

Literature Review: Data Mining For Student Data Classification

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Abstract

The abundance of student data and student graduation number data, hidden information can be found by processing student data to be useful to the university. The processing of student data needs to be done to uncover important information in the form of new knowledge (knowledge discovery) such as information on student data classification based on profile and academic data. Therefore, in this research, the researcher plans to conduct a literature review related to data mining for student data classification with the aim of finding out about data mining data processing classification and collecting all designs used in identifying data starting from problems, methodology, equations and results. For this research, researchers used historical data from students from 2007 to 2011 who had graduated. There were 9 research journals that researchers managed to find, each of which used different algorithms or classification techniques. To conduct a literature review, researchers conducted a journal review using PICOT. The results of this research are the success of researchers in classifying student data using data mining techniques.

Keywords: Classification, Students, Data Processing.

1. INTRODUCTION

Higher education today must gain a competitive advantage using every available resource. In addition to infrastructure and human resources, information systems are one of the resources that can be used to collect, process and disseminate information to support day-to-day operations as well as outsourcing activities. strategic decisions. Education has a very important role to improve and prepare superior and highly competitive human resources (HR). This is where the role of higher education institutions becomes very important in creating experts who are able to develop knowledge and contribute to development. Higher education as one of the educational institutions, is required to be able to provide quality and quality education to its stakeholders[1]. The abundance of student data and student graduation count data, hidden information can be found by processing student data to be useful to higher education institutions. The processing of student data should be performed to uncover important information in the form of new knowledge (knowledge discovery), for example information on classification of student data by profile and academic data. This new knowledge could help universities rank student graduation rates to determine strategies to increase graduation rates in later years. Data mining is a series of processes that aim to manually extract added value in the form of unknown information from a database by extracting patterns from data with the aim of manipulating the data to obtain More valuable insights by extracting and recognizing important patterns. or dig into the data contained in the database [2].

In relation of Data Mining, there are several previous research that have become references for researchers. The first is research from Cornelia Selvi Diana, Latifah Hanum and Saut Parsaoran Tamba, who in this research implemented data mining using the K-means algorithm to determine the title of the thesis and research journal by making FTIK UNPRI the research object. The results of this application itself certainly provide convenience or solutions for students and their scope to find out ideas for thesis titles and research journals. The second is research conducted by Ni Luh Putu Purnama Dewi, I Nyoman

Purnama and Nengah Widya Utami. In this research, data mining was applied to cluster lecturer performance assessments using the K-means algorithm at STMIK Primakara. The results obtained from this research are research on lecturer performance based on student satisfaction, namely very good cluster 312 (31.74%) student data, good cluster 401 (40.79% student data, quite good cluster 189 (19.23%) student data and the less good cluster was 81 (8.24%). The DBI accuracy level was 0.270 or 27%. The last research was conducted by Hozairi, Anwari and Syarifu Alim, in this research implementing orange data mininh to classify student graduation using a model. K-Nearest Neighbor, Decision Treee and Naive Bayes in the Informatics Engineering Study Program, Madura Islamic University, class of 2016. In this study, a comparison of classification models was carried out with the results of K-Nearest Neighbor having an accuracy level of 77%, Decision Tree with an accuracy level of 74%. % and Naive Bayes is 89%, thus the most recommended model is Naive Bayes.

From these three research, it can certainly be seen that there are many classification models or classification algorithms for data mining. So in this research, researchers will conduct a literature review related to data mining for student data classification with the aim of finding out about data mining classification for data processing and collecting all designs used in identifying data starting from problems, methodology, equations and results. For this research, researchers used historical data from students from 2007 to 2011 who had graduated. Regarding the classification algorithm consists of 6 parts consisting of: Decision Tree Analysis, is a technique that belongs to the machine-learning family, arguably the most popular classification technique in the data mining area. Statistical Analysis [3] Statistical engineering was the source of popular classification algorithms for many years until the emergence of machine learning techniques. Statistical classification techniques include logistic regression and discriminant analysis, both of which assume that the relationship between input and output variables is essentially linear, that the data are normally distributed, and that the data are normally distributed. The variables are neither interdependent nor independent of each other. The nature of these questionable assumptions eventually led to the shift towards machine learning techniques. Neural Networks is one of the most popular techniques in Machine-Learning that can be used for classification problems. Case-Based Reasoning, this approach uses historical cases to recognize similarities to determine a new case into the most probable category. Bayesian Classifiers, this approach uses probability theory to create classification models based on past events that can place a new instance into the most probable class. Genetic Algorithms, using analogies to natural evolution to create a purposeful search-based mechanism for classifying data samples.

