

# Applied Data Science and Analysis Vol. **(2023)**, 2023, **pp**. 16-41

DOI: <a href="https://doi.org/10.58496/ADSA/2023/002">https://doi.org/10.58496/ADSA/2023/002</a>; ISSN: <a href="mailto:xxxxxx">xxxxxx</a> <a href="https://mesopotamian.press/journals/index.php/ADSA">https://mesopotamian.press/journals/index.php/ADSA</a>



# Research Article

# Fuzzy Decision-Making Framework for Sensitively Prioritizing Autism Patients with Moderate Emergency Level

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## **ARTICLE INFO**

#### Article History

Received 02 Feb 2023 Accepted 13 Mar 2023

#### Keywords

Autism

ASD

MCDM

FWZIC

Fuzzy Decision Making



## Abstract

Autism Spectrum Disorder (ASD) is a complex neurodevelopmental disorder that requires careful assessment and management. The prioritization of ASD patients involves navigating through complexities such as conflicts, trade-offs, and the importance of different criteria. Therefore, this study focuses on prioritizing patients with ASD in the healthcare setting through an evaluation and benchmarking framework. The aim of this study is to develop a framework that utilizes Multi-Criteria Decision Making (MCDM) methods to assist healthcare professionals in prioritizing ASD patients. particularly those with moderate injury levels. The methodology of the framework outlines several phases, including dataset identification, development of a decision matrix, weighting of 19 ASD criteria using the FWZIC method, ranking 432 patients using the VIKOR method, and evaluating the proposed framework using four sensitivity analysis scenarios. Among the 19 ASD criteria, the criterion 'verbal communication' obtained the highest weight. Additionally, criteria such as 'laughing for no reason', 'nodding', 'patient movement at home', and 'pointing with the index finger' obtained similar higher weights, indicating their potential impact on ASD patients. The experimental results highlight the significance of adjusting ASD weights in influencing the final rankings obtained through the VIKOR method. This emphasizes the need for careful consideration when assigning weights to the 19 ASD criteria to ensure accurate prioritization. Moreover, the framework provides valuable insights into improving the care and support provided to individuals with autism in Iraq. The findings contribute to the existing body of knowledge in the field of autism care prioritization and pave the way for future research and interventions aimed at enhancing the quality of care for individuals with autism in Iraq.

#### 1. Introduction

Autism Spectrum Disorder (ASD) is a complex neuropsychiatric disorder that is influenced by genetic factors. The World Health Organization (WHO) estimates that there are 67 million individuals worldwide affected by ASD [1][2]. Children with autism exhibit various symptoms, including repetitive behaviors, abnormalities in facial and vocal expressions, delayed speech learning, and difficulties in communication [3]. These symptoms typically emerge within the first two years of life and persist over time. The exact causes of ASD are still not fully understood, but they are believed to involve a combination of genetic mutations, such as "de novo or transmitted loss of function (LOF) mutations," as well as environmental factors that may contribute to genetic abnormalities [4][5]. In some cases, autism can be inherited from one or both parents, and it is more prevalent in males than females, with a fourfold higher occurrence [6]. Early diagnosis is crucial in managing and limiting the impact of autism [7]. However, traditional diagnostic methods, such as the Autism Diagnostic Observation Schedule (ADOS) and the Autism Diagnostic Interview-Revised (ADI-R), can be time-consuming. Many researchers have focused on the genetic aspects of ASD, attempting to identify specific gene sets and their priority in order to achieve accurate diagnoses. Given the complexities and various perspectives surrounding ASD, there is a need to explore different approaches, such as decision-making techniques, to develop effective prioritization methods for ASD patients.

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Triage is a crucial process used in medical settings to assign degrees of urgency to wounds or illnesses and determine the order of treatment for a large number of patients [8]. It involves identifying the most urgent cases and deciding who should receive treatment and in what order [9][10]. Triage can be applied to ASD patients to determine their level of urgency and prioritize their needs [11]. This helps in reducing waiting times and effectively utilizing available resources. ASD patients can be categorized into different triage levels, such as severe, moderate, and mild cases [6]. Triage-based questionnaires can be used to identify patients who require priority diagnostic evaluation [9]. Certain autism centers have implemented criteria-based prioritization, considering factors like age, urgency of needs, and additional complexities that may impact treatment decisions [9]. Prioritization approaches in the context of genetic studies have also been utilized to identify genes that contribute to ASD. These methods prioritize genes based on their importance in specific cell types, which may have etiological relevance to ASD [12]. In summary, triage and prioritization tools are valuable in ASD diagnosis, supporting evidence-based decision-making in medical tests and treatment strategies. They contribute to improving the accuracy of autism medical contexts and interventions [10]. ASD triage involves assessing and prioritizing individuals with ASD based on the severity of their condition and healthcare needs. It facilitates resource allocation, patient flow management, and timely provision of interventions [13]. Healthcare professionals use triage systems to determine the appropriate level of care and intervention for each ASD patient, taking into account medical urgency, symptom severity, and available resources [14]. These systems often incorporate standardized assessment tools and protocols to guide clinicians in prioritizing patients effectively [15].

The prioritization of ASD patients is a complex decision-making process that involves various challenges and considerations. One of the key issues encountered is the presence of conflicts and trade-offs among different criteria used for prioritization. Healthcare professionals and decision-makers must navigate through conflicting factors and make difficult choices when determining the priority of ASD patients [13]. Additionally, the importance of different ASD criteria adds another layer of complexity to the prioritization process. Each criterion holds a certain level of significance and contributes to the overall assessment of a patient's condition. Balancing the importance of various criteria and weighting them appropriately is crucial to ensure fair and accurate prioritization [14]. Addressing these issues requires the application of robust decision-making methodologies, such as Multi-Criteria Decision-Making (MCDM) techniques. MCDM methods allow for the systematic evaluation of multiple criteria and facilitate the resolution of conflicts and trade-offs. By incorporating the perspectives and expertise of healthcare professionals, a comprehensive and well-informed prioritization approach can be developed [15]. These methods provide a structured framework for assessing and ranking ASD patients based on their individual needs, considering factors such as symptom severity, available resources, and urgency of care. Because they can handle complicated decision issues with numerous objectives, MCDM approaches are operational research techniques that are extensively employed in many different domains [16]-[19]. MCDM is a decision theory extension that enables the assessment of options based on numerous criteria, sometimes with competing aims [20]. These methods require a number of processes, such as structure, planning, and resolving dilemmas utilizing various standards [21]. To evaluate how each alternative performs against each criterion and how important each criterion is in relation to the overall aim, decision-makers may use qualitative or quantitative methods [22]-[24]. There are several MCDM techniques available for evaluating and weighting criteria in different fields, including the Analytical Hierarchy Process (AHP) [25], Weighted Product Model (WPM) [26], Hierarchical Adaptive Weighting (HAW) [27], Best-Worst Method (BWM) [28], Multiplicative Exponential Weighting (MEW) [29], Weighted Sum Model (WSM) [30], Simple Additive Weighting (SAW) [31], and Analytical Network Process (ANP) [32]. However, when the number of criteria rises, several of these approaches, such as AHP, ANP, and BWM, may experience consistency problems. It gets more difficult to maintain consistency in the assessments as the number of pairwise comparisons increases. Complete consistency in the AHP technique has been argued to be almost unachievable when there are more than nine criteria By breaking down the criterion into sub-criteria, this problem may be solved, but it complicates the model even more. This inconsistency issue can be resolved using a relatively recent technique dubbed the Fuzzy-Weighted Zero-Inconsistency (FWZIC) approach [33]. Regardless of the quantity of criteria, the FWZIC technique consistently calculates the weight coefficients of the criterion. In order to determine each criterion's importance level throughout the decision-making process, it depends on variations in expert preferences. Different MCDM techniques have been created and used in many sectors to address the evaluation and benchmarking challenge [34][35]. The VlseKriterijumska Optimizacija I Kompromisno Resenje (VIKOR), which chooses the best choice from a group of options using quantitative or qualitative data, is one widely utilized technique. However, VIKOR relies on external weighing techniques since it lacks a means for allocating weights to the criterion. To solve this issue, the comprehensive FWZIC technique was developed [33]. It enables the consistent estimation of important weights for criterion. The primary objective of this study is to develop a framework for the evaluation and benchmarking of ASD patients with moderate injury levels. By integrating MCDM methods, such as FWZIC, with VIKOR, this framework aims to provide various solutions for prioritizing and managing ASD patients based on important criteria that can influence their condition. To this end, the study has the following objectives:

- 1. Develop a decision matrix for evaluation and benchmarking ASD patients.
- 2. Prioritize ASD patients using new fuzzy decision-making framework by integrated FWZIC-VIKOR methods.
- 3. Evaluate the developed framework results through sensitivity analysis.

