



Vol. 17 No.2 2023

Submit :

Accept :

Publish:

30/06/2023

28/06/2023

06/12/2022

COMPARATIVE ANALYSIS OF K-NEAREST NEIGHBOR AND NAÏVE BAYES CLASSIFIER ALGORITHMS FOR CLASSIFICATION OF FEASIBILITY OF ACCEPTANCE OF PAWN GOODS

Mely Syafira^{*1)}, Mariza Devega²⁾ ¹²Universitas Lancang Kuning Pekanbaru, Indonesia *Email: <u>melysfr@gmail.com</u>

Abstract

One of the problems faced by companies engaged in pawn services is that they must always pay attention to the appropriateness of the value of the loan based on the goods being pawned. Sometimes the pawnbrokers do not come back after depositing their goods with the company and do not pay them off. If the value of the goods does not match and the amount of money lent is too large compared to the value of the pawned goods, the company will lose money. There needs to be a way to determine whether or not it is appropriate for a pawned item to be accepted by the company and the value of the loan given to the pawnbroker. The way to find out can be classified by data mining classification techniques using the K-Nearest Neighbor method and the Naïve Bayes method. The analysis carried out by manual calculations and testing with Rapidminer tools resulted in the accuracy of the comparison of the two methods of the K-Nearest Neighbor and Naïve Bayes algorithms in predicting the classification of the feasibility of receiving pawned goods using the Confusion Matrix formula which resulted in an accuracy rate of the K-Nearest Neighbor method of 40%. while the Naïve Bayes method has an Accuracy Value of 20%, so we can determine that the accuracy of the prediction value using the K-Nearest Neighbor method is better than the Naïve Bayes method.

Keywords: Data Mining, Classification, K-Nearest Neighbor, Naïve Bayes, Rapidminer

© 2023 Lembaga Layanan Pendidikan Tinggi Wilayah X. This is an open-access article under the CC Attribution 4.0 license (<u>https://creativecommons.org/licenses/by/4.0/</u>).

http://publikasi.lldikti10.id/index.php/jit

DOIs: https://doi.org/10.22216/jit.v17i2.1827

PAGE: 401-406

ISSN : 1979-9292 / 2460-5611 © 2023 Higher Education Service Institute Region X

This is an open access article under the CC Attribution 4.0 license (https://creativecommons.org/licenses/by/4.0/). JIT is Sinta 3 Journal (<u>https://sinta3.kemdikbud.go.id/journals/profile/2143</u>) accredited by the Ministry of Research & Technology, Republic of Indonesia

ISSN : 1979-9292



INTRODUCTION

CV. Malau Jaya Gadai is one of the pawn services in Pekanbaru City. The company in the field of loan services runs with a pawn scheme. The types of goods that can be mortgaged are cellphones, laptops, motorcycles, moto bpkb, car bpkb, house certificates, gold and others. The loan price list is determined by the company for mortgagers who want to pawn their goods, of course, the goods to be mortgaged by the provisions of their value. One of the problems faced by companies engaged in pawn services is that they must always pay attention to the feasibility of borrowing value based on the goods being mortgaged. Sometimes, the mortgagers do not return after entrusting the goods to the company and not paying them off. If the value of the goods does not match and the amount of money lent is too large when compared to the value of the pawn goods, the company will lose money. There needs to be a way to find out whether or not a lien is accepted by the company and the value of the loan that will be given to the mortgagee.

The way to find out can be classified with data mining classification techniques. Data mining is a science used to mine from several data to produce useful information [1]. This research uses the Naïve Bayes method and also the K-Nearest Neighbor method to dig up hidden information from existing data and analyze it to get information.

Naïve Bayes is a simple probabilistic classification method [2]. This method will calculate a set of probabilities by summing the frequency and combination of values from a given dataset [3]. The naïve Bayes method assumes all attributes in each category have no dependencies on each other (independent) [4]. The advantage of using Naive Bayes is that it only requires a small amount of training data to determine the mean parameters and variance of the variables required for classification [5]. Naive Bayes is a supervised document classification method, that requires training data before the classification process [6].