2. RESEARCH METHODOLOGY

In this research, the researcher conducted a literature review, in which case the researcher reviewed journals that matched the PICOT and search terms for journals through MESH (Medical Subject Heading), limitations taken by journals and other things. The journal is used in literature reviews obtained through databases that provide Indonesian scientific journals through Google Scholar and websites such as Garuda Kemdikbud.

The researcher wrote down the keywords according to the MESH (Medical Subject Heading) namely "processing", "data", "students" and selected the full text to appear 100 findings, then narrowed down to Dissertations and Theses and found the next 9 findings sorted from the most recent. Regarding the choice of language, it was not carried out because all the journals found used Indonesian. Each of these questions has followed the PICOT where in each of these questions there is P = Problem/Student/Population, I/E= Implementation /Intervention/Exposure, C=Control/Comparative Intervention, O=Results and T=Time. It is relevant that the author used to get journals about the classification of Data Mining processing student data. The author takes all the designs in the research that are used in identifying student data

3. **RESULTS AND DISCUSSION**

In this research, using history data from students from 2007 to 2011 who have graduated with a total of 377 data with 72 attribute values of the course and 1 target class in the form of study period. This research was conducted by following the stages of data mining work which refers to the process of knowledge discovery in databases. The data mining application was successfully built with experimental results showing that the best study period classification results were obtained by selecting attributes

from all elective courses with accuracy values. Based on data obtained from the Department of Computer Science/Informatics, Diponegoro University, students from the 2007 to 2011 class year with a total of 377 graduates obtained information that the average student study period is still over 4 years. Inna Alvi Nikmatum (2019) and Indra are alert. There are several data mining tools including Rapid Miner, Orange, KNIME, Weka, Keel and R. WEKA is GUI-based so that it minimizes the use of coding which can make it easier for system users. In this study using WEKA tools. The data processed at WEKA has ARFF and CSV formats. Based on the experimental results, it was concluded that using three model criteria, namely the Ratio Gain, Information Gain and Gini Index. The highest accuracy results are found in the Gini Index criterion model, namely 92.18%. The highest result of the three feature selections is the Information Ratio Gain with a value of p = 0.6 and the accuracy results are 92.46. Feature selection is the process of selecting the right features to be used in the classification or clustering process. The independent variable that gives the greatest t value is taken as X1 provided that H0 is rejected. The two taken from the independent variable are taken as X2 provided that H0 is rejected. Wiwit Suprivanti (2018), and Miss Puspitasari in the number of questionnaire results that were successfully collected were 981 samples. Of the 981 data collected, there are 948 valid data that can be used and 33 data that are invalid due to an unbalanced dataset, so the dataset is adjusted randomly to 360 data with an ouas value and 350 data with a dissatisfied value. The validity and readability tests were carried out on 30 randomly selected data samples, meaning that each statement given by the respondent was correlated with the total score and all were declared valid ...

From the results of the implementation of the confusion matrix calculation, the accuracy output of the classification model for the three algorithms is obtained as shown in Table 1. The three algorithms, decision tree C4.5, SVM and Naïve Bayes, show quite good accuracy in classifying correctly both for satisfied classes and the dissatisfied class is quite good, namely above 80%, where the accuracy of the C4.5 algorithm and the SVM algorithm is better than the accuracy provided by the Naïve Bayes algorithm. These results are consistent with previous studies which claim that each algorithm has good performance on the dataset used. The results of the questionnaire show that 61% of students answered they were satisfied with the learning facilities and infrastructure and 39% answered they were dissatisfied. C4 Decision Tree Algorithm. 98% compared to the Naïve Bayes and Support Vector Machine algorithms when modeled from training data and tested using test data from the student satisfaction questionnaire dataset for learning facilities and infrastructure Elga Mariati, Ariesta Lestari, and Widiatry (2020).

