Implementation of Simple Additive Weighting Method in Determination of Aid for Students From Low Socio-Economic Status Family

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Abstract. Assistance for underprivileged students (BSM) is a program from the government given to underprivileged students who excel, the goal is to provide relief in paying school fees. But there are problems in determining BSM, including the distribution of scholarships not on target. Therefore, to obtain better results and overcome these problems, a decision support system is used in determining BSM criteria. This study uses the method of Simple Additive Weighting (SAW), the method is often known as the weighted sum method. By applying the SAW method, it can help in making decisions for prospective poor student beneficiaries appropriately. The final results of this study found that the SAW method was able to overcome the problems in determining BSM with 100% accuracy.

Keywords: Decision Support System, Scholarship, Criteria, Simple Additive Weighting.

1 Introduction

Poor Student Aid is a form of material which aims to provide relief in paying school fees for underprivileged students, and to make poor students stay in school, prevent dropouts, attract poor students to stay in school, help students fulfill their learning needs, supporting a nine-year educational program, and helping the smooth running of school program[1]. Some of the results of the research carried out and continued studies in the BSM program show the weaknesses of this program, which found obstacles in distribution so that it was not in line with the main objectives of providing poor students assistance and the accuracy of the target of poor student assistance, where many non-poor households were found accept BSM and inadequate number of students[2]. The target of BSM recipients is still weak where many BSM recipients who are not from poor families are found and many students from poor families do not receive BSM benefits and the manuals used in BSM recipient ranking are many BSM recipients and many criteria indicators[3]. therefore, to obtain better results and overcome these problems, namely by developing an application that implements methods that can facilitate decision making and can help in increasing the targeting accuracy of BSM program recipients. In this study the author tries to use the Simple Additive Weighting method which method is often known as the addition method[4]-[7].

2 SAW Method Data Calculation

Here are the following criteria that required for taking decision in the process of selection of the underprivileged student scholarship receiver that drawn in Table 3.1:

Table. 1 Criteria Table

Criteria	Note
Presence/Attendance	C1
Status of Student Activeness	C2
Orphan Status	C3
Underprivileged Official Reference	C4
Parents Salary	C5

In the following table below contains weighted values of criteria that has determined by the decision maker. It can be changed as well as necessary and suitable with recent valid provision[5], [8]. The determined values are conduct to put implementation in the under way current system.

Table. 2 Weight Table

Criteria	Weight
C1	15 %
C2	15 %
C3	20 %
C4	25 %
C5	25 %

From the criteria above, it has been made some tiers of criteria significance based on alternative values that had determined into point value[9]–[12]. Henceforth, the compatibility rank in each alternative is displayed in the following table 3:

Table. 3 Compatibility Rate

Value	C1	C2	C3	C4	C5
10	100%	Active	Yes	Orphan	>= Rp. 2.000.000,-
8	>=90%			Orphan	Rp. 1.500.000,- s.d Rp. 2.000.000,-
6	>=75%				Rp. 1.000.000,- s.d Rp. 1.500.000,-
4	>=70%				Rp. 500.000,- s.d Rp. 1.000.000,-
0	<70%	Non active	No	Not Orphan	<= Rp. 500.000,-

According to the criteria and compatibility rate in each alternative toward determined criteria, resulting next phase that is to alternate delineation in every criteria that has converted with point values[11], [13].

Here are the following manual calculation based on the sample case. There are five alternatives contained data as described by the table below:

Table. 4 Alternative Data

Student	Attendance	Status of Students	Underprivileged	Orphan	Parents Salary
		Activeness	official reference	Condition	
A1	>=90%	Active	Yes/Have	Orphan	<= Rp. 500.000,-
A2	100%	Active	No/Does not have	Not orphan	Rp. 500.000,- s.d Rp. 1.000.000,-
A3	100%	Active	Yes/Have	Not orphan	<= Rp. 500.000,-
A4	>=70%	Active	Yes/Have	Orphan	<= Rp. 500.000,-
A5	0%	Not Active	No/Does not have	Not Orphan	Rp. 500.000,- s.d Rp. 1.000.000,-

Here are the step-by-step completion phases of SAW method[14], [15]:

a. Determine the criteria for decision-making. (Criteria)

- b. Determine the compatibility rates in each alternative of detremined criteria
- c. Making the criteria decision matrix (Ci)
- d. Matrix normalization from its similiarity is adjusted into its kind of attribute (profit/max or cost/min) in order to obtain normalized matrix as R. The sum of R normalized matrix multiplication with weight vector could obtain biggest values that used by proper alternative from other existing alternative (Ai) as the problem solver.

