

Hybrid Association Rules to Classify and Discovering Item sets based on User Knowledge

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ABSTRACT

Increase large luminous data is posing research concept to predict and effective analysis of data in different types of business oriented applications. Intelligence of business maintenance is an aggressive concept to provide analysis customers, employees and suppliers for more effective decision making. It is basic representation to update and enhance their business in different formats like quality of service, and achieve profitability for business organizations. To provide analysis of data effectively, traditionally some of the data mining approaches like Classification, Clustering and Association were used to provide and increase the efficiency in real time business organizations. Based on in depth analysis of data relates to data delivery, reporting and predictive analysis, data mining techniques have failure to elaborate business in real works. So it aims to explore the advanced and statistical techniques apart from existing data mining or modeling approaches. In this paper, we propose and implement a novel HDM approach (which consist Association rules and Classification rules) for effective analysis of large data sets. Our approach is increases the services like quality of service, customer service, and report generations in business organizations. Furthermore, hybrid data mining approach is designed to assist to user throughout data analyzing task. By applying our new approach over large data to integrate domain expert knowledge in post processing to reduce set of rules in business intelligence service analysis. Our experimental results show effective maintenance of customer service to reduce quality of filtered rules by domain expert knowledge with interactive process in real world business organizations.

Keywords: Intelligence of Business, Data mining, Hybrid approach, Classification, Clustering, Customer services, Quality of service and crystal reports.

I. INTRODUCTION

In business oriented organizations, mining information is the data innovation for different web application developments. Data mining is the way to retrieve data based on usage with respect to data relations present in data base. Knowledge based data discovery with respect to associate attributes. Association rule mining sequences is one of fundamental issue knowledge discovery procedure to define different data item sets while processing different transactional data bases, for effective decision maker to large data bases is an implicative tendency in that can be process valuable data.

To define data implication between different data sources, association is interesting measure, association between two data sources X and Y i.e. $X \rightarrow Y$ defined by two interesting parameters with following relations $X \cap Y = \phi$. Apriori [1] was the first calculation method to define above association between different data sources and then some of the other calculation methods are proposed from Apriori. For efficient relation between data sources association rules are maintain threshold based support and confidence. Mining algorithms that can find the different association rules based on several types of attributes with respect to different types of transactions. To extract valuable information is often

from unrelated association relations associate with users because of low support efficiency in transactional data sources. If we want to increase threshold value then more and efficient algorithms and more finding rules are required and then they were helped to define association procedure and interest to end user, further more support and other values associate support values at lower support values consists large volume of rules; it is not comfortable for decision making to analyze data mining result. For efficient decision making with an efficient approach to reduce association rules is an aggressive concept as a result. To achieve this limitation, several techniques are proposed in literature, from one perspective, diverse calculations were acquainted with diminishing the quantity of item-sets by creating shut [4], maximal [5] or ideal item-sets [6], and a few calculations to lessen the quantity of rules, utilizing non-redundant rules [7], [8], or pruning methods [9]. Then again, post-processing techniques can enhance the determination of found principles. Extraordinary reciprocal post-processing techniques might be utilized, such as pruning, outlining, gathering, or perception [10]. Evaluating techniques comprises in evacuating irrelevant or excess rules. In condensing, brief arrangements of standards are created. Gatherings of guidelines are created in the gathering procedure, and the perception enhances the lucidness of an extensive number of principles by utilizing adjusted graphical representations based on different data sources.

Traditionally, numbers of approaches were introduced based on conceptual data sets, to extract in-depth data analysis by considering user interestingness is not possible in these traditional techniques. In different types of business intelligence applications like semantic web data extraction and data analysis related applications, presentation of user knowledge sequences. The more extract user information in a flexible, expressive and formalize presentations based on different rules selection is efficient. In data extraction, User Knowledge is the most aggressive concept with respect to user knowledge; different types' data mining applications were introduced to define user specifications. For effective data analysis in business intelligence oriented applications. So that we propose novel Hybrid data mining approach (which consist Association rules and Classification rules) for effective

analysis of large data sets to explore semantic b data extraction and visualization. For that, first, we propose domain User Knowledge to integrate the user knowledge in post processing representation. Secondly, we propose rule schema classification by extending user beliefs and expectations use of the data analyzing task.

