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Implementation of ID3 algorithm classification using webbased weka

A S Fitrani^{*}, M A Rosid, Y Findawati, Y Rahmawati, A K Anam

Program Studi Informatika, Universitas Muhammadiyah Sidoarjo, Jl. Raya Gelam 250 Candi Sidoarjo, Jawa Timur, Indonesia

asfjim@umsida.ac.id

Abstract. The Bangil District Court is an IB class court that handles a large number of case cases. Every year more and more case cases are included in the Bangil District Court, but not all case cases are in a mutation status. By using classification techniques that can process large amounts of data to find patterns that occur in case data. Data processing is used to predict case minutation with the decision tree method using ID3 algorithm. Case data has 8 attributes and has been classified into 6 parts, namely division based on Case Type, Register, Case Classification, Length of Process, Public Prosecutor and Decision with a goal of Mutation Status. Weka 3.6 is an API that is used to build rules / rule bases. The rule that was formed was then implemented in the making of a case status prediction application in the web-based Bangil District Court.

1. Introduction

The Bangil District Court which is an IB class court that along with the increasing number of case cases in the Bangil District Court, the number of incoming case data automatically increases, but not all case cases are in a mutation status. This case data will be processed using data mining techniques that will generate new knowledge and information from patterns or models formed on large data.

By using classification techniques that can process large amounts of data[1][2][3] where it will be explained about the model produced by the data that has a data class on case data in Pengadialan Negeri Bangil. Data processing is used to predict case minutation with the decision tree method using ID3 algorithm.

Case data taken is data from the website of the Case Search Information System (SIPP) in the Bangil District Court. Data will be processed using the iterative classification method dichotomiser 3 (Id3) to find out predictions of case mutations from 2011 to 2015 in the Bangil District Court.

The impact of this application is that the IT Staff of the Bangil District Court is easier and more efficient in terms of completing the case data prediction process to find out the status of the mutation with ID3 algorithm.

2. Theoretical foundation

2.1 Definition of Case Search Information System

The Case Search Information System (SIPP) is an official web-based application made by the Supreme Court of the Republic of Indonesia (MARI) to support and facilitate justice seekers in monitoring the process of handling criminal cases[4].

This application is intended for administration and tracking of first-degree court case data. Case Phase, Case Status, Case Costs, Session Schedule, Case Statistics, and History, including information services contained in this application and complete information, has been recorded.

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All information can be accessed by the general public for free and realtime. This SIPP application allows access to case files. The community can directly access and search for the subject matter [5]

2.2 Definition of Data Mining

Data Mining is a process of collecting data that is processed with various methods. Another term for data mining is knowledge-discovery in a database (KDD). The purpose of data mining is to utilize data and process it to get new and useful information.

2.3 Group Data Mining

There are several types of data mining groupings, according to [6]:

- 1. Description, to describe the pattern in data.
- 2. Estimates, the goal is more numerical than category, and this estimate is almost similar to classification.
- 3. Prediction, the result of a prediction is to show something that hasn't happened yet
- 4. Classification, the purpose of this classification is categorical
- 5. Clustering, clustering is more towards grouping records or cases that have similar data
- 6. Association, to identify the relationship between events that occur at one time.

2.4 Stages of Data Mining

The data mining stage is interactive, and the stages of data mining are divided into several stages including. [6]



Figure 1. Stages of Data Mining

- 1. Cleaning (Cleaning)
- 2. Is the process of removing/cleaning inconsistent data.
- 3. Integration (integration)
- 4. Is the merging of data from various databases into one new database
- 5. Selection
- 6. Data selection (data selection), not all data contained in the database is used; the data taken is only suitable for analysis.
- 7. Transformation, data will be converted into a format that is compatible with the data mining process.
- 8. Data mining process, is the most important process in the method used to find knowledge.
- 9. Evaluate patterns (pattern evaluation), Identify a unique pattern and draw it into the knowledge base

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10. Presentation of knowledge (knowledge presentation), is a presentation of knowledge from the method used to obtain knowledge obtained by the user.

2.5 Decision Tree Method

Decision Tree is an approach that is widely used to solve clarification problems, and this method is used to estimate discrete values and target functions. The Decision Tree is a set of IF-THEN rules that each line or path is connected to the rule. In the decision, the tree can be used to map decision [7].

2.5.1 ID3 algorithm

ID3 algorithm is one method in the decision tree model, where estimating the discrete value of a function from a decision tree produces an if-then rule so that it is easier to understand. The following is an example of the Iterative Dichonomiser 3 or ID3 algorithm system [8][9].

