# A SYSTEM TO PREDICT PHARMACEUTICAL STUDENT PERFORMANCE USING MACHINE LEARNING

# NURUL NAZIHAH BINTI SUKOR

# BACHELOR OF COMPUTER SCIENCE UNIVERSITI MALAYSIA PAHANG

## A SYSTEM TO PREDICT PHARMACEUTICAL STUDENT PERFORMANCE USING MACHINE LEARNING

#### NURUL NAZIHAH BINTI SUKOR

Thesis submitted in fulfillment of requirements for the award of the degree of Bachelor of Computer Science (Software Engineering)

Faculty of Computer Systems & Software Engineering

UNIVERSITI MALAYSIA PAHANG

MAY 2019



## UNIVERSITI MALAYSIA PAHANG

#### DECLARATION OF THESIS AND COPYRIGHT

#### TITLE: A SYSTEM TO PREDICT PHARMACEUTICAL STUDENT PERFORMANCE USING MACHINE LEARNING

ACADEMIC SESSION: SEMESTER 2 SESSION 2018/2019

#### I AM NURUL NAZIHAH BINTI SUKOR

I acknowledge that University Malaysia Pahang reserves the following rights:

- 1. The Thesis is the Property of University Malaysia Pahang
- 2. The Library of University Malaysia Pahang has the right to make copies of the thesis for the purpose of research only.
- 3. The Library has the right to make copies of the thesis for academic exchange.
- 4. I declare that this thesis is classified as:

□ CONFIDENTIAL	(Contains confidential information under the Official
	Secret Act 1997)*
□ RESTRICTED	(Contains restricted information as specified by the
	organization where research was done)*
□ OPEN ACCESS	I agree that my thesis to be published as online open access (Full Text)

Certified by:

(Student's Signature)

(Supervisor's Signature)

Student's Name: NURUL NAZIHAH BINTI SUKOR Date: 14<sup>th</sup> JUNE 2019 Supervisor's Name: DR AHMAD FIRDAUS BIN ZAINAL ABIDIN Date: 14<sup>th</sup> JUNE 2019

NOTE: \* If the thesis is CONFIDENTIAL or RESTRICTED, please attach a thesis declaration letter.

### **STUDENT'S DECLARATION**

I hereby declare that the work in this thesis is based on my original work except for quotations and citations which have been duly acknowledged. I also declare that it has not been previously or concurrently submitted for any other degree at University Malaysia Pahang or any other institutions.

Signature	:
Name	: NURUL NAZIHAH BINTI SUKOR
ID Number	: CB14I32
Date	: 14 <sup>th</sup> JUNE 2019

### SUPERVISOR'S DECLARATION

I hereby declare that I have read this thesis and in my opinion this thesis/report is sufficient in terms of scope and quality for the award of the degree of Bachelor of Computer Science (Software Engineering).

Signature:Supervisor Name: DR AHMAD FIRDAUS BIN ZAINAL ABIDINDate: 14<sup>th</sup> JUNE 2019

#### ACKNOWLEDGEMENT

In the Name of Allah s.w.t, the Most Beneficent and the Most Merciful

I would like to thank to Allah s.w.t for his blessed during completing this research. Alhamdulillah, I will not be able to finish this research without his blessed.

First and foremost, our deepest gratitude I bid to our ever so helpful, Dr Ahmad Firdaus Bin Zainal Abidin for his all effort and guidance which leads us to complete my research based. Without his guidance, I will not be completion of this research papers in PSM 1 and PSM 2.

Last but not least, thank you and my highest appreciation is for my beloved family, especially my father, Mr Sukor Bin Awang, my mother, Mrs Zainon Bte Haji Hitam and my sibling for their continuous support, encouragement during the completion of this research papers in PSM 1 and PSM 2.

Special thanks to my friend also my senior with same course and major in Software Engineering is Nur Shahirah Binti Mohamad Faizal, Wan Noorafirah Binti Wan Ramli and all friend who had helped, give me strength and supported us throughout the creation and completion o of this research papers in PSM 1 and PSM 2.

#### ABSTRACT

Recently, pharmacy course is the important course as it involves human lives. Moreover, predicting of student performance is most of higher learning institutions in Malaysia. The main objective of this paper is to provide an overview on the Machine Learning technique that has been used to propose to improve student achievement and to predict the pharmacy student will be quit or graduate of a University. There are three reasons why this is happening relate in the problem statement. Firstly is the lack of pharmacy student in the sector in the pharmaceutical industry. Second is increase the number of student from local IPT to offer a pharmacy programed and lastly is the result of pharmacist shortage that student needed. As well know, if they take the pharmaceutical subject whether they have basic or not in the pharmaceutical, the result will display many pharmacy students is failed. This is because, they still don't know that they still suitable with the course or not. Here, the research is very important and suitable to needs to implement a system to predict pharmaceutical student performance using machine learning. This paper also focuses on how prediction uses Machine Learning classifier and finalized model to make a prediction on new data before the student registering the pharmaceutical course from WEKA to Java code. It could bring the benefit and impact to the student. This study proposes a Machine Learning technique to predict either the pharmacy student will be quit or graduate. The classifiers are multilayer perceptron (MLP) to predict pharmaceutical student performance. Among the MLP classifier, the outstanding outcome acquire is the MLP, which achieves through 80% accuracy. Based on result from the simulation used to display the train use test with cross validation is 10 fold in MLP is 29 for 63% in correctly classified instances (true positive rate for graduate) and 17 for 36% in incorrectly classified instances (false negative rate for quit).

#### ABSTRAK

Baru-baru ini, kursus farmasi adalah kursus penting kerana ia melibatkan kehidupan manusia. Selain itu, meramalkan prestasi pelajar adalah kebanyakan institusi pengajian tinggi di Malaysia. Objektif utama ini adalah untuk memberikan gambaran mengenai teknik Pembelajaran Mesin yang telah digunakan untuk mencadangkan untuk meningkatkan pencapaian pelajar dan untuk meramalkan bahawa pelajar farmasi akan berhenti atau lulus dari Universiti. Terdapat tiga sebab mengapa ini berlaku berkaitan dengan pernyataan masalah. Pertama adalah kekurangan pelajar farmasi dalam sektor industri farmaseutikal. Kedua adalah meningkatkan jumlah pelajar dari IPT tempatan untuk menawarkan program farmasi dan akhirnya adalah hasil kekurangan ahli farmasi yang diperlukan oleh pelajar. Serta tahu, jika mereka mengambil subjek farmaseutikal sama ada mereka mempunyai asas atau tidak dalam farmaseutikal, hasilnya akan memaparkan banyak pelajar farmasi gagal. Ini kerana, mereka masih tidak tahu bahawa mereka masih sesuai dengan kursus atau tidak. Di sini, penyelidikan sangat penting dan sesuai untuk keperluan untuk melaksanakan sistem untuk meramalkan prestasi pelajar farmaseutikal menggunakan pembelajaran mesin. Makalah ini juga menumpukan pada bagaimana ramalan menggunakan pengeluar Pembelajaran Mesin dan model yang telah diselesaikan untuk membuat ramalan pada data baru sebelum pelajar mendaftar kursus farmaseutikal dari WEKA ke kod Java. Ia boleh membawa manfaat dan impak kepada pelajar. Kajian ini mencadangkan teknik Pembelajaran Mesin untuk meramalkan sama ada pelajar farmasi akan berhenti atau lulus. Pengelas adalah perceptron multilayer (MLP) untuk meramalkan prestasi pelajar farmaseutikal. Di antara pengelas MLP, hasil yang cemerlang diperoleh adalah MLP, yang mencapai ketepatan 80%. Berdasarkan keputusan dari simulasi yang digunakan untuk memaparkan ujian penggunaan kereta api dengan pengesahan silang adalah 10 kali lipat dalam MLP adalah 29 untuk 63% dalam keadaan yang diklasifikasikan dengan benar (tingkat positif untuk lulusan) dan 17 untuk 36% dalam keadaan yang tidak benar (kadar negatif palsu untuk berhenti).