No.	Writer	Research Title	Year	Method	Results
1.	Inna Alvi	Implementation	2019	K-Nearest	Student data and student graduation data can
	Nikmatum	of Data Mining		Neighbor	produce abundant information, hidden information
	& Indra	for			can be found by managing the data. Based on data
	Waspada	Classification of			obtained from the Department of Computer
		Student Study			Science/Informatics, Diponegoro University,
		Period Using the			students from the 2007 to 2011 class year with a
		K-Nearest			total of 377 graduates obtained information that the
		Neighbor			average student study period is still over 4 years
		Algorithm			[4]. [5]
2.	Marina	Performance	2019	Decision	One of the factors affecting the quality of a higher
	Windarti &	Comparison of 6		Tree,	education institution is student learning outcomes,
	Agustinus	Data Mining		Bayesian	which can be measured over time of study. The
	Suradi	Classification		Network,	acquisition of knowledge in a database is often
		Algorithms for		K-Nearest	referred to as data mining or data mining. This
		Prediction of		Neighbors,	study aims to determine the performance of the six
					classification algorithms used, which are Decision

 Table 1 : Table literature review

		Student Study		Naïve	Trees C4.5, Bayesian Networks, Nearest
		Period		Bayes,	Neighbors, Naïve Bayes, Neural Networks, and
				Neural	SVMs. The acquisition of knowledge in a database
				Network,	is often referred to as data mining. Decision Tree
				Svm	C4. Bayesian Network has the best performance
					with an accuracy value of 80.615%, precision and
					recall values of 0.785 and 0.806, for AUC values
					included in the good category, namely 0.837 [5].
3.	Wiwit	Implementation 20	018	K-Nearest	The experimental results concluded that using
	Supriyanti	of Forward		Neighbor	three criteria models namely gain ratio,
	& Nona	Selection			information gain and Gini index, the highest
	Puspitasari	Feature			accuracy results were found in the Gini index
		Selection			criterion model, namely 92.18%. The highest result
		Techniques in			of the three feature selections is the information
		the Minig Data			gain ratio with a value of p=0.6 and the accuracy
		Classification			results are 92.46. Feature selection is the process of
		Algorithm for			selecting the right features to be used in the
		Predicting Study			classification or clustering process. The
		Period of			independent variable that gives the greatest t value
		Indonusa			is taken as X1 provided that H0 is rejected. The two
		Surakarta			taken from the independent variable are taken as
		Polytechnic			X2 provided that H0 is rejected [6]
		Students			
		Students			
4.	Elga	Engineering 20	020	Decision	Of the 16 attributes used, only 8 attributes affect
4.	Elga Mariati,	Engineering 20 Student	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they
4.	Elga Mariati, Ariesta	Engineering 20 Student Satisfaction	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they have a higher presentation of answers, the rest have
4.	Elga Mariati, Ariesta Lestari &	Engineering 20 Student Satisfaction Classification	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they have a higher presentation of answers, the rest have relatively small presentations. The use of the naïve
4.	Elga Mariati, Ariesta Lestari & Widiatry	Engineering 20 Student Satisfaction Classification Model for	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they have a higher presentation of answers, the rest have relatively small presentations. The use of the naïve Bayes algorithm in this study resulted in an
4.	Elga Mariati, Ariesta Lestari & Widiatry	Engineering 20 Student Satisfaction Classification Model for Learning	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they have a higher presentation of answers, the rest have relatively small presentations. The use of the naïve Bayes algorithm in this study resulted in an accuracy of 74.46% which stated that the amount
4.	Elga Mariati, Ariesta Lestari & Widiatry	Engineering 20 Student Satisfaction Classification Model for Learning Facilities Using	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they have a higher presentation of answers, the rest have relatively small presentations. The use of the naïve Bayes algorithm in this study resulted in an accuracy of 74.46% which stated that the amount of correct data was greater than the wrong data.
4.	Elga Mariati, Ariesta Lestari & Widiatry	Engineering 20 Student Satisfaction Classification Model for Learning Facilities Using Data Mining	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they have a higher presentation of answers, the rest have relatively small presentations. The use of the naïve Bayes algorithm in this study resulted in an accuracy of 74.46% which stated that the amount of correct data was greater than the wrong data. Input attributes of customer satisfaction used in