1. Preliminaries

For efficient data exploration, association and classification of different approaches with different attributes. Let us consider $K = \{k_1, k_2, \dots, k_n\}$ be set of items i.e. literals, and let $T = \{t_1, t_2, \dots, t_m\}$ are number of different transactions K. K is non-empty then it is defined $X = \{k_1, k_2, \dots, k_n\}$, in this item set, number of item sets are analyzed based on transactions are utilized in different item sets. Each transaction (t_i) consists different items like k_1, k_2, \dots, k_n with n number of items for each transactions t. Relationship between data items based on classification and association between attributes in large data sets organized with different definitions:

Definition 1: let $X \subseteq K$ & $D \subseteq T$, based on these relations, define set of transactions with different item sets X.

$$t : K \rightarrow T, t(X) = \{t \in D \mid X \subseteq t\}$$

Similarly, describe different item sets combined with transactions T by

$$k : D \rightarrow K, i(T) = \{x \in K \mid \forall t \in T, x \in t\}$$

Definition 2: $X \rightarrow Y$ is implication representation where X and Y 2 different item sets $X \cap Y$, the former, Y is major concept and X is indecent relation. A concept $X \rightarrow Y$ is the two factor representations with statistical relative attributes.

With respect to associated item sets, can be defined as $\sup(X \rightarrow Y) = \sup(X \cup Y) \cup$ are different transactions which contain $X \cup Y$, if $\sup(X \rightarrow Y) = s$ then s% of item transactions with respect to item sets $X \cup Y$.

Based on support representation, confidence can be calculated as follows: $\text{conf}(X \rightarrow Y) = \sup(X \rightarrow Y) / \sup(X) = \sup(X \cup Y) / \sup(X) = c$ and c% is the number of transactions that contain X contain Y.

II. Description of Hybrid Approach

The proposed work in this paper is hybrid approach to analyze user knowledge based on different item set formations and associations with semantic data analysis. Hybrid approach (Association and Classification) consists mainly two steps described in figure 1. First, the user knowledge defines customer information and objectives. Sector knowledge offers a common perspective of customer information in the database domain and customer objectives show the first user knowledge of the found guidelines. Secondly, the pre-process or post-process comprises in implementing iteratively based on filter representations overproduced guidelines in order to draw out exciting rules: enhancement restriction narrow, item-relatedness filter, concept schema filters/pruning.

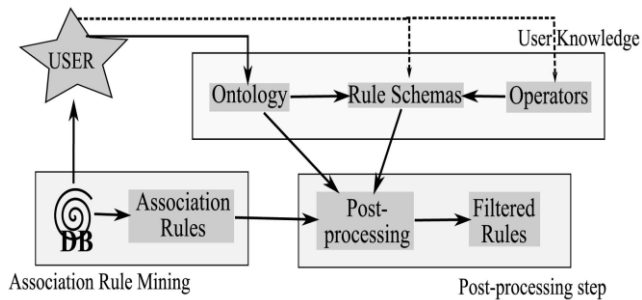


Figure 1. Hybrid approach procedure for analyzing data.

Novelty in this approach is supervising the user knowledge finding process based on conceptual data representations based on user knowledge with several schemes and generalizing iterative user process. Procedure of the hybrid data mining for interactive process and analyzing data consists following modules.

Post-processing Interactive Process

Hybrid approach proposes user interactive process to discovery rules shown in figure 2 by taking user feedbacks to revise functional attributes to contribute intermediate data analyzed results.

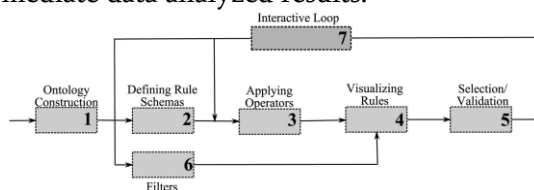


Figure 2. Interactive process description for different operators

The following several steps are required to do hybrid data mining approach effectively.

1. Development of user knowledge — beginning with data base relations, and in the long run, from existing web ontology descriptions, the client builds up a metaphysics on database things;

2. Characterizing different rules (as GIs and RPCs) — the client communicates nearby objectives and desires concerning the affiliation decides that he/she needs to discover;

3. Picking the correct administrators to be connected over the administer outlines made, and after that, applying the administrators;

4. Envisioning the outcomes—the separated affiliation rules relate to the client;

5. Choice/approval—beginning with preparatory outcomes, the client can approve the outcomes can reexamine his/her data;

6. We introduce to the client two channels effectively delivered in the writing. Two channels can be connected over tenets at whatever point the client needs them with the fundamental objective of decreasing the quantity of standards; and

7. The intuitive circle licenses to the client to change the data that he/she proposed. In this manner, he/she can come back to stage 2 keeping in mind the end goal to adjust the control patterns, or he/she can come back to stage 3 with a specific end goal to change the administrators. In addition, in the intelligent circle, the client could choose to predefined channels stage 6.

III. Improved Impressions based on User Knowledge

Impressions are generally represents vague feelings of user based on precise knowledge in different data analysis with different attributes. It is very flexible, formalize and comprehensible for every user based on different representations. General syntax for effective impressions in user view with respect to support and confidence

$$gi(< S_1, \dots, S_m >)[sup, conf]$$

S_i is the source of data representations using $*+=?/?$ Providers and assistance and confidence limits are optionally available. For best impressions in generally, proposed approach consists following rules.

