

## Chapter 4.22

# Traversal Pattern Mining in Web Usage Data

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### ABSTRACT

Web usage mining is to discover useful patterns in the web usage data, and the patterns provide useful information about the user's browsing behavior. This chapter examines different types of web usage traversal patterns and the related techniques used to uncover them, including Association Rules, Sequential Patterns, Frequent Episodes, Maximal Frequent Forward Sequences, and Maximal Frequent Sequences. As a necessary step for pattern discovery, the preprocessing of the web logs is described. Some important issues, such as privacy, sessionization, are raised, and the possible solutions are also discussed.

### INTRODUCTION

Web usage mining is to discover useful patterns in the web usage data, i.e., web logs. The web

logs record the user's browsing of a web site, and the patterns provide useful information about the user's browsing behavior. Such patterns can be used for web design, improving web server performance, personalization, etc.

Several different types of traversal patterns have been proposed in the literature, namely, Association Rules, Sequential Patterns, Frequent Episodes, Maximal Frequent Forward Sequences, and Maximal Frequent Sequences. These patterns differ in how the patterns are defined, and they can be used for different purposes. This chapter examines these patterns and the related techniques used to uncover them.

One important issue about mining traversal patterns, or about web usage mining in general, is the preprocessing of the web logs. Since the web logs are usually not in a format for web usage mining, preprocessing is needed. Such preprocessing becomes complicated or problematic by the current use of the Web. The details about the

problems and possible solutions are discussed in this chapter.

The rest of the chapter is organized as follows: The second section, *Web Usage Data*, gives the background to web usage mining. The third section, *Preprocessing*, describes the web log preprocessing. The different types of traversal patterns are described in the fourth section, *Pattern Discovery*. The fifth section, *Pattern Analysis and Applications*, describes the analyses and applications of these patterns. The sixth section, *Conclusion*, concludes the chapter.

### Web Usage Data

To characterize the web usage data, the terms defined by the W3C Web Characterization Activity (WCA) (<http://www.w3c.org/WCA>) are adopted. A *user* is defined as an individual who is accessing the Web through a browser. A *user session* is a delimited set of user clicks across one or more web servers. A click corresponds to a page on the web server, which is uniquely identified by a URI (Universal Resource Identifier). A *server session* is a collection of user clicks to a single web server during a user session.

The web usage data can be collected from different sources, e.g., the *server* side, the *client* side (Catledge & Pitkow, 1995) and the *proxy* side (Cohen et al., 1998). The server usage data correspond to the logs that are collected at a web server. They provide an aggregate view of the usage of a web site by all users. Such web server log data may not be entirely reliable due to the presence of various levels of caching (e.g., client caching by the browser, proxy caching by the proxy server) within the Web environment. The client usage data can be collected by using a remote agent, e.g., Java Applets, or by asking the user to use a specialized browser. The client side data can potentially capture every click of the user, but it requires the user's cooperation to collect. The proxy usage data are collected at a proxy server, which acts as an intermediate level

of caching between the client browsers and web servers. Such data may capture the browsing behavior of a group of anonymous users sharing a common proxy server.

The focus of this chapter is on the usage data at a web server, since all the traversal patterns target such web server logs. The information that a typical log on a web server contains is shown in Table 1.

The IP address is the address of the client machine from which the request is made. The user ID is relevant only when the user logs in to the web server. The time field shows when the page is accessed. The method/URI/protocol records which page (identified by a URI) is accessed, and the method and protocol used for the access. The status field and the size field show the access status (e.g., 200 for success) and the size of the page. The referrer field is from the referrer log, which indicates the page the user was on when he or she clicked to come to the accessed page. If the referrer is not empty (i.e., it is not '-' in the log), it usually means there is a hyperlink from the referrer page to the accessed page. This is useful for preprocessing, as indicated below. The user agent field shows whatever software the user used to access the web server, which is typically a web browser.

### PREPROCESSING

The tasks of preprocessing for web usage mining include extraction, cleansing, transformation, sessionization, etc. Extraction is for selecting the related fields from the web logs. Traversal patterns typically require three fields: IP address, access time, and the page accessed. Other fields, such as referrer and user agent, can be used in cleansing and sessionization. Transformation converts the fields to the format required by specific pattern discovery algorithm. The other two important tasks, cleansing and sessionization, are described.

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