By achieving these objectives, the study intends to provide healthcare professionals with a systematic and reliable approach to prioritize ASD patients. The framework incorporates decision-making techniques and addresses challenges such as conflicting criteria and the importance of different factors. The study contributes to improving the care and support provided to individuals with autism and paves the way for future research in autism care prioritization.

#### 2. Literature Review

Patient prioritization involves determining the order in which patients receive healthcare interventions based on their individual needs and available resources. The goal is to attend to those with the most urgent or critical conditions first. In the context of ASD, prioritization focuses on determining the order in which individuals with ASD receive interventions based on the severity of their condition and the urgency of their healthcare needs. This ensures that those requiring immediate or specialized attention are given priority. Various factors are considered in the prioritization process, including symptom severity, functional impairments, safety concerns, medical emergencies, and resource availability. The aim is to allocate resources and interventions effectively and efficiently, especially when resources are limited or waiting lists are long. Standardized assessment tools, clinical judgment, and established criteria specific to ASD may be utilized in the prioritization process. Prioritizing ASD patients is crucial to ensure timely access to appropriate interventions, maximize outcomes, and address the specific needs and challenges faced by individuals with ASD and their families. In a study [13], a triage method was developed for early autism diagnosis and clinical treatment, utilizing fuzzy MCDM techniques and considering multidimensional criteria. The authors selected four medical criteria and gave significant attention to sociodemographic criteria, resulting in a total of 19 criteria for correlation analysis. Another study [36] aimed to develop a hybrid model for predicting and diagnosing autism by intersecting nine machine learning methods and eight feature selection techniques. The model focused on effective sociodemographic and medical factors and aimed for high accuracy by applying rigorous methodological standards. In another study [37], a model combining MCDM and machine learning was developed in three phases. The first phase involved handling imbalanced ASD datasets through preprocessing stages such as imputing missing values and feature selection of sociodemographic and family characteristics, resulting in a balanced dataset of 107,573 cases. The second phase applied the proposed complex T-spherical fuzzy-weighted zeroinconsistency (CT-SFWZIC) method for model development.

The mentioned studies provide valuable insights but also face some criticisms and limitations regarding the prioritization of autism patients. These limitations include the lack of transparency and reproducibility in describing the specific methodology used, limited validation and external application of the prioritization framework, insufficient consideration of a moderate emergency level which is the most frequent condition among autistic patients according to clinical expertise, and inadequate emphasis on sensitivity analysis. Addressing these limitations is crucial for improving the reliability, applicability, and ethical considerations of the fuzzy decision-making framework. Therefore, the development of a dynamic framework for prioritizing ASD patients provides a systematic and comprehensive approach to evaluate and benchmark individuals on the autism spectrum. By leveraging various methodologies and decision-making techniques, the framework aims to enhance the accuracy, reliability, and understanding of ASD patient evaluations.

# 3. Fuzzy Decision-Making Framework

This section provides a comprehensive overview of the dynamic framework for prioritizing ASD patients. The framework consists of four distinct phases that guide the entire process, starting from dataset identification and culminating in the ranking of patients. By integrating various methodologies and decision-making techniques, the framework enhances the accuracy and reliability of the evaluation process, ultimately contributing to a better understanding and support for individuals on the autism spectrum. Phase 1 focuses on the identification of an appropriate ASD dataset. This involves carefully selecting a dataset that contains relevant information and characteristics of ASD patients. The criteria for dataset selection are meticulously considered to ensure its suitability for the subsequent evaluation and benchmarking process. Phase 2 centers around the development of a decision matrix, which serves as a powerful tool for organizing and analyzing the data collected from the ASD dataset. The decision matrix captures key parameters and encompasses 19 criteria related to ASD patients, facilitating a systematic evaluation and comparison of 432 moderate injury patients. Phase 3 introduces the FWZIC method, which plays a crucial role in assigning appropriate weights to the evaluation criteria. By incorporating fuzzy decision-making techniques, the FWZIC method ensures a fair and balanced assessment of the criteria's significance. This phase significantly contributes to the overall evaluation process. In the fourth phase, the VIKOR method is employed to rank ASD patients based on the weights obtained from the FWZIC method. VIKOR, as a multi-criteria decision-making

approach, facilitates the identification of the most suitable patients according to predefined criteria. This phase enables the prioritization of ASD patients who require immediate attention or specific interventions. Lastly, the fifth phase involves conducting a sensitivity analysis to evaluate the robustness and reliability of the proposed framework. This analysis ensures that the framework's performance remains consistent and reliable under varying conditions and input parameters.

#### 3.1. PHASE 1: ASD Dataset Identification

ASD raw data was obtained from two sources, namely [13] and [36]. To ensure the confidentiality and privacy of the patients, the data was anonymized. The dataset, consisting of 538 patients diagnosed with ASD, underwent pre-processing as described in the study conducted by [17]. This pre-processing step aimed to include 19 criteria related to the emergency triage level of the patients. For the triage process, the study utilized the Processes for Triaging Autism Patients (PTAP) method, which was developed by [17]. This method enabled the classification of patients into three severity levels: minor injury, moderate injury, and urgent injury. Figure 1 provides an overview of the distribution of patients across these severity levels. Notably, a significant proportion of patients, specifically 432 individuals, were assigned the moderate injury level. This level was specifically chosen to address the predefined problem and serve as a proof of concept.

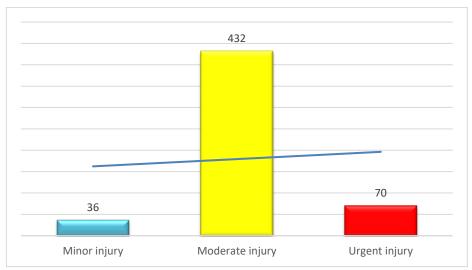


Figure 1 the result of Triage ASD patients on the real dataset [36]

To further illustrate the dataset and its features, Table 1 presents a sample of the moderate injury patients along with their corresponding dataset features.

Table 1 Sample of autism patients triaged with moderate injury level

0u	yes	yes	yes	no	no	yes	no	yes	no	yes	yes	male	no	no	yes	poog	yes	no	Moderate injury
yes	yes	yes	yes	no	yes	no	no	yes	no	110	yes	male	yes	yes	yes	poog	yes	yes	Moderate injury
n0	yes	no	yes	no	no	yes	no	yes	yes	yes	ou	male	yes	no	yes	pood	no	yes	Moderate injury
no	yes	yes	yes	no	no	yes	no	no	yes	yes	yes	male	yes	no	no	poog	yes	no	Moderate injury
yes	yes	no	yes	no	yes	yes	no	yes	yes	yes	ou	male	yes	no	yes	poog	yes	yes	Moderate injury

# 3.2. PHASE 2: Development of Decision Matrix

This section presents the developed dynamic decision-making (DM) approach used for evaluating and benchmarking ASD patients. DM is a crucial component of the assessment and benchmarking technique, as indicated in previous studies [38]–[43]. The DM framework consists of two main elements: the evaluation criteria and the alternatives. The evaluation criteria represent the 19 ASD criteria utilized to benchmark the 432 patients, who serve as the alternatives. The construction of the DM involved a series of steps, which are outlined in detail in Table 2. These steps encompass the necessary procedures for creating a robust and comprehensive DM model for evaluating and benchmarking ASD patients.

Table 2 DM

Altern	atives/Criteria	ASD Criteria										
AS	D Patients	C1	C1 C2 C3 C4 C5									
A1	Patient#1	C1-A1	C2-A1	C3-A1	C4-A1	C5-A1	•••	C19-A1				
A2	Patient#2	C1-A2	C2-A2	C3-A2	C4-A2	C5-A2	•••	C19-A2				
A3	Patient#3	C1-A3	C2-A3	C3-A3	C4-A3	C5-A3	•••	C19-A3				
A4	Patient#4	C1-A4	C2-A4	C3-A4	C4-A4	C5-A4	•••	C19-A4				
	•	•				•		•				
•	•	•	•	•		•		•				
•	•	•	•	•		•		•				
A432	A432 Patient#432 C1-A432 C2-A432 C3-A432 C4-A432 C5-A432 C19-A432											
C= Cri	C= Criteria, A= Alternative, C= Criteria											

# 3.3. PHASE 3: FWZIC method for Weighting Criteria

The FWZIC method is employed as an MCDM technique for weighting the 432 ASD criteria. Figure 2 depicts the step-by-step process of FWZIC, highlighting the five essential processes that should be applied to the 19 ASD criteria. These five steps are outlined below:



Figure 2 FWZIC methodology for weighting the 19 criteria [36]

**Step 1**: Establish the set of evaluation criteria: The planned set of assessment features for ASD is investigated and provided in the first phase of the evaluation and benchmarking procedure.