## TABLE OF CONTENTS

CONTENT	Page
STUDENT'S DECLARATION	ii
SUPERVISOR'S DECLARATION	iii
ACKNOWLEDGEMENT	iv
ABSTRACT	v
ABSTRAK	vi
TABLE OF CONTENTS	vii
LIST OF TABLES	xi
LIST OF FIGURES	xii
LIST OF ABBREVIATIONS	xiv

## CHAPTER 1 INTRODUCTION

1.1	INTRODUCTION	1
1.2	PROBLEM STATEMENT	2
1.3	OBJECTIVE	4
1.4	SCOPE	4

## CHAPTER 2 LITERATURE REVIEW

2.1	INTRODUCTION	5
2.2	EXISTING TECHNOLOGIES USING MACHINE LEARNING	6
2.3	RELATED WORKS	7
	2.3.1 Machine Learning Classifier	7
	2.3.1.1 Supervised Learning	7
	2.3.1.2 Unsupervised Learning	7
2.4	PREDICTION USE ALGORITHM / METHOD /	
	TECHNIQUE	8
	2.4.1 Multilayer Perceptron (MLP)	8
2.5	EVICTING EEATUDES	
2.3	EXISTING FEATORES	9
	2.5.1 Comparison between Existing Features of	9
	Prediction	
2.6	MACHINE LEARNING CLASSIFIER	9
		,
	2.6.1 Example of Machine Learning Classifier	9

## CHAPTER 3 METHODOLOGY

2
2

5

3.2	RESEARCH METHODOLOGY	13
	<ul> <li>3.2.1 Phase 1: Preliminary Studies</li> <li>3.2.2 Phase 2: Literature Review</li> <li>3.2.3 Phase 3: Knowledge Analysis</li> <li>3.2.4 Phase 4: Methodology Phase</li> <li>3.2.5 Phase 5: Prototype Design</li> <li>3.2.6 Phase 6: Validate Prediction</li> </ul>	14 14 15 15 16 17
3.3	SYSTEM DESIGN FOR MACHINE LEARNING IN WEKA TOOLS	17
3.4	SYSTEM DESIGN FOR MACHINE LEARNING IN JAVA PROGRAMMING	22
	<ul><li>3.4.1 Use Training Test</li><li>3.4.2 Predict New Samples using Exported</li></ul>	23
	Model Object	25
3.5	HARDWARE AND SOFTWARE DEVELOPMENT	29
	351 Hardware Development	29
	3.5.2 Software Development	30
3.5	GANTT CHART	30

## CHAPTER 4 RESULT AND DISCUSSION

4.1	INTRODUCTION	31

4.2	RESULT OUTPUT PREDICTION	31
	4.2.1 Result Prediction from WEKA Explorer	31
	4.2.2 Result Prediction in JAVA Programming	33
	4.2.3 Result Predict New Sample use classifier	33
	4.2.4 Result Output After Prediction	35
4.3	TEST CASE SPECIFICATION	36
CHAPTER 5	CONCLUSION	
5.1	INTRODUCTION	42
5.2	RESEARCH CONSTRAINT	43

5.3	FUTURE WORK	44

REFERENCES 45

APPENDICES

## LIST OF TABLES

Table No.	Title	Page
2.1	Previous Thesis by Different Technologies using Machine Learning	10
2.2	Result related the features of Student Performance	11
2.3	Types of Machine Learning Classifier in the Previous Studies	14
3.1	Show the hardware development used in the research	29
3.2	Show the software development used in the research	30
4.1	Declaration to Predict New Sample in JAVA	35
4.2	Test Case Specification	37

## LIST OF FIGURES

Figure No.	Title	Page
3.1	Flowchart of research Methodology	13
3.2	WEKA GUI Chooser Interface	18
3.3	The WEKA start up box	18
3.4	Finalize a Machine Learning Model in WEKA	19
3.5	Starting the modeling process and saving a model	20
3.6	WEKA model load a finalized model ready for use	21
3.7	WEKA dataset copy edit using Notepad++	21
3.8	WEKA dataset for make prediction one new data by Loaded Model	22
3.9	Coding for programming in JAVA to use training set to WEKA	23
3.10	Edit with Notepad ++ to use Training Set	24
3.11	Header files for WEKA in JAVA Programming	25
3.12	Class name for WEKA in JAVA Programming	25
3.13	WEKA Attributes in JAVA Programming	26

3.14	WEKA [Array List] for each attributes in JAVA	27
3.15	Unpredicted data set and last features in JAVA	27
3.16	Create new instances from data model in JAVA	28
3.17	Declaration of instances to use in prediction in JAVA	28
3.18	Declaration of reference to dataset in JAVA	29
4.1	Result from use training set use in WEKA	32
4.2	Result from use training set in JAVA	33
4.3	Declaration to ready trained model in JAVA	33
4.4	Declaration to predict new sample in JAVA	34

## LIST OF ABBREVIATION

MLP	Multilayer Perceptron
ML	Machine Learning
PC-SAS	Personal Computer Statistical Analysis Software
URL	Uniform Resource Locator

#### **CHAPTER 1**

#### INTRODUCTION

#### **1.1 INTRODUCTION**

Nowadays, the growth of information technology has ensured that matching performance to aims has become more complex and highly competitive in the environment. It is undeniable that the Institute of Higher Learning has an effective wardrobe, the professional instructors, the latest teaching and learning facilities will be more competitive in the face of increasingly challenging global competition.

Generally, student performance is most of higher learning institutions in Malaysia. This is because one of the criteria of higher learning institution that suitable based on the previous literature. Although, the student performance has good basic and knowledge, some researchers want to predicting student performance by using machine learning to evaluate and predict student performance to know the status that suitable if the pharmacy student whether quit or graduated at the University that they studied. So, they can predict themselves when they know the result, it is not impossible that they can improve their learning and get the best result, no matter where they come from groups before they registering the pharmaceutical course. Therefore, there are many techniques to evaluate student performance by using various machine learning techniques classifier that are appropriate for educational purposes. One of them is to identify every student either in high risk status or not. For example, they predict student performance through existing data set. This is because they can also be the way they want to identify whether these characteristics can affect the students' performance [1].

#### **1.2 PROBLEM STATEMENT**

Nowadays, the level of learning and performance of graduates in Higher Institutions such as STPM, Diploma and degree level is particularly alarming. However, after they have finished studying, their latest results determine their performance from the University to real-world employment.

The pharmaceutical course is the most important and to evaluate for pharmaceutical studies is important to maintain the student performance and effectiveness of the learning process. Firstly, the problem statement related to the case study is the lack of pharmacy student in the sector in the pharmaceutical industry. However, they have a problem is exceeding difficult with the contribution of Saudi pharmacist to local the pharmaceutical industry is very useful and important has affected the sector in the pharmaceutical industry.

Besides, secondly is, the student still does not know why will be learning the subject because increasing the number of students from local IPT University who wish to offer a pharmacy program dramatically. They do not have the flexibility to choose the tutors who are teaching the subject. They just don't know how to predict on themselves when they know the result; it is not impossible that they can improve in the pharmaceutical course

Lastly, achieving the high learning in performance is crucial attracting the result of this serious pharmacist shortage which is the pharmacy student is needed. Based on the Medical and Health field plans a great opportunity for career opportunities in the private sector. For example, latest for the case study that happens in area Malaysia such as there have a job vacancy to anyone want to apply as Lecturer University at UMP in major pharmaceutical such as pharmaceutical formulation or pharmaceutical manufacturing, technology and so on. This is because there is the best platform provided for gathering the student to apply for the job. (Refer Appendix A).

#### **1.3 OBJECTIVE**

The goals to be achieved in this research are the following:

- i. To study and analyse the performance of the student in Pharmaceutical course
- ii. To design and develop a system (prototype) using WEKA technique in JAVA programming codes.
- To apply the finalized model to make prediction on new data before the student registering the pharmaceutical course whether quit or graduated.
- iv. To evaluate the machine learning classifier to detect the pharmaceutical of student performance.

#### 1.4 SCOPE

The scope of this research is discussing are described as follow:

- i. Scope User friendly and feasibility
  - Can load your finalized model and use it tom make prediction new data
  - To train to test the dataset for your machine learning model from WEKA in the JAVA programming
- ii. User Student
  - All student takes subject in pharmaceutical course.