K-Nearest Neighbor is a method of finding cases by calculating the proximity between new cases and old cases based on the weight match of several existing features [7]. To define the distance between two points in the training data (x) and points in the testing data (y) the formula is used euclidean distance [8].

By conducting data mining analysis using the two methods above to classify whether or not a pawn item is accepted by the company. And also aims to be able to help the company in the decision-making process related to mortgage issues.

RESEARCH METHODS

This study used two algorithms, namely Naïve Bayes and K-Nearest Neighbor. The Naive Bayes algorithm is a classification with probability and statistical methods while the K-Nearest Neighbor algorithm is a classification method based on attributes and training samples. Here is the general equation of the Naïve Bayes algorithm [9] :

$$(Ci|X) = P(X|Ci) P(Ci) P(X)$$
(1)

LLDIKTI Region X



ISSN : 1979-9292



JURNAL IPTEKS TERAPAN Research of Applied Science and Education V17.i2 (401-406)

E-ISSN: 2460-5611

Here is the general equation of the K-Nearest Neighbor algorithm [10] :

$$d(x,y) = \sqrt{\sum_{i=1}^{r} (Xik - Yik)^2}$$
(2)

Where :

d = is the euclidean distance between points in the training data

X = data training

Y = data testing

R = number of variables

RESULTS AND DISCUSSION

In preprocessing this data is done by taking raw data derived from the CV. Malau Jaya Gadai is then recapitulated in the form of data stored in Excel (.xls) format, which will later be used as new data consisting of item type data (Var. A), Loan Category (var. B), and Timeframe (Var. C) as a calculation material.

No	Item	Loan	Tenor	Rate
	SMARTPHONE	2.000.000	30	260.000
1	SMARTPHONE	1.000.000	15	130.000
	SMARTPHONE	1.200.000	15	156.000
	LAPTOP/NOTEBOOK	1.500.000	15	195.000
2	LAPTOP/NOTEBOOK	2.000.000	30	260.000
	BARANG ELEKTRONIK	800.000	15	104.000
	SMARTPHONE	1.200.000	30	156.000
424	SMARTPHONE	4.000.000	30	520.000
	SMARTPHONE	1.300.000	30	169.000
	Table 1. Pre	liminary	Data	

Table 1. Preliminary Data

Testing

The testing process is a very important process to find out the extent to which Data Mining designs can be tested using hardware or software.

Rapidminer's Initial Look 1.

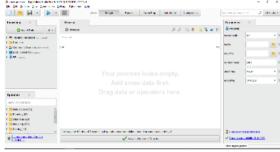


Figure 1. Initial Look

Import Database 2.

Data import is done to enter the data to be tested in Excel format.

) 💻 🖬 🔹 🕨 🔹	In part data - tanks de sid	in head came				X		ell 2 sale
onling X		Sele	ect litre data i	laculion.		Process	×	
Connected and the solution	10/m · · · · · · · · · · · · · · · · · · ·						**	
Campline Company (a Samples personne)	Protects	No. Lana	Bar MAR	State FOR Day	Local Marchinel Marcolo, 2010	in the second se		-
Local (Repeated Grand		 March 1980 March 1980 March 1980 March 1980 	2184	Monator chief wear wear	Aur 7 2022	eestie .		
		(i) conserve to (ii) constant quest	NO MAN MAN	Menterhühler Des W. Manser (Charles et W.		10.000-0000	2001	
		and the period	90 G.		The of 2000	-	care.	
						Another a	OPOIDA	
annan x								
INCOMPANY INCOMPANY						v		
Outer Access phile	data tananyaka					- 11		
Charles (27)	Al Plan			Transfer to Transfer	a ny data. Kaodi 🦲	÷		
Manalog (152) Paralog (152)				e- N-Aria			- coding and sold	
Concentration for the state	1		Advert of the last of					

