

Analysis of Decreased Public Awareness in the Application of Health Protocols with the C4.5 . Algorithm

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Abstract

The purpose of this study is to determine the dominant factors that affect the decline in public awareness on the application of health protocols using the C4.5 Algorithm. Sources of data used in this study obtained by conducting observations and interviews. The variables used include (1) Employment, (2) Environment, (3) Sanctions and (4) Concern. The research test process uses RapidMiner software to create a decision tree. The results obtained 6 rules with 4 rules decreasing status and 2 rules increasing status. The level of accuracy obtained is 100%. The results of this study are expected to be input for the surrounding community to better understand the importance of implementing health protocols at this time, so that they can help the Government to succeed in the health protocol awareness program in inhibiting the spread of Covid-19 in Indonesia.

Keywords: C4.5 Algorithm, Covid-19, Data Mining, Society

1. Introduction

The implementation of health protocols is an important thing that must be done in preventing or protecting themselves from the spread of the Corona virus, but so far most people think this is not an important thing that must be done continuously. There are several possibilities that influence the community to no longer be disciplined in applying health protocols, such as in terms of work, the environment, the application of sanctions in the community and public concern in responding to the Covid-19 cases that occur. During the determination of Covid-19 as a pandemic in Indonesia, many people experienced fear, their health declined, and an excessive sense of worry made their immune system weak. Below are various government policies to prevent the spread of the corona virus. Social distancing policy, physical distancing policy, personal protective equipment (mask) policy, personal hygiene policy (hand washing), work/study from home policy, postponing policy all activities that gather large crowds, the Large-Scale Social Restriction Policy (PSBB); until the last, New Normal policy implementation[1]. Health protocols that are enforced in almost every place are rules that need to be obeyed by the community so that people can carry out their activities safely. Adhering to this protocol does not completely protect the public from the Corona virus but can prevent exposure to the virus itself. The high number of transmission of Covid-19 cases in Indonesia is still caused by various problems. These problems are especially common among the community. One of them is the lack of concern, awareness, and knowledge of the public about the threat of the Covid-19 pandemic[9].

However, people are starting to ignore health protocols. For example, roaming outside without wearing a mask, crowding around without thinking about the importance of keeping a distance, shopping at supermarkets or eating outside the home and ignoring the health protocols that have been provided on the spot and some accidentally don't wear masks for some reason. The 3M health protocol aims to the community can still carry out their activities safely, without endangering the health and safety of themselves and others. In its implementation, the implementation of the 3M health protocol has various responses from residents, many of whom comply with it with full awareness, but some do not care. It is the people's disobedience that exacerbates the situation[10]. In this study, one of the Data Mining techniques is used, namely the C4.5 classification algorithm. C4.5 algorithm. is a group of decision tree algorithms (decision tree). This algorithm has input in the form of training samples and samples. Training samples in the form of sample data that will be used to build a tree that has been tested for truth[2]. While the decision tree is the result of the C4.5 algorithm where the decision tree is useful for exploring data, finding hidden relationships between a number of candidate input variables and a target variable.[3]. The decision tree is able to make complex decisions simpler, so that decision makers will better interpret solutions to problems [8]. Previous researchers who have conducted a study using the C4.5 Algorithm with the title C4.5 Classification Analysis of the Factors Causing the Decline of Student Learning Achievement in the Pandemic Period. This study was conducted to determine the obstacles in the learning process that affect student achievement.[4].

2. Results and Discussion

2.1. Data Mining

Data mining is a process that employs one or more computer learning techniques (machine learning) to analyze and extract knowledge automatically.[5]. The C4.5 algorithm is one of the methods in data mining. In general, the steps of the C4.5 algorithm to create a decision tree is to start by selecting the attribute as the root (rootnode).
basic formula for entropy.'

$$Entropy(S) = - \sum p_i \log_2(p_i) \quad (1)$$

Where :

S : case set

n : number of partitions S

p_i : the proportion of S_i to S

then calculate the gain for object separation.