The formula in the process of normalization is in the following:

$$r_{ij} = \begin{cases} \frac{x_{ij}}{\max x_{ij}} & \text{if } j \text{ atribut benefit} \\ \frac{\min x_{ij}}{x_{ij}} & \text{if } j \text{ atribut cost} \end{cases}$$
(1)

Notes:

 R_{ij} = Normalized implementation rate

 $Max_{ij} = Maximum$ value from each column and row

 $Min_{ij} = Minimum$ value from each column and row

 X_{ij} = Column and row of matrix

In addition, Rij as the normalized implementation rate from Ai alternative in attribut Cj: i=1,2...m dan j=1,2...n

Preference values for each alternative (Vi) are given half :

$$V_i = \sum_{j=1}^n W_j r_{ij} \tag{2}$$

Where :

 V_i = Final value from alternative

 $W_j = Determined weight$

 $R_{ij} = Normalization$

In addition, the bigger Vi value indicates that Ai alternative is more selected.

5 Result and Discussion

The following table below depicts criteria that required in the decision-making of the selection process of underprivileged student scholarship receiver. The determined caracteria are:

Tab	le.	5	Criteria

Nama Kriteria	Keterangan
Presence/Attendance	C1
Status of Student Activeness	C2
Orphan Condition	C3
Underprivilege Official Reference	C4
Parents Salary	C5

In addition, the table below shows the weight value of criteria that had determined by the decision-maker that can changed as well as the current valid provision. The values has determined to put implementation in the under way system ahead.

Table. 6 Weight

Criteria	Weight
C1	15 %
C2	15 %
C3	20 %
C4	25 %
C5	25 %

Table. 7 Compatibility rate

Value	C1	C2	C3	C4	C5
10	100%	Active	Yes/have	Orphan	>= 2.000.000,-
8	>=90%			Orphan	Rp.1.500.000,- s.d Rp. 2.000.000,-
6	>=75%				Rp. 1.000.000,- s.d Rp. 1.500.000,-
4	>=70%				Rp. 500.000,- s.d Rp. 1.000.000,-
0	<70%	Not active	No/Does not have	Not orphan	<= Rp. 500.000,-

Based on the criteria and compatibility rate in each alternative towards determined criteria that it has drawn the alternative description in each converted criteria with the point values.

Here are the following manual calculation based on the case sample. There are five alternatives had data as described below:

Table. 8 Compatibility Rate from each alternatives in criteria

Alternative	C1	C2	C3	C4	C5
A1	8	10	10	8	2
A2	10	10	2	2	4
A3	10	10	10	2	2
A4	4	10	10	8	2
A5	2	2	2	2	4

Its calculation based on matrix formula : From C1 :

$$A1 = \frac{8}{Max \{ 8,10,10,4,2 \}} = \frac{8}{10} = 0.8$$

$$A2 = \frac{10}{Max \{ 8,10,10,4,2 \}} = \frac{10}{10} = 1$$

$$A3 = \frac{10}{Max \{ 8,10,10,4,2 \}} = \frac{10}{10} = 1$$

$$A4 = \frac{4}{Max \{ 8,10,10,4,2 \}} = \frac{4}{10} = 0.4$$

$$A5 = \frac{2}{Max \{ 8,10,10,4,2 \}} = \frac{2}{10} = 0.2$$

From C2 :

$$A1 = \frac{10}{Max \{ 10,10,10,4,2 \}} = \frac{10}{10} = 1$$

$$A2 = \frac{10}{Max \{ 10,10,10,4,2 \}} = \frac{10}{10} = 1$$

$$A3 = \frac{10}{Max \{ 10,10,10,4,2 \}} = \frac{10}{10} = 1$$

$$A4 = \frac{10}{Max \{ 10,10,10,4,2 \}} = \frac{10}{10} = 1$$

$$A5 = \frac{2}{Max \{ 10,10,10,4,2 \}} = \frac{2}{10} = 0.2$$

From C3 :