#### **1.5 THESIS ORGANIZATION**

This thesis consists of five chapters including chapter one until chapter five. Based on chapter 1, we shall discuss an introduction to the project to describe briefly about the project related to the issues. The next part is we shall discuss the problem statement related to the suitable issue before you trying to solve the problem. Next part is we shall discuss the objective to be achieved with the goal of the project. Another part is we shall discuss the scope which is the boundary of the research and users. Then lastly part is thesis organization to the summary of each chapter.

Based on chapter 2, we shall discuss the literature review. We must make reviews and explain this system. This chapter is to explain in details and make a comparison between the techniques or method or hardware or technologies which are suitable to be adapted into the research, and also discuss related works such as a domain, technique, methods and framework.

Based on chapter 3, we shall discuss the methodology. This chapter is to discuss the overall approach to covering the method, technique to be used. Therefore, the project implementation will be shown including system flow.

Based on chapter 4, we shall discuss result and discussion. This chapter is to discuss the development process between the data collection and analysis phase. This part can perform testing on every module and carried the result and knowledge analysis including development process, data collection that used to prediction, results from the output from WEKA to Java highlights, test use specification will be explained more details include with result and discussion.

Based on chapter 5, we shall discuss the final conclusion for the research. This chapter is to discuss concerning overall conclusion concerning this research. The outline can conclude the benefits and quality of the research.

#### **CHAPTER 2**

#### LITERATURE REVIEW

#### 2.1 INTRODUCTION

Machine Learning is a set of techniques used to apply the machine learning algorithm to predict the performances of students before registering the pharmaceutical course. The most important factors are to identify for predicting of pharmaceutical student performance as well as we know more considerable interest to pharmacy educator [2].

The prediction method to predict student performance must use the several machine learning algorithms that has three types such as classification, regression, and category. It seems to be mostly as chosen as a classification that suitable to apply the machine learning algorithm [3]. There are several method algorithms will use that have been applied to predict student performance using Machine Learning is Multilayer Perceptron (MLP).

#### 2.2 EXISTING TECHNOLOGIES USING MACHINE LEARNING

This research to discuss the existing technologies using machine learning for previous studies in technology, literature, there are three examples of existing technologies using machine learning to predict student performance and previous thesis by technologies using Machine Learning

**Title of Thesis** Criteria Details No. References Predicting To predict the final grade of Predict Postgraduate postgraduate students Student's Naïve Bayes and 1-NN [2] Method Performance Outcomes Result - total accuracy. To predict the student is successful Predict Using Machine or not, based on other properties Linear Regression, Decision Tree Learning to Method [4] Predict Student & Naïve Bayes Performance Result - the effectiveness of Outcomes different algorithm and method То predict pharmacy student Predicting Predict academic performance using many Academic independent variables Performance of Data collection (Questionnaire) Pharmacy Method [5] Data analysis (PC-SAS) Students : Result - Correlation matrix of Demographic overall student, correlation analysis Outcomes Comparisons academic success demographic To prediction student performance Predict with pharmacy student whether the Predicting of student is quit or graduate. Student Machine Learning Performance Method [This study] **Classifier : Multilayer Perceptron** using Machine Result - Display status prediction Learning Outcomes (graduated or quit) from loaded model object (WEKA Explorer and JAVA)

Table 2.1: Previous thesis by different technologies using Machine Learning

#### 2.3 RELATED WORKS

This thesis briefly to describe the effectiveness of different machine learning algorithm and method will use that have been applied to the prediction of student performance using Machine Learning.

#### 2.3.1 Machine Learning Classifier

Machine Learning is a set of techniques used to apply the machine learning algorithm to predict the performances of students in the pharmaceutical course depending on demographic variables. In order, the machine learning algorithm is divided into two learning types are supervised and unsupervised learning.

#### 2.3.1.1 Supervised Learning

Based on the supervised learning, the classification, whereas for each data set of training data are labeled with the category also is known as input data. The task of the algorithm by creating a model, it can predict one property by other properties. After this created, the algorithm can use to process data that has the same class structure as input data [4]

#### 2.3.1.2 Unsupervised Learning

In contrast with supervised learning, the unsupervised learning just test data only demand input data does not have a class structure that has not been labelled, classified or categorized. Instead of the task of the algorithm is to reveal a structure in the data [10]

#### 2.4 PREDICTION USE ALGORITHM /METHOD /TECHNIQUE

In educational by machine learning method, this reason is to predict of the modelling is usually from the student in the pharmaceutical course to make a prediction of student performance [3]. There are types of the method and algorithm that used in prediction of student performance is such as Multilayer Perceptron.

#### 2.4.1 Multilayer Perceptron (MLP)

Some researchers have used this technique because this is suitable for classification functions to study focus on classification based on an algorithm on pharmaceutical's course for student data. Most importantly, when a classification function can be considered a return on prediction, the multilayer perceptron to have a measure of how the relationship between sets with the attribute values can be determined into one or more classes when used in multilayer perceptron to predict student performance [7].

#### 2.5 EXISTING FEATURES

#### 2.5.1 Comparison between existing features of prediction

This section briefly describes the features of student performance in previous work. The purpose of my research the aim is to find the best method and algorithm for make prediction of student performance in the pharmaceutical course [4]. This thesis also consists of many similarities of the previous work a study was made by researchers (See Table 2.2)

#### 2.6 MACHINE LEARNING CLASSIFIER

#### 2.6.1 Example of Machine Learning Classifier

This section briefly describes the comparison between different existing techniques and example of machine learning classifier for prediction of student performance in previous work. This thesis also consists of many similarities of the previous work studies were made by researchers (See Table 2.3).

## Table 2.2 Result related the Features of Student Performances

Motivation	List of the Problem	Objectives	Purpose	Result	Authors
Pharmacy student's academic is the most important in the pharmacy programs.	To predict pharmacy students' academic performance using many independent variables	To compare the predictive power of these independent variables and demographic variables	Apply the Data collection such as questionnaire, Demographic variables and Data Analysis – PCSAS (Personal Computer Statistical Analysis Software)	Correlation analysis of academic success by demographics	[5]
In education, the application of machine learning algorithm to predict student performance.	To predict whether the student is successful or not, based on other properties.	To find the best method to evaluate a variable approach for predicting student dropout or performance.	Apply the Machine Learning to predict a student will be successful or not.	To evaluate the effectiveness of different machine learning algorithm and methods	[4]
Pharmacy course is the important course as it involve human lives	Many pharmacy student is failed. Student still don't know that they still suitable with the course or not. In case, many patient increase even the have new technology (Moore' Law)	To identify and make prediction student performance whether the student is quit or graduate.	Apply the Machine Learning the pharmacy student is quit or graduate.	For this result by the accuracy from WEKA's were used. Predict result from WEKA to JAVA code after the loaded model.	[This Study]

Table 2.3Example for Machine Learning Classifier in the Previous Studies

References	Machine Learning Classifier	Predict Student for	Year of Journal
[1]	Decision Tree (DT), Artificial Neural Network (ANN), Naïve Bayes (NB), Bayesian (B), KNearest Neighbor (KNN), Ruler Based Learner (JRIP) & Perceptron based Neutral Network [WINNOW]	Student in Academic Organization	2015
[2]	Naïve Bayes (NB), K-Nearest Network (K-NN), Sequentially Minimal Optimization(SMO), J48, Random Forest (RF), Repeated Incremental Pruning to Produce Error reduction (RIPPER) & Ruler Based Learner (JRIP)	Final grade Postgraduate student	2003 - 2006
[4]	Linear Regression (LR), Decision Tree (DT) & Naïve Bayes (NB)	Student	2017
[This Study]	Multilayer Perceptron (MLP)	Pharmacy Student	2019

#### **CHAPTER 3**

#### METHODOLGY

#### 3.1 INTRODUCTION

This section will be focused on the methodology that suitable to survey in research that related to predicting student performance using machine learning. In other words, this research can show the overall workflow of the study experiment.

In this chapter, it will discuss all the necessary information need in this research. The methodology used in Predicting of Student Performance in WEKA and approach in JAVA development tools will discuss in this chapter. The Gantt chart show the duration and the research phase also present in this chapter.



Figure 3.1 Flowchart of Research Methodology

Figure 3.1 shows the research methodology for this research from phase 1 until phase 5. There are preliminary studies performed in Phase 1, literature review that performed in Phase 2, knowledge analysis that performed in Phase 3, methodology stages that performed in Phase 4 and validate prediction that performed in Phase 5. These will discuss in details about it.