By using the following formula:

$$(S, A) = Entropy(S) - \sum |S_i|/|S| \times n \times Entropy(S_i) \quad (2)$$

Where :

S : data set

A : attribute

n : number of attribute partition A

$|S_i|$: number of cases on partition i

$|S|$: number of cases in S

Then create a branch and split the cases in branches

2.2. Decision Tree

The decision tree is the result of the Entropy and information gain calculation process, after repeated calculations until all tree attributes have classes and the calculation process can no longer be carried out[2].

2.3. Rapid Miner

RapidMiner is an open source application that can be used to perform data mining processes[6]. Rapid Miner is a science data software platform developed by the company of the same name that provides an integrated environment for machine learning, deep learning, text mining, and predictive analytics.[7].

2.4. Research data

In this study, the data used were the results of the distribution of questionnaires given to the community as many as 100.

In determining the classification of the class value on each attribute can be seen in the table below.

Table 1: Data Class Type

| Attribute | Field Name | Data Class |
|-----------|-------------|------------------------------------|
| C1 | Work | Disturbed - Not Disturbed |
| C2 | Environment | Support Not Support |
| C3 | Penalty | Firm - Less Firm |
| C4 | Concern | Care - Careless - Don't not mather |

The following is a sample of the data resulting from the questionnaire recapitulation processed in Microsoft Excel in Table 3.1.

Table 2: Questionnaire Recapitulation Data

| No | PROFESSION | ENVIRONMENT | PENALTY | CONCERN | Awareness |
|------|---------------|------------------|--------------|-----------------|-----------|
| R1 | Disturbed | Does not support | Lack of firm | Does not matter | Decrease |
| R2 | Disturbed | Does not support | Firm | Careless | Decrease |
| R3 | Not disturbed | Support | Firm | Care | Increase |
| R4 | Disturbed | Does not support | Lack of firm | Careless | Decrease |
| R5 | Not disturbed | Support | Firm | Careless | Decrease |
| R6 | Disturbed | Does not support | Lack of firm | Careless | Decrease |
| R7 | Disturbed | Does not support | Lack of firm | Does not matter | Decrease |
| R8 | Disturbed | Does not support | Lack of firm | Careless | Decrease |
| R9 | Disturbed | Support | Firm | Careless | Decrease |
| R10 | Disturbed | Support | Lack of firm | Does not matter | Decrease |
| R11 | Disturbed | Does not support | Lack of firm | Careless | Decrease |
| R12 | Disturbed | Support | Firm | Careless | Decrease |
| R13 | Disturbed | Support | Firm | Does not matter | Decrease |
| R14 | Disturbed | Does not support | Firm | Care | Decrease |
| R15 | Disturbed | Does not support | Lack of firm | Careless | Decrease |
| R16 | Not disturbed | Does not support | Lack of firm | Careless | Decrease |
| R17 | Disturbed | Does not support | Lack of firm | Does not matter | Decrease |
| R18 | Disturbed | Does not support | Lack of firm | Careless | Decrease |
| R19 | Disturbed | Does not support | Lack of firm | Does not matter | Decrease |
| R20 | Disturbed | Support | Lack of firm | Careless | Decrease |
| ... | ... | ... | ... | ... | ... |
| ... | ... | ... | ... | ... | ... |
| ... | ... | ... | ... | ... | ... |
| R100 | Disturbed | Does not support | Lack of firm | Careless | Decrease |

2.5. C4.5 Algorithm Calculation

Calculating the total entropy:

$$\begin{aligned} \text{Entropy [Total]} &= \left(-\frac{87}{100} \times \log_2 \left(\frac{87}{100} \right) \right) + \left(-\frac{13}{100} \times \log_2 \left(\frac{13}{100} \right) \right) \\ &= 0.557438185 \end{aligned}$$

