

Decision Support System for Plantation Land Recommendations in Mandailing Natal Regency Using The TOPSIS Method


Lidya Rosnita^{1*}, Bustami², Dini Kairiyah Samosir³, Hafizh Al Kautsar Aidilof⁴

^{1,2,3,4}Department of Informatic Engineering, Faculty of Engineering, Malikussaleh University, Indonesia

ABSTRACT

Plantations have great potential to be developed because they are a source of income for the community, farmers and PTPN in the area because Indonesia has the largest plantation land in the world. Mandailing Natal Regency is the district with the largest area in North Sumatra province, but Mandailing Natal has not been able to surpass crop production from plantation land in North Sumatra. The method for determining plantation land to produce good harvests is the TOPSIS algorithm for recommending plantation land. In this research, recommendations were made for plantation land with the aim of finding out what land is suitable in each sub-district in Mandailing Natal Regency. The data method for this plantation land was taken at the Mandailing Natal Central Statistics Agency with the variables used, namely land area, area height, topography and rainfall. In implementing The decision support system using the TOPSIS algorithm on plantation land in Mandailing Natal Regency, the Recommendation results based on the type of plantation land in each sub-district are as follows: Rubber Plantation Land is highly recommended in Siabu District with a preference value of 0.56938682730811 and highly not recommended in Penyabungan District with preference value 0.33537499293319. Then, Palm Oil Plantation Land is highly recommended in Batang Natal sub-district with a preference value of 0.53467087652891 and not highly recommended in Tambangan sub-district with a preference value of 0.33181406496882. And the last one is Cocoa Farm which is highly recommended in Tambangan District with a preference value of 0.62855110465075 and highly not recommended in East Panyabungan District with a preference value of 0.25592982445435.

Keyword : Plantation; Recommendations; TOPSIS.

 This work is licensed under a Creative Commons Attribution-ShareAlike 4.0 International License.

Corresponding Author:

Lidya Rosnita,
Department of Informatic Engineering
Malikussaleh University
Jl. Batam, Blang Pulo, Muara Satu, Lhokseumawe, Aceh, 24351, Indonesia.
Email : lidyarosnita@unimal.ac.id

Article history:

Received Dec 23, 2024
Revised Jan 15, 2025
Accepted Mar 10, 2025

1. INTRODUCTION

Plantations are an important sector in Indonesia's economy, playing a role in boosting economic growth, creating jobs, and meeting the public's needs for both food and non-food products. Indonesia has vast plantation potential with various types of crops such as oil palm, coffee, tea, rubber, and others. However, the conditions and characteristics of plantations in each region are very diverse and require different development strategies.

Mandailing Natal Regency is one of the areas in North Sumatra with significant plantation potential. In addition, Mandailing Natal Regency also produces other plantation commodities. However, it is often the case that plantations are established in areas that are not suitable for them, resulting in unsatisfactory yields. To avoid harm to the community from poor plantation results, the people need to know how to determine which crops are suitable for their area. Therefore, data mining using the TOPSIS algorithm is required to recommend plantation land in Mandailing Natal Regency (Agricultural Statistics, 2014).

Land recommendations for plantations in Mandailing Natal Regency are crucial for sustainable plantation management. Good recommendations can assist in decision-making for plantation management, including the development and improvement of production. Plantation area recommendations can also help address issues such as climate change, plant diseases, and natural disasters.

In the context of Mandailing Natal Regency, a Decision Support System that recommends plantation land using the TOPSIS method can help understand the characteristics and potential of plantation lands, as well as assist in developing better and more sustainable plantation development strategies. With effective and accurate plantation land recommendations, it is hoped that the productivity and quality of plantation yields, as well as the welfare of the people in Mandailing Natal Regency, can be improved.

2. Research Stages

Technique for Order Preference by Similarity to Ideal Solution (TOPSIS) is one of the methods in the MADM (Multi-Attribute Decision Making) decision model. The TOPSIS method uses the concept where the selected best alternative not only has the shortest distance from the positive ideal solution but also has the longest distance from the negative ideal solution. The concept is simple, easy to understand, computationally efficient, and has the ability to measure relative performance, making this method widely used in MADM models (Khomsatun et al., 2020). It should be noted that even though an alternative has the smallest distance from the positive ideal solution, it does not always mean that the alternative will have the largest distance from the negative ideal solution.