2.5.2 Entropy

Entropy is a measure of data that has a measure of information theory that can determine the characteristics of data impurity and feature equations of members of an attribute[1][8]. To calculate the entropy value using the formula:

Entropy (S) = $\sum_{i=1}^{n} -pi * \log_2 pi$

(1)

(2)

Information: S: a set of cases n: number of S partitions Pi: the proportion of Si to S

2.5.3 Gain

After the entropy value is known, the next step is to find the gain value of an attribute in classifying the data[1][10].

To calculate the Gain value using the formula:

 $Gain(S,A) = Entropy (S) - \sum_{i=1}^{n} * \frac{|S_i|}{s} * Entropy (Si)$

Information: S: a set of cases A: Attribute n: number of partition attributes A | Si |: number of cases on a partition to i | S |: number of cases in S

3. System Design

The design of the application of data mining prediction on minutation cases can be divided into the following stages, namely:

- 1. Classification Diagram
- 2. Data Attributes
- 3. Flowchart
- 4. Designing the User Interface

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3.1 Classification Diagram



Figure 2. Classification Diagram

3.2 Flowchart

The following is a description of the flowchart classification method with algorithm ID3 to predict case status minutation, namely:



Figure 3. Flowchart Data Training

The description of figure 3 is:

- 1. Enter training data
- 2. Calculate the entropy and gain values
- 3. Create a root node (root node) based on the highest Gain value.
- 4. Calculate the Entropy value based on the attribute on the selected Gain.
- 5. Calculate all entropy values until you have one answer or leaf node if you have not found one answer, it needs to be recalculated by choosing the largest gain value remaining. After all, are counted, create a branch node (internal node) based on the results of the calculation of entropy.
- 6. After Gain and Entropy have calculated all, make a decision tree and Rule from the decision tree that has been made.

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Figure 4. Flowchart Data Testing

The information from figure 4 is:

- a. Input testing data
- b. The testing process is based on the decision tree from the training data
- c. Get the results of testing data

4. Discussion

4.1 Input

Data obtained from the Bangil District Court totaled 998 case data. The data will be tested using the Decision Tree method with algorithm ID3 using WEKA tools and website[2]. Data collection was 65% of 649 training data and 35% of 349 testing data. Following are some examples of case data, namely:

| Table1. Dataset | | | | | | | | |
|-----------------|------------------------------|---------------|----------|---------------------|--------------|---------------|------------|-----------------|
| No | Terdakwa/Pemohon | Jenis Perkara | Register | Klasifikasi Perkara | Lama Proses | Penuntut Umum | Putusan | Status Minutasi |
| 1 | IRA SUGIARTI | Pdt.G | Α | Ganti Rugi | Sangat Lama | TD | Dikabulkan | Belum |
| 2 | PT. BPR PANDAAN ARTA JAYA | Pdt.G | Α | Wanprestasi | Sangat Lama | TD | Dikabulkan | Belum |
| 3 | WARTI DAN PINI | Pdt.G | Α | Wanprestasi | Sangat Lama | TD | Dicabut | Sudah |
| 4 | M. SYAIFUDDIN NUR | Pid.B | Α | Pembunuhan | Lama | TD | Penjara | Belum |
| 5 | YOYOK SAPTOADI JOKO SUPRIOKO | Pid.B | Α | Penggelapan | Cepat | TD | Penjara | Belum |
| 6 | DAIM | Pid.B | Α | Narkotika | Lama | TD | Penjara | Belum |
| 7 | FATHUR ROHMAN | Pid.B | Α | Penggelapan | Lama | TD | Penjara | Belum |
| 8 | RAKHMAT HIDAYAT | Pid.B | Α | Narkotika | Lama | TD | Penjara | Belum |
| 9 | H. NUR HASAN | Pid.B | Α | Pencurian | Lama | TD | Penjara | Belum |
| 10 | INUNG | Pid.B | Α | Narkotika | Sangat Lama | TD | Penjara | Belum |
| 11 | IDA BAGUS KOKO | Pid.B | Α | Lain-Lain | Sangat Lama | TD | Bebas | Belum |
| 12 | IRA SUGIARTI | Pdt.G | А | Ganti Rugi | Sangat Lama | TD | TDD | Belum |
| 13 | Disamarkan | Pdt.G | В | Penceraian | Cepat | D | Dikabulkan | Sudah |
| 14 | Disamarkan | Pdt.G | В | Penceraian | Sangat Cepat | D | Dikabulkan | Sudah |
| 15 | Disamarkan | Pdt.G | В | Penceraian | Cepat | D | Dikabulkan | Sudah |
| 16 | PT. BPR GUNUNG ADI DANA | Pdt.G | В | Lain-Lain | Sangat Lama | TD | Dikabulkan | Belum |
| 17 | PT. CIPTA MAKARYA INDAH | Pdt.G | В | Lain-Lain | Lama | TD | Dicabut | Sudah |
| 18 | GINTEN | Pdt.G | В | Lain-Lain | Sangat Lama | TD | Dikabulkan | Behum |
| 19 | Disamarkan | Pdt.G | В | Penceraian | Sangat Cepat | D | Dikabulkan | Sudah |
| 20 | LILA HERDIANTI | Pdt.G | В | Ganti Rugi | Cepat | TD | Dikabulkan | Sudah |

