

# **PREDICTION OF USERS' PERFORMANCE ON A MARKETING WEBSITE**

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Abstract - Online shopping is becoming more and more common in our daily lives. Understanding users' interests and behavior is essential to adapt e-commerce websites to customers' requirements. The information about users' behavior is stored in the web logs. The analysis of such information has focused on applying data mining techniques, where a rather static characterization is used to model users' behavior, and the sequence of the actions performed by them is not usually considered. Therefore, incorporating a view of the process followed by users during a session can be of great interest to identify more complex behavioral patterns. To address this issue, we are implementing data mining algorithms and perform analysis of structured e-commerce web logs. Then identify different behavioral patterns that consider the different actions performed by a user will help to improve the website as whole.

*Key Words*: Data mining, online shopping, sequence pattern, behavioural pattern.

### **1. INTRODUCTION**

The increase in popularity of the Internet and the rapid development of E-commerce, Internet-based businesses' websites are facing increasing competition. E-commerce sites generate large amounts of data daily, and these data include potential consumer-related information that is valuable for market analysis and prediction. E-commerce business analysts require to know and understand consumers' behaviour when those navigate through the website, as well as trying to identify the reasons that motivated them to purchase, or not, a product[1]. Therefore, the most important challenge of E-commerce is to elucidate customers' wants, love, and value orientation as much as possible to ensure competitiveness in the E-commerce era.

### **2. LITERATURE SURVEY**

Data mining (DM) is used to attain knowledge from available information in order to help companies make weighted decisions. Data mining is field that focuses on access of information useful for high level decisions and to help online shopping stores to indentify online customer behaviour to recommend for him the appropriate products he/she is interesting to them [3].

Various algorithms are used to find the pattern or frequent sequence in the data. Thus, the most efficient algorithms are extracted and used for our system. Moreover, an organization needs to invest only on the group of products which are frequently purchased by its customers as well as price them appropriately in order to attain maximum customer satisfaction [8].

Hidden relationships in sales data can be discovered from the application of data mining techniques [1]. Thus, techniques for identifying the areas of improvement or the area which is liked or disliked are needed by data analyst or owner of ecommerce website.

### 2. PROPOSED SYSTEM



Fig . Architecture of proposed system

The user will have to login to the system. The selection of database file is done by the user.

On the selected dataset the SPADE and ID3 algorithms are processed applied.

The SPADE algorithm will do the generation of frequent sequences.

After that its minimum support will be calculated. Graphical view shows that which item or combinations of items were selling frequently. Result gives proper combinations to the user.

The ID3 algorithm will do the extraction of sales prediction of selected product.

The result shows the predicted sell of an item according to the views, ratings given by the user.

### **4. RELATED WORK**

### 4.1 ID3 (ITERATIVE DICHOTOMISER 3)

ID3 is a simple decision tree learning algorithm developed by Ross Quinlan (1983). The basic idea of ID3 algorithm is to construct the decision tree by employing a top-down, greedy search through the given sets to test each attribute at every tree node[1]. In order to select the attribute that is most useful for classifying a given sets, we introduce a metric---information gain.

#### Algorithm:

ID3 (Learning Sets S, Attributes Sets A, Attributes values V) Return Decision Tree.

Begin Load learning sets first, create decision tree root node 'rootNode', add learning set S into root node as its subset.

For rootNode, we compute Entropy(rootNode.subset) first

If Entropy(rootNode.subset)==0, then rootNode.subset consists of records all with the same value for the categorical attribute, return a leaf node with decision attribute:attribute value;

If Entropy(rootNode.subset)!=0, then compute information gain for each attribute left(have not been used in splitting), find attribute A with Maximum(Gain(S,A)). Create child nodes of this rootNode and add to rootNode in the decision tree.

For each child of the rootNode, apply ID3(S, A, V) recursively until reach node that has entropy=0 or reach leaf node.

End ID3.

### **Attribute Selection**

How does ID3 decide which attribute is the best? A statistical property, called information gain, is used. Gain measures how well a given attribute separates training examples into targeted classes. The one with the highest information (information being the most useful for classification) is selected. In order to define gain, we first borrow an idea from information theory called entropy. Entropy measures the amount of information in an attribute.

Given a collection S of c outcomes

 $Entropy(S) = S - p(I) \log 2 p(I)$ 

where p(I) is the proportion of S belonging to class I. S is over c. Log2 is log base 2.

Note that S is not an attribute but the entire sample set.

# 4.2 SPADE (Sequential Pattern Discovery using Equivalence Class)

SPADE algorithm was introduced in 2001 by M.J.Zaki. It makes use of Apriori vertical formatting approach [4]. The original sequence database is transformed into vertical id-list data format, in which each id-list associates with the corresponding items (SID) and time stamp (TID). The aim of this algorithm is to find frequent sequences using efficient lattice search techniques and simple join operations. It requires only three database scans to discover all the sequences.

SPADE algorithm working is as follows:

Step 1: Scan the database once and discover frequent sequences of length one by using Apriori property.

Step 2: Generation of candidate sequences set of length two by joining all pairs of frequent items.

a) Check if the two items have the same SID and is in sequential order of events.

b) A list of frequent sequences of length two is discovered and finalized.

Step 3: Traversing the lattice for support count and enumeration of frequent sequences.

a) Lattice is traversed in either breadth first search or depth first search. It is quite large to be filled in main memory. So decomposition of lattice into equivalence classes by the prefix of sequence. Sequences that are in same class have a common prefix.

# **5. COMPARISON OF METHODS**

Table -1: Comparison

PARAMETERS	ID3	SPADE Sequential pattern mining	
Algorithm	Classification algorithm		
Method	Decision Tree	Time series analysis	
Advantages	Easy to understand and implement; Classification results with strong interpretability; Does not need complex data pre-processing	It can find the potential connection in the sequence with certain order.	
Disadvantage s	It can be difficult to control the size of the tree; In some complex cases, splitting data into classes might not be helpful;	Often there are ways around getting a model that is time- series based where the predictions are almost as good and is faster to implement.	

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# 6. RESULTS

8	\$ - <b>-</b> ×
	Registration Form
	ID: 3
	Username :
	Password :
	E-mail ID :
	Already Register?
	Register

### Fig – 1: Registration form

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	Log in			
	User Name :	ketan		
	Password :	••••		
		Register	Login	

# Fig – 2: Login form

1D3 Apriori Algo				- ¤ ×
	Item Name SWAG rating total price user viewed	GA Women Clogs Get Prediction usin	v ng ID3	

Fig - 3: ID3 page(1)

<b>.</b>	- 🗆 X
ID3 Apriori Algo	
Item Name	SWAGGA Women Clogs
rating	3.9
Total price	15000
user viewed	20000
	Get Prediction using ID3
	predicted sales is :10906



Browse File	
Apply Algorithm	
- Recult Evaluation	
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Select mCount 👻	
Get Graph	
	Browse File Apply Algorithm Result Evaluation Select mCount

**Fig - 5**: Spade(1)

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ID3 Aprice Algo	Result Copen Look pr. Contex Conte	yootoos-data
	File Name:	demo.dat
1	Files of Type:	/ All Files

Fig - 6: Spade(2)



Fig – 7: minimum support is 1



# Fig – 8: minimum support is 2



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Fig – 10: minimum support is 4

# 7. CONCLUSIONS

In this paper, a detailed study based on data mining techniques was conducted in order to extract knowledge in a data set with information about user's history associated to an e-commerce website[8].

The set of descriptive data mining techniques are applied that allows data analyst working at ecommerce companies make strategic decisions to boost their sales as well as provide effective customer service.

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