

APPLICATION OF NAÏVE BAYES CLASSIFIER METHOD FOR PREDICTING CLAIMS IN AUTOMOBILE INSURANCE

UNDERGRADUATE THESIS

Submitted as one of the requirements to obtain

Sarjana Aktuaria

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FACULTY OF BUSINESS

ACTUARIAL SCIENCE STUDY PROGRAM

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NOVEMBER 2023

PANEL OF EXAMINERS APPROVAL

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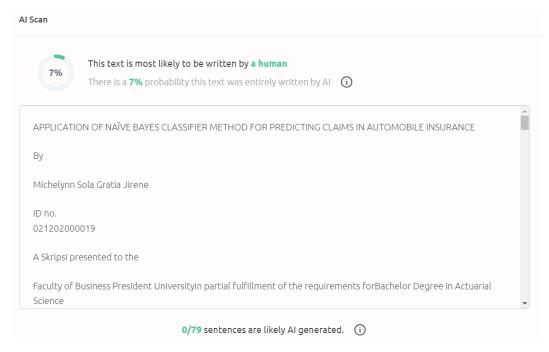
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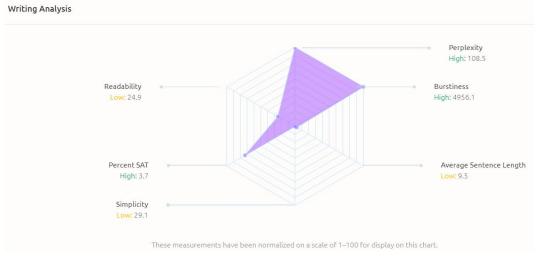
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ABSTRACT

Insurance companies often experience difficulties in planning funds. This is caused by the risk of uncertainty in life. Not least, the insurance company went bankrupt because they could not pay their obligations. For this reason, insurance companies need to know the right strategy for setting up reserve funds. One solution that can help insurance companies make decisions and determine strategies is to make claim prediction. In this study, the author will use the Naïve Bayes Classifier method to predict claims in automobile insurance. The Naïve Bayes Classifier itself is a simple probability method where the calculations are based on Bayes' Theorem. The data used is secondary data from Kaggle.com where this data consists of 10,000 samples with 19 features. The prediction results will be divided into two results, namely 0 and 1 where 0 means "no" and 1 means "yes". "yes" or "no" label will inform us whether the customer will claim or not. The data will go through preprocessing in python so that the format is appropriate. The model will be built without feature selection and with feature selection then being compared to determine the best model. Predicted data result will be compared to the actual data and the accuracy of the best model is 82%. Other evaluation method was applied to evaluate how well the model performed by using ROC – AUC score which has a score of 0.87 and 10-fold cross validation which has an average score of 80%. The result of the prediction will help insurance company with underwriting decision and financial approach or planning.

Keywords: Claim Prediction, Naïve Bayes Classifier, Automobile Insurance

ABSTRAK

Perusahaan asuransi umumnya mengalami kesulitan dalam perencanaan dana. Hal ini disebabkan oleh resiko ketidakpastian dalam hidup. Tidak sedikit, Perusahaan asuransi mengalami kebangkrutan karena tidak dapat membayar kewajiban mereka. Karena itu, Perusahaan asuransi perlu untuk mengetahui strategi yang tepat dalam mencadangkan dana. Salaah satu solusi yang dapat membantu perusahaan asuransi dalam mengambil keputusan dan menetapkan strategi adalah dengan melakukan prediksi klaim. Dalam penelitian ini, penulis akan menggunakan metode Naïve Bayes Classifier untuk memprediksi klaim pada asuransi automobile. Naïve Bayes metode Classifier sendiri merupakan probabilitas sederhana perhitungannya berdasarkan Teorema Bayes. Data yang digunakan adalah data sekunder dari Kaggle.com dengan 10.000 sampel dan 19 fitur. Hasil prediksi akan dibagi menjadi 2 hasil yaitu 0 dan 1 dimana 0 berarti "tidak" dan 1 berarti "ya". Label "ya" dan "tidak" akan memberikan informasi apakah pelanggan akan melakukan klaim atau tidak. Data akan melalui proses preproses di python agar format sesuai. Model akan dibuat tanpa seleksi fitur dan dengan seleksi fitur lalu dibandingkan untuk menentukan model terbaik. Hasil prediksi data akan dibandingkan dengan data aktual dan akurasi pada model terbaik adalah 82%. Metode evaluasi lain digunakan untuk mengevaluasi seberapa baik performa model dengan menggunakan nilai ROC - AUC dengan nilai 0,87 dan 10-fold cross validation dengan nilai rata-rata 80%. Hasil prediksi akan membantu prerusahaan asuransi dalam keputusan underwriting dan rencana pendekatan finansial.

