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Chapter XV

Algorithms for Data Mining

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Abstract

In this chapter, we present an overview of some common data mining algorithms. Two techniques are considered in detail. The first is association rules, a fundamental approach that is one of the oldest and most widely used techniques in data mining. It is used, for example, in supermarket basket analysis to identify relationships between purchased items. The second is the maximum sub-array problem, which is an emerging area that is yet to produce a textbook description. This area is becoming important as a new tool for data mining, particularly in the analysis of image data. For both of these techniques, algorithms are presented in pseudo-code to demonstrate the logic of the approaches. We also briefly consider decision and regression trees and clustering techniques.

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Introduction

Data mining is often used to extract useful information from vast volumes of data, typically contained within large databases. In this context "useful information" usually means some interesting information that realistically can only be found by analyzing the database with a computer and identifying patterns that an unaided human eye would be unable to ascertain. Applications of data mining occur in a wide variety of disciplines — the database could contain the sales data of a supermarket, or may also be image data such as a medical x-rays. Interesting information could then be customers' purchasing behavior in the sales database, or some abnormality in the medical image. As the size of these databases is measured in gigabytes and they are stored on disk, algorithms that deal with the data must not only be fast, but also need to access the disk as few times as possible.

One of the oldest and most widely used data mining techniques involves the identification of association rules. For example, mining an association rule in a sales database can involve finding a relationship between purchased items that can be expressed in terms such as: "A customer who buys cereal is likely to buy milk." In the following discussion we use a simple example to illustrate a number of issues with association rule mining and to assist in the outline of data mining algorithms. Figure 1 illustrates a simple record of sales at a food supermarket, including a list of items purchased by specific customers, as well as some known attributes of the customers.

Customer	Items		Total amount spent		
1	ham, cheese,	cereal, milk	\$42		
2	bread, cheese, milk		\$22		
3	ham, bread, cheese, milk		\$37		
4	bread, milk		\$12		
5	bread, cereal,	, milk	\$24		
6	ham, bread, cheese, cereal		\$44		
Customers Customer	Name	Gender	Age	Annual income	Address
Customer					Address suburb A
Customer 1	Name	Gender	Age 33 45	\$20000	
Customer	Name	Gender female	33		suburb A
Customer 1 2	Name Anderson Bell	Gender female female	33 45	\$20000 \$35000	suburb A suburb A
Customer 1 2 3	Name Anderson Bell Chen	Gender female female male	33 45 28	\$20000 \$35000 \$25000	suburb A suburb A suburb B

Figure 1. Example transaction and customer databases

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