#### **3.2.1** Phase 1: Preliminary Studies

This phase 1 has discussed to identify the research for introduction, problems statement, objective and scope of the study. It also discussed how they predict student performance by using machine learning to evaluate and predict student performance to know the status that suitable if the pharmacy student whether quit or graduated. So, they can predict themselves when they know the result, it is not impossible that they can improve their learning and get the best result, no matter where they come from groups before they registering the pharmaceutical course.

#### **3.2.2** Phase 2: Literature Review

This phase 2 has discussed to identify the main research concept, to analysis and compare the several techniques or method target variable, algorithm, method and related work in predicting student of performance. Next, phase 2 also to study and understand all the previous work studies including all the technique that used in previous work. Lastly, phase 2 is to make a review and analyse existing technologies using Machine Learning and how they work it. Review previous research finding of prediction student performance using Machine Learning still in the literature review phase.

#### 3.2.3 Phase 3: Knowledge Analysis

This phase 3 has discussed to define the technique of classifier machine learning and software or hardware that suitable for making the prediction of student performance. This phase also makes analysis phase which is analysed from WEKA to generate the result that approach in Java Programming which means only one Machine Learning classifier chosen to prove the efficiency

#### 3.2.4 Phase 4: Methodology Phase

This phase 4 has discussed will be focused on the methodology that suitable to survey in research related to predicting student performance using machine learning. In other words, we can show the overall the workflow of the study experiment. There are many stages that used in predicting student performance such as data collection, features investigation, Machine Learning classification and performance prediction to evaluate these characteristics used by the machine learning. The stages of details of the research are described as follows:

#### i) Data Collection

- It focused on how the classification based on the algorithm on pharmaceutical's course can be used to identify the attributes in student data (dataset).
- A dataset that contains student data and other information just recently in real-world data were collected relate UMP student (FTEK Faculty)

- Selection of student data also is random. It organized into divided into certain types of data structures to help the study experiment which to examine the application.
- Example of data collection is gender, early qualification, gender, state, and result MUET.
- ii) Features Investigation
  - It focused on the process of selecting the variable to improve data result from the Machine Learning from WEKA to JAVA highlight.
  - To optimize the detection to select unique features because they use the features when the Machine Learning makes predicting student performance.
  - Example of features investigation is when you set to value 0 for data automatically they will predict as "graduated" and when you set value is 1 for data automatically they will predict as "quit". This means the features can be done when observing from the output display whether the pharmacy student will be quit or graduate.

#### **3.2.5** Phase 5: Prototype Design

This phase 5 has discussed to develop a prototype based on machine learning classifier was chosen that approach in JAVA programming. During the prototype designing the flow for the algorithm, actually, the research must need to analyse and identify the step by step from the process initial until the process finish. The first step is, getting input from the user and next is generate the resulting output. The prototype design is developed using JAVA programming (NetBeans) and actually WEKA tools also in JAVA language

#### 3.2.6 Phase 6: Validate Prediction

This phase 6 has discussed to predict new samples using pre-computed and exported from a loaded model from WEKA to JAVA programming. The validation process is to evaluate the final result to know the status that suitable if the pharmacy student whether quit or graduated.

## 3.3 SYSTEM DESIGN FOR MACHINE LEARNING IN WEKA TOOLS

As being we have mentioned in Chapter 3 in performance prediction, WEKA and JAVA are being used for developed and create a prototype. This software is an essential part to proven research to implement from WEKA to Java Programming environment with WEKA to the prototype to show how to store and load model, then manipulate them and use them to evaluate data to make prediction pharmacy student performance. There are the figures of WEKA GUI Chooser as below:



Figure 3.2: WEKA GUI Chooser Interface

Figure 3.2 shows that the interface main for WEKA GUI Chooser. When we open the WEKA, the user is given a small window with four buttons labelled application.

Wieka Explorer	and the state and the state			- ¤ ×
Open file. Open Ufit.		rate	n (. 6.8.	1. Sec.
itor				
Choose None				APP: U.B.P.
arrest relation		Selected attribute		
Relation: None Instances: None	Attributes: None Sum of weights: None	Name None Masing None	Weight None Distinct None	Type: None Unique: None
ttributes				
	nuve			Visualize J
status				
Welcome to the Weika Explorer				L00.

Figure 3.3: The WEKA start up box

Figure 3.3 shows a WEKA interface after the user selecting Explorer to describes the various data modelling process.



Figure 3.4: Finalize a Machine Learning Model in WEKA

Figure 3.4 WEKA also can read in a variety of file types including CSV files. The pharma student is a dataset that used and available as ARFF file also make the file the current dataset in WEKA. Just only one dataset can be in memory at a time.
Weka Explorer									-		~
Preprocess Classify Cluster Au	sociale   Select attribute	is Viouaio	10								
Classifier		111	ni.								
Choose Logistic -R 1 DE 8 M -1	num decimal places 4										
Test options	Classifier output										
Use having set     Supplechest set     Gross-validation     Percentage apit     More options	Correctly Class Incorrectly Class Nappa statistic Hean absolute o Boot mean squar Relative absolu Boot relative a Total Bumber of	ified Inst solfled In rror ed error guared err Instances	iances latances	35 11 0.51 0.20 0.30 57.61 70.63 46	55 64 17 05 8 105 8	76,087 28,518	:				Î
(Nom) STATUS		TUTNEY By TUTNEY 0.763 0.760 0.761 atris sssified sssified sssified sssified	TF Rate 0.240 0.220 0.229	Precision 0,727 0,792 0,763	Recall 0.762 0.761 0.761	<b>F-Heasure</b> 0.744 0.776 0.761	HCC 0.510 0.520 0.520	ROC Area 0.870 0.870 0.870	FRC Ares 0.040 0.923 0.024	CLASS GRADUATE QUIT	
States	. [.t.										
OK									L	-	8

Figure 3.5 starting the modeling process and saving a model

Figure 3.5 show the contents of WEKA once the user makes train initial model use the current dataset and then the user clicks the "Classify" and choose the model type is Logistics and click start to start the modeling the process. Save the model as "data1.model" on your disk.

Weka Explorer		-		×
Preprocess Classify Cluster Associate Set	ect attributes [ Visuatue ]			
lassifier				
Choose ZoroR				
wat options	Classifier output			
C Use training set				14
C Supplied test and Sul				1
	Odds Ratios			- 11
Croos-validadoli Folde III	NOTATION INC.	Class		- 82
🕗 Percentage split 🛸 👐	Vatiable	ORADUATES		- 83
thus antions	ENTRY QUALIFICATION-MATRICULATION	8983603.7037		- 81
	ENTRY_OCALIFICATION-DIPLONA	0		- 8
	ENTRY_QUALIFICATION+STEN	#714927.4127		- 8
(Nom) STATUS	HESOLI MAL	0.2343		- 81
	GENDER-F	2.7832		- 81
Gitart 11000	STATESTATEST	0.0472		- 81
would live transfel other for and seen	STATE-Panang	3.6663519919013495813		- 8
and the fright cack for special	STATE-Selangur	0.0407		- 6
18 29:48 - functions Logistic from file 'data model'	STATE+Johor	0.1130		
A CONTRACTOR OF	STATE-Degeri Semmilan	0.0338		
	STATE=Serevek	0		
	STATE=Poleu Finang	0.0644		
	STATE=Kelenten	0.0430		
	STATE-Sedan	0,0125		
	STATE-Sabab	0.0176		
	BUBLIC CONTRACTOR	0.6427		1
			_	
	iliux.	11		2.20
Tatus				
		Log	10000	10

Figure 3.6 WEKA model Load a Finalized Model ready for use

Figure 3.6 shows to load saved WEKA models from your file. We can use the loaded the model to make a prediction for new data. A user just right clicks on the Result List and click "Load Model", select the model that saved in the previous section as "data1.model.



Figure 3.7 WEKA dataset copy with edit Notepad++

Figure 3.7 shows WEKA dataset for making a new prediction on new data using edit Notepad++. Users just keep one record only and move down1lines only. Then the class output (output variable) and replace with a question mark as a symbol (?).