Step 2: Using structured expert judgment (SEJ), members of the expert team are found and chosen from appropriate medical specialties, mostly psychiatrists. The selection and nomination procedure then starts, leading to the creation of the SEJ panel. The language scale is converted to the matching numerical scale in order to speed up the panelists' judgment and ability to reach consensus. The process of this conversion is illustrated in Table 3. The SEJ team also creates an assessment form to record their collective agreement on each ASD criteria.

Table 3 Five-point	Likert scal	e and equival	lent numerical	scale
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Linguistic terms	Numerical scoring scale
Not important	1
Slight important	2
Moderately important	3
Important	4
Very important	5

Step 3: Creating the Expert Decision Matrix (EDM): In the step before, the list of chosen experts and their selections based on predetermined criteria were defined. Building the EDM is the main objective of this phase. The alternatives (ASD criteria) and decision criteria are the main elements of the EDM, as illustrated in Table 4. Each selective expert (Ei), who has assessed the level of significance for each criterion, interacts with each criterion (Cj) in the attribute (representing the ASD criteria).

		Table 4 EDM		
Criteria / Experts	C1	<i>C</i> 2	•••	Cn
E1	Imp (E1/C1)	Imp (E1/C2)		Imp (E1/Cn)
<i>E2</i>	Imp (E2/C1	Imp (E2/C2)		Imp (E2/Cn)
<i>E3</i>	Imp (E3/C1)	Imp (E3/C2)		Imp (E3/Cn)

•••			 
Em	Imp (En/C1)	Imp (En/C2)	 Imp (Em/Cn)

<sup>\*\*</sup>Imp represents the importance level.

**Step 4:** Application of a fuzzy membership function: A fuzzy membership function and defuzzification technique are used to improve the precision and usability of the data in the EDM for further analysis. It is hard to give accurate preference ratings to each criterion in the context of Multiple Criteria Decision Making (MCDM) since the situation is frequently unclear and imprecise. The fuzzy approach allows for the management of ambiguous and imprecise circumstances by using fuzzy numbers rather than exact numbers to compute the relative value of each characteristic (criteria) [44]–[46]. Triangular fuzzy numbers (TFNs), one of several forms of fuzzy numbers, are frequently employed in fuzzy MCDM. A = (a, b, c) is used to express them, with a, b, and c denoting the lower, middle, and upper values, respectively. TFNs are appropriate for use in practical applications due to their conceptual and computational simplicity [47]. The triangle membership function used to describe TFNs in this work is shown in Figure 3.

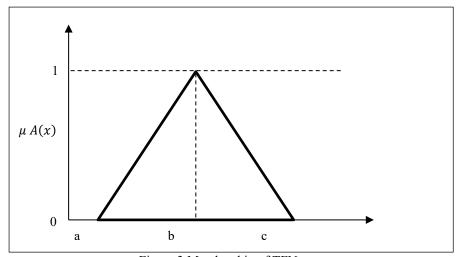


Figure 3 Membership of TFNs

The membership function (x) of TFN A is given by Equation 1

$$\mu A(x) = \begin{cases} 0 & \text{if } x < a \\ \frac{x-a}{b-a} & \text{if } a \le x \le b \\ \frac{c-x}{c-b} & \text{if } b \le x \le c \\ 0 & \text{if } x > c \end{cases}$$
 where  $a \le b \le c....(1)$ 

Remark: Let  $\tilde{x} = (a1, b1, c1)$  and  $\tilde{y} = (a2, b2, c2)$  be two non-negative TFNs and  $\alpha \in \mathbb{R}_+$ . Following the extension principle, the arithmetic operations are defined as follows:

$$\begin{array}{lll} 1. \ \tilde{x} + \tilde{y} = (a1 + a2, b1 + b2, c1 + c2), & \dots (2) \\ 2. \ \tilde{x} - \tilde{y} = (a1 - c2, b1 - b2, c1 - a2), & \dots (3) \\ 3. \ \alpha \tilde{x} = (\alpha a1, \alpha b1, \alpha c1), & \dots (4) \\ 4. \ \tilde{x} - 1 \cong (1/c1, 1/b1, 1/a1), & \dots (5) \\ 5. \ \tilde{x} \times \tilde{y} \cong (a1a2, b1b2, c1c2), & \dots (6) \\ 6. \ \tilde{x}/\tilde{y} \cong (a1/c2, b1/b2, c1/a2). & \dots (7) \end{array}$$

The value of each Numerical term with TFN is shown in Table 5.

Table 5 Numerical terms and their equivalent TFNs

Numerical scoring scale	TFNs
1	(0.00,0.10,0.30)
2	(0.10, 0.30, 0.50)

3	(0.30,0.50,0.75)
4	(0.50, 0.75, 0.90)
5	(0.75, 0.90, 1.00)

For each expert and set of criteria, the conversion of linguistic variables into TFNs is shown in Table 5. Using language factors, psychiatric expert N evaluates the relative weight of each evaluation criterion. Given the fuzziness and ambiguity of the situation, the linguistic variables are transformed into TFNs to express the judgments. The assessments may now be represented in a more precise and quantitative manner, allowing for more in-depth analysis and decision-making during the evaluation and benchmarking process.

Step 5: Computation of the final values of the weight coefficients of the evaluation criteria: In this stage, the final values of the weight coefficients for the evaluation criteria  $(w1, w2, ..., w19)^T$  are determined using the fuzzy data for the criterion from the previous step.

1) By using Equation 8, the ratio of fuzzification data is determined. As demonstrated in Table 6, the preceding equations are employed with TFNs [47].

$$\frac{Imp(\widetilde{E1}/C1)}{\sum_{j=1}^{n}Imp(\widetilde{E1}/C_{1j})} \qquad \dots (8)$$

where Imp(E1/C1) represent the fuzzy number of Imp (E1/C1).

Table 6 Fuzzy EDM (EDM) [47]

Criteria\Experts	$\widetilde{c}$ 1	$\widetilde{c2}$		$\widetilde{Cn}$
E1	$Imp(\widetilde{E1}/C1)$	$Imp(\widetilde{E1}/C2)$	•••	$Imp(\widetilde{E1}/Cn)$
	$\overline{\sum_{j=1}^{n} Imp(\widetilde{E1}/C_{1j})}$	$\overline{\sum_{j=1}^{n} Imp(\widetilde{E1}/C_{1j})}$		$\overline{\sum_{j=1}^{n} Imp(\widetilde{E1}/C_{1j})}$
E2	$Imp(\widetilde{E2}/C1)$	$Imp(\widetilde{E2}/C2)$	•••	$Imp(\widetilde{E2}/Cn)$
	$\overline{\sum_{j=1}^{n} Imp(\widetilde{E2}/C_{2j})}$	$\overline{\sum_{j=1}^{n} Imp(E2/C_{2j})}$		$\overline{\sum_{j=1}^{n} Imp(E2/C_{2j})}$
<i>E3</i>	$Imp(\widetilde{E3}/C1)$	$Imp(\widetilde{E3}/C2)$	•••	$Imp(\widetilde{E3}/Cn)$
	$\overline{\sum_{j=1}^{n} Imp(\widetilde{E3}/C_{3j})}$	$\overline{\sum_{j=1}^{n} Imp(\widetilde{E3}/C_{3j})}$		$\overline{\sum_{j=1}^{n} Imp(\widetilde{E3}/C_{3j})}$
<b>E4</b>	$Imp(\widetilde{E4}/C1)$	$Imp(\widetilde{Em}/C2)$	•••	$Imp(\widetilde{E4}/Cn)$
	$\overline{\sum_{j=1}^{n} Imp(\widetilde{E4}/C4_{j})}$	$\sum_{j=1}^{n} Imp(\widetilde{Em}/C_{4j})$		$\overline{\sum_{j=1}^{n} Imp(\widetilde{E4}/C_{nj})}$

2) To determine the final fuzzy values of the weight coefficients of the evaluation criteria  $(\widetilde{w1}, \widetilde{w2}, ..., \widetilde{wn})^T$ , The mean values are determined. The Fuzzy EDM ( $\widetilde{\text{EDM}}$ ) is utilized to calculate the final weight value of each criterion using Equation 9.

$$\widetilde{w_{j}} = \left(\sum_{i=1}^{m} \frac{Imp(\widetilde{E_{ij}}/C_{ij})}{\sum_{j=1}^{n} Imp(\widetilde{E_{ij}}/C_{ij})}\right) / m \text{ ), for } i = 1,2,3,..m \text{ and } j = 1,2,3,..n$$
... (9)

3) The most common defuzzification method is the centroid approach, which is used to determine the final weight. The mathematical formula for this process using TFNs is ((a+b+c))/3. Prior to determining the final values of the weight coefficients, each criterion should be given a weight of importance based on the sum of all weights for the rescaling purpose employed in this phase.