Figure 2. Import Database

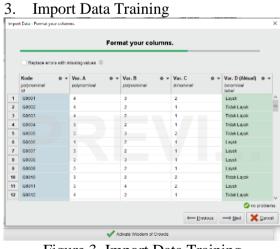


Figure 3. Import Data Training







JURNAL IPTEKS TERAPAN Research of Applied Science and Education V17.i2 (401-406)

	Format your columns.							
	Kode polynominal Id	ð .	Var. A polynominal	ð Ŧ	Var. B polynominal	ð .	Var. C &	 Var. D (Aktual) & binominal label
1	G0921		1		з		2	Tidak Layak
2	G0922		3		2		1	Layak
3	C0923		3		4		1	Tidak Layak
4	G0924		2		2		2	Layak
5	G0925		4		4		1	Tidak Layak

Figure 4. Import Data Testing

5. Repository

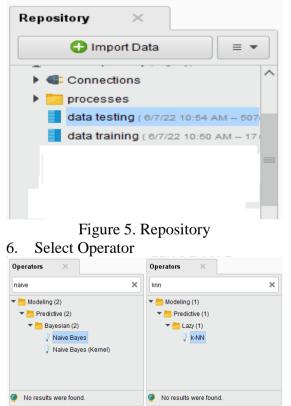


Figure 6. Select Operator

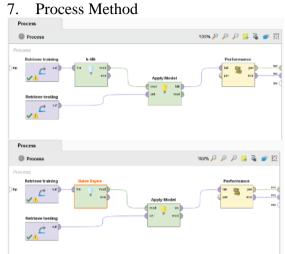


Figure 7. Process Method

8. Setting Parameters

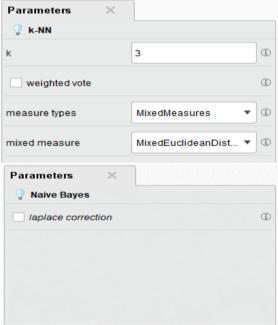


Figure 8. Setting Parameters

9. Example Set K-Nearest Neighbor





JURNAL IPTEKS TERAPAN Research of Applied Science and Education V17.i2 (401-406)

E-ISSN : 2460-5611



Neighbor



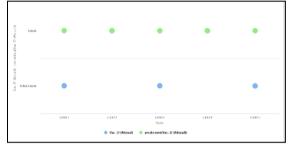


Figure 10. Scatter Plot K-Nearest Neighbor

11. Performance Vector K-Nearest Neighbor

Views D	icsign Results	Turbo Prep	Auto Model	Deployments	Find dela, operator
(App y Mozel) 🛛 👋 🧏 Perfor	manceVector (Performanc	ə ×			
Table View O Plot View					
accuracy: 40.00%					
accuracy: 40.00%	hue Indak Layak		hus Lay	at	dasa precsion
accuracy: 40.00%	Frue Indek Lagak D		thus Lay	at	data precaron 0.00%
-				at	

Figure 11. Performance Vector K-Nearest Neighbor

12. Example Set Naïve Bayes



Figure 12. Example Set Naïve Bayes

13. Scatter Plot Naïve Bayes

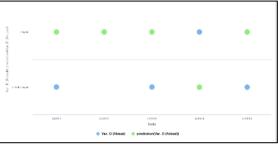


Figure 13. Scatter Plot Naïve Bayes

14. Performance Vector NB

Table View 🔘 Plot View			
accuracy: 20.00%			
	bua Tidak Layak	tua Layak	dasa precisio
pred. Tidak Layak	0	4	0.00%
prod. Layak	3	1	25.00%
	0.00%	50.00%	

Figure 14. Performance Vector Naïve Bayes

CONCLUSION

Based on the research that has been carried out, the author concludes that to see the level of accuracy of the comparison of the two methods of the K-Nearest Neighbor and Naïve Bayes Algorithms in predicting the classification of the feasibility of receiving pawn goods, the Confusion Matrix formula is used which produces an accuracy rate of the K-Nearest Neighbor method of 40% while the Naïve Bayes method has an Accuracy Value of 20%, so we can determine that the accuracy of the accuracy value from the prediction with the K-Nearest Neighbor method is a better one when compared to the Naïve Bayes method.