This concept is widely used in several MADM models to solve decision problems practically. This is because the concept is simple and easy to understand, computationally efficient, and has the ability to measure the relative performance of decision alternatives in a simple mathematical form (Wahyuni, 2017).

TOPSIS follows the steps as follows:

1. Construct the normalized decision matrix:

The first step is to determine the normalized decision matrix using the following formula:

$$r_{ij} = \frac{x_{ij}}{\sqrt{\sum_{i=1}^m x_{ij}^2}} \quad (1)$$

Where : x_{ij} = decision matrix

r_{ij} = normalized matrix

2. Construct the weighted normalized decision matrix:

After constructing the normalized decision matrix, the next step is to determine the weighted normalized matrix, using the following formula:

$$y_{ij} = w_i \cdot r_{ij} \quad (2)$$

Where : y_{ij} = weighted normalized matrix

w_i = weighted vector from the data process

r_{ij} = normalized matrix

3. Determine the positive ideal solution matrix and the negative ideal solution matrix:

After determining the weighted normalized matrix, the next step is to determine the positive and negative ideal solutions. The positive ideal solution A^+ and the negative ideal solution A^- can be determined based on the normalized weighted matrix (y_{ij}), using the following formula:

- a. Positive Ideal Solution

$$A^+ = (y_1^+, y_2^+, \dots, y_n^+) \quad (3)$$

- b. Negative Ideal Solution

$$A^- = (y_1^-, y_2^-, \dots, y_n^-) \quad (4)$$

Where : $y_j^+ = \begin{cases} \max y_{ij}, & \text{if } j \text{ is a benefit attribute} \\ \min y_{ij}, & \text{if } j \text{ is a cost attribute} \end{cases}$

$y_j^- = \begin{cases} \min y_{ij}, & \text{if } j \text{ is a benefit attribute} \\ \max y_{ij}, & \text{if } j \text{ is a cost attribute} \end{cases}$

4. Determining the distance between the value of each alternative and the positive and negative ideal solution matrices.

After determining the positive and negative ideal solutions, the next step is to determine the distance between the positive and negative ideal solutions using the following formula:

- a. Positive Distance

$$D_i^+ = \sqrt{\sum (y_i^+ - y_{ij}^+)^2} \quad (5)$$

- b. Negative Distance

$$D_i^- = \sqrt{\sum (y_i^- - y_{ij}^-)^2} \quad (6)$$

Where : D_i = distance to the ideal solution

y_{ij} = value of the ideal solution

y_{ij}^+ = value of the positive ideal solution

y_{ij}^- = value of the negative ideal solution

5. Determining the preference value for each alternative.

After all the steps have been determined, the next step is to determine the preference value using the following formula:

$$v_i = \frac{D_i^-}{D_i^- + D_i^+} \quad (7)$$

Where : V_i = preference value
 D_i^- = negative ideal distance
 D_i^+ = positive ideal distance

6. Determining the Rank.
 After completing all the steps, the final step is to determine the rank from first to last based on the number of reference values.

The following is the system schema used in the TOPSIS method:

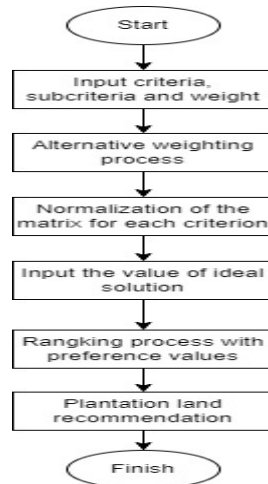


Fig 1. System Schema

The collected data set will be analyzed using the TOPSIS algorithm. The analysis will be performed using four attributes, namely Land Area, Elevation, Topography, and Rainfall in 2022. These attributes are initialized in the following table:

Table 1. Attribute Initialization

Attribute	Initialization
Land Area	K1
Elevation	K2
Topography	K3
Rainfall	K4

In this research, the alternatives provided are used, and these alternatives represent plantation land from the case study research. There are 4 criteria and data from the research results:

Table 2. Subcriteria Option for Land Area

Land Area	Grade
10-20 (Ha)	1
21-30 (Ha)	2
31-60 (Ha)	3

Table 3. Selection of The Elevation Criteria

Elevation	Grade
0-300 (mdpl)	1
200-600 (mdpl)	2
600-1.300 (mdpl)	3

Table 4. Selection of The Topography Criteria

Topography	Grade
Hilly to mountainous	1
Wavy to hilly	2
Flat to hilly	3
Flat to wavy	4

Table 5. Selection of The Rainfall Criteria

Rainfall	Grade
112-300 (mm)	1
125-722 (mm)	2
127-400 (mm)	3
500-1481 (mm)	4

3. RESULTS AND DISCUSSION

Plantations in Mandailing Natal Regency are one of the main sources of income for the local community. Mandailing Natal Regency has a climate that is very suitable for agriculture and plantations, making it one of the main producers of agricultural and plantation commodities in North Sumatra. The area of Mandailing Natal is 4,372.86 km², consisting of 30,749 hectares of rice fields, 363,439.14 hectares of non-rice agricultural land, and 43,286 hectares of non-agricultural land. Agriculture is the driving force behind the economic development of Mandailing Natal Regency, as reflected in the composition of the Gross Regional Domestic Product (GRDP) in 2021, where the agricultural, forestry, and fisheries sectors contributed the largest share of 51.34 percent to the economy of Mandailing Natal. Generally, plantation companies have their own processing units, so the products marketed are already in the form of processed goods. Processed products are primary productions that have been processed into finished or semi-finished goods, thereby increasing their economic value (BPS, 2021).

Oil palm plantations in Mandailing Natal are spread across several subdistricts. These plantations are generally managed by large companies that employ many workers. The management of oil palm plantations in Mandailing Natal is carried out intensively with proper irrigation systems to ensure that the oil palm plants grow optimally. In addition to being a source of income for the companies, oil palm plantations also provide significant economic benefits to the local community. Jobs in the oil palm plantations can absorb a large workforce and provide stable income for workers. Oil palm plantations in Mandailing Natal are an important sector for the region's economy. However, their management must still consider environmental and social aspects to ensure that they provide sustainable benefits to the community and the surrounding environment.

Mandailing Natal Regency has a climate that is very suitable for rubber cultivation, making it one of the main rubber producers in North Sumatra. Rubber plantations in Mandailing Natal are generally managed by smallholder farmers who own rubber plantations of varying sizes. These smallholder farmers often collaborate with rubber agents or factories to supply their harvests. Generally, rubber farmers in Mandailing Natal manage their plantations in a traditional manner. The process of creating planting holes, applying fertilizer, and planting rubber seedlings is still done manually. Once the rubber seedlings are planted, maintenance tasks such as pruning, weed clearing, and irrigation are also done manually by the farmers. During the harvest season, rubber is collected by tapping the latex from the trees and collecting it in containers. The latex is then processed into raw rubber material, which is ready to be sold to rubber processing factories.

Although rubber plantations in Mandailing Natal provide significant economic benefits for farmers and the surrounding area, their management also raises some social issues. Common issues include the low selling price of rubber, which affects the welfare of farmers, and the lack of knowledge and technology, which leads to low productivity in the rubber plantations. To address these issues, the government and relevant parties have taken various measures, such as providing assistance and training on technology for rubber farmers, opening new market access, and establishing modern and efficient rubber production centers. Overall, rubber plantations remain an important sector for the regional economy in Mandailing Natal. However, their management needs to be continuously improved to enhance the welfare of rubber farmers and ensure that the surrounding environment is preserved.

The recommendation for plantation land using the TOPSIS algorithm has been successfully carried out, categorizing the land into three types of plantations: cocoa, oil palm, and rubber. The following is the data from each district that will be used in the calculation.