Kata kunci: Prediksi Klaim, Naïve Bayes Classifier, Asuransi Automobile

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TABLE OF CONTENTS

COVERi
PANEL OF EXAMINERS APPROVAL ii
STATEMENT OF ORIGINALITY iii
SCIENTIFIC PUBLICATION APPROVAL FOR ACADEMIC INTEREST iv
ADVISOR APPROVAL FOR JOURNAL/INSTITUTION'S REPOSITORY . v
PLAGIARISM REPORTvi
ANTI-PLAGIARISM REPORTvii
ABSTRACTix
ACKNOWLEDGEMENTxi
TABLE OF CONTENTSxii
LIST OF TABLESxv
LIST OF FIGURESxvii
LIST OF ABBREVIATIONSxvii
CHAPTER I INTRODUCTION
1.1. Research Background
1.2. Problem of the Research
1.3. Research Question
1.4. Research Objective
1.5. Research Scope and Limitation
1.6. Research Benefit
1.7. Research Outline
CHAPTER II LITERATURE REVIEW7
2.1. Insurance
2.2. Data Mining8
2.3. Machine Learning9
2.3.1. Machine Learning Definition
2.3.2. Supervised and Unsupervised Machine Learning9
2.3.3. Types of Supervised and Unsupervised Learning Algorithm 11
2.3.4. Data Preprocessing

	2.3.5. Training and Testing	. 13
	2.4. Probability	. 14
	2.5. Conditional Probability	. 16
	2.6. Bayes' Theorem	. 16
	2.6.1. Naïve Bayes Classifier	. 18
	2.6.2. Advantage and Disadvantage	. 22
	2.7. Confusion Matrix	. 22
	2.8. ROC - AUC Curve	. 24
	2.9. K-fold Cross Validation	. 26
	2.10. Previous Research and Research Gap	. 28
CF	HAPTER III METHODOLOGY	. 30
	3.1. Research Design	. 30
	3.2. Sampling Plan	. 30
	3.3. Instrument/Operational Definition	. 30
	3.4. Data Collection Design	. 31
	3.5. Flowchart	. 32
CF	HAPTER IV ANALYSIS AND RESULT	. 34
	4.1. Data Preparation	. 34
	4.2. Import Data	. 34
	4.3. Check and Fill Null Data	. 37
	4.4. Feature Plotting	. 41
	4.5. Transform Data into Numerical	. 54
	4.6. Prediction	. 57
	4.7. Evaluation	. 58
	4.8. Categorical Feature Model Prediction and Evaluation	60
	4.9. Numerical Feature Model Prediction and Evaluation	. 63
	4.10. 5 Feature Selection Model Prediction and Evaluation	. 66
	4.11. 7 Feature Selection Model Prediction and Evaluation	. 68
	4.12. 9 Feature Selection Model Prediction and Evaluation	. 71
Cŀ	HAPTER V CONCLUSION	. 74
	5.1. Conclusion	. 74
	5.2. Recommendation	. 75

