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Note: This letter should be written by the supervisor, addressed to PSZ and a copy attached to the thesis

KNOWLEDGE DISCOVERY FOR LARGE DATABASES IN EDUCATION INSTITUTES

ROBAB SAADATDOOST

A dissertation submitted in partial fulfillment of the requirements for the award of the degree of Master of Information Technology-Management

Faculty of Computer Science and Information Systems Universiti Teknologi Malaysia

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This dissertation is dedicated to my Parents who have never failed to give me every support, and my supervisor for giving all my need during the time we researched and for teaching me that even the largest task can be accomplished if it is done one step at a time.

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.

ABSTRACT

This project presents the patterns and relations between attributes of Iran Higher Education (Iran Higher Education) data gained from the use of data mining techniques to discover knowledge and use them in decision making system of IHE. Large dataset of IHE is difficult to analysis and display, since they are significant for decision making in IHE. This study utilized the famous data mining software, Weka and SOM to mine and visualize IHE data. In order to discover worthwhile patterns we used clustering techniques and visualized the results. The selected dataset includes data of five medical university of Tehran as a small data set and Ministry of Science - Research and Technology's universities as a larger data set. Knowledge discovery and visualization are necessary for analyzing of these datasets. Our analysis reveals some knowledge in higher education aspect related to program of study, degree in each program, learning style, study mode and other IHE attributes. This study helps to IHE to discover knowledge in a visualize way; our results can be focused more by experts in higher education field to assess and evaluate more.

ABSTRAK

Disertasi ini merupakan penyelidikan terhadap pola dan hubungan antara atribut Institusi Pengajian Tinggi Iran dengan menggunakan teknik "data mining" dalam pengumpulan data untuk eksplorasi dan pencarian pengetahuan serta menggunakannya dalam sistem membuat keputusan IHE. Dataset IHE yang besar dan sukar untuk dianalisis dan paparan, kerana ia adalah penting untuk membuat keputusan dalam IHE. Penyelidikan ini menggunakan perisian "data mining" (pelombongan data) yang terkenal iaitu Weka dan SOM untuk melombong (mine) dan visualisasi data IHE. Dalam rangka mencari pola-pola yang berharga, penyelidik menggunakan teknik pengugusan, pengkelasan dan divisualisasikan hasilnya. Dataset yang dipilih merangkumi data dari lima universiti perubatan Teheran sebagai kumpulan data kecil dan Kementerian Sains - universiti Penyelidikan dan Teknologi sebagai data yang lebih besar ditetapkan. Penemuan pengetahuan dan visualisasi adalah perlu dalam proses untuk menganalisis dataset ini. Hasil analisis menunjukkan pengetahuan dalam aspek pendidikan tinggi adalah berkaitan dengan program pengajian, ijazah dalam setiap program, kaedah pembelajaran, mod pengajian dan atribut IHE yang lain. Penyelidikan ini membantu untuk IHE dalam mencari pengetahuan dengan cara pemvisualisasian dan hasil kajian boleh difokuskan lebih oleh pakar-pakar di bidang pengajian tinggi untuk menilai dan menilai lebih.

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CHAPTER 1

INTRODUCTION

1.1 Introduction

Nowadays each organization deals with some data about their area and during time it increases, one of these organizations that includes big volume of data is higher education institute that has always held much data about universities, students and teachers. Thus, it is possible for us to discover some worthwhile relations or patterns that can be useful for making decision. For examples, planning the future development of a university and identifying the cluster of students who required more attentions. Management faces many challenges particularly in planning and for this purpose it needs some facts extracted from data. In our rapidly changing world, every year we accumulate data and add it to our data sets so after several years we will have a massive databank, in this environment every year our data volume increases so we need some tools to analysis this data for extracting some valuable outcome from it. Data mining has many techniques that can apply and facilitate analyzing of data.

Data mining has many definitions and almost all of them point to the discovery of patterns, and the analysis of some relations between variables in data. It does not limit to collecting and managing data; it also includes analysis. In this study, we intend to use historical data as the basis of discovering hidden relations. We intend to perform data mining techniques to discover knowledge. There are some examples, such as:

- Mining of statistical data of one university to discover successful students(Venus Shokorniaz & akbari, 2008)(Venus Shokorniaz & akbari, 2008).
- Mining on students and discovering groups of students those are available from the data and their relations (Yghini, Akbari, & Sharifi, 2008)(Yghini, Akbari, & Sharifi, 2008).

In this project, we applied data mining techniques on data related to Iran Higher Education Institute to discover some relations and patterns that are useful in decision making system of higher education.