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4.	Elga Mariati, Ariesta Lestari & Widiatry	Engineering 20 Student 20 Satisfaction 1 Classification 1 Model for Learning Facilities Data Mining 1	020	Decision Tree	Of the 16 attributes used, only 8 attributes affect the level of customer satisfaction because they have a higher presentation of answers, the rest have relatively small presentations. The use of the naïve Bayes algorithm in this study resulted in an accuracy of 74.46% which stated that the amount of correct data was greater than the wrong data. Input attributes of customer satisfaction used in this study include price, facilities, service and loyalty. The implementation of the support vector machine algorithm gives a fairly good accuracy value of 80%. Each of the previous studies only implemented one algorithm, then stated that the algorithm used had given good results in terms of
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data and tested using test data from	m the student
satisfaction questionnaire dataset	for learning
facilities and infrastructure [7]	for loaning
5 Eka Sahaa Data Mining 2021 Naïva The results of implementing data mi	ning using the
5. Eka Saolia Data Mining 2021 Natve The results of hipfementing data him	
& Yuda with 2 (Two) Bayes, Rapidminer software were carried	out on two
Irawan Student classification algorithm models, nan	hely C4.5 and
Performance NBC, then entering datasets as test m	aterial for the
Prediction two models, which contained expe	rimental data
Classification and test data. By using the Asso	ocia on Rule
Models Method. in the Naive Bayes Class	ssifier (NBC)
algorithm model is 80% while	in the C4.5
algorithm model it is 60%. It can be c	concluded that
the best accuracy value from the	results of the
comparison of the two algorithm	n models is
obtained by the Naïve Bayes algorith	m model with
an accuracy value of 80% [8].	
6. Resti, Dodo Application of 2021 Naïve Bayes This research used 100 student data	for 2015 and
Zaenal Outstanding 2016 respectively, so that the total of	lata used was
Abidin, & Classification 200. These attributes will be applied	l to the Naive
Errissya Data Mining at Bayes method for students majoring	in Informatics
Rasywir Stikom Engineering class of 2015-2016 as	many as 200
Dinamika data to predict student potential. ach	ievers, which
Bangsa Jambi are categorized into 3 namely V	ery Potential,
Using the Naïve Potential, and Potential Enough. The	results of this
Bayes Method study used the Use Training Set with	th a Correctly
Classified Instances accuracy perce	ntage of 87%
and 13% Incorrectly Classified Insta	ances. 5-cross
validation Correctly with a percentag	ge accuracy of
Correctly Classified Instances of	f 77% and
Incorrectly Classified Instances of	23%. 10-Fold
Cross Validation with a Correct	
Instances percentage of 78% ac	ly Classified
	curacy, 22%
Incorrectly Classified Instances [9].	curacy, 22%
7. Anggi Application of 2022 Algoritma Algorithm C4.5 Calculation Process	curacy, 22%
7. Anggi Application of 2022 Algorithm Algorithm C4.5 Calculation Process Trifani, C4.5 C4.5 Algorithm C4.5 to determine the dot	Calculation of ninant factors
7. Anggi Application of 2022 Algoritma Algorithm C4.5 Calculation Process 7. Anggi Application of 2022 Algorithm Algorithm C4.5 Calculation Process 7. Trifani, C4.5 C4.5 Algorithm C4.5 to determine the dom Agus Classification of causing Stress can be described as for	Calculation of ninant factors llows: Step 1:
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7.AnggiApplication of C4.52022AlgoritmaAlgorithm C4.5 Calculation Process7.AnggiApplication of C4.52022AlgoritmaAlgorithm C4.5 Calculation ProcessTrifani,C4.5C4.5C4.5Algorithm C4.5 to determine the dor causing Stress can be described as for Count the number of cases, the numbPerdanaData Mining in Windarto,CerrminingHighly Stressful decisions, the numbHendryStress Levels of QurniawanNot Stressed decisions. Step 2: Calculation	Calculation of ninant factors llows: Step 1: er of cases for er of cases for alculating the ed by attribute
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7.AnggiApplication of C4.52022AlgoritmaAlgorithm C4.5 Calculation Process Algorithm C4.5 Calculation Process Calculation Process C4.57.AnggiApplication of C4.52022AlgoritmaAlgorithm C4.5 Calculation Process Causing Stress can be described as fo Count the number of cases, the numb Highly Stressful decisions, the numb Hendry QurniawanHendryStress Levels of Final StudentsNot Stressed decisions. Step 2: Ca entropy of all cases and cases divide class with equation (1) then calculatin each attribute with equation (2). C4 Generates 20 rules and the accuracy	Calculation of ninant factors llows: Step 1: er of cases for er of cases for alculating the ed by attribute ng the gain for .5 Algorithm. rate produced