At this point and after calculating the weights for 19 criteria.

## 3.4. PHASE 4: VIKOR for Ranking Patients

We employ the VIKOR method to evaluate and benchmark the 432 alternatives for the ASD patients as follows:

STEP 1: Mark the worst  $f^-i$  and best  $f^*i$  values for the overall ASD criteria, as i=1; 2; ...; n. If the *i*th criteria function act as a benefit, then

$$\int_{i}^{*} = max_{j} \int ij, \quad \int_{i}^{-} = min_{j} \int ij, \dots (10)$$

**STEP 2:** The weights for each criterion (FWIZC weights) are introduced to VIKOR throughout this phase. The decision-maker's set of weights,  $w = w_1, w_2, w_3, \cdots, w_j, \cdots, w_n$ , is accommodated in the DM and is equal to 1. Calculating the resultant matrix is also possible, as shown in Equation 11 below:

$$WM = wi * (f^*i - fij)/(f^*i - f^-i)...(11)$$

Upon applying the above process, it will produce a new weighted matrix as follows:

$$\begin{bmatrix} w_{1}(f^{*}1-f11)/(f^{*}1-f^{-}1) & w_{2}(f^{*}2-f12)/(f^{*}2-f^{-}2) & \dots & w_{i}(f^{*}i-fij)/(f^{*}i-f^{-}i) \\ w_{1}(f^{*}1-f21)/(f^{*}1-f^{-}1) & w_{2}(f^{*}2-f22)/(f^{*}2-f^{-}2) & \dots & w_{i}(f^{*}i-fij)/(f^{*}i-f^{-}i) \\ \vdots & \vdots & \vdots & \vdots \\ w_{1}(f^{*}1-f31)/(f^{*}1-f^{-}1) & w_{2}(f^{*}2-f32)/(f^{*}2-f^{-}2) & \dots & w_{i}(f^{*}i-fij)/(f^{*}i-f^{-}i) \end{bmatrix}$$
(12)

**STEP 3:** Compute the values S<sub>j</sub> and R<sub>j</sub>, j=1,2,3,...,J, i=1,2,3,...,n by using the following equations:

$$Sj = \sum_{i=1}^{n} wi * (f^*i - fij) / (f^*i - f^{-}i)$$

$$Rj = \max_{i} wi * (f^*i - fij) / (f^*i - f^{-}i)$$
(13)

Where wi are the weights of criteria expressing their relative importance

**STEP 4:** Calculate the values  $Q_{j,j} = (1,2,\dots,J)$  using Equation 15:

$$Q_{\rm j} = \frac{{\rm v}(S_{\rm j} - S^*)}{S^- - S^*} + \frac{(1 - {\rm v})(R_{\rm j} - R^*)}{R^- - R^*}$$
 (15)

Where

$$\begin{split} S^* &= \min_j S_j \;, \quad S^- &= \max_j S_j \\ R^* &= \min_j R_j \;, \quad R^- &= \max_j R_j \end{split}$$

v is presented as the weight of the strategy of 'the majority of criteria' or 'the maximum group utility'; in this research, v = 0.5.

STEP 5: The alternative set, which consists of ASD patients, is sorted in ascending order based on the value Q. The lowest value obtained from each detection model indicates a higher emergency level for the patient. After individual rankings are completed, variations may be observed among different decision makers involved in the research. To combine the individual decisions and obtain an aggregated ranking, the scores of each alternative in Group Decision Making (GDM) are calculated using an arithmetic mean. The alternative with the highest mean value is considered the best alternative and represents the most critical emergency situation.

#### 3.5. PHASE 5: Evaluation Framework

In this phase, a sensitivity analysis was performed to investigate the effect of different weight allocations on the prioritization outcomes. The purpose of the sensitivity analysis was to assess the robustness of the decision framework and understand how changes in criteria weights could impact the final rankings of autism patients' emergency levels. The 19 criteria were divided into two groups: the first group included 10 criteria, and the second group included 9 criteria, as described in Equation 16. The objective was to examine how variations in weight allocations within these groups would influence the prioritization results.

$$wf = (wo * ws) ... (16)$$

Furthermore, to determine the weight of each criterion, the original weight assigned to the criterion (wo) was adjusted through the sensitivity analysis. The adjusted weight (ws) was calculated using Equation 17.

$$ws = (m/n) * 100 ... (17)$$

where m represents the chosen percentage for the weight allocation scenario, and n represents the total number of criteria in the group. The sensitivity analysis was conducted using four weight allocation scenarios: 60%:40%, 40%:60%, 70%:30%, and 30%:70%. By implementing these different weight allocations, we examined how the changes affected the

final rankings of autism patients' emergency levels. The new scores for each alternative were calculated using the VIKOR method, and the results were compared across the scenarios.

#### 4. Results and Discussion

This section provides a comprehensive overview of the results achieved through the proposed framework for prioritizing ASD patients with moderate injury levels. The section presents the results of the decision matrix, highlighting the priority weights assigned to the evaluation criteria using the FWZIC method. It also showcases the prioritization results obtained through the VIKOR method. Additionally, this chapter discusses the evaluation framework for the proposed methodology, including a sensitivity analysis using four different weight allocation scenarios. These findings contribute to a better understanding of the patients' status and help identify cases that require immediate attention and intervention.

#### 4.1. Decision Matrix Results

The decision matrix, which represents the ASD dataset for the 432 patients, is presented in Table 7.

4th patient 3rd patient 2nd patient 1st patient Alternatives 

Table 7 Samples of four patients within the developed decision matrix

In this step, the evaluation of the decision matrix is carried out using the ASD data of the 19 criteria. The weights for each criterion are required, as discussed in the next section.

# 4.2. Weights Result

The FWZIC method, as described in a previous study [13], was used to determine the priority weights for each criterion in the decision matrix. In [13], four experts with expertise in ASD were selected to provide their subjective judgments and weight the 19 criteria based on their experience. The experts used the Five-point Likert scale and the corresponding numerical scale presented in Table 3 in Section 3 to provide their judgments for the criteria. These judgments are reflected in the EDM presented in Table 8. It is important to note that the presentation of these weights, which were already constructed in [13], is done to establish a connection between the results of different phases and to provide clarity for the readers.

							Tab	ole 8	EDM	1 resu	lts of	[13]							
Criteria \ Experts	C1	C2	ຮ	C4	C5	92	C7	C8	C9	C10	C11	C12	C13	C14	C15	C16	C17	C18	C19
1st Expert	4	4	4	4	3	2	3	4	5	4	1	5	1	4	4	5	2	4	2
2 <sup>nd</sup> Expert	2	3	2	3	2	1	3	4	5	5	4	5	2	5	4	5	4	4	4
3 <sup>rd</sup> Expert	4	4	3	4	3	3	4	3	3	3	3	5	3	2	2	2	3	4	3
4 <sup>th</sup> Expert	3	2	3	4	2	1	3	5	4	3	4	4	2	4	4	5	4	5	3

Depending on the EDM Table, the weights result of the 19 criteria based on the FWZIC method for the four experts are presented in Table 9.

Table 9 Result of weighting 19 criteria of ASD patients [13]

ASD Criteria	FWZIC Weights
C1=Verbal communication	0.07298
C2=Laughing for no reason	0.06666
C3=Nodding	0.06528
C4=Patient movement at home	0.06330
C5=Pointing with the index finger	0.06232
C6=Complications of childbirth	0.05976
C7=Spinning round things	0.05822
C8=Bathroom skills	0.05711
C9=Wave	0.05412
C10=Unnecessary drug	0.05185
C11=Maternal diseases during pregnancy	0.05117
C12=Afraid of loud sounds	0.05073
C13=Patient's Gender	0.04969
C14=Notice the sound of the bell	0.04713
C15=Premature baby	0.04707
C16=Crying for no reason	0.04675
C17=Marital Relationship	0.03853
C18=Taste the food	0.03037
C19=Consanguinity	0.02687

As shown in Table 9, the criterion 'verbal communication' obtained the highest weight of 0.07298183, indicating its significant impact on ASD patients. Several other criteria, including 'laughing for no reason', 'nodding', 'patient movement at home', and 'pointing with the index finger', also obtained relatively high weights, suggesting their importance in assessing ASD. On the other hand, the criterion 'consanguinity' received the lowest weight of 0.026873862, indicating that it may have lower priority in relation to ASD patients. The results of the criteria weighting were discussed with the four experts, who provided their analysis and positive opinions on using these weights in their diagnosis through a questionnaire. The ranking results of the VIKOR method for ASD patients will be presented in the following section, considering the weightings obtained through the FWZIC method.