Calculating entropy and gain (C1) :

$$\begin{aligned} \text{Entropy [Disturbed]} &= \left(-\frac{69}{72} \times \log_2 \left(\frac{69}{72} \right) \right) + \left(-\frac{3}{72} \times \log_2 \left(\frac{3}{72} \right) \right) \\ &= 0.249882293 \end{aligned}$$

$$\begin{aligned} \text{Entropy [Not disturbed]} &= \left(-\frac{18}{28} \times \log_2 \left(\frac{18}{28} \right) \right) + \left(-\frac{10}{28} \times \log_2 \left(\frac{10}{28} \right) \right) \\ &= 0.940285959 \end{aligned}$$

$$\begin{aligned} \text{Gain} &= 0.557438185 - \left(\left(\frac{72}{100} \times 0.249882293 \right) + \left(\frac{28}{100} \times 0.940285959 \right) \right) \\ &= 0.114242866 \end{aligned}$$

Calculating entropy and gain (C2) :

$$\begin{aligned} \text{Entropy [Support]} &= \left(-\frac{14}{25} \times \log_2 \left(\frac{14}{25} \right) \right) + \left(-\frac{11}{25} \times \log_2 \left(\frac{11}{25} \right) \right) \\ &= 0.989587521 \end{aligned}$$

$$\begin{aligned} \text{Entropy [No Support]} &= \left(-\frac{73}{75} \times \log_2 \left(\frac{73}{75} \right) \right) + \left(-\frac{2}{75} \times \log_2 \left(\frac{2}{75} \right) \right) \\ &= 0.177389453 \end{aligned}$$

$$\begin{aligned} \text{Gain} &= 0.557438185 - \left(\left(\frac{25}{100} \times 0.989587521 \right) + \left(\frac{75}{100} \times 0.177389453 \right) \right) \\ &= 0.176999215 \end{aligned}$$

Calculating entropy and gain (C3) :

$$\begin{aligned} \text{Entropy [Firm]} &= \left(-\frac{13}{24} \times \log_2 \left(\frac{13}{24} \right) \right) + \left(-\frac{11}{24} \times \log_2 \left(\frac{11}{24} \right) \right) \\ &= 0.994984828 \end{aligned}$$

$$\begin{aligned} \text{Entropy [Lack of firm]} &= \left(-\frac{74}{76} \times \log_2 \left(\frac{74}{76} \right) \right) + \left(-\frac{2}{76} \times \log_2 \left(\frac{2}{76} \right) \right) \\ &= 0.175565026 \end{aligned}$$

$$\begin{aligned} \text{Gain} &= 0.557438185 - \left(\left(\frac{24}{100} \times 0.994984828 \right) + \left(\frac{76}{100} \times 0.175565026 \right) \right) \\ &= 0.185212407 \end{aligned}$$

Calculating entropy and gain (C4) :

$$\begin{aligned} \text{Entropy [Care]} &= \left(-\frac{7}{20} \times \log_2 \left(\frac{7}{20} \right) \right) + \left(-\frac{13}{20} \times \log_2 \left(\frac{13}{20} \right) \right) \\ &= 0.934068055 \end{aligned}$$

$$\begin{aligned} \text{Entropy [Careless]} &= \left(-\frac{64}{64} \times \log_2 \left(\frac{64}{64} \right) \right) + \left(-\frac{0}{64} \times \log_2 \left(\frac{0}{64} \right) \right) \\ &= 0 \end{aligned}$$

$$\begin{aligned} \text{Entropy [Does not matter]} &= \left(-\frac{16}{16} \times \log_2 \left(\frac{16}{16} \right) \right) + \left(-\frac{0}{16} \times \log_2 \left(\frac{0}{16} \right) \right) \\ &= 0 \end{aligned}$$

$$\begin{aligned} \text{Gain} &= 0.557438185 - \left(\left(\frac{20}{100} \times 0.934068055 \right) + \left(\frac{64}{100} \times 0 \right) + \left(\frac{16}{100} \times 0 \right) \right) \\ &= 0.3706246 \end{aligned}$$

Attribute C4 gets the highest gain so that it is used as a node. Because the Peduli attribute class has not yet obtained the results, further calculations must be carried out. Meanwhile, the Less Caring and Not Caring attribute classes have obtained the same result, namely Decreasing. The following are the results of the calculation of the entropy and gain values described in Table 3.