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| | |

| | | 14 | 010 2. | Training Data | | | | |
|----|------------------------------|---------------|----------|---------------------|--------------|---------------|------------|-----------------|
| No | Terdakwa/Pemohon | Jenis Perkara | Register | Klasifikasi Perkara | Lama Proses | Penuntut Umum | Putusan | Status Minutasi |
| 1 | IRA SUGIARTI | Pdt.G | Α | Ganti Rugi | Sangat Lama | TD | Dikabulkan | Belum |
| 2 | PT. BPR PANDAAN ARTA JAYA | Pdt.G | Α | Wanprestasi | Sangat Lama | TD | Dikabulkan | Belum |
| 3 | YOYOK SAPTOADI JOKO SUPRIOKO | Pid.B | Α | Penggelapan | Cepat | TD | Penjara | Belum |
| 4 | DAIM | Pid.B | Α | Narkotika | Lama | TD | Penjara | Belum |
| 5 | FATHUR ROHMAN | Pid.B | Α | Penggelapan | Lama | TD | Penjara | Belum |
| 6 | RAKHMAT HIDAYAT | Pid.B | Α | Narkotika | Lama | TD | Penjara | Belum |
| 7 | H. NUR HASAN | Pid.B | А | Pencurian | Lama | TD | Penjara | Belum |
| 8 | INUNG | Pid.B | Α | Narkotika | Sangat Lama | TD | Penjara | Belum |
| 9 | IDA BAGUS KOKO | Pid.B | Α | Lain-Lain | Sangat Lama | TD | Bebas | Belum |
| 10 | IRA SUGIARTI | Pdt.G | А | Ganti Rugi | Sangat Lama | TD | TDD | Belum |
| 11 | Disamarkan | Pdt.G | В | Penceraian | Cepat | D | Dikabulkan | Sudah |
| 12 | Disamarkan | Pdt.G | В | Penceraian | Sangat Cepat | D | Dikabulkan | Sudah |
| 13 | Disamarkan | Pdt.G | В | Penceraian | Cepat | D | Dikabulkan | Sudah |
| 14 | PT. BPR GUNUNG ADI DANA | Pdt.G | В | Lain-Lain | Sangat Lama | TD | Dikabulkan | Belum |
| 15 | PT. CIPTA MAKARYA INDAH | Pdt.G | В | Lain-Lain | Lama | TD | Dicabut | Sudah |
| 16 | GINTEN | Pdt.P | В | Lain-Lain | Sangat Lama | TD | Dikabulkan | Belum |
| 17 | Disamarkan | Pdt.G | В | Penceraian | Sangat Cepat | D | Dikabulkan | Sudah |
| 18 | LILA HERDIANTI | Pdt.G | В | Ganti Rugi | Cepat | TD | Dikabulkan | Sudah |
| 19 | LILA HERDIANTI | Pdt.G | В | Ganti Rugi | Lama | TD | Dikabulkan | Sudah |
| 20 | NINDIA AVISTA CITRAYANTI | Pdt.P | В | PAKT | Sangat Cepat | TAO | Dikabulkan | Sudah |

