Chapter IV

A Lattice-Based Framework for Interactively and Incrementally Mining Web Traversal Patterns

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Abstract

Web mining is one of the mining technologies, which applies data mining techniques in large amounts of Web data to improve the Web services. Web traversal pattern mining discovers most of the users’ access patterns from Web logs. This information can provide the navigation suggestions for Web users such that appropriate actions can be adopted. However, the Web data will grow rapidly in the short time, and some of the Web data may be antiquated. The user behaviors may be changed when the new Web data is inserted into and the old Web data is deleted from Web logs. Besides, it is considerably difficult to select a perfect minimum support threshold during the mining process to find the interesting rules. Even the experienced experts also cannot determine the appropriate minimum support. Thus, we must constantly
adjust the minimum support until the satisfactory mining results can be found. The
essences of incremental or interactive data mining are that we can use the previous
mining results to reduce the unnecessary processes when the minimum support is
changed or Web logs are updated. In this chapter, we propose efficient incremental
and interactive data mining algorithms to discover Web traversal patterns and make
the mining results to satisfy the users’ requirements. The experimental results show
that our algorithms are more efficient than the other approaches.

Introduction

With the trend of the information technology, huge amounts of data would be eas-
ily produced and collected from the electronic commerce environment every day.
It causes the Web data in the database to grow up at amazing speed. Hence, how
should we obtain the useful information and knowledge efficiently based on the
huge amounts of Web data has already been the important issue at present.

Web mining (Chen, Park, & Yu, 1998; Chen, Huang, & Lin, 1999; Cooley, Mobasher,
& Srivastava, 1997; EL-Sayed, Ruiz, & Rundensteiner, 2004; Lee, Yen, Tu, & Hsieh,
2003, 2004; Pei, Han, Mortazavi-Asl, & Zhu, 2000; Yen, 2003; Yen & Lee, 2006)
refers to extracting useful information and knowledge from Web data, which applies
data mining techniques (Chen, 2005; Ngan, 2005; Xiao, 2005) in large amount of
Web data to improve the Web services. Mining Web traversal patterns (Lee et al.,
2003, 2004; Yen, 2003) is to discover most of users’ access patterns from Web logs.
These patterns can not only be used to improve the Web site design (e.g., provide
efficient access between highly correlated objects, and better authoring design for
Web pages, etc.), but also be able to lead to better marketing decisions (e.g., put-
ing advertisements in proper places, better customer classification, and behavior
analysis, etc.)

In the following, we describe the definitions about Web traversal patterns: Let \( I = \{x_1, x_2, \ldots, x_n\} \) be a set of all Web pages in a Web site. A traversal sequence \( S = <w_1, w_2, \ldots, w_m> \) \((w_i \in I, 1 \leq i \leq m)\) is a list of Web pages, which is ordered by tra-
versal time, and each Web page can repeatedly appear in a traversal sequence, that
is, backward references are also included in a traversal sequence. For example, if
there is a path which visits Web page \(<B>\), and then go to Web page \(<G>\) and \(<A>\)
sequentially, and come back to Web page \(<B>\), and then visit Web page \(<C>\). The
sequence \(<BGABC>\) is a traversal sequence. The length of a traversal sequence \( S \) is
the total number of Web pages in \( S \). A traversal sequence with length \( l \) is called an
\( l \)-traversal sequence. For example, if there is a traversal sequence \( \alpha = <AGDFAB>\),
the length of \( \alpha \) is 6 and we call \( \alpha \) a 6-traversal sequence. Suppose that there are two
traversal sequences \( \alpha = <a_1, a_2, \ldots, a_m> \) and \( \beta = <b_1, b_2, \ldots, b_n> \) \((m \leq n)\), if there
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