#### 4.3. VIKOR Prioritization Result

In this stage, the VIKOR method is applied to benchmark the 432 patients and identify high emergency cases based on the weights assigned to the evaluation criteria. The overall weights obtained from Table 9 are incorporated into the VIKOR method, utilizing the configurations derived from the FWZIC method. The patients are ranked according to their value Q, with the ranking presented in ascending order in Table 10. Additionally, a visualization of the first 100 ranks is shown in Figure 4. These rankings provide valuable insights into the priority levels of the ASD patients, helping to identify those requiring immediate attention and intervention.

Table 10 VIKOR results of the benchmarking of moderate emergency level for ASD patients

Patients	Q Value	Ranking Order	Patients	Q Value	Ranking Order	Patients	Q Value	Ranking Order	Patients	Q Value	Ranking Order
P1	0.6189	230	P109	0.6268	224	P217	0.6743	183	P325	0.9147	27
P2	0.4705	311	P110	0.5594	267	P218	0.6534	200	P326	0.6843	173
Р3	0.4662	312	P111	0.7251	152	P219	0.1196	429	P327	0.6541	197
P4	0.4804	308	P112	0.6056	237	P220	0.9350	17	P328	0.8918	39
P5	0.5016	296	P113	0.9345	18	P221	0.2742	403	P329	0.7548	138
P6	0.4260	332	P114	0.7662	125	P222	0.4276	328	P330	0.4103	347
<b>P7</b>	0.4026	352	P115	0.8345	74	P223	0.6329	219	P331	0.6609	190
P8	0.3017	389	P116	0.3419	376	P224	0.8713	51	P332	0.5106	292
P9	0.4142	343	P117	0.5439	274	P225	0.5023	295	P333	0.2993	394
P10	0.0523	431	P118	0.2566	411	P226	0.1849	421	P334	0.5700	253
P11	0.4068	349	P119	0.4003	355	P227	0.4937	302	P335	0.7763	113
P12	0.3554	373	P120	0.8623	61	P228	0.7739	115	P336	0.4087	348
P13	0.5298	281	P121	0.5051	294	P229	0.2131	418	P337	0.6887	169
P14	0.7837	110	P122	0.7491	141	P230	0.4943	301	P338	1.0000	1
P15	0.7231	153	P123	0.8066	86	P231	0.5990	241	P339	0.6655	188
P16	0.6881	170	P124	0.7512	139	P232	0.9067	31	P340	0.3266	382
P17	0.4170	341	P125	0.4436	320	P233	0.9119	28	P341	0.8219	81
P18	0.4544	316	P126	0.5147	288	P234	0.8658	59	P342	0.7886	98
P19	0.4147	342	P127	0.8711	54	P235	0.1848	422	P343	0.7347	151
P20	0.9296	22	P128	0.8050	87	P236	0.6360	215	P344	0.8867	41
P21	0.7670	123	P129	0.4718	310	P237	0.7844	107	P345	0.2455	412
P22	0.6235	227	P130	0.7004	166	P238	0.8976	36	P346	0.8668	58
P23	0.7740	114	P131	0.2428	414	P239	0.7572	128	P347	0.6155	232
P24	0.9039	33	P132	0.7550	137	P240	0.9387	15	P348	0.9678	10
P25	0.6445	206	P133	0.4841	307	P241	0.6328	220	P349	0.5125	290
P26	0.4270	331	P134	0.3633	368	P242	0.2911	399	P350	0.3621	370

P27	0.9271	23	P135	0.5393	276	P243	0.7498	140	P351	0.6221	228
P28	0.4854	306	P136	0.7552	136	P244	0.7498	43	P352	0.9029	34
P29	0.6387	214	P137	0.6570	193	P245	0.7982	91	P353	0.1811	424
P30	0.3700	366	P138	0.8520	71	P246	0.5894	245	P354	0.6113	235
P31	0.8074	85	P139	0.7035	165	P247	0.8183	83	P355	0.3305	380
P32	0.5473	273	P140	0.7723	117	P248	0.8828	45	P356	0.9221	26
P33	0.6319	221	P141	0.6451	205	P249	0.7854	105	P357	0.8267	78
P34	0.3992	356	P142	0.7678	121	P250	0.7733	116	P358	0.6506	203
P35	0.5491	271	P143	0.7075	163	P251	0.8521	70	P359	0.6600	191
P36	0.4219	336	P144	0.7382	150	P252	0.5163	286	P360	0.7859	104
P37	0.7173	154	P145	0.6871	171	P253	0.7912	96	P361	0.1968	420
P38	0.8694	56	P146	0.3880	359	P254	0.6855	172	P362	0.4895	303
P39	0.4105	346	P147	0.6748	182	P255	0.8945	38	P363	0.7714	118
P40	0.7673	122	P148	0.7850	106	P256	0.8577	65	P364	0.6520	201
P41	0.6197	229	P149	0.6588	192	P257	0.6285	223	P365	0.5493	270
P42	0.2010	419	P150	0.4644	313	P258	0.6676	185	P366	0.6643	189
P43	0.3217	383	P151	0.6422	209	P259	0.2862	400	P367	0.4981	297
P44	0.6347	218	P152	0.2697	404	P260	0.5611	265	P368	0.3812	361
P45	0.4129	344	P153	0.5286	282	P261	0.7663	124	P369	0.6749	181
P46	0.8888	40	P154	0.4042	350	P262	0.7767	112	P370	0.3783	362
P47	0.8729	46	P155	0.6809	178	P263	0.5399	275	P371	0.8700	55
P48	0.7865	102	P156	0.5637	261	P264	0.8537	66	P372	0.8621	62
P49	0.5662	259	P157	0.7458	145	P265	0.2298	416	P373	0.2317	415
P50	0.6685	184	P158	0.5263	283	P266	0.4979	298	P374	0.7108	158
P51	0.1705	425	P159	0.3938	358	P267	0.5872	246	P375	0.2433	413
P52	0.4196	338	P160	0.4226	334	P268	0.7430	146	P376	0.1672	426
P53	0.4459	319	P161	0.7944	95	P269	0.4433	321	P377	0.9565	11
P54	0.7396	147	P162	0.9787	8	P270	0.9894	4	P378	0.6670	186
P55	0.5649	260	P163	0.9796	6	P271	0.5685	258	P379	0.1644	427
P56	0.2635	407	P164	0.4304	327	P272	0.7382	149	P380	0.3311	379
P57	0.8728	47	P165	0.8712	53	P273	0.6548	196	P381	0.1847	423
P58	0.5689	255	P166	0.7712	120	P274	0.6511	202	P382	0.5215	284
P59	0.7885	99	P167	0.5686	257	P275	0.8672	57	P383	0.2156	417
P60	0.6567	194	P168	0.6761	180	P276	0.8721	50	P384	0.6119	234
P61	0.8089	84	P169	0.6537	199	P277	0.5145	289	P385	0.7958	93
P62	0.6470	204	P170	0.8725	49	P278	0.9338	19	P386	0.2989	396
P63	0.4887	304	P171	0.3159	385	P279	0.5167	285	P387	0.3005	392
P64	0.3323	378	P172	0.8521	70	P280	0.6812	177	P388	0.6388	213
P65	0.8342	75 25	P173	0.4019	354	P281	0.3872	360	P389	0.3560	372
P66	0.9024	35	P174	0.3759	364	P282	0.0499	432	P390	0.7562	132
P67	0.4397	323	P175	0.7568	130	P283	0.3016	390	P391	0.7077	162
P68	0.5694	254	P176	0.5969	243	P284	0.1079	430	P392	0.2599	410
P69	0.8531	67 222	P177	0.7092	160	P285	0.6093	236	P393	0.2639	406
P70	0.6291	222	P178	0.4968	300	P286	0.1570	428	P394	0.7839	109
P71	0.8191	82	P179	0.9236	25	P287	0.5484	272	P395	0.7144	156