Figure 3.8 WEKA dataset for make prediction on new data by a Loaded Model

Figure 3.7 shows shows WEKA dataset to making new data on "unseen" data with no unknown output to make a prediction. User will select "Supplied Test Set" and click "Open File" to add new dataset with the name "data1.arff" and close. Then, the user must click the "More Option" button to uncheck the information and click OK. So the output displays the predictions for each test instance are then listed in the "Classifier Output" pane

# 3.4 SYSTEM DESIGN FOR MACHINE LEARNING IN JAVA PROGRAMMING

In this section, we have mentioned in Chapter 3 in performance prediction, Java programming is also being used for developed and create a prototype to make prediction student performance. This software is an essential part to proven research to implement an approach for programming in Java to WEKA environment that used in the research prototype to print out prediction with WEKA in Java Programming such as, train and test the data and predict the new data. There are the figures of a framework for Java programming as below:

#### 3.4.1 Use Training Test



Figure 3.9 (i) Coding for programming in Java to use training set to WEKA



Figure 3.9 (ii) Coding for programming in Java to use training set to WEKA

Figure 3.9 (i) and (ii) shows coding for programming in Java to use the training set to WEKAto know how to interact with the WEKA API for the first time with a simple Java code. In this code, we have loaded an ARFF file called 'pharma student.arff' and then used Multilayer Perceptron classifier with a 10 fold CV setup. So, we showed the standard output of WEKA on the JAVA output as well as the F-score, precision, and recall of the 10 fold CV.



Figure 3.10 Edit with NotePad ++ to Use Training Set

Figure 3.10 shows coding for programming in Java to use the training set to WEKA to know how to interact with the WEKA API for the first time with a simple Java code

#### 3.4.2 Predict New Samples using Exported Model Object

```
package wekatester;

import java.util.ArrayList;

import java.util.List;

import weka.classifiers.Classifier;

import weka.core.Attribute;

import weka.core.DenseInstance;

weka.core.Instances;
```

Figure 3.11 Header files for WEKA in Java script Programming

Figure 3.11 shows the header files for WEKA that used in Javas script programming such as WEKA classifier, WEKA attributes, WEKA dense instance and WEKA instance. User must right click button the "Libraries" from the "Project" and add "JAR/Folder" name as weka.jar.

```
public class Wekatester
{
  public static void main(String[] args)
  {
    new Wekatester().main();
}
```

Figure 3.12 Class name for WEKA in Java script Programming

Figure 3.12 shows the class name for WEKA that used in Java script programming name as "WekaTester". So, the package name and the class name is same name variable.

```
public void main()
{
    // we need those for creating new instances later
    final Attribute attributeEntryQualification = new
    Attribute("ENTRYQUALIFICATION");
    final Attribute attributeResultMuet = new
Attribute("RESULTMUET");
    final Attribute attributeGender = new Attribute("GENDER");
    final Attribute attributeState = new Attribute("STATE");
    final Attribute attributeBumiputra= new Attribute("BUMIPUTRA");
    final List<String> status = new ArrayList<String>() {
        { add("GRADUATED");
        }
};
```

Figure 3.13 WEKA attributes in Java script Programming

Figure 3.13 shows the WEKA attributes for WEKA that used in Java script programming. The user make create the new instance or attributes from current dataset such as entry qualification, result MUET, gender, state, "bumiputra" and status

```
// Instances(...) requires ArrayList<> instead of List<>...
ArrayList<Attribute> attributeList = new ArrayList<Attribute>(2){{
    add(attributeEntryQualification);
    add(attributeResultMuet);
    add(attributeGender);
    add(attributeState);
    add(attributeBumiputra);
    Attribute attributeStatus = new Attribute("@@status@@",
    status);
    add(attributeStatus);
    } };
```

Figure 3.14 WEKA [Array List] for each attributes in Java Programming

Figure 3.14 shows the WEKA [arrayList] for each attributes for WEKA that used in Java script programming. The user must added all attribute instances requires to the [arrayList].

// unpredicted data sets (reference to sample structure for new
instances)

Instances dataUnpredicted = new

Instances("TestInstances",attributeList, 1);

// last feature is target variable
dataUnpredicted.setClassIndex(dataUnpredicted.numAttributes() - 1);

Figure 3.15 Unpredicted data sets and last feature in Java Programming

Figure 3.15 shows the unpredicted data sets for reference to sample structure for new instances and last feature (target variable) in Java script Programming.

Figure 3.16 Create new instances from data model in Java Programming

Figure 3.16 shows the user want to create new instances and set value is 0 for each attribute lines. This is because the Java script programming used [numAttributes()].

// instance to use in prediction
DenseInstance newInstance = newInstanceStatus;

Figure 3.17 Declaration of instance to use in prediction in JAVA

Figure 3.17 shows the declaration of instance to use in prediction and reference to dataset in Java script Programming use import WEKA [DenseInstance] in Java script programming. // reference to dataset

newInstance.setDataset(dataUnpredicted);

Figure 3.18 Declaration of reference to dataset in JAVA

Figure 3.18 shows the declaration of reference to dataset in Java script Programming use import WEKA [NewInstance] and [setDataset] in Java script programming.

# 3.5 HARDWARE AND SOFTWARE DEVELOPMENT

## 3.5.1 Hardware Development

This thesis will focus about the hardware development used in the research (Refer Table 3.2)

|--|

Type of Hardware	Purpose
Personal computer (PC) /	Device to write the research and develop the
Laptop	system
Printer	Device to print the search paper
Pendrive / Hardisk	Devices to save the journal and research paper

This thesis will focus about the software development. There are list of software development that uses in the thesis. (Refer Table 3.3)

Table 3.3 shows the software used in the research

Type of Software	Purpose
Microsoft Office Word 2010	Application to write the research and develop
	the system
Windows 7 Ultimate	Type of operating system
Microsoft Project 2010	Application to develop the Gantt Chart
	Application to collect the student data and
Microsoft Excel 2010	other information dataset use number and
	string
WEKA tools	Application to predict the data
Mendeley Desktop	Application to insert citation to make reference
Grammarly / Grammar	Application to check grammar in the thesis
Checker Ginger	
Netbeans	Application to make a prototype

### 3.6 GANTT CHART

The Gantt chart has shown the estimate duration from the research start until the research end. (Refer Appendix B) **CHAPTER 4** 

#### **RESULT AND DISCUSSION**

### 4.1 INTRODUCTION

Continuously from the previous chapter, this chapter is presented to display the result of the finding based on the prototype like a system to predict student performance to predict a new samples using pre computed and exported model object from WEKA in Java development and testing that have been done to help achieved this research objective. Besides, this research also has a test case specification was conducted on the developed the research prototype

## 4.2 **RESULT OUTPUT PREDICTION**

### 4.2.1 Result Prediction from WEKA Explorer

assifier output									
Stratified o	eross-vali	dation	-						
Summary									
Correctly Class	fied Inst	ances	29		63.0435				
Incorrectly Clas	sified In	stances	17		36.9565				
Kappa statistic			0.25	24					
Mean absolute en	ror		0.37	87					
Root mean square	ed error		0.55	12					
Relative absolut	e error		76.11	86 1					
Root relative squared error			110.4069 %						
Total Number of	Instances		46						
Detailed Acc	suracy By	Class	č.						
	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.571	0.320	0.600	0.571	0.585	0.253	0.677	0.627	GRADUATED
	0.680	0.429	0.654	0.680	0.667	0.253	0.677	0.759	QUIT
Weighted Avg.	0.630	0.379	0.629	0.630	0.630	0.253	0.677	0.699	10100-0010
Confusion Me	trix ===								
$\begin{array}{cccc} a & b & < cla \\ 12 & 9 &   & a & = GI \end{array}$	Assified a								
8 17   b = Q0	JIT								

Figure 4.1 Result from use training set use in WEKA

Figure 4.1 shows the result from use training set use in WEKA. Actually, the results also display with same result from the WEKA to Java Programming.

#### 4.2.2 Result Prediction in JAVA Programming



Figure 4.2 Result from use training set use in Java Programming

Figure 4.2 shows the result from use training set use that approach in JAVA programming. Actually, the results also display with same result from the WEKA Explorer.

