Single valued neutrosophic numbers (SVNNs) |  |  |  |

| Very Wicked   | <0.15,0.8,0.8>   |
|---------------|------------------|
| Wicked        | <0.25,0.7,0.7>   |
| Medium Wicked | <0.35,0.6,0.6>   |
| Medium        | <0.45,0.5,0.45>  |
| Medium Moral  | <0.6,0.4,0.35>   |
| Moral         | <0.75,0.35,0.25> |
| Very Moral    | <0.85,0.2,0.2>   |

| Table 2. The weights of main and sub criteria. |         |                 |          |  |  |  |
|--|---------|-----------------|----------|--|--|--|
| Main Criteria                                  | Weights | Sub Criteria    | Weights  |  |  |  |
|  |         | C11             | 0.161754 |  |  |  |
| $C_1$  | 0.369   | C12             | 0.103247 |  |  |  |
|  |         | C13             | 0.061742 |  |  |  |
|  |         | C14             | 0.042257 |  |  |  |
|  |         | C21             | 0.066624 |  |  |  |
| C2   |         | C22             | 0.025625 |  |  |  |
|  | 0.252   | C <sub>23</sub> | 0.010752 |  |  |  |
|  |         | C31             | 0.145019 |  |  |  |
| C <sub>3</sub>                                 |         | C32             | 0.070173 |  |  |  |
|  | 0.169   | C33             | 0.027164 |  |  |  |
|  |         | C34             | 0.009644 |  |  |  |
|  |         | C41             | 0.10184  |  |  |  |
| C4   |         | C42             | 0.040736 |  |  |  |
|  |         | C43             | 0.019872 |  |  |  |
|  | 0.103   | C44             | 0.006551 |  |  |  |
| <b>C</b> 5                                     |         | C51             | 0.02644  |  |  |  |
|  | 0.067   | C52             | 0.01356  |  |  |  |
|  |         | C61             | 0.039717 |  |  |  |
| <b>C</b> <sub>6</sub>                          |         | C62             | 0.019219 |  |  |  |
|  | 0.04    | C63             | 0.008064 |  |  |  |

Applying the VIKOR method for ranking the risks of GSC. Start with building the decision matrix between criteria and risks with the SVNNs in Table 1 by opinions of three experts by Eq. (6). Then covert the SVNNs to the crisp value by using Eq. (1). Then combine the three decision matrix into one matrix by using Eq. (2) in Table 3. Then the best and worst solution is obtaining by using Eqs. (7,8), the procurement criteria are the negative criteria and the rest of the criteria is positive criteria. The value of  $g_m$ ,  $h_m$  is obtained by Eqs. (9,10) in Table 4. Then the value of  $Z_m$  is obtained by using Eq. (11) in Table 4. Finally, the risks of GSC is ranking according to ascending order of  $Z_m$  in Table 4.

As result of VIKOR, the  $R_7$  is the highest rank and the  $R_3$  is the lowest rank. Fig 6. Show the rank of risks by VIKOR method.



Fig 6. The rank of Risks GSC

| Criteria/Risks | C11    | C12    | C13    | C14    | C <sub>21</sub> | C22    | C23    | C31    | C32   | C33   |
|----------------|--------|--------|--------|--------|-----------------|--------|--------|--------|-------|-------|
| R1             | 0.6055 | 0.4277 | 0.2833 | 0.6778 | 0.6778          | 0.6055 | 0.6778 | 0.6444 | 0.427 | 0.283 |
| R2             | 0.6444 | 0.7833 | 0.5    | 0.6778 | 0.7500          | 0.6055 | 0.6055 | 0.7500 | 0.427 | 0.5   |
| R3             | 0.3555 | 0.4277 | 0.5333 | 0.5722 | 0.6444          | 0.5    | 0.6055 | 0.5333 | 0.427 | 0.783 |
| $\mathbb{R}_4$ | 0.2833 | 0.7833 | 0.7833 | 0.4277 | 0.7500          | 0.6778 | 0.5333 | 0.4611 | 0.677 | 0.783 |
| R5             | 0.7833 | 0.6055 | 0.6055 | 0.5    | 0.4277          | 0.4277 | 0.7500 | 0.5722 | 0.750 | 0.711 |
| R <sub>6</sub> | 0.711  | 0.677  | 0.2833 | 0.283  | 0.5333          | 0.427  | 0.355  | 0.283  | 0.572 | 0.750 |
| R7             | 0.816  | 0.750  | 0.6055 | 0.750  | 0.7500          | 0.750  | 0.355  | 0.638  | 0.427 | 0.783 |
| R8             | 0.677  | 0.427  | 0.7833 | 0.2833 | 0.7500          | 0.533  | 0.716  | 0.783  | 0.750 | 0.283 |
| R9             | 0.677  | 0.572  | 0.5    | 0.5722 | 0.3555          | 0.750  | 0.711  | 0.572  | 0.5   | 0.355 |
| R10            | 0.716  | 0.283  | 0.4277 | 0.7830 | 0.7167          | 0.816  | 0.283  | 0.427  | 0.355 | 0.750 |
|                |        |        |        |        |                 |        |        |        |       |       |
|                | C34    | C41    | C42    | C43    | C44             | C51    | C52    | C61    | C62   | C63   |
| $R_1$          | 0.644  | 0.605  | 0.461  | 0.750  | 0.355           | 0.783  | 0.750  | 0.427  | 0.605 | 0.750 |
| R2             | 0.783  | 0.427  | 0.355  | 0.677  | 0.533           | 0.644  | 0.355  | 0.283  | 0.605 | 0.572 |
| R3             | 0.533  | 0.750  | 0.605  | 0.716  | 0.5             | 0.5    | 0.711  | 0.605  | 0.5   | 0.638 |
| $R_4$          | 0.605  | 0.283  | 0.783  | 0.5    | 0.750           | 0.283  | 0.750  | 0.572  | 0.750 | 0.638 |
| R5             | 0.572  | 0.605  | 0.644  | 0.283  | 0.783           | 0.427  | 0.427  | 0.750  | 0.783 | 0.572 |
| R <sub>6</sub> | 0.355  | 0.283  | 0.283  | 0.283  | 0.533           | 0.716  | 0.283  | 0.750  | 0.427 | 0.533 |
| R7             | 0.283  | 0.716  | 0.750  | 0.355  | 0.783           | 0.750  | 0.427  | 0.572  | 0.644 | 0.783 |