Table 3: Node Calculation Results 1.

| Node 1 | | Amount | Decrease | Increase | Entropy | Gain |
|-----------|------------------|--------|----------|----------|-------------|------------------|
| | | 100 | 87 | 13 | 0.557438185 | |
| C1 | | | | | | 0.114242866 |
| | Disturbed | 72 | 69 | 3 | 0.249882293 | |
| | Not disturbed | 28 | 18 | 10 | 0.940285959 | |
| C2 | | | | | | 0.176999215 |
| | Support | 25 | 14 | 11 | 0.989587521 | |
| | Does not support | 75 | 73 | 2 | 0.177389453 | |
| C3 | | | | | | 0.185212407 |
| | Firm | 24 | 13 | 11 | 0.994984828 | |
| | Lack of firm | 76 | 74 | 2 | 0.175565026 | |
| C4 | | | | | | 0.3706246 |
| | Care | 20 | 7 | 13 | 0.934068055 | |
| | Careless | 64 | 64 | 0 | 0 | |
| | Does not matter | 16 | 16 | 0 | 0 | |

Table 4: Node Calculation Results 1.1

| Node 1.1 | | Amount | Decrease | Increase | Entropy | Gain |
|-----------|------------------|--------|----------|----------|-------------|-----------------|
| Care | | 20 | 7 | 13 | 0.934068055 | |
| C1 | | | | | | 0.162281 |
| | Disturbed | 8 | 5 | 3 | 0.954434003 | |
| | Not disturbed | 12 | 2 | 10 | 0.650022422 | |
| C2 | | | | | | 0.590176 |
| | Support | 11 | 0 | 11 | 0 | |
| | Does not support | 9 | 7 | 2 | 0.764204507 | |
| C3 | | | | | | 0.0172375 |
| | Firm | 16 | 5 | 11 | 0.896038233 | |
| | Lack of firm | 4 | 2 | 2 | 1 | |

Table 5: Node Calculation Results 1.1.1

| Node 1.1.1 | | Amount | Decrease | Increase | Entropy | Gain |
|--------------------|---------------|--------|----------|----------|-------------|------------------|
| Caring-Not Support | | 9 | 7 | 2 | 0.764204507 | |
| C1 | | | | | | 0.3197601 |
| | Disturbed | 5 | 5 | 0 | 0 | |
| | Not disturbed | 4 | 2 | 2 | 1 | |
| C3 | | | | | | 0.0928885 |
| | Firm | 7 | 5 | 2 | 0.863120569 | |
| | Lack of firm | 2 | 2 | 0 | 0 | |

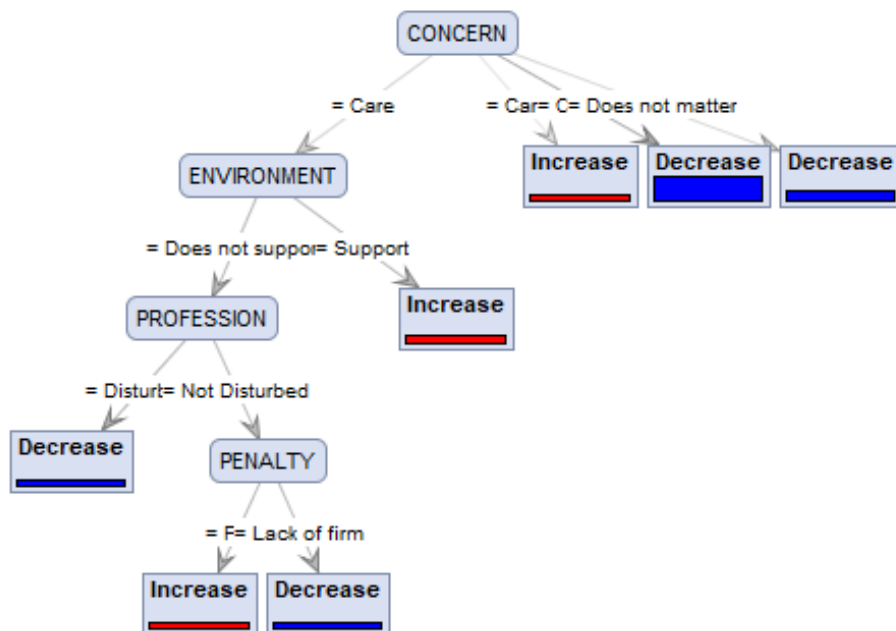
The following are the results of the calculations on the last node that obtained the decision to Increase in the Assertive attribute class and the decision to Decrease in the Less Assertive attribute class. Thus Node 1.1.1.1. is the last node formed.