#### 4.2.3 Result Predict New Sample use the classifier

```
// import ready trained model
Classifier cls = null;
try {
  cls = (Classifier) weka.core.SerializationHelper.read("d:/new1.model"); }
  catch (Exception e) {
    e.printStackTrace(); }
    if (cls == null)
return;
```



Figure 4.3 shows the declaration of instance to use in prediction and reference to dataset in JAVA Programming. The following example in JAVA programming has two aspects of this process, such as loading such a model object, how to format a new, yet unclassified instance so that its target variable can be predict using the loaded model

```
// predict new sample
try{
    int result = (int) cls.classifyInstance(newInstanceStatus);
    System.out.println("Index of predicted class label: " + result + ", which
    corresponds
    To status: " + status.get(new Double(result).intValue())); }
    catch (Exception e) {
        e.printStackTrace(); }
    }
}
```

Figure 4.4 Declaration to predict new sample in JAVA Programming

Figure 4.4 shows the declaration to classify the instance after loaded model so that its target variable can be predict using the loaded model in JAVA Programming. The following example in JAVA programming has two aspects of this process, such as loading such a model object and how to format a new sample for the make prediction.

# 4.2.4 Result Output After Prediction

Table 4.1 shows the result output after make prediction. After the user click the "Run" button, the output display is "Index of predicted class label: 0, which corresponds to status: GRADUATED". The result just displays one lines result because the data model just one keep record only. (Refer table 4.1)

Table 4.1 Declaration to predict new sample in JAVA Programming

	Output - JavatesterWeka (run) 🕷
First Trial using JAVA	<pre>run: The Number of Instances: 0 The status you have: [Graduated, Quit] Predict of your result is: Graduated BUILD SUCCESSFUL (total time: 0 seconds)</pre>
First Trial using WEKA	<pre>=== Predictions on test set === inst# actual predicted error prediction 1 1:? 1:GRADUATED 0.518 === Evaluation on test set ===</pre>
	Output - JavatesterWeka (run) %
Second Trial using JAVA	run:         The Number of Instances: 0         The status you have: [Graduated, Quit]         Predict of your result is: Graduated         BUILD SUCCESSFUL (total time: 0 seconds)

### 4.3 TEST CASE SPECIFICATION

This test specification yield the actual result based on WEKA and Java Programming. Table 4.1 shows the list of test case specification of research prototype using WEKA and Java. This software is an essential part to proven research to implement from WEKA to Java Programming environment.

Table 4.2 Test Case Specification

Test Case	Test	Test Case	Pre -	Tost Stop	Expected	A atual Dagult	Status
Scenario	Case ID	Test Case	Condition	Test Step	Result	Actual Kesult	Status
Predict of	TC01	Check the search	Check	1. Search WEKA GUI	Search WEKA	Search	FAILED
student		functionality when	menu	Chooser platform	functionality	functionality	
performance		you open file and	search has	2. Click the "Explorer"	should not	not work when	
using WEKA		entered file saved	been	3. Click the open file	work when file	file cannot	
platform		in *.xlsl (excel) or	opened in		cannot entered	entered to the	
		*.txt (text)	user's		to the WEKA	WEKA	
			desktop		Explorer	Explorer	
	TC02	Check the search		1. Search WEKA GUI	Search WEKA	Search	PASS
		functionality when		Chooser platform	functionality	functionality	
		you open file and		2. Click the "Tool"	should can	can work when	
		entered data files		3. Click the "ArffViewer"	work when	data files can	
		in saved in *.arff		4. Click "File" and	data files can	entered to the	
		(Attribute-		"Open" the dataset.	entered to the	WEKA	
		Relation File		5. Click "Save as" in arff	WEKA	Explorer	
		Format)		data files (*.arff)	Explorer		

Test Case	Test	Test Case	Pre -	Test Stop	Expected	A stual Desult	Status
Scenario	Case ID	Test Case	Condition	Test Step	Result	Actual Result	Status
Predict of	TC03	Check the results	Check	1.Search WEKA GUI	Total number of	Total number of	PASS
student		to be displayed in	menu	Chooser platform.	results should be	results less than	
performance		first predict new	search has	2.Click the "Explorer"	less than 1 line	1 line record for	
using WEKA		sample after load	been	3.Open files dataset	record for first	first predict new	
platform		models	opened in	4. Click the "Classify"	predict new	sample in the	
-			user's	5.Right clicked from the	sample in the	WEKA platform	
			desktop	"Result List". Click	WEKA platform		
				"Load Model" and add			
				"data.model"			
				7.Click "Supplied Test			
				Set" and click "Set". Then			
				click "Open File" entered			
				new dataset and close.			
				8.Click "More Options"			
				and uncheck list			
				a) Output model			

				b) Output per-class-stats			
				c)Output confusion matrix			
				d)Store prediction for			
				visualization			
				9.Choose the "Output			
				Predictions" from Null to			
				Plain text. Click OK			
				10. Click "Start"			
Predict of	TC04	Check the header	Check	1.Open a Netbean	The developed	The developed	FAILED
student		files that suitable	menu	2.Not entered keyword for	prototype	prototype	
performance		for the Java script	search has	the header files that	should not	should not	
using JAVA		programming	been	related to the Javascript	work on the	work when	
platform			opened in	programming	header files to	header files	
			user's		the Java Script	cannot add	
			desktop	Keyword = "weka"	Programming	keyword to the	
						Java Script	
						Programming	

Predict of	TC05	Check the header	Check	1. Open a Netbean	The developed	The developed	PASS
student		files that suitable	menu	2.Entered keyword for the	prototype	prototype	
performance		for the Java script	search has	header files to the	should can	should can	
using JAVA		programming	been	Javascript programming.	work on the	work when	
platform			opened in	Keyword = "weka"	header files to	header files	
			user's	3.Right click the	the Java Script	can add	
			desktop	"Libraries" and click	Programming	keyword to the	
				"JAR/Folder" and choose		Java Script	
				the "weka.jar"		Programming.	
				4. Click OK			
Predict of	TC06	Check the results	Check	1. Add 1 record data only	Total number	Total number	PASS
student		to be displayed in	menu	2.Set value 0 for each	of results	of results less	
performance		predict a new	search has	attributes	should be less	than 1 line	
using JAVA		samples using pre	been	3.Declare the classifier	than 1 line	record for first	
platform		computed and	opened in	from exported data model	record for first	predict new	
1		exported model	user's	1	predict new	sample in the	
		1			1	1	
		object from	desktop		sample in the	research	
		object from WEKA in Java.	desktop		sample in the prototype	research prototype	
		object from WEKA in Java.	desktop		sample in the prototype (JAVA)	research prototype (JAVA)	
		object from WEKA in Java.	desktop		sample in the prototype (JAVA)	research prototype (JAVA)	

Λ	1
+	T.

TC07	Check the results	Check	1. Add more than one	Total number	Total number	FAILED
	to be displayed in	menu	record data.	of results	of results more	
	predict a new	search has	2.Set value 0 for each	should be more	than 1 line	
	samples using pre	been	attributes	than 1 line	record for first	
	computed and	opened in	3.Declare the classifier	record for first	predict new	
	exported model	user's	from exported data model	predict new	sample in the	
	object from	desktop		sample in the	research	
	WEKA in Java.			research	prototype	
				prototype	(JAVA)	
				(JAVA)		
	TC07	TC07 Check the results to be displayed in predict a new samples using pre computed and exported model object from WEKA in Java.	TC07Check the results to be displayed in predict a new search has samples using pre computed and exported model object from WEKA in Java.Check menu menu been opened in desktop	TC07Check the results to be displayed in predict a new samples using pre computed and object from WEKA in Java.Check menu menu search has been opened in desktop1. Add more than one record data.TC07Check the results menu search has to be displayed in search has been opened in desktop1. Add more than one record data.TC07Check the results menu search has to be displayed in search has to be of the classifier from exported data model	TC07Check the results to be displayed in predict a new samples using pre opened and object from WEKA in Java.Check menu1. Add more than one record data.Total number of results1. Add more than one record data.Total number of of resultsOf results1. Add more than one record data.Total number of of results1. Add more than one record data.Total number of of results1. Add more than one record data.Total number of of results2. Set value 0 for each attributesshould be more than 1 line record for first from exported data model1. Add more than one samples using pre object from WEKA in Java.Image: Set value 0 for each attributes1. Add more than one attributesImage: Set value 0 for each attributes1. Add more than one samples using pre opened in attributesImage: Set value 0 for each attributes1. Add more than one attributesImage: Set value 0 for each attributes1. Add more than one attributesImage: Set value 0 for each attributes1. Add more than 1 line record dataImage: Set value 0 for each attributes1. Add more than 1 line record dataImage: Set value 0 for each attributes1. Add more than 1 line record dataImage: Set value 0 for each attributes1. Add more than 1 line record dataImage: Set value 0 for each attributes1. Add more than 1 line record for first protocybe (JAVA)1. Add more than 1 line record for first protocybe (JAVA)	TC07Check the resultsCheck1. Add more than oneTotal numberTotal numberto be displayed in predict a new samples using pre computed and opened in opened in 3.Declare the classifier object from WEKA in Java.1. Add more than one record data.Total number of results of results more should be more than 1 line record for first predict new sample in the research prototype (JAVA)

Table 4.1 shows the final test case after predict a new samples using pre computed and exported model object from WEKA Explorer in the Java programming on the development and testing that have been done to help achieved this research objective in the developed as prototype name as 'WEKATESTER.java. There are two test case scenarios with seven test cases that have been listed out. Each test case has been provided precondition, test step, expected result and actual result.