Table 3. The decision matrix between criteria and alternatives.

| R8  | 0.355 | 0.750 | 0.355 | 0.427 | 0.750 | 0.5   | 0.427 | 0.711 | 0.355 | 0.2833 |
|-----|-------|-------|-------|-------|-------|-------|-------|-------|-------|--------|
| R9  | 0.750 | 0.716 | 0.572 | 0.750 | 0.750 | 0.605 | 0.716 | 0.605 | 0.605 | 0.2833 |
| R10 | 0.427 | 0.461 | 0.605 | 0.750 | 0.572 | 0.783 | 0.644 | 0.5   | 0.750 | 0.7167 |

| Risks             | $g_m$   | $h_m$    | z <sub>m</sub> | Rank |
|-------------------|---------|----------|----------------|------|
| R1                | 0.46789 | 0.07342  | 0.400352       | 5    |
| $\mathbb{R}_2$    | 0.38739 | 0.070318 | 0.287022       | 7    |
| R <sub>3</sub>    | 0.50237 | 0.139849 | 0.760587       | 3    |
| R4                | 0.47686 | 0.161754 | 0.834264       | 2    |
| <b>R</b> 5        | 0.33777 | 0.061229 | 0.182796       | 9    |
| R <sub>6</sub>    | 0.61233 | 0.145019 | 0.919877       | 1    |
| R7                | 0.20363 | 0.057324 | 0.000001       | 10   |
| R8                | 0.30981 | 0.07342  | 0.206966       | 8    |
| R9                | 0.4121  | 0.066624 | 0.299557       | 6    |
| $\mathbf{R}_{10}$ | 0.47943 | 0.103247 | 0.557282       | 4    |

**Table 4.** The value of  $g_m$ ,  $h_m$ ,  $z_m$  and rank of risks

## 5. Sensitivity Analysis

The weights of criteria affect the rank of risks. So, this paper introduces seven scenarios for changing the weights of criteria in Table 5. Then the weights of sub-criteria are changed in Table 6. Fig 7. shows the rank of risks under different scenarios.

The next step combines the rank with different scenarios into one rank. First, the highest rank takes 10 points and the next take 9 points, and so on [31]. Then calculate the total points. Table 7. Show the aggregation rank under different scenarios. Fig 8. shows the Final rank under different scenarios.



Fig 7. The rank of risks under seven scenarios



Fig 8. The Aggregation rank risks under different scenarios

| Scenarios  | Operational | Financial | Supply | Production Recovery | Government | Demand |
|------------|-------------|-----------|--------|---------------------|------------|--------|
| Scenario 1 | 1/6         | 1/6       | 1/6    | 1/6                 | 1/6        | 1/6    |
| Scenario 2 | 0.5         | 0.1       | 0.1    | 0.1                 | 0.1        | 0.1    |
| Scenario 3 | 0.1         | 0.5       | 0.1    | 0.1                 | 0.1        | 0.1    |
| Scenario 4 | 0.1         | 0.1       | 0.5    | 0.1                 | 0.1        | 0.1    |
| Scenario 5 | 0.1         | 0.1       | 0.1    | 0.5                 | 0.1        | 0.1    |
| Scenario 6 | 0.1         | 0.1       | 0.1    | 0.1                 | 0.5        | 0.1    |
| Scenario 7 | 0.1         | 0.1       | 0.1    | 0.1                 | 0.1        | 0.5    |

Table 5. Seven scenarios of weights changes

Table 6. Seven scenarios of sub criteria weights

| Criteria        | Scenario 1 | Scenario 2 | Scenario 3 | Scenario 4 | Scenario 5 | Scenario 6 | Scenario 7 |
|-----------------|------------|------------|------------|------------|------------|------------|------------|
| C11             | 0.219178   | 0.043836   | 0.043836   | 0.043836   | 0.043836   | 0.043836   | 0.219178   |
| C12             | 0.139901   | 0.02798    | 0.02798    | 0.02798    | 0.02798    | 0.02798    | 0.139901   |
| C13             | 0.083661   | 0.016732   | 0.016732   | 0.016732   | 0.016732   | 0.016732   | 0.083661   |
| C14             | 0.057259   | 0.011452   | 0.011452   | 0.011452   | 0.011452   | 0.011452   | 0.057259   |
| C <sub>21</sub> | 0.064683   | 0.323416   | 0.064683   | 0.064683   | 0.064683   | 0.064683   | 0.064683   |
| C22             | 0.024878   | 0.124391   | 0.024878   | 0.024878   | 0.024878   | 0.024878   | 0.024878   |
| C23             | 0.010439   | 0.052193   | 0.010439   | 0.010439   | 0.010439   | 0.010439   | 0.010439   |
| C31             | 0.057547   | 0.057547   | 0.287737   | 0.057547   | 0.057547   | 0.057547   | 0.057547   |
| C32             | 0.027846   | 0.027846   | 0.139232   | 0.027846   | 0.027846   | 0.027846   | 0.027846   |
| C33             | 0.010779   | 0.010779   | 0.053897   | 0.010779   | 0.010779   | 0.010779   | 0.010779   |

| C <sub>34</sub> | 0.003827 | 0.003827 | 0.019135 | 0.003827 | 0.003827 | 0.003827 | 0.003827 |
|-----------------|----------|----------|----------|----------|----------|----------|----------|
| C41             | 0.060261 | 0.060261 | 0.060261 | 0.301303 | 0.060261 | 0.060261 | 0.060261 |
| C42             | 0.024104 | 0.024104 | 0.024104 | 0.120521 | 0.024104 | 0.024104 | 0.024104 |
| C43             | 0.011759 | 0.011759 | 0.011759 | 0.058794 | 0.011759 | 0.011759 | 0.011759 |
| C44             | 0.003876 | 0.003876 | 0.003876 | 0.019382 | 0.003876 | 0.003876 | 0.003876 |
| C51             | 0.066101 | 0.066101 | 0.066101 | 0.066101 | 0.330503 | 0.066101 | 0.066101 |
| C52             | 0.033899 | 0.033899 | 0.033899 | 0.033899 | 0.169497 | 0.033899 | 0.033899 |
| C61             | 0.05928  | 0.05928  | 0.05928  | 0.05928  | 0.05928  | 0.296398 | 0.05928  |
| C62             | 0.028685 | 0.028685 | 0.028685 | 0.028685 | 0.028685 | 0.143423 | 0.028685 |
| C63             | 0.012036 | 0.012036 | 0.012036 | 0.012036 | 0.012036 | 0.060178 | 0.012036 |

| <b>Table 7.</b> The aggregation ran | k under different scen | arios |
|-------------------------------------|------------------------|-------|
|-------------------------------------|------------------------|-------|

| Risks                 | Total Points | Rank |
|-----------------------|--------------|------|
| $R_1$                 | 42           | 3    |
| $R_2$                 | 34           | 8    |
| <b>R</b> <sub>3</sub> | 35           | 7    |
| $\mathbb{R}_4$        | 25           | 9    |
| $R_5$                 | 41           | 4    |
| $R_6$                 | 16           | 10   |
| R7                    | 66           | 1    |
| $R_8$                 | 48           | 2    |
| R9                    | 38           | 6    |
| <b>R</b> 10           | 40           | 5    |

## 6. Conclusions

GSC plays a vital part in enhancement the ecological performance of companies. But the GSC has many risks. So, these risks need to rank for companies and organizations. This work proposed a hybrid neutrosophic MCDM for ranking the risks of GSC using SWARA and VIKOR methods under a neutrosophic environment. The SVNSs are proposed to overcome the uncertainty of information. The SWARA is used to calculate the weights of criteria and the VIKOR method is used to rank the risks of GSC. The proposed methodology is tested by a numerical example with twenty criteria and ten risks (alternatives).

The main contributions in this study proposed a neutrosophic environment to deal with indeterminacy value due to no previous study deal with the indeterminacy value. The SWARA and VIKOR method not used in previous research. In the future study, used other MCDM methods like PRPMETHEE II and ELECTRE.

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