Table 6. Data by Distric

Number	Distric	Land area (Ha)	Elevation (mdpl)	Topography	Rainfall (mm)
1	Kotanopan	1	2	1	3
2	Pakantan	3	3	1	3
3	Bukit Malintang	3	2	3	4
4	Batahan	1	1	4	2
5	Sinunukan	1	1	4	2
6	Batang Natal	3	3	2	2
7	Lingga Bayu	3	3	2	2

8	Ranto Baek	3	3	2	2
9	Ulu Pungkut	1	2	1	3
10	Tambangan	1	1	1	3
11	Lembah Sorik Marapi	2	2	1	3
12	Puncak Sorik Marapi	3	3	1	3
13	Muara Sipongi	2	3	1	3
14	Payabungan	2	3	3	1
15	Payabungan Selatan	2	3	1	1
16	Payabungan Barat	2	1	3	1
17	Payabungan Utara	2	1	3	1
18	Payabungan Timur	3	2	1	1
19	Huta Bargot	2	2	3	1
20	Natal	1	1	4	4
21	Muara Batang Gadis	2	2	4	4
22	Siabu	2	2	2	4
23	Naga Juang	3	2	3	4
	Weight	0,2	0,3	0,3	0,2

The final result of the ideal solution for the rubber plantation is as follows:

Table 7. The Preference Value of The Rubber Plantation

Distric	Value	Ranking
Siabu	0.56938682730811	1
Lembah Sorik Marapi	0.562357948894	2
Tambangan	0.56175943623131	3
Bukit Malintang	0.55544108297678	4
Naga Juang	0.55544108297678	5
Kotanopan	0.53744446250792	6
Ulu Pungkut	0.53744446250792	7
Pakantan	0.51065684291359	8
Puncak Sorik Marapi	0.51065684291359	9
Muara Sipongi	0.51022253844963	10
Muara Batang Gadis	0.50440288626509	11
Natal	0.48190444992693	12
Batang Natal	0.4377233768058	13
Lingga Bayu	0.4377233768058	14
Ranto Baek	0.4377233768058	15
Payabungan Timur	0.41459532382605	16
Batahan	0.37302178979839	17
Sinunukan	0.37302178979839	18
Payabungan Barat	0.36500573405374	19
Payabungan Utara	0.36500573405374	20
Payabungan Selatan	0.36346913658877	21
Huta Bargot	0.35207310508025	22
Panyabugan	0.33537499293319	23

From the calculation, we can determine that the area highly recommended is Siabu district, while the area not recommended is Payabugan district for rubber plantation. The final result of the ideal solution for the oil palm plantation is as follows:

Table 8. The Preference Value of The Oil Palm Plantation

Distric	Value	Ranking
Batang Natal	0.53467087652891	1
Lingga Bayu	0.53467087652891	2
Ranto Baek	0.53467087652891	3
Payabungan Timur	0.53465365229525	4
Huta Bargot	0.5168613898434	5
Panyabugan	0.50266042948436	6
Payabungan Barat	0.50168345455916	7
Payabungan Utara	0.50168345455916	8
Payabungan Selatan	0.48866190830181	9
Pakantan	0.4750449890252	10
Puncak Sorik Marapi	0.4750449890252	11
Bukit Malintang	0.46278273759928	12
Naga Juang	0.46278273759928	13
Muara Sipongi	0.42941964760573	14
Lembah Sorik Marapi	0.42801801835755	15
Muara Batang Gadis	0.41660326776286	16
Batahan	0.41060883445268	17
Sinunukan	0.41060883445268	18
Siabu	0.39609742209913	19

Natal	0.36519193188955	20
Kotanopan	0.35812012822587	21
Ulu Pungkut	0.35812012822587	22
Tambangan	0.33181406496882	23

From the calculation, we can determine that the area highly recommended is Batang Natal district, while the area not recommended is Tambangan district for oil palm plantation. The final result of the ideal solution for the cocoa plantation is as follows:

Table 9. The Preference Value of The Cocoa Plantation

Distric	Value	Ranking
Tambangan	0.62855110465075	1
Siabu	0.62322981566	2
Kotanopan	0.62187907090507	3
Ulu Pungkut	0.62187907090507	4
Natal	0.57912238180828	5
Lembah Sorik Marapi	0.5768513625647	6
Muara Batang Gadis	0.57150665614346	7
Bukit Malintang	0.56384724704411	8
Naga Juang	0.56384724704411	9
Muara Sipongi	0.53362843402673	10
Pakantan	0.47706367108479	11
Puncak Sorik Marapi	0.47706367108479	12
Batahan	0.42406994061151	13
Sinunukan	0.42406994061151	14
Batang Natal	0.36873077076193	15
Lingga Bayu	0.36873077076193	16
Ranto Baik	0.36873077076193	17
Payabugan	0.32144373987322	18
Huta Bargout	0.32049790893688	19
Payabugan Selatan	0.31970147322884	20
Payabugan Barat	0.31964042463454	21
Payabugan Utara	0.31964042463454	22
Payabugan Timur	0.25592982445435	23

From the calculation above, we can determine that the area highly recommended is Tambangan district, while the area not recommended is Payabugan Timur district for cocoa plantation. The preference results for each land in every subdistrict are as follows:

Table 10. The Recommendation Results from Each District

Distric	Rubber Preference result	Oil Palm Preference result	Cocoa Preference result	Ranking
Tambangan	0.5693868273	0.628551104	0.62855110465	Cocoa
Siabu	0.5623579488	0.623229815	0.62322981566	Rubber
Kotanopan	0.5617594362	0.6218790709	0.62187907090	Cocoa
Ulu Pungkut	0.5554410829	0.6218790709	0.62187907090	Cocoa
Natal	0.5554410829	0.579122381	0.57912238180	Cocoa
Lembah Sorik Marapi	0.5374444625	0.576851362	0.57685136256	Oil Palm
Muara Batang Gadis	0.53744446250	0.5715066561	0.57150665614	Oil Palm
Bukit Malintang	0.51065684291	0.5638472470	0.56384724704	Oil Palm
Naga Juang	0.51065684291	0.5638472470	0.56384724704	Cocoa
Muara Sipongi	0.51022253844	0.5336284340	0.53362843402	Cocoa
Pakantan	0.50440288626	0.4770636710	0.47706367108	Cocoa
Puncak Sorik Marapi	0.48190444992	0.4770636710	0.47706367108	Rubber
Batahan	0.43772337680	0.4240699406	0.42406994061	Cocoa
Sinunukan	0.43772337680	0.4240699406	0.42406994061	Oil Palm
Batang Natal	0.43772337680	0.3687307707	0.36873077076	Oil Palm
Lingga Bayu	0.41459532382	0.3687307707	0.36873077076	Oil Palm
Ranto Baik	0.37302178979	0.3687307707	0.36873077076	Oil Palm
Payabugan	0.37302178979	0.3214437398	0.32144373987	Oil Palm
Huta Bargout	0.36500573405	0.3204979089	0.32049790893	Oil Palm
Payabugan Selatan	0.36500573405	0.3197014732	0.31970147322	Cocoa
Payabugan Barat	0.36346913658	0.3196404246	0.31964042463	Cocoa
Payabugan Utara	0.3520731050	0.31964042463	0.31964042463	Cocoa
Payabugan Timur	0.3353749929	0.25592982445	0.25592982445	Cocoa

4. CONCLUSION

In its application, the Decision Support System using the TOPSIS algorithm for plantation land in Mandailing Natal Regency yielded the following recommendations based on the type of plantation land in each district:

1. Rubber plantation land is highly recommended in Siabu District with a preference value of 0.56938682730811.
2. Oil palm plantation land is highly recommended in Batang Natal District with a preference value of 0.53467087652891.
3. Lastly, cocoa plantation land is highly recommended in Tambangan District with a preference value of 0.62855110465075.