### **CHAPTER 5**

### CONCLUSION

### 5.1 INTRODUCTION

:

This chapter discussed that features involve with the methodology stages such as data collection, feature investigation, Machine Learning classification and performance prediction that have been done, conclusion for research work will presented , future suggestion and enhancement of implementing a prototype design using WEKA tools to approach in JAVA highlight..

The aim for the research was to develop a research prototype with implementing using JAVA code from WEKA tool which is the result of prediction is same and both use the classifier to predict of student performance. Research for the existing tool for the suggestion also helped to develop the prototype for this research. The thesis use multilayer perceptron to make prediction. Then, the thesis also makes many references and the example to know how to save machine learning model and make prediction use WEKA to approach in JAVA programming code

The methodology, system design and implementation of prediction of student performance using machine learning from WEKA tools to JAVA programming code technique in the developed the simple a prototype to prove and predict new sample that used in exported loaded model that already save from WEKA tool to approach in JAVA programming code work as expected in this thesis. The testing includes the machine learning classifier suggestion from the existing tool. As a conclusion as been mentioned from the previous chapter which are literature review such as existing technology in machine learning, comparison between existing use features investigation to prediction of student performance and machine learning classifier that used in the thesis. Lastly, the result obtained show that the JAVA programming code technique was correctly to implement to predict new sample that used in exported loaded model that already save from WEKA tool to approach in JAVA programming.

### 5.2 RESEARCH CONSTRAINT

A research constraint is Constraints in research are a restriction that defines a project's limit. in other words, the limit of the project. For example, the scope includes one example of the limit in which the thesis expects before implementing the prototype. There are two of research constraint. First constraint for this research is limited time for time constraints. For example, we not enough the time to develop a prototype to link to the GUI (Graphical User Interface). The research takes long time to added linked to the GUI which is have many errors when the prototype wants to runs application. Second constraint for this research is configure the parameter to the JAVA code for reliability of the data. Lastly, all of the constraint for this research will be recommendation for improvement for future work.

### 5.3 FUTURE WORK

There are part discussion suggestion that can be taken into consideration to enhancement of the research including knowledge and contribution to the University, faculty, society or the researcher throughout the research. There are several enhancements that can be carried out for future improvement of prediction of student performance as follow:

- a) Develop in GUI (Graphical User Interface) for software development.
- b) Implementing in mobile application. This function is all student from different faculty can access the application every time and everywhere.
- c) Create an application to connect with the server CLOUD. This function is can add many data and feature for your future work such as result for another subject. Advantages is safe security and faster

#### REFERENCES

A. H., G. F., & C. C. (2010). From Search Engine Optimization to Search Engine Marketing Management: development of a new area for information systems research. *Information Systems, Organisation and Society (ISOS) research centre*.

Al-Qesi, K., Dennis, C., Alamanos, E., & Jayawardhena, C. (2014). Website design quality and usage behavior: Unified Theory of Acceptance and Use of Technology. *Journal of Business Research*, 2282-2290

Beal, V. (2016). White Hat SEO (search engine optimization). Webomedia

Chen, C. Y., Shih, B. Y., Chen, Z. S., & Chen, T. H. (2011). The exploration of onternet marketing strategy by search engine optimization: A critical review and comparison. *African Journal of Business Mnagement*, 4644-4649

Cheng-Jhe Luh, S.-A. Y., & D. H. (2016). Estimating Google's search engine ranking function from a search engine optimization perspective. *Online Information Review*, 239-255

Fiorini, P. M., & Lipsky, L. R. (2012). Search marketing traffic and performance models. *Computer Standards & Interfaces*, 517-526

Ghose, A., & S. Y. (2009). An Empirical Analysis of Search Engine Advertising: Sponsored Search in Electronic Markets. *Management Science*, 1605-1622.

Hillard, D., Schroedl, S., Manavoglu, E., Raghavan, H., & Leggetter, C. (2010). Improving Ad Relevance in Sponsored Search . *Proceedings of the third ACM international conference on We search and data mining*, 361-370. Katona, Zsolt, & Sarvary, M. (2010). The Race for Sponsored Links Bidding Patterns for Search Advertising. *Marketing Science*, 199-215.

Langville, L.N. and Meyer, C.D. (2006), *Google's PageRank and Beyond: The Science of Search Engine Rankings*, Princeton University Press, Princeton and Oxford.

Lewandowski, D. (2011). The retrieval effectiveness of search engines on navigational queries. *Aslib Proceedings*, 354-363

Lieberman, M. (2013). Onsite VS Offsite Search Engine Optimiation-What's The Right Mix? *Inbound-The Blog:The Secrets Behind Click To Close*.

M. C., & Wong, C. H. (2010). Designing the user interface and functions of a search engine development tool. *Decision Support System*, 369-382.

Macey, A. (2016). The Differences Between Direct and Organic Search Traffic Sources. *Smart Bug Media* 

Malaga, R. A. (2007). The Value of Search Engine Optimization: An Action Research Project at a New E-Commerce Site. *Journal of Electronic Commerce in Organizations*, 68-82

Malaga, R. A. (2010). Search Engine Optimization-Back and White Hat Approches. 1-39

McCarthy, B. (2015). An introduction to the 3 different types of search marketing. *The Direct Responce Coach* 

McIntyre, E. S. (2015). Search Engine Optimization. MyScienceWork

Nasomyont, T. (2014). A Study on the Relationship between Search Engine Optimization Factors and Rank on Googgle Search Result Page. *Advance Materials Research*, 1462-1466

Nikitha, S. R. (2015). Search Engine Optimization. vRock.

Pawar, V. B., Patil, S. P., & Patil, A. S. (2013). Search Engine Optimization: A Study. *Research Journal of COmputer and Information Technology Sciences*, 10-13

R.S, N. (2015). Search Engine Optimization. vRock.

Ramos, A., & Cota, S. (2009). Search Engine Marketing. New York: McGraw-Hill

Regoloni, A. G. (2011). Web site search engine optimization : a case study of Fragfornet. *Library Hi Tech News*, 6-13

Sohail, A. (n.d.). Search Engine Optimization Methods & Search Engine Indexing for CMS Applications. *Master's Thesis*.

Spink, A., & M. Z. (2008). Web Search. Brisbane: Spinger.

Yalcin, N., & U. K. (2010). What is search engine optimization: SEO? *Procedia Social and Behavioral Sciences*, 487-493

Yang, Z., Shi, Y., & Wang, B. (2015). Search Engine Marketing, Financing Ability and Firm Performance in E-commerce. *Proceedia Compter Science*, 1106-1112.

#### **APPENDICES**

#### **APPENDIX** A

## (Part I)



#### **APPENDICES**

# APPENDIX A

# (Part II)

Keutamaan kepada calon:
Mempunyai kemahiran dalam bidang:
1) Pharmacy
Klasifikasi Bidang:
Syarat Tambahan :
1. Calon mestilah mempunyai Ijazah Doktor Falsafah (PhD) dalam bidang farmasi yang diiktiraf
oleh
Kerajaan daripada institusi-institusi pengajian tinggi tempatan atau kelayakan yang diiktiraf setaraf
dengannya serta berdaftar penuh dengan Lembaga Farmasi Malaysia.
2. Calon hendaklah merupakan Ahli Farmasi Berdaftar yang diiktiraf oleh Lembaga Farmasi Malaysia
<ol><li>Calon mestilah mempunyai kepakaran dalam bidang berikut:</li></ol>
(i) Drug delivery system;
(ii) Pharmaceutical Formulation;
(iii) Quality Management System;
dan mempunyai sub bidang kepakaran berikut:
(i) Clinical Pharmacy
(ii) Pharmaceutical Manufacturing/ Technology/ Engineering/ Sciences
1. Mempunyai pengalaman dalam industri Farmaseutikal sekurang-kurangnya dua (2) tahun; atau
mempunyai pengalaman mengajar di peringkat pengajian tinggi sekurang-kurangnya dua (2) tahun dan ke atas.
2. Mempunyai kemahiran penyelidikan dan mempunyai rekod penerbitan yang cemerlang
3. Mempunyai kemahiran berkomunikasi yang tinggi

# APPENDIX B

# (Gantt Chart)

ID	0	Task Mode	Task Name	Duration	Start	Finish	01 July 01 Octobe	r 01 January	01 April	01 July
1		*	The Research Methodology	238 days	Tue 14-08-18	Thu 11-07-19	•		1	-
Z	1118	13	Planning	28 days?	Tue 14-08-18	Thu 20-09-18				
3		*	Meeting wih Superv	visor 1 day	Fri 21-09-18	Fri 21-09-18	Ĩ			
4		1	Plemininary Studies	7 days	Sat 22-09-18	Sat 29-09-18	0			
5		*	Identify introduction problems statement, objective and scope	7 days	Sat 22-09-18	Sat 29-09-18				
6		*	Meeting with Superv	isor 1 day	Mon 08-10-18	8 Mon 08-10-18	I			
7		*	Literature Review	15 days	Mon 08-10-18	8 Fri 26-10-18				
8		*	Identify the main res concept, target varia algorithm, method a related work.	earch 15 days ble, nd	Mon 08-10-18	Fri 26-10-18				
9		*	Meeting with Superv	isor 1 day	Fri 12-10-18	Fri 12-10-18	I			
10		*	Review the previous	thesis 5 days	Fri 12-10-18	Thu 18-10-18				
11		*	Submission Part Chaj and Chapter 2	oter 1 day	Fri 19-10-18	Sat 20-10-18	I			
12		*	Correction Chapter 1 Chapter 2	and 6 days	Sat 20-10-18	Fri 26-10-18				
13		*	Knowledge Analysis	27 days	5at 27-10-18	Mon 03-12-18	Council of	1)		
14		*	Define the technique classifier machine lea	of 6 days Irning	5at 27-10-18	Fri 02-11-18	٥			
15		*	Meeting with Superv	isor 1 day	Mon 05-11-18	Man 05-11-18	I			
16		*	Methodology Stages	14 days	Tue 06-11-18	Fri 23-11-18	63			
17		*	Submission Abstract Supervisor	to 1 day	Wed 28-11-18	Wed 28-11-18		8		
12			Task		19	External Mileston	e 🔶	Manual Summary	Rollup -	
			Split			Inactive Task	C)	Manual Summary	φ	
Projec	t: gant	ttchart naz	iha Milestor	e 🔶		Inactive Mileston	e O	Start-only	E	
Date:	Fri O3-	05-19	Summar	y 🗭	~	Inactive Summary	0-0	Finish-only	3	
			Project S	ummary 🛡		Manual Task	6 3	Deadline	4	
			External	Tasks 🔤		Duration-only	1811	Progress	_	
						Page 1				

D	0	Task Mode	Task Name		Duration	Start	Finish	01 July 01-07	01 October 12-08 23-09 04-11	01 January 16-12 27-01	01 April 10-03 21-04	02-06	1 July 14-07
18		*	Submission Supervisor	Part Chapter 3 to	2 days	Sat 01-12-18	Mon 03-12-18		I				
19		*	Submissio FSKKP	n PSM 1 to	1 day	Tue 04-12-18	Tue 04-12-18		I				
20	1	*	Prototype I	Design	25 days	Tue 08-01-19	Mon 11-02-19						
21		*	Make predi	ction from WEKA	24 days	Tue 08-01-19	Fri 08-02-19			<b>c</b> 3			
22		*	Design prot	otype using JAVA	24 days	Tue 08-01-19	Fri 08-02-19			c 3			
23	t .	*	Meeting wi	th Supervisor	1 day	Mon 11-02-19	Mon 11-02-19			I			
24		*	Designing L	/ser interface	12 days	Tue 12-02-19	Wed 27-02-19						
25		*	Meeting wi	th Supervisor	1 day	Thu 28-02-19	Thu 28-02-19			I			
26		*	Developme	nt	27 days	Fri 01-03-19	Mon 08-04-19						
27		*	Meeting wi	th Supervisor	1 day	Fri 01-03-19	Fri 01-03-19			I			
28		*	Validate Pri	ediction	46 days	Sat 02-03-19	Fri 03-05-19						
29		*	Predict a us WEKA to JA	e train test from VA programming	27 days	Sat 02-03-19	Sun 07-04-19			6			
30		*	Meeting wi	th Supervisor	1 day	Mon 08-04-19	Mon 08-04-19				I		
31		*	Unit testing code JAVA	; use Weka in my	8 days	Tue 09-04-19	Thu 18-04-19				•		
32	1	*	Meeting wi	th Supervisor	1 day	Fri 19-04-19	Fri 19-04-19				I		
33		*	Fix defect a	nd error	8 days	Fri 19-04-19	Tue 30-04-19						
34		*	Meeting wi	th Supervisor	1 day	Mon 22-04-19	Mon 22-04-19				I		
35		19	Prototype	Maintenance									
36		*	Submission	Part Chapter 4	1 day	Fri 26-04-19	Fri 26-04-19				Ŧ		
				Task	1	0	External Milestor	ne	•	Manual Summa	ry Rollup 📟		_
				Split	75		Inactive Task			Manual Summa	γ 🕶		
anies			iha I	Milestone		•	Inactive Mileston	ie:	0	Start-only	E		
Date: I	r. gant Fri 03-0	05-19	in a	Euromani	-		Inactive Summer	-	00	Finish-only	-		
0.000	55 3 12 10 3	600 T 11.		Project Summan			Manual Task	y		Deadline	-		
				External Tasks			Duration-only			Progress	_		_
							10						

D	-	Task	Task Name	Duration	Start	Finish	01 July		01 October	r 01 Janua	ary 01	April	01 July
37	0	Mode	Submission Poster DSM3	1 days	Tue 20.04.40	Tue 20.04.10	01-07	12-08	23-09 04-1	1 16-12 27-	01 10-03	21-04 02	-06 14-07
38		+	Meeting with Supervisor	1 day	Thu 02-05-19	Thu 02-05-19						-	
39		+	Submission Full P5M 2	1 day	Fri 03-05-19	Fri 03-05-19						1	
			Task	-	3	External Milesto	he	Φ.		Manual Sum	imary Rollu	p	
			• • • • • • • • • • • • • • • • • • •		*********	Inactive Task		<u>(</u>		Manual Sum	mary		
			Split										
Projec	t: gant	ttchart naz	Split iha Milestone	٠		Inactive Milestor	e	Φ.		Start-only		E	
Projec Date:	t: gant Fri 03-1	ttchart naz 05-19	iha Milestone Summary	*		Inactive Milestor Inactive Summar	ie V	ф Ф	0	Start-only Finish-only		5	
Projec Date:	t: gant Fri 03-i	ttchart naz 05-19	Split iha Milestone Summary Project Summar	φ φ		Inactive Milestor Inactive Summar Manual Task	ie V	0 0	0	Start-only Finish-only Deadline		c ⊐ ₽	
Projec Date:	t: gant Fri 03-1	ttchart naz 05-19	Split iha Milestone Summary Project Summar External Tasks	۰ ۲ ۲		Inactive Milestor Inactive Summar Manual Task Duration-only	v Y	• • •	•	Start-only Finish-only Deadline Progress		: ; *	_