Table 2. Training Data

Table 3. Testing Data

| No | Terdakwa/Pemohon | Jenis Perkara | Register | Klasifikasi Perkara | Lama Proses | Penuntut Umum | Putusan | Status Minutasi |
|----|-------------------------------|---------------|----------|---------------------|--------------|---------------|------------|-----------------|
| 1 | WARTI DAN PINI | Pdt.G | Α | Wanprestasi | Sangat Lama | TD | Dicabut | Sudah |
| 2 | M. SYAIFUDDIN NUR | Pid.B | Α | Pembunuhan | Lama | TD | Penjara | Belum |
| 3 | BAMBANG APRIYANTO | Pid.B | В | Penggelapan | Cepat | TD | Penjara | Sudah |
| 4 | KHOLIK | Pdt.G | B | Lain-Lain | Lama | TD | Dikabulkan | Sudah |
| 5 | SUNARKO Als REJES Bin DASIRUN | Pid.B | B | Penggelapan | Cepat | TD | Penjara | Sudah |
| 6 | H.HILMI | Pdt.P | B | PAKT | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 7 | DIANTO | Pdt.P | B | PAKT | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 8 | NURJANAH | Pdt.P | B | WDIJ | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 9 | MAWARDI | Pdt.P | B | WDIJ | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 10 | Disamarkan | Pdt.G | B | Perceraian | Sangat Cepat | D | Dikabulkan | Sudah |
| 11 | TRI SETYO JANUARIFIN | Pdt.P | B | PAKT | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 12 | SRI RAHAYU YULIARTI | Pid.B | B | Pemalsuan | Cepat | TD | Penjara | Belum |
| 13 | Terdakwa | Pid.B | B | Lain-Lain | Cepat | TD | Penjara | Sudah |
| 14 | NURIYANTO | Pdt.P | B | PAKT | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 15 | SAHUR BIN MARSAN | Pid.B | B | Penganiayaan | Sangat Cepat | TD | Penjara | Sudah |
| 16 | RUHANAH | Pdt.P | B | PAKT | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 17 | YAYUK SUPRAPTI BROWNE | Pdt.P | В | PAKT | Cepat | TAO | Dikabulkan | Sudah |
| 18 | LOLOK WIJAYANTI | Pdt.G | В | Lain-Lain | Sangat Lama | TD | Dikabulkan | Sudah |
| 19 | MITA FITRIA | Pdt.P | В | PAKT | Sangat Cepat | TAO | Dikabulkan | Sudah |
| 20 | USMANTO BIN NGADIARTO | Pid.B | В | Pencurian | Cepat | TD | Penjara | Sudah |

4.1.1 Data

The dataset used for the calculation process is 998 case data. The dataset is divided into two parts, namely training data and testing data. Training data is used to process a calculation by forming a classifier model. While data testing is used to measure the extent to which the classifier successfully classifies correctly.

4.1.2 Pre Processing Data

The 1,000 datasets, which were the initial data of this study experienced pre-processing to 998 data. Where the pre-processing techniques in preparing data are discarding duplicate data, discarding data inconsistencies, and correcting data errors.

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| Journal of Physics: Conference Series | 1381 (2019) 012036 | doi:10.1088/1742-6596/1381/1/012036 |

4.2 Proses

The process used at this stage is to find out the results of the ID3 classification calculation on the Website.

4.2.1 Website

1. After that, take the dataset with the extension .csv / .arff

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Figure 5. Display of Dataset Files

2. Select the classify method, select the ID3 algorithm and then select the dataset and Testing data and then select submit

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Figure 6. Classify Display

3. Wait for a while until the classification results appear

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Figure 7. Display of ID3 Classification Results

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Figure 8. Display of Testing Results

4. Display prediction menu.

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Figure 9. Display Prediction

5. Wait for a while until the prediction results appear

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| Algoritma Id3 | | | | |
| Heal Tealing Predictions on test data | | | | |
| inst# actual predicted error p 1 2:5udah 2:5udah 1 | refliction | | | |

Figure 10. Display of Prediction Results

4.3 Output

4.3.1 Classification Results

1. Predicted results on testing with a total testing data of 35% of a total dataset of 998.

| 1 | realetions | on test data | |
|-------|------------|---------------|----------------|
| inst# | actual p | predicted err | ror prediction |
| 1 | 2:Sudah | 1:Belum | + 1 |
| 2 | 1:Belum | ?? | • |
| 3 | 2:Sudah | ?? | |
| 4 | 2:Sudah | 2:Sudah | 1 |
| 5 | 2:Sudah | ?? | |
| 6 | 2:Sudah | 2:Sudah | 1 |
| 7 | 2:Sudah | 2:Sudah | 1 |
| 8 | 2:Sudah | 2:Sudah | 1 |
| 9 | 2:Sudah | 2:Sudah | 1 |
| 10 | 2:Sudah | 2:Sudah | 1 |
| | | | |
| wit | h sino | le data | a testing |

2. Prediction results on testing with single data

actual predicted error prediction 1:Belum 2:Sudah + 1 inst#

4.3.2 Percentage Results

Percentage of processes to determine accuracy. The following are the results of the percentage test accuracy Table 4 :

| Table 4. Percentage Results | | | | | | | | |
|-----------------------------|--------|---------|-------|--------|--|--|--|--|
| Data | Result | | | | | | | |
| Testing | True | % | False | % | | | | |
| 988 | 326 | 93.41 % | 23 | 6.59 % | | | | |

5. Conclusion

Based on the results of the study, it can be concluded that the algorithm ID3 can be considered to be used in the case data prediction process to determine the mutation status. This is supported by the results of testing the system using algorithm ID3 in the Bangil District Court, which has an average accuracy of 93.41% and has an error rate of 6.59%.

Suggestions that can be given for this research are expected to be able to use other methods. So, can know the comparison between the percentage results in the calculation process according to the method used.

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