REFERENCES

- Abidin, N. A. S. Z., Avila, R. D., Hermatyar, A., & Rismayani, R. (2022). Comparison of K-Means and K-Medoids Algorithms for Grouping Cocoa Production Areas. In *JUTISI (Jurnal Teknik Informatika dan Sistem Informasi)*, 8(2), 383-391. <https://doi.org/10.28932/jutisi.v8i2.4897>
- Fadlisyah, F., Rosnita, L., & Pane, M. W. A. (2023). Decision Support System in Determining High School Student Specialization Using MOORA Method. In *JICS (Journal of Informatics and Computer Science)*, 9(1), 65-70. <https://doi.org/10.33143/jics.v9i1.2945>
- Fajriana, F. (2021). Analysis of the K-Medoids Algorithm in The Capture Fisheries Production Clustering System of North Aceh Regency. In *JEPIN (Jurnal Edukasi dan Penelitian Informatika)*, 7(2), 263-269. <https://dx.doi.org/10.26418/jp.v7i2.47795>
- Harahap, L. M., Fuadi, W., Rosnita, L., Darnila, E., & Meiyanti, R. (2022). Clustering of Featured Vegetables Using The K-Means Algorithm. In *JUTISI (Jurnal Teknik Informatika dan Sistem Informasi)*, 8(3), 567-579. <https://doi.org/10.28932/jutisi.v8i3.5277>
- Karlina, L., & Nurdiawan, O. (2023). Application of K-Medoids in Classifying The Distribution of Critical Land in West Java Based on District/City. In *JATI (Jurnal Mahasiswa Teknik Informatika)*, 7(1), 527-532. <https://doi.org/10.36040/jati.v7i1.6348>
- Khomsatun, K., Ikhsan, D., Ali, M., & Kursini, K. (2020). Decision Support System for Selecting Planting Areas in Wonosobo Regency Using K-Means Clustering and TOPSIS. In *Jurnal Nasional Pendidikan Teknik Informatika (JANAPATI)*, 9(1), 55. <https://doi.org/10.23887/janapati.v9i1.23073>
- Marlina, D., Lina, N., Fernando, A., & Ramadhan, A. (2018). Implementation of K-Medoids and K-Means Algorithm for Grouping of Defect in Children. In *Jurnal CoreIT: Jurnal Hasil Penelitian Ilmu Komputer dan Teknologi Informasi*, 4(2), 64. <https://doi.org/10.24014/coreit.v4i2.4498>
- Maulida, R. (2018). Application of Data Mining in Grouping Tourist Visits to Top Tourist Attractions in DKI Jakarta Province Using K-Means. In *JISKA (Jurnal Informatika Sunan Kalijaga)*, 2(3), 167. <https://doi.org/10.14421/jiska.2018.23-06>
- Mubarok, A., Suherman, H. D., Ramdhani, Y., Topiq, S. (2019). Decision Support System for Credit Granting Feasibility Using The TOPSIS Method. In *Jurnal Informatika*, 6(1), 37-46. <https://doi.org/10.31311/ji.v6i1.4739>
- Risawandi, R., Rosnita, L., & Putra, R. K. (2023). Decision Support System for Predicting The Graduation of Informatics Students Using the SAW Method. In *Journal of Informatics and Computer Science (JICS)*, 9(1), 58-64. <https://doi.org/10.33143/jics.v9i1.2944>
- Rosnita, L., Yunizar, Z., & Ananda, E. F. (2024). Implementation of Data Mining in Determining Drug Purchase Patterns Using the Apriori Method. In *Journal Serambi Engineering (JSE)*, 9(3), 9459-9466. <https://doi.org/10.32672/jse.v9i3.1679>
- Sadli, M., Fajriana, F., Fuadi, W., Ermatita, E. & Pahendra, I. (2018). Application of The K-Nearest Neighbors Model in Classifying Electricity Demand for Each Area in Lhokseumawe City. In *Jurnal ECOTIPE*, 5(2), 11-18. <https://doi.org/10.33019/ecotipe.v5i2.64668>
- Sulistiyawati, A. A. D., & Sadikin, M. (2021). Application of the K-Medoids Algorithm to Determine Customer Segmentation. In *Sistemasi*, 10(3), 516. <https://doi.org/10.32520/stmsi.v10i3.1332>
- Wahyuni, E. G. (2017). Decision Support System for Employee Recruitment Using The TOPSIS Method. In *Jurnal Sains, Teknologi Industri*, 14(1), 108-116.
- Wira, B., Budianto, A. E., & Wiguna, A. S. (2019). Implementation of K-Medoids Clustering Method to Identify Study Program Selection Patterns of New Students in 2018 at Kanjuruhan University Malang. In *RAINSTEK : Jurnal Terapan Sains & Teknologi*, 1(3), 53-68. <https://doi.org/10.21067/jtst.v1i3.304>