					using the $C4.5$ Algorithm the most dominant
					using the C4.5 Algorithm, the most dominant
					factor is interpersonal with a gain value of
					0.328180116 [10].
8.	Arief	Main Student	2020	Algoritma	Student main data at the preparatory stage was
	Jananto,Sula	Data as a		CART	taken from the Academic Administration Bureau
	stri, Eko Nur	Predictor of			of the Faculty who had graduated in the graduation
	Wahyudi	Timeliness of			data in the form of student name, gender, city of
	Dan Sunardi	Graduation			birth, date of birth, name of school of origin for 5
		Using the CART			years distributed in each semester. From this
		Algorithm for			process the data obtained amounted to 1151
		Data Mining			records. By implementing categorical data mining
		Classification			techniques and the CART algorithm using repeated
		Data Mining			binary partitioning, it is expected to obtain a
					decision tree that can be used to predict class
					timeliness of graduation from active students. By
					using graduation data and student main data of
					1018 records, a decision tree model is obtained
					with an accuracy rate of 63% from the data testing
					[11].
9.	Ni Luh	Optimization of	2019	Algoritma	The results of the classification on the training data
	Ratniasih	Data Mining		C4.5 Dan	are the GPA attribute as the root of the decision
		Using Naïve		KNN	tree, while the other attributes are as child nodes.
		Bayes and C4.5			From the training data with a total of 50 data
		Algorithms for			produced 5 rules. The rules that have been obtained
		Classification of			from the training data can be used as rules to
		Student			determine whether graduation is on time or not for
		Graduations			STMIK STIKOM Bali students. 4 predictor
					attributes and 1 target attribute produce 5 rules in
					the decision tree so that these rules can be used in
					determining timely graduation for STMIK
					STIKOM Bali students. The results of the analysis
					using the Naïve Bayes method obtained an
					accuracy of 89.27% where the performance results
					are accurate [12].

Based on the data in Table 1, it can be said that the emphasis on the problems raised in Inna Alvi Nikmatum and Indra Waspada's research has problems with the timeliness of students in completing their studies and the proportion of students who complete their studies within the study period are included in the element of accreditation assessment. In Marina Windarti and Agustinus Suradi's research, the problem that can affect the quality of a tertiary institution is student performance which can be measured through the length of study period. In Wiwit Supriyanti's research, and Miss Puspitasari have problems processing student data needs to be done to find out important information in the form of new knowledge (knowledge discovery), for example information about classifying student data based on profiles and academic data. This new knowledge can help universities to classify student graduation rates in order to determine strategies to increase graduation in the following years. In the research by Elga Mariati, Ariesta Lestari, and Widiatry, assessment of student satisfaction with facilities and

learning at the Faculty of Engineering had been carried out before, but the assessment was still carried out partially and the results of data collection on satisfaction assessment had not been evaluated before. This study uses data mining techniques in classifying.

Next, namely the research steps or methods used, based on Table 1. In the research of Inna Alvi Nikmatum and Indra Waspada using the K-Nearest Neighbor. Marina Windarti and Agustinus Suradi used this study to understand the performance of the six classification algorithms used, namely C4.5 decision tree, Bayesian network, KNearest neighbor, Naïve Bayes, neural network and SVM. There are several data mining tools including Rapid Miner, Orange, KNIME, Weka, Keel and R. WEKA is GUI-based so that it minimizes the use of coding which makes it easier for system users. In this study using WEKA tools. Data processed at WEKA has ARFF and CSV formats. In the research by Wiwit Supriyanti, and Nona Puspitasari using the Information Gain Feature Selection Technique for Predicting Student Academic Performance in mathematics. In the research of Elga Mariati, Ariesta Lestari, and Widiatry. In this study, the method from the data mining approach was applied to classify whether students were satisfied or not with the quality of learning facilities at the Faculty of Engineering. This study compares three data mining algorithms, namely Decision Tree C4.5, Support Vector Machine, and Naïve Bayes to get the best algorithm for prediction systems.

4. CONCLUSION

Based on the research conducted by the researcher, it can be concluded that from the 9 studies presented, static classification techniques include logistic regression and discriminant analysis. Both assume that the relationship between input and output variables is basically linear, the data is normally distributed, and the variables are not interrelated and independent of one another. Assessment of student satisfaction with facilities and learning at the Faculty of Engineering has been carried out before, but the assessment is still carried out partially and the results of data collection on satisfaction assessment have never been evaluated before. This research uses data mining techniques in classifying. The research that has been carried out is considered insufficient so it is advisable to classify student data in order to be able to manage the data as a whole consisting of student personal data in the form of name, name, student address, parents' names and have student GPA data so that it can be processed appropriately with the like that, it can be seen how student performance, student accuracy, understanding of lessons and can see graduation in a timely manner or not.

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