- . Accommodating standards—affiliation decides that are adjusting to the predefined convictions;

. Surprising forerunner rules—affiliation decides that are unforeseen in regards to the predecessor of the predefined convictions;

. Unforeseen subsequent guidelines—affiliation decides that is surprising with respect to the subsequent of the determined convictions; and

Both side surprising standards—affiliation decides that are startling in regards to both the forerunner and the resulting in the predetermined convictions.

To enhance affiliation control determination, we propose another control sifting model, based on Schema relations. A manage pattern depicts, in a run like formalism, the client desires in terms of fascinating/clear principles. Accordingly, Rule Mappings go about when in doubt gathering, characterizing guideline families.

$$RS(< X_1, X_2, \dots, X_n \rightarrow Y_1, Y_2, \dots, Y_m >)$$

Where $X_i, Y_j \in C$ off $O\{C, R, I, H, A\}$ and the implication “ \rightarrow ” is optionally available. Basically, we can remember that the proposed formalism brings together Common Opinions and Reasonably Precision Concepts (RPC).

Description of Association Mapping

To describe user knowledge effectively concerning in data base in our proposed approach. In our implementation, major relations based on different ideas $O = (C, R, I, H, A)$ to data base, different item sets with different things associated with conceptual relations. To this end, we present sorts with different ideas with different concepts from connections i.e < or > with ontology relations.

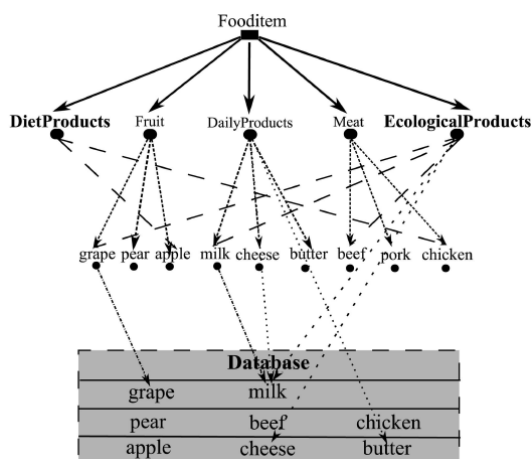


Figure 3. Description of different item sets in User Knowledge.

Keeping in mind the end goal to continue with the meaning of each sort of ideas, let us help that a set to

remember things in a database is characterized as $K = \{k_1, k_1, \dots, k_n\}$.

Associated leaf connection C_0

$$C_0 = \{c_0 \in C \mid \hbar \leq c_0\}$$

Associated with database C_0 with different products :

$$f_0 : C_0 \rightarrow I, \forall c_0 \in C_0, \exists i \in K, k = f_0(c_0)$$

Generalized Concepts (C_1) are depicted as the ideas that subsume different ideas in the philosophy. A summed up idea is associated with the database through its restricted concepts with different item sets shown in figure 3.. This implies, recursively, just the leaf-concepts subsumed by the summed up idea add to its database association:

$$f : C_1 \rightarrow 2^K$$

$$\forall c_1 \in C_1, f(c_1) = \bigcup_{c_0 \in C_0} c_0 \leq c_1\}$$

Restricted data relations using different expression based on architectural C2 implications, in this region, we define different data relations with variable in logical data relations.

Classify Rule Schema Operations

The concept schema narrow is dependent on providers used over rule schemas enabling the customer to carry out several actions over the found guidelines. We recommend two important operators: trimming and filtration providers. Countering unexpectedness and examination with different relations different on filtering methodologies introduced by Liu et.al based on post and pre-processing task representations as follows: Pruning, Confirming and Exceptions.

Pruning. The pruning administrator permits to the client to expel groups of principles that use different data relations. In databases, there exist, much of the time, data relations which are reliable to known conditions. In this manner, it isn't helpful to discover data relations found different affiliations. The displaying organizer connected over a manage construction, P (RS), wipes out all affiliation rules coordinating the govern mapping. To separate every one of the tenets coordinating a run blueprint, the adjusting administrator is utilized.

Confirming. The accommodating administrator connected over a manage pattern, C(RS), affirms a suggestion or finds the suggestion between a few ideas. Therefore, rules coordinating every one of the components of a non-relative impressions govern composition are separated. For

relative impressions, the condition furthermore, the finish of the affiliation lead should coordinate those of the pattern.

Exceptions. Lastly, the exemption owner defined relative impressions (i.e., RS1) and guidelines based on the following new implicative concept schema: $X \wedge Z \rightarrow Y$ where Z is the item sets.

Let us consider the example to do effective data presentation for implicative rule schema $RS : Fruits \rightarrow Eco\ logical\ Pr\ oducts$