P72	0.7460	144	P180	0.6051	239	P288	0.4206	337	P396	0.9311	21
P73	0.3667	367	P181	0.4534	318	P289	0.8005	90	P397	0.5616	264
P74	0.6661	187	P182	0.6824	176	P290	0.4397	322	P398	0.6347	217
P75	0.2965	397	P183	0.9081	30	P291	0.4173	340	P399	0.5734	251
P76	0.4035	351	P184	0.7462	143	P292	0.4380	324	P400	0.3938	358
P77	0.4361	325	P185	0.6267	226	P293	0.4189	339	P401	0.4226	334
P78	0.5149	287	P186	0.5834	248	P294	0.3119	386	P402	0.7944	95
P79	0.7062	164	P187	0.4271	330	P295	0.7560	133	P403	0.9787	8
P80	0.5377	277	P188	0.6425	208	P296	0.5328	280	P404	0.9796	6
P81	0.5931	244	P189	0.7556	135	P297	0.2963	398	P405	0.4304	327
P82	0.9368	16	P190	0.6889	168	P298	0.6119	234	P406	0.8712	53
P83	0.7383	148	P191	0.9900	3	P299	0.7958	93	P407	0.7712	120
P84	0.4730	309	P192	0.6565	195	P300	0.2989	396	P408	0.5686	257
P85	0.7136	157	P193	0.7615	127	P301	0.3005	392	P409	0.6761	180
P86	0.3621	370	P194	0.5726	252	P302	0.6388	213	P410	0.6537	199
P87	0.5350	278	P195	0.8295	76	P303	0.3560	372	P411	0.8725	49
P88	0.9457	13	P196	0.5091	293	P304	0.7562	132	P412	0.3159	385
P89	0.3543	374	P197	0.3003	393	P305	0.7077	162	P413	0.8521	70
P90	0.6392	211	P198	0.5575	268	P306	0.2599	410	P414	0.4019	354
P91	0.8459	73	P199	0.4225	335	P307	0.2639	406	P415	0.3759	364
P92	0.2793	401	P200	0.7889	97	P308	0.7839	109	P416	0.7568	130
P93	0.8947	37	P201	0.5608	266	P309	0.7144	156	P417	0.5969	243
P94	0.4574	315	P202	0.6830	174	P310	0.9311	21	P418	0.7092	160
P95	0.8025	88	P203	0.9456	14	P311	0.5616	264	P419	0.4968	300
P96	0.5106	292	P204	0.4122	345	P312	0.6347	217	P420	0.9236	25
P97	0.8834	44	P205	0.3437	375	P313	0.5734	251	P421	0.6051	239
P98	0.8275	77	P206	0.4606	314	P314	0.7881	100	P422	0.4534	318
P99	0.4855	305	P207	0.9040	32	P315	0.5334	279	P423	0.6824	176
P100	0.8243	79	P208	0.3048	387	P316	0.3299	381	P424	0.9081	30
P101	0.5633	262	P209	0.3392	377	P317	0.8856	42	P425	0.7462	143
P102	0.8627	60	P210	0.6041	240	P318	0.9708	9	P426	0.6267	226
P103	0.2601	408	P211	0.7864	103	P319	0.6399	210	P427	0.5834	248
P104	0.5742	249	P212	0.8620	63	P320	0.8468	72	P428	0.4271	330
P105	0.3748	365	P213	0.9538	12	P321	0.8599	64	P429	0.6425	208
P106	0.3017	388	P214	0.5534	269	P322	0.2749	402	P430	0.7556	135
P107	0.8008	89	P215	0.8237	80	P323	0.7648	126	P431	0.6889	168
P108	0.7880	101	P216	0.6184	231	P324	0.7832	111	P432	0.9900	3
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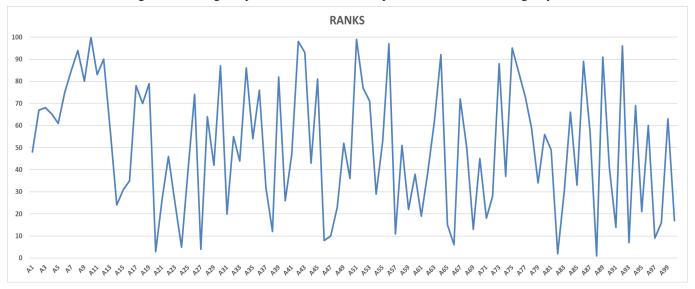


Figure 4 Ranking samples of the first 100 ASD patients of moderate emergency level

In Table 11, the rankings of the first three emergency cases among the ASD patients are presented. These patients have been identified as requiring immediate attention based on their evaluation scores and prioritization using the VIKOR method. Table 12, on the other hand, displays the rankings of the last three patients, indicating those with the lowest priority levels among the evaluated cases. These tables provide a snapshot of the extreme cases in terms of emergency levels, aiding in the identification of patients who may need urgent interventions or can be managed with lower priority.

Table 11 First three ranks of ASD patients of moderate emergency level according to VIKOR method

Patients / Criteria	Patient Gander	Marital Relationship	Consanguinity	Unnecessary drugs	Maternal diseases during	Complications of childbirth	premature baby	Taste the food	Wave	Patient movement at home	Frightened by loud noises	Laughing for no reason	Crying for no reason	No verbal communication	Pointing with the index finger	Notice the sound of the bell	Bathroom skills	Nodding	Spinning round things
1 <sup>st</sup>	no	yes	yes	ou	yes	yes	no	no	yes	yes	no	yes	yes	yes	yes	yes	no	yes	no
2 <sup>nd</sup>	yes	no	yes	ou	no	no	no	yes	yes	yes	yes	yes	yes	yes	no	yes	yes	yes	yes
3 <sup>rd</sup>	yes	no	yes	ou	no	no	no	yes	yes	yes	yes	yes	yes	yes	no	ou	yes	yes	yes

Table 12 Last three ranks of ASD patients of moderate emergency level according to VIKOR method

Patients / Criteria	Patient Gander	Marital Relationship	Consanguinity	Unnecessary drugs	Maternal diseases during pregnancy	Complications of childbirth	premature baby	Taste the food	Wave	Patient movement at home	Frightened by loud noises	Laughing for no reason	Crying for no reason	No verbal communication	Pointing with the index finger	Notice the sound of the bell	Bathroom skills	Nodding	Spinning round things
100 <sup>th</sup>	yes	no	no	ou	00	yes	yes	no	yes	no	yes	no	no	no	yes	yes	no	no	no
99 <sup>th</sup>	yes	yes	yes	ou	no	no	no	yes	yes	yes	no	yes	no	no	no	yes	no	no	no
98 <sup>th</sup>	yes	no	ou	no	yes	ou	ou	no	no	yes	yes	ou	ou	no	yes	yes	yes	no	no

In Table 11, it can be observed that the top three ranked patients share common high-weighted criteria such as 'no verbal communication', 'laughing for no reason', 'nodding', and 'patient movement at home'. However, there may be slight differences in some criteria with lower weights, such as 'frightened by loud noises', 'taste the food', and 'notice the sound of the bell'. On t-he other hand, Table 12 represents the last three ranked patients, and when compared to the top three ranks, there are differences in the criteria with high weights. The top three ranks have a value of 'yes' for criteria like 'no verbal communication', 'laughing for no reason', and 'nodding', while the last three ranks have a value of 'no'. This highlights the importance of weight allocation and how it influences the ranking of patients. The benchmarking results obtained from the VIKOR method demonstrate the effectiveness of the prioritization approach in ranking and evaluating the 432 ASD patients based on the evaluation criteria and their assigned weights. The application of VIKOR allows for the identification of high emergency cases among the patients, enabling timely intervention and care.

# 4.4. Evaluation Framework

This section provides the results of the sensitivity analysis conducted in Phase 5. Four different weight allocation scenarios were implemented based on Equation 16 and 17, and the outcomes are presented in Table 12.

Table 12 Sensitively weights result for the four scenarios

Weights/Scenario	Patient's Gander	Marital Relationship	Consanguinity	Unnecessary drugs	Maternal diseases during pregnancy	Complications of childbirth	premature baby	Taste the food	Wave	Patient movement at home	Frightened by loud noises	Laughing for no reason	Crying for no reason	No verbal communication	Pointing with the index finger	Notice the sound of the bell	Bathroom skills	Nodding	Spinning round things
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Scenario 4	Scenario 3	Scenario 2	Scenario 1
0.0270	0.0686	0.0368	0.0576
0.0209	0.0532	0.0285	0.0446
0.0146	0.0371	0.0199	0.0311
0.0282	0.0716	0.0384	0.0601
0.0278	0.0707	0.0379	0.0593
0.0325	0.0826	0.0442	0.0692
0.0256	0.0650	0.0348	0.0545
0.0165	0.0420	0.0225	0.0352
0.0294	0.0748	0.0401	0.0627
0.0344	0.0875	0.0468	0.0733
0.0715	0.0334	0.0626	0.0435
0.0939	0.0439	0.0822	0.0572
0.0659	0.0308	0.0577	0.0401
0.1029	0.0480	0.0900	0.0626
0.0878	0.0410	0.0769	0.0535
0.0664	0.0310	0.0581	0.0404
0.0805	0.0376	0.0704	0.0490
0.0920	0.0429	0.0805	0.0560
0.0821	0.0383	0.0718	0.0500

The first scenario involved assigning weights of 60% to the first group and 40% to the second group. This allocation aimed to emphasize the importance of the criteria in the first group while considering the relevance of the criteria in the second group. In the second scenario, weights of 40% were assigned to the first group and 60% to the second group. This weight allocation aimed to give greater weightage to the criteria in the second group, acknowledging their potential impact on the prioritization outcomes. The third scenario allocated weights of 70% to the first group and 30% to the second group. This adjustment sought to increase the significance of the criteria in the first group in the prioritization process while downplaying the influence of the criteria in the second group. Lastly, the fourth scenario involved assigning weights of 30% to the first group and 70% to the second group. This distribution placed more emphasis on the criteria in the second group, recognizing their critical role in the decision-making process.

