

**DETERMINING THE EFFECT OF CURRICULUM AND  
FACILITIES ON ACADEMIC ACHIEVEMENT USING DATA  
MINING APPROACH**

**A thesis submitted to the Faculty of Information Technology in partial  
Fulfillment of the requirement for the degree  
Master of Science (Intelligent System)  
Universiti Utara Malaysia**

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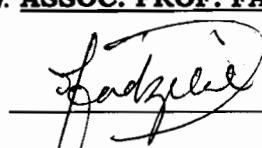
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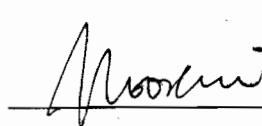
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## **ABSTRAK**

### **(BAHASA MELAYU)**

*Bidang pendidikan mempunyai pelbagai aplikasi perlombongan data yang menarik dan mencabar, serta dikenalpasti sebagai satu teknik yang berpotensi digunakan untuk membantu tenaga pengajar dan pelajar, dan memperbaiki kualiti sistem pendidikan. Berdasarkan masalah yang telah ditemui teknik perlombongan data dengan mengaplikasikan moden statistik dah teknologi pengkomputeran dapat menyelesaikan masalah pencarian maklumat yang penting di dalam pangkalan data yang besar. Kajian ini telah mengaplikasikan teknik perlombongan data dengan tujuan untuk mengkaji maklumat yang tersembunyi dan penting yang boleh menjelaskan pencapaian akademik dikalangan responden. Responden yang terlibat didalam kajian ini merupakan responden dari universiti awam yang telah menamatkan pengajian pada tahun 2007. Soal selidik yang digunakan untuk kajian ini berdasarkan soal selidik Kajian Pengesanan Graduan Kementerian Pengajian Tinggi dan kajian ini hanya menganalisa soalan di bahagian latarbelakang responden, kurikulum dan kemudahan yang telah disediakan di setiap institusi. Objektif utama kajian ini adalah untuk mengkaji sama ada kurrikulum dan kemudahan yang telah disediakan di setiap institusi mempengaruhi keputusan akademik pelajar atau tidak. Bilangan responden yang terlibat didalam kajian ini adalah seramai 55,315 orang untuk bahagian pendekatan penerangan data dan 39,801 untuk bahagian pendekatan ramalan. Pendekatan ramalan akan menggunakan dua kaedah yang berbeza iaitu kaedah statistik yang menggunakan perisian “Statistical Pakages for Sosial Science (SPSS)” dan kaedah rangkaian neural yang menggunakan perisian “Statistical Analytical Software (SAS)”. Setelah semua ujian untuk melihat kesan kurikulum dan kemudahan di IPTA menggunakan kedua-dua kaedah selesai. Keputusan yang diperolehi menunjukan rangkaian neural memberikan peratusan tertinggi berbanding statistik. Berdasarkan keputusan yang telah diperolehi menunjukkan bahawa rangkaian neural memperolehi peratusan tertinggi iaitu 89.47 % dengan*

*mengukur item kurikulum dan latarbelakang dengan pencapaian akademik responden. Secara keseluruhannya, model rangkaian neural memperolehi peratusan lebih daripada 80 % manakala statistik hanya menghasilkan kurang daripada 50%.*

## ABSTRACT (ENGLISH)

*Education domain offers many interest and challenge in data mining applications that potentially identified as a tool to help both educators and students to improve the quality of education system. Data Mining applies modern statistical and computational technologies to the problem of finding useful hidden patterns within large databases. Thus, this study applied data mining technique to identify the hidden information that affects the academic achievement among respondents. The respondents for this study are consists of all public university students which complete their study within year 2007. The questionnaire that has been used in this study was adopted from Kajian Pengesanan Graduan, Kementerian Pengajian Tinggi and it focuses on curriculum and facilities that have been provided by universities. The aims of this study is to determine whether the curriculum and facilities that provided by university has an effect on students academic achievement. 55,315 respondents data were used for descriptive task while 39,801 data for predictive task. Both data mining approaches, namely the descriptive and predictive have been utilized to perform the analysis prior to build the model. For descriptive purposes, frequency, cross tabulation and correlation coefficients were computed to check whether significant correlation exists. For predictive modeling, logistic regression and neural network were used. Statistical Pakages for Sosial Science (SPSS) was used for regression technique and Statistical Analytical Software (SAS) for Neural Network modeling. Then, the online questionnaire was integrated with Neural Network model to predict future student academic achievement. The findings in this study suggest neural network is the best model compared to logistic regression to measure the effect of curriculum and facilities on academic achievement. The highest accuracy from neural network is 89.47%, when demographics and curriculum become the contributing variables to academic achievement. Most of the neural network model accuracy is over than 80% while logistic regression accuracy is below than 50 %.*

## **ACKNOWLEDGEMENT**

In the name of Allah, Most Gracious, Most Merciful. Peace upon the prophet, Muhammad S.A.W, a foremost praise and thankful to Allah for His blessing, giving me the strength in completing this research.

On the other hand, my special thanks to Universiti Utara Malaysia (UUM) for facilities and resource that has been provided to all students. With all the facilities that have been provided, it helped me to make my research completed and succeeded.

I would like to extend my thanks and gratitude to my supervisor Associate Professor Fadzilah Siraj and Miss Nooraini Yusoff for the guidance, patience, encouragements, advice and flourish of knowledge during completing this research.

My warm appreciation dedicates to the Ministry of Higher Education who has gives their cooperation during this research.

Last but not least, a lasting heartfelt to my family and all my friends for all of the love and support in completing this course, as well as to my siblings.

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# **CHAPTER ONE**

## **INTRODUCTION**

This chapter presents the overview of the study that involves the curriculum and facilities are effect the academic achievement is applying the data mining technique. In addition, this chapter also includes the problem statement, objectives, scope, and significance of study and thesis organization.

### **1.1 BACKGROUND**

Data mining is defined as the extraction from large amount of data. It is a powerful new technology with great potential to help companies focus on the most important information in their data warehouse (Berson, 2000). According to Tsantis and Castellani (2001), data mining applies modern statistical and computational technologies to the problem of finding useful patterns hidden within large database. In addition, data mining uses a combination of an explicit knowledge base, sophisticated analytical skills, and domain knowledge to uncover hidden trends and

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