REFERENCES	76
APPENDICES	79

LIST OF TABLES

Table 2.1 Sample Table for Naïve Bayes Classifier Example	. 20
Table 2.2 x and f(x) Approximate Value	. 25
Table 4.1 Raw Data 1	. 34
Table 4.2 Raw Data 2	. 35
Table 4.3 Raw Data 3	. 36
Table 4.4 Raw Data Info	. 36
Table 4.5 Sum of Null Data	. 37
Table 4.6 Columns of Filled Data 1	. 38
Table 4.7 Columns of Filled Data 2	. 39
Table 4.8 Columns of Filled Data 3	. 39
Table 4.9 Recheck Sum of Null Data	. 40
Table 4.10 New Data Info with Filled Columns	. 41
Table 4.11 Updated Data 1	. 51
Table 4.12 Updated Data 2	. 52
Table 4.13 Updated Data 3	. 53
Table 4.14 Updated Data Info	. 53
Table 4.15 Unique Value	. 54
Table 4.16 Transformed Data 1	. 55
Table 4.17 Transformed Data 2	. 56
Table 4.18 Transformed Data Info	. 56
Table 4.19 Ten Rows of Probability	. 57
Table 4.20 Classification Report	. 59
Table 4.21 Categorical Features Data Info	. 61
Table 4.22 Categorical Feature Model Classification Report	. 62
Table 4.23 Numerical Features Data Info	. 63
Table 4.24 Numerical Feature Model Classification Report	. 65
Table 4.25 5 Feature Selection Model Data Info	. 66
Table 4.26 5 Feature Selection Model Classification Report	. 67
Table 4.27 7 Feature Selection Model Data Info	. 68

Table 4.28 7 Feature Selection Model Classification Report	. 70
Table 4.29 9 Feature Selection Model Data Info	.71
Table 4.30 9 Feature Selection Model Classification Report	. 72

LIST OF FIGURES

Figure 3.1 Flowchart	32
Figure 4.1 Count Plot Between Age and Outcome	42
Figure 4.2 Count Plot Between Gender and Outcome	42
Figure 4.3 Count Plot Between Race and Outcome	43
Figure 4.4 Count Plot Between Driving Experience and Outcome	43
Figure 4.5 Count Plot Between Education and Outcome	44
Figure 4.6 Count Plot Between Income and Outcome	45
Figure 4.7 Count Plot Between Vehicle Ownership and Outcome	45
Figure 4.8 Count Plot Between Vehicle Year and Outcome	46
Figure 4.9 Count Plot Between Married and Outcome	46
Figure 4.10 Count Plot Between Children and Outcome	47
Figure 4.11 Count Plot Between Vehicle Type and Outcome	48
Figure 4.12 Box Plot Between Speeding Violations and Outcome	48
Figure 4.13 Box Plot Between DUIS and Outcome	49
Figure 4.14 Box Plot Between Past Accidents and Outcome	50
Figure 4.15 Box Plot Between Credit Score and Outcome	50
Figure 4.16 Box Plot Between Annual Mileage and Outcome	51
Figure 4.17 Confusion Matrix Plot	59
Figure 4.18 ROC – AUC Curve	60
Figure 4.19 Categorical Feature Model Confusion Matrix Plot	62
Figure 4.20 Categorical Feature Model ROC – AUC Curve	63
Figure 4.21 Numerical Feature Model Confusion Matrix Plot	64
Figure 4.22 Numerical Feature Model ROC – AUC Curve	65
Figure 4.23 5 Feature Selection Model Confusion Matrix Plot	67
Figure 4.24 5 Feature Selection Model ROC – AUC Curve	68
Figure 4.25 7 Feature Selection Model Confusion Matrix Plot	69
Figure 4.26 7 Feature Selection Model ROC – AUC Curve	70
Figure 4.27 9 Feature Selection Model Confusion Matrix Plot	72
Figure 4.28 9 Feature Selection Model ROC – AUC Curve	73

LIST OF ABBREVIATIONS

AI : Artificial Intelligence

ML : Machine Learning

NBC : Naïve Bayes Classifier

ROC : Receiver Operator Characteristic Curve

TP : True Positive

TN : True Negative

FP : False Positive

FN : False Negative

TPR : True Positive Rate

FPR : False Positive Rate

OJK : Otoritas Jasa Keuangan

KDD : Knowledgeable Discovery in Data

AUC : Arena Under the Curve

DUIS : Driving Under Influences

PDF : Probability Density Function

CDF : Cumulative Distribution Function