om C3 :

$$A1 = \frac{10}{Max \{ 10,2,10,10,2 \}} = \frac{10}{10} = 1$$

$$A2 = \frac{2}{Max \{ 10,2,10,10,2 \}} = \frac{2}{10} = 0.2$$

$$A3 = \frac{10}{Max \{ 10,2,10,10,2 \}} = \frac{10}{10} = 1$$

$$A4 = \frac{10}{Max \{ 10,2,10,10,2 \}} = \frac{10}{10} = 1$$

$$A5 = \frac{2}{Max \{ 10,2,10,10,2 \}} = \frac{2}{10} = 0.2$$

From C4 :

$$A1 = \frac{8}{Max \{ 8,2,2,8,2 \}} = \frac{8}{8} = 1$$

$$A2 = \frac{2}{Max \{ 8,2,2,8,2 \}} = \frac{2}{8} = 0.25$$

$$A3 = \frac{2}{Max \{ 8,2,2,8,2 \}} = \frac{2}{8} = 0.25$$

$$A4 = \frac{2}{Max \{ 8,2,2,8,2 \}} = \frac{8}{8} = 1$$

$$A5 = \frac{2}{\text{Max}\left\{8,2,2,8,2\right\}} = \frac{2}{8} = 0.25$$

From C5:

$$A1 = \frac{\min\{2,4,2,2,4\}}{2} = \frac{2}{2} = 1$$

$$A2 = \frac{\min\{2,4,2,2,4\}}{4} = \frac{2}{4} = 0.5$$

$$A3 = \frac{\min\{2,4,2,2,4\}}{2} = \frac{2}{2} = 1$$

$$A4 = \frac{\min\{2,4,2,2,4\}}{2} = \frac{2}{2} = 1$$

$$A5 = \frac{\min\{2,4,2,2,4\}}{4} = \frac{2}{4} = 0.5$$

Then, it acquires normalized matrix values as follows:

	F 0.8	1	1	1	ן 1	
	1	1	0.2	0.25	0.5	
R =	1	1	1	0.25	1	
	0.4	1	1	1	1	
	L.02	0.2	0.2	0.25	0.5	

The next stage is to conduct preference values for each alternative (Vi) by sum the result from normalized matrix multiplication (R) with the weight value (W). Then, the bigger Vi value indicates that (Ai) alternative is the most selected.

 $W = [0.15 \ 0.15 \ 0.20 \ 0.25 \ 0.25]$

Then, the preference values as follows :

$$\begin{aligned} & V1 = (0.8x0.15) + (1x0.15) + (1x0.20) + (1x0.25) + (1x0.25) = 0.97 \\ & V2 = (1x0.15) + (1x0.15) + (0.2x0.20) + (0.25x0.25) + (0.5x0.25) = 0.528 \\ & V3 = (1x0.15) + (1x0.15) + (1x0.20) + (0.5x0.25) + (1x0.25) = 0.8113 \\ & V4 = (0.4x0.15) + (1x0.15) + (1x0.20) + (1x0.25) + (1x0.25) = 0.91 \\ & V5 = (0.2x0.15) + (0.2x0.15) + (0.2x0.20) + (0.25x0.25) + (0.5x0.25) = 0.288 \end{aligned}$$

The final step is the ranking process. Here are the following results that acquired some of the biggest values:

Table.	9 Ran	king	Resul	t
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Ranking	Alternative (student)	Value
1	A1	0.97
2	A4	0.91
3	A3	0.811
4	A2	0.528
5	A5	0.288

6 Conclusion

In sum up, after implementation of Simple Additive Weighting (SAW) method in the support system of decision determination regarding underprivileged student aid, it could be concluded that this system is produce decision in the form of student ranking list which is contain some of the list of student that proper to get the aid as well as suitable with the determined criteria. Based on the previous conducted test result, from 20 people as the sample data which consists of underprivileged aid receiver candidates, shows that the accuracy level up to 100%. This result however illustrated that this implementation of Simple Additive Weighting (SAW) method in recent system is reach its succeed-rate and it works well.

Last but not least, this support system of decision determination in selection of underprivileged student aid is still far from perfect because it is using only one method as well as its existing system still manual. The developer could make some system that compares one method with another. Further conclusion and suggestion are highly required to this research.

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