By examining the results of the sensitivity analysis presented in Table 12, we can observe the variations in the rankings of the ASD patients across the different weight allocation scenarios. This analysis provides valuable insights into how changes in weight allocations can impact the prioritization outcomes. It helps us understand the robustness and stability of the decision-making process and identify the criteria that have the most significant influence on the final rankings. The sensitivity analysis enhances our understanding of the decision model and its reliance on the assigned weights. It provides valuable information for decision-makers to consider when using the prioritization framework in practice.

Table 13 and Table 14 present the results of the sensitivity analysis, showcasing the changes in rankings and their implications for selecting the most emergency patients across the four weight allocation scenarios using the VIKOR method.

Table 13 Ranks by four sensitivity analysis using phase 5 weights with the VIKOR method (patient 1 to 216)

Patients	Original Rank	Scenario 1	Scenario 2	Scenario 3	Scenario 4	Patients	Original Rank	Scenario 1	Scenario 2	Scenario 3	Scenario 4
P1	230	140	169	151	146	P109	224	318	227	339	88
<b>P2</b>	311	234	104	273	48	P110	267	219	109	217	188
P3	312	340	97	353	89	P111	152	213	293	170	349
P4	308	225	142	228	96	P112	237	288	238	277	306
P5	296	323	116	337	91	P113	18	419	400	417	399
P6	332	286	94	322	51	P114	125	236	321	219	375
<b>P7</b>	352	148	102	135	99	P115	74	133	364	18	315
P8	389	229	45	247	47	P116	376	14	89	16	132
P9	343	232	139	222	43	P117	274	375	182	375	50

P10	431	4	5	26	14	P118	411	141	21	181	19
P11	349	3	133	12	85	P119	355	28	99	46	276
P12	373	124	57	138	53	P120	61	120	414	83	361
P13	281	291	144	326	149	P121	294	12	137	6	243
P14	110	97	313	80	308	P122	141	334	288	305	354
P15	153	220	265	218	352	P123	86	275	351	252	402
P16	170	387	289	392	136	P124	139	64	349	51	381
P17	341	305	61	333	24	P125	320	154	180	100	137
P18	316	235	77	299	69	P126	288	15	152	7	332
P19	342	17	135	24	267	P127	54	314	398	278	411
P20	22	377	401	381	366	P128	87	282	344	260	295
P21	123	261	280	268	216	P129	310	386	100	396	80
P22	227	144	174	153	187	P130	166	208	258	180	178
P23	114	376	263	384	325	P131	414	109	27	127	44
P24	33	365	385	369	248	P132	137	246	275	251	271
P25	206	204	148	236	113	P133	307	228	141	237	74
P26	331	210	86	223	160	P134	368	128	66	136	18
P27	23	382	393	391	326	P135	276	90	93	111	247
P28	306	202	140	184	103	P136	136	62	363	57	374
P29	214	51	187	55	161	P137	193	47	260	20	342
P30	366	82	82	77	162	P138	71	396	361	390	338
P31	85	289	335	276	321	P139	165	61	264	9	302
P32	273	378	184	376	142	P140	117	260	294	258	277
P33	221	121	226	90	215	P141	205	143	222	126	303
P34	356	108	90	95	122	P142	121	238	326	220	350
P35	271	250	173	264	115	P143	163	196	281	146	390
P36	336	346	72	363	57 226	P144	150	418	336	416	272
P37	154	319	259	301	336	P145	171	57 25	256	54	227
P38	56 246	335	373	334	305	P146	359	25	87	45	100
P39 P40	346	114	98	99 270	118	P147	182	183 245	231	165 221	317 278
P40 P41	122 229	374	262	379	185	P148	106		348		
P41 P42	419	48 96	171 12	60 129	78 6	P149 P150	192 313	193 379	194 95	205 395	173 40
P43	383	218	32	263	25	P151	209	168	189	175	83
P44	218	132	214	112	138	P152	404	76	19	98	35
P45	344	163	58	201	131	P153	282	11	196	2	189
P46	40	356	375	359	403	P154	350	182	75	200	86
P47	46	336	378	332	391	P155	178	54	250	53	281
P48	102	89	337	72	398	P156	261	136	200	186	125
P49	259	34	257	11	194	P157	145	364	232	377	231
P50	184	184	215	176	68	P158	283	268	170	281	119
P51	425	86	4	133	8	P159	358	26	91	47	104
P52	338	173	106	169	111	P160	334	265	112	282	107
P53	319	351	111	352	64	P161	95	293	299	306	384
P54	147	254	241	292	184	P162	8	428	415	425	427
P55	260	24	164	23	109	P163	6	420	425	421	368
P56	407	145	22	193	46	P164	327	130	114	119	167
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P57	47	350	357	368	228	P165	53	406	345	414	343
P58	255	98	134	103	197	P166	120	84	322	92	312
P59	99	287	301	296	268	P167	257	269	190	284	127
P60	194	166	218	145	285	P168	180	206	212	208	206
P61	84	381	327	366	291	P169	199	49	219	49 22 <i>5</i>	319
P62	204	142	225	124	242	P170	49 295	337	376 47	335	282 81
P63 P64	304 378	205 209	143 53	187 224	95 135	P171	385 70	185 401	47 339	202 404	237
P65	75	117	384	107	413	P172 P173	354	179	73	404 194	37
P66	35	410	383	412	293	P174	364	102	63	108	120
P67	323	304	96	311	28	P175	130	239	296	229	213
P68	254	115	110	144	219	P176	243	126	145	140	220
P69	67	322	367	321	307	P177	160	199	283	147	299
P70	222	129	208	113	265	P178	300	391	127	397	251
P71	82	309	312	331	405	P179	25	197	423	122	378
P72	144	430	360	423	199	P180	239	105	201	86	208
P73	367	137	62	149	150	P181	318	164	120	154	133
P74	187	397	287	383	182	P182	176	55	252	61	345
P75	397	226	35	215	42	P183	30	366	387	370	396
P76	351	244	103	257	102	P184	143	69	324	68	330
P77	325	170	147	134	63	P185	226	38	239	29	235
P78	287	21	105	41	153	P186	248	31	178	32	210
P79	164	217	251	213	193	P187	330	174	118	160	180
P80	277	20	151	15	141	P188	208	36	278	3	323
P81	244	42	163	44	362	P189	135	248	273	254	266
P82 P83	16 148	352 216	428	314 172	421 304	P190	168 3	298 422	242 430	261 419	392 356
P84	309	221	310 136	226	98	P191 P192	3 195	187	430 198	196	261
P85	157	332	221	344	124	P193	127	233	315	216	234
P86	370	150	40	190	21	P194	252	35	138	42	289
P87	278	23	122	22	322	P195	76	110	380	110	355
P88	13	413	422	403	353	P196	293	252	157	267	76
P89	374	107	76	94	106	P197	393	171	38	192	41
P90	211	46	236	36	335	P198	268	19	205	8	203
P91	73	147	365	128	269	P199	335	347	34	364	54
P92	401	66	44	59	129	P200	97	357	338	318	241
P93	37	370	368	385	309	P201	266	33	124	40	174
P94	315	214	131	204	151	P202	174	189	246	166	382
P95	88	300	305	308	157	P203	14	416	419	407	324
P96	292	73	83	81	250	P204	345	243	79	233	79
P97	44	327	397	310	389	P205	375	201	81	185	65 204
P98	77	320	317	341	294	P206	314	215	129	211	284
P99	305	383	125	387	155	P207	32	399	420	354	341
P100	79 262	301 29	352	286 5	260 275	P208	387	176 8	42 78	197	29 75
P101 P102	262 60	29 324	149 371	3 325	275 240	P209	377 240	8 41	78 188	21 39	75 246
P102 P103	408	324 65	3/1 29	63	240 101	P210 P211	103	283	295	39 294	380
F 103	408	03	29	03	101	Г411	103	203	293	∠ <b>9</b> 4	300

P251

P359

P104	249	58	230	70	116	P212	63	306	394	274	425
P105	365	169	46	207	20	P213	12	380	421	367	404
P106	388	78	39	79	87	P214	269	292	199	303	218
P107	89	385	304	380	328	P215	80	302	350	295	290
P108	101	363	331	330	339	P216	231	122	203	104	232