Table 6: Node Calculation Results 1.1.1.1

| Nodes 1.1.1.1 | | Amount | Decrease | Increase | Entropy | Gain |
|-------------------------------------|--------------|--------|----------|----------|---------|------|
| Caring-Not Supporting-Not Disturbed | | 4 | 2 | 2 | 1 | |
| C3 | | | | | | 1 |
| | Firm | 2 | 0 | 2 | 0 | |
| | Lack of firm | 2 | 2 | 0 | 0 | |

2.6. Rapid Miner Test Results

The results of data processing with a decision tree model according to the Rapid Miner software, can be seen in Figure 3.1. as follows :

**Figure 1:** Result of decision tree

Furthermore, the final decision tree pattern is obtained after calculating and testing data on each attribute with the C4.5 Algorithm as follows:

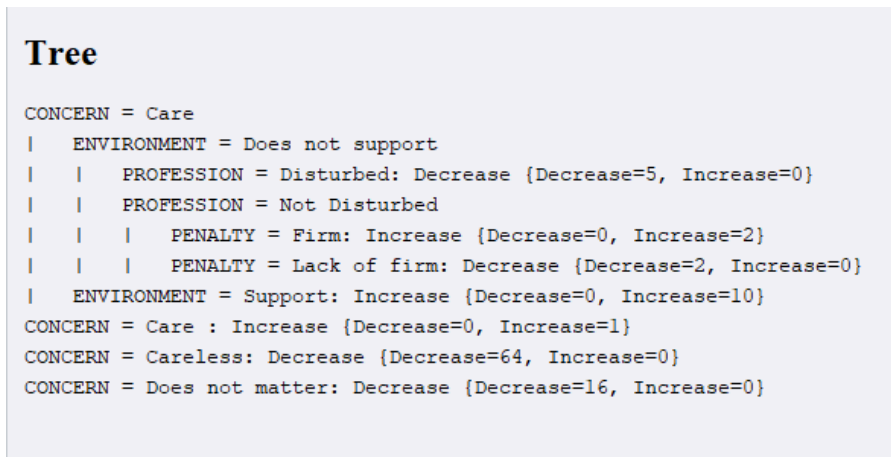


Figure 2: Description of decision tree

Based on data processing using RapidMiner software, the system accuracy value is 100%. Where the model that has been formed is tested for accuracy by using test data derived from training data with split validation on the rapidminer 5.3 application.

PerformanceVector

PerformanceVector:

accuracy: 100.00%

ConfusionMatrix:

| | | |
|-----------|----------|----------|
| True: | Decrease | Increase |
| Decrease: | 26 | 0 |
| Increase: | 0 | 4 |

Figure 3: Value of Performance Vector Algorithm C4.5

3. Conclusion

Based on all the results of the research stages that have been carried out on the application of the C4.5 Algorithm in the analysis of the decline in public awareness in the application of the COVID-19 health protocol, it can be concluded that the problem of determining the dominant factor of decreasing public awareness can be solved using data mining techniques, namely the C4.5 Algorithm. Produce 6 rules and the level of accuracy generated by this method is 100% and from calculations with the C4.5 algorithm, the most dominant factor is Concern (C4) with a gain value of 0.3706246.

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