BIBLIOGRAPHY

[1] Y. Apridonal M, W. Choiriah, and A. Akmah, "Penerapan Data Mining Menggunakan Metode Association



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

405

ISSN : 1979-9292



JURNAL IPTEKS TERAPAN Research of Applied Science and Education V17.i2 (401-406)

Rule Dengan Algoritma Apriori Untuk Analisa Pola Penjualan Barang," *JURTEKSI*, vol. V, no. 2, pp. 193–198, 2019.

- [2] D. Remawati, R. D. Nugroho, and P. Harsadi, "Penerapan Sistem Penunjang Keputusan Menggunakan Algoritma Naive konsep Bayes Pada Human Information Resource System (Studi (HRIS) kasus : Penerusan Kontrak Kerja Karyawan di PT. XYZ)," J. Ilm. SINUS, vol. 18, no. 1, pp. 63–74, 2020.
- [3] R. M. Sari and Y. Apridonal M, "Data Mining Implementation For Printer Sales Prediction Using Naive Bayes Method," *ICoSSIT.*, vol. 1, no. 1, pp. 215–220, 2020.
- [4] R. Yendra, L. Marifni, and I. Suryani, "Klasifikasi Data Mining Untuk Seleksi Penerimaan Calon Pegawai Negeri Sipil Tahun 2017 Menggunakan Metode Naïve Bayes," J. Sains Mat. dan Stat., vol. 6, no. 1, pp. 65–78, 2020.
- [5] R. Jayadi, H. M. Firmantyo, M. T. J. Dzaka, M. F. Suaidy, and A. M. Putra, "Employee Performance Prediction Using Naïve Bayes," *Int. J. Adv. Trends Comput. Sci. Eng.*, vol. 8, no. 6, pp. 3031–3035, 2019.
- [6] C. Fadlan, S. Ningsih, and A. P. Windarto, "Penerapan Metode Naïve Bayes Dalam Klasifikasi Kelayakan Keluarga Penerima Beras Rastra," J. Tek. Inform. Musirawas, vol. 3, no. 1, p. 1, 2018.
- [7] A. D. Afifaturahman and M.

Firmansyah, "Perbandingan Algoritma K-Nearest Neighbour (KNN) dan Naive Bayes pada Intrusion Detection System (IDS)," *Innov. Res. Informatics*, vol. 3, no. 1, pp. 17–25, 2021.

- [8] R. N. Devita, H. W. Herwanto, and A. P. Wibawa, "Perbandingan Kinerja Metode Naive Bayes dan K-Nearest Neighbor untuk Klasifikasi Artikel Berbahasa indonesia," J. Teknol. Inf. dan Ilmu Komput., vol. 5, no. 4, pp. 427–434, 2018.
- [9] A. S. Fitrani, F. Fajrillah, and W. Novarika, "Implementation of Data Mining Using Naïve Bayes Classification Method To Predict Participation of Governor And Vocational Governor Selection in Jemirahan Village, Jabon District," *Int. J. Informatics Comput. Sci.*, vol. 3, no. 2, pp. 66–79, 2019.
- [10] A. Indriani, "Analisa Perbandingan Metode Naïve Bayes Classifier Dan K-Nearest Neighbor Terhadap Klasifikasi Data," *Sebatik*, vol. 24, no. 1, pp. 1–7, 2020.

LLDIKTI Region X

