# LOG MINING USING GENERALIZED ASSOCIATION RULES

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Master of Science (Intelligent Knowledge Based System)

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by Mohd Helmy Abd Wahab

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### ABSTRACT (ENGLISH)

Explosive growth in size and usage of the World Wide Web has made it necessary for Web site administrators to track and analyze the navigation patterns of Web site visitors. To achieve this goal, the use of web mining tool is necessary. Web mining can be defined as the use of data mining techniques to automatically discover and extract information from web documents. Since Data Mining is primarily concerned with the discovery of knowledge and aims to provide answers to questions that people do not know how to ask, it is not an automatic process. Rather one has to exhaustively explores very large volumes of data to determine otherwise hidden relationships. The process extracts high quality information that can be used to draw conclusions based on relationships or patterns within the data. However, data mining techniques are not easily applicable to Web data due to problems both related with the technology underlying the Web and the lack of standards in the design and implementation of Web pages. Information collected by the Web servers are kept in the server log is the main source of data for analyzing user navigation patterns. Once logs have been preprocessed and sessions have been obtained, there are several kinds of access pattern mining that can be performed depending on the needs of the analyst. Since the method use in this study relied on relatively simple techniques therefore the information gathered is adequate for real user profile data due to the noise in the data has to be first tackled. In this study, Data Mining techniques known as generalized association rules was used in order to get some insights into website usage pattern. For the purpose of this study, server logs from tutor.com portal were retrieved, pre-processed and analyzed. An important finding from this study is that Mathematics subject generally popular from UPSR, PMR and UPSR levels. On the contrary, arts subjects are not popular to Tutor.com users. The system administrator may consider evaluating the content and the link for such subjects, so that the real problem can be identified.

## ABSTRACT (BAHASA MELAYU)

Perkembangan yang pesat di dalam penggunaan World Wide Web (www) menyebabkan perlunya pentadbir laman web menjejaki dan menganalisis corak pelayaran pengunjung laman web. Untuk mencapai matlamat ini, penggunaan alat bantu pengguna web amat diperlukan oleh pentadbir laman web. Perlombongan web didefinisikan sebagai penggunaan teknik perlombongan data untuk memenuhi dan memperolehi maklumat daripada dokumen web. Perlombongan Data adalah tertumpu kepada penemuan pengetahuan dan juga bertujuan untuk menyediakan jawapan kepada soalan pengguna, teknik ini tidak dapat melakukan pemprosesan secara automatik. Sebaliknya, pelayan web perlu meneroka data yang banyak untuk menentukan perkaitan antara data. Proses ini mengekstrak maklumat berkualiti yang boleh digunakan untuk membuat sesuatu kesimpulan berdasarkan hubungan atau paten yang terdapat didalam data berkenaan. Walaubagaimanapun, teknik perlombongan data tidak mudah diaplikasikan kepada data web disebabkan masalah yang berkaitan dengan teknologi web dan kekurangan keseragaman dalam rekabentuk dan implementasi laman web. Maklumat yang terkumpul oleh pelayan web akan disimpan didalam log pelayan yang mana merupakan sumber utama data bagi menganalisis corak pelayaran pengguna. Setelah log pelayan di pra-proses dan sesi pengguna diperolehi, perlombongan paten capaian boleh dilaksanakan dengan beberapa kaedah. Ini bergantung kepada matlamat penganalisis log pelayan. Kaedah yang digunakan dalam kajian ini adalah berdasarkan kepada teknik yang lazim digunakan, maka maklumat yang dikumpulkan tidak mencukupi untuk menjadikan data profil pengguna kerana "noise" yang terdapat dalam data perlu di pra-proses terlebih dahulu. Teknik perlombongan data yang dikenali sebagai generalized association rules digunakan untuk mendapatkan paten penggunaan laman web. Untuk kajian ini, log pelayan dari portal Pendidikan Utusan telah diperolehi, diproses dan telah dianalisis menggunakan alat bantu perlombongan data. Dapatan kajian menunjukkan, bahawa Matematik merupakan subjek yang popular untuk peringkat UPSR, PMR dan peringkat UPSR.. Sebaliknya, subjek sastera tidak begitu popular kepada pengguna Portal Pendidikan. Oleh sebab itu, penyelenggara Portal Pendidikan Utusan perlu mempertimbangkan kandungan serta pautan kepada setiap subjek di dalam Portal Pendidikan supaya masalah yang sebenarnya dapat diatasi.

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# CHAPTER 1

# INTRODUCTION

### 1.1 Context of the study

With the explosive growth of data available on the World Wide Web (WWW), discovery and analysis of useful information from the World Wide Web becomes a practical necessity. Data Mining is primarily concerned with the discovery of knowledge and aims to provide answers to questions that people do not know how to ask. It is not an automatic process but one that exhaustively explores very large volumes of data to determine otherwise hidden relationships. The process extracts high quality information that can be used to draw conclusions based on relationships or patterns within the data.

Using the techniques used in Data Mining, Web Mining applies the techniques to the Internet by analyzing server logs and other personalized data collected from customers to provide meaningful information and knowledge. Web access pattern, which is the sequence of accesses pursued by users frequently, is a kind of interesting and useful knowledge in practice (Pei, 2000). Today web browsers provide easy access to myriad sources of text and multimedia data. More than 1 000 000 000 pages are indexed by search engines, and finding the desired information is not an easy task (Pal et al., 2002). Web Mining is now a popular term of techniques to analyze the data from World Wide Web (Pramudiono, 2004). A widely accepted definition of the web mining is the application of data mining techniques to web data. With regard to the type of web data, web mining can be classified into three types: Web Content Mining, Web Structure Mining and Web Usage Mining.

# The contents of the thesis is for internal user only

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