P252	286	262	161	279	192	P360	104	285	290	298	145
P253	96	259	342	239	316	P361	420	88	13	116	27
P254	172	191	247	173	373	P362	303	284	168	235	226
P255	38	333	406	304	417	P363	118	83	318	67	370
P256	65	162	372	130	288	P364	201	408	267	411	258
P257	223	146	177	157	311	P365	270	224	217	199	286
P258	185	317	311	289	179	P366	189	326	276	327	253
P259	400	71	33	76	67	P367	297	331	160	357	56
P260	265	354	195	345	223	P368	361	149	71	164	60
P261	124	79	316	65	358	P369	181	59	224	73	212
P262	112	344	328	315	327	P370	362	227	85	245	147
P263	275	279	186	290	233	P371	55	172	382	189	377
P264	66	321	370	313	224	P372	62	343	347	362	198
P265	416	116	16	152	26	P373	415	212	3	250	2
P266	298	237	159	243	117	P374	158	63	266	58	254
P267	246	30	193	28	200	P375	413	13	56	17	11
P268	146	257	244	297	229	P376	426	75 202	7	91	61
P269	321	192	132	174	257	P377	11	393	408	399	420
P270	4	431	418	427	310	P378	186	315	309	288	126
P271	258	92	150	88	159	P379	427	72	6	89	4
P272	149	253	235	293	183	P380	379	297	26	323	5
P273	196	181	204	178	163	P381	423	80	9	101	22
P274	202	177	197	182	264	P382	284	267	166	275	52
P275	57 50	404	356	408	410	P383	417	190	11	232	12
P276 P277	50 289	310 194	402 211	270 137	409	P384	234 93	312 280	206	346 271	204 255
P277 P278	289 19	411	412	402	337 359	P385 P386	93 396	152	319 51	158	32
P279	285	10	167	1	230	P387	390	156	54	162	93
P280	177	188	245	168	186	P388	213	394	248	393	139
P281	360	258	60	316	23	P389	372	118	69	117	190
P282	432	7	10	19	16	P390	132	389	353	388	347
P283	390	222	43	227	31	P391	162	424	285	428	143
P284	430	18	28	31	17	P392	410	111	36	114	9
P285	236	123	175	125	70	P393	406	67	24	74	72
P286	428	139	1	206	1	P394	109	100	307	131	364
P287	272	241	192	241	172	P395	156	426	291	430	333
P288	337	255	117	269	148	P396	21	371	409	355	414
P289	90	368	343	340	292	P397	264	273	233	248	201
P290	322	362	88	365	36	P398	217	44	228	37	195
P291	340	359	59	382	77	P399	251	93	155	84	164
P292	324	251	162	238	170	P400	358	26	91	47	104
P293	339	52	153	64	84	P401	334	265	112	282	107
P294	386	178	49	188	169	P402	95	293	299	306	384
P295	133	384	369	373	262	P403	8	428	415	425	427
P296	280	303	130	350	71	P404	6	420	425	421	368
P297	398	2	68	10	112	P405	327	130	114	119	167
P298	234	312	206	346	204	P406	53	406	345	414	343

P299	93	280	319	271	255	P407	120	84	322	92	312
P300	396	152	51	158	32	P408	257	269	190	284	127
P301	392	156	54	162	93	P409	180	206	212	208	206
P302	213	394	248	393	139	P410	199	49	219	49	319
P303	372	118	69	117	190	P411	49	337	376	335	282
P304	132	389	353	388	347	P412	385	185	47	202	81
P305	162	424	285	428	143	P413	70	401	339	404	237
P306	410	111	36	114	9	P414	354	179	73	194	37
P307	406	67	24	74	72	P415	364	102	63	108	120
P308	109	100	307	131	364	P416	130	239	296	229	213
P309	156	426	291	430	333	P417	243	126	145	140	220
P310	21	371	409	355	414	P418	160	199	283	147	299
P311	264	273	233	248	201	P419	300	391	127	397	251
P312	217	44	228	37	195	P420	25	197	423	122	378
P313	251	93	155	84	244	P421	239	105	201	86	208
P314	100	272	314	266	280	P422	318	164	120	154	133
P315	279	242	154	256	97	P423	176	55	252	61	345
P316	381	9	65	27	171	P424	30	366	387	370	396
P317	42	342	386	343	424	P425	143	69	324	68	330
P318	9	388	427	378	408	P426	226	38	239	29	235
P319	210	43	255	34	407	P427	248	31	178	32	210
P320	72	277	405	212	426	P428	330	174	118	160	180
P321	64	307	389	280	274	P429	208	36	278	3	323
P322	402	22	31	43	114	P430	135	248	273	254	266
P323	126	432	306	432	318	P431	168	298	242	261	392
P324	111	230	366	179	423	P432	3	422	430	419	75
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Upon examining the rankings in Table 13 and Table 14, several observations can be made:

- Scenario 1: In this scenario, the first group of criteria is assigned a higher weight (60%) compared to the second group (40%). As a result, the rankings of some patients have changed compared to the original ranking. For example, patient A1 is ranked higher in Scenario 1 compared to the original ranking. This indicates that the criteria in the first group have a more significant impact on the prioritization outcomes in this scenario.
- Scenario 2: In this scenario, the weight allocation is reversed, with the second group of criteria receiving a higher weight (60%) and the first group a lower weight (40%). This adjustment leads to further changes in the rankings of the patients. For instance, patient A2 is ranked higher in Scenario 2 compared to the original ranking. This suggests that the criteria in the second group play a more prominent role in determining the emergency levels of the patients in this scenario.
- Scenario 3: Here, the weight allocation is skewed towards the first group of criteria, with a weight of 70%, while the second group receives a weight of 30%. This adjustment emphasizes the importance of the first group in the prioritization process. As a result, some patients experience significant changes in their rankings. For example, patient A3 is ranked considerably higher in Scenario 3 compared to the original ranking, indicating the increased significance of the criteria in the first group.
- Scenario 4: In this scenario, the weight allocation is reversed compared to Scenario 3. The second group of criteria is assigned a higher weight (70%), while the first group receives a lower weight (30%). The rankings of the patients are again affected by this adjustment. Patient A10, for instance, is ranked higher in Scenario 4 compared to the original

ranking. This suggests that the criteria in the second group have a greater influence on the prioritization outcomes in this scenario.

By comparing the rankings across the four scenarios, it becomes apparent that different weight allocations result in varying rankings for the ASD patients. This demonstrates the sensitivity of the prioritization process to changes in criteria weights. It also highlights the importance of carefully considering the weight allocation and its implications for identifying high emergency cases. In conclusion, the results in Table 13 and Table 14 provide insights into the impact of different weight allocation scenarios on the rankings of ASD patients. This information can assist in understanding the robustness and stability of the decision-making process and aid in the selection of the most critical cases for immediate attention and intervention.

#### 5. Conclusion

The integration of the FWZIC and VIKOR methods in this study has proven to be effective in prioritizing the emergency levels of autism patients with a moderate injury level. The developed framework provides a systematic and reliable approach for assessing and ranking patients based on their emergency needs. The experimental results and performance evaluation demonstrate the accuracy and effectiveness of the integrated methods in accurately prioritizing autism patients. The sensitivity analysis conducted in this study has highlighted the significance of weight configurations in the prioritization process. It emphasizes the need for careful consideration and selection of appropriate weights to ensure accurate and reliable rankings. This insight is crucial for healthcare professionals and decision-makers involved in prioritizing autism patients and allocating resources accordingly. The findings of this study contribute valuable insights to the field of autism patient prioritization, specifically for patients with a moderate injury level. By enabling early identification and intervention for these patients, healthcare professionals can take proactive measures to prevent their conditions from worsening. This framework addresses the needs of healthcare professionals and supports ongoing efforts to optimize patient care and resource allocation in the field of ASD medicine. However, it is important to acknowledge the limitations of this research. The study focused solely on autism patients with a moderate injury level and did not include patients with other levels of severity. Therefore, the proposed framework and results may not be applicable to patients in different severity categories. Additionally, the framework was developed and tested using a specific dataset, and its effectiveness may vary when applied to different datasets or populations. Further research and validation on diverse datasets are necessary to assess the generalizability of the framework.

In terms of future development, there is potential to create a real-time web-based application based on the proposed framework. Such an application would provide a user-friendly interface for healthcare professionals to input patient data and generate prioritization results efficiently. This would enhance the accessibility and usability of the framework. Furthermore, extending the framework to include other severity levels of injury, such as urgent injury and minor injury autistic patients, would be beneficial. Adapting and expanding the framework to cater to a broader range of cases would enhance its applicability and relevance in the field of ASD patient care. Overall, this study has contributed to the understanding of prioritizing emergency levels in autism patients and has provided a foundation for further research and development in this area.

#### **Conflicts of Interest**

The authors declare that there is no conflict of interests regarding the publication of this paper.

#### Funding

The authors receive no funding for this work.

#### Data availability

The datasets generated during and/or analysed during the current study are available from the corresponding author upon reasonable request.

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