1.2 **Problem Background**

We have chosen this topic because of government and management of universities need to plan before an event occurrence. We face huge data and need to analysis them to reach some knowledge. For this purpose we need some techniques that data mining helps us on this way, data mining has two common techniques that are classification and clustering. In this project we study about these techniques and choose one of them in our project.

Clustering is a data mining technique that is a division of data elements into groups of similar objects without advance knowledge of the group definitions. In addition it is a tool for data analysis, which solves classification problems. ln clustering, there are strong associations between members of each group and according to the type of clustering; Clustering algorithms have 4 types: exclusive, overlapping, Hierarchical, and Probabilistic. We may find some associations between different groups. Some of these associations are strong and some of them are weak. For example exclusive algorithm has weak association and overlapping has strong (Berkhin, 2006). Clustering is a discovery tool that may discover associations and patterns in data which is not previously obvious. In short: clustering attempts to find some groups of elements, based on some similarities (Ong, 2000). One of the cluster analyses is SOM (self organizing method) that is one of the most important algorithms in data visualization and exploration. Visualization transforms from the invisible to the visible (Alhoniemi, et al., 2002,2003). SOM is a particular type of neural network used in clustering. It maps high dimensional input onto two dimensional.

Classification is a data mining technique that predicts data elements' group, for example we can predict the weather of a day will be sunny, rainy or cloudy. In classification we have predefined classes that classification is a task to assign instances to these classes opposite of clustering that we don't have knowledge about group definitions. In clustering we cluster elements based on their attribute on the contrary in classifying we classify elements into groups by recognizing pattern.

Our concern in this study is finding a way to discover exciting knowledge for universities management to achieve an appropriate plan to improve society. Each society can be affected by higher education; its economic, political and scientific improvement can be resulted by advanced higher education, when we develop our programs and management part in universities so its output will be successful graduated students that can be helpful in every part of one society.

1.3 **Problem Statement**

The main problem is the volume of data that we have in higher education and government needs to analysis and discover fast and correct knowledge from them. "How we can discover it?" is our significant question. Our endeavour was proposing a methodology to discover knowledge that reveals some patterns and relations between data. We have huge databases about universities but most of time we cannot use them efficiently because we don't have appropriate pattern for our executive system whereas we have data to this purpose.

We encounter many challenges in higher education such as allocating budget to universities in start of new semester without delay, accepting accurate number of students for every semester, finding ways for effective teaching. In this study our intension is discover some findings that help us in this way to improve higher education decisions and policies.

1.4 **Project Objectives**

Objectives of this study are:

- To study for an appropriate data mining approach suitable for analysing Higher Education Institute of Iran.
- To analyse data on medical universities to discover patterns usable for managerial decision making.
- To generalise steps for discovery on larger number of universities.
- To suggest an appropriate software for the analyses of this research.

1.5 **Project Scope**

The study is analyse huge volume of data in institute of research and planning for higher education (IRPHE) in Iran. We choose two dataset; the smaller dataset is medical universities of Tehran and the larger one is Ministry of Science - Research and Technology's universities. We proposed a methodology that can be used in similar institute in Iran with large volume of data.

1.6 **The Project Importance**

Knowledge discovery is necessary for most of our plantings. Nowadays planning for higher education has significant impact on developing of one society, successful planning needs to analysis some huge and historical data that is available in higher education institutes; most of the time there is not any correct and precise analysis whereas this analysis can be helpful for managers, researchers to plan, report and discover some knowledge.

The other importance of this project is collection of data with large volume that relate to many universities during many years so it has high probability to discover some exciting knowledge. This study can be used in every university, because it faces much data about students, teachers, staff and financial resources and most of the time, this data includes information and worthwhile patterns.

The methodology that will be suggested is helpful for higher education institute in collecting data and to discover some knowledge for improving management decisions. Furthermore it can be useful for researchers that study and research about higher education. This knowledge leads us to development in technical, scientific and economic aspects according to type of data that we will analyse.

1.7 Summary

This chapter discussed the overview of this study which is knowledge discovery in higher education institute with statistical data and other associated issue in data mining techniques that helps us in discovering relations and patterns that are useful in decision making in higher education. Higher education usually does not have appropriate methodology to analysis huge volume of data and thus most of time it ignores much knowledge that can be very helpful in many decisions to improve quality of education. There are four objectives that were achieved successfully in order to improve higher education decision making systems and their policies that certainly affect on society. At first we studied knowledge discovery and then analysed Iran higher education data to find some relations and patterns. Finally we proposed a methodology to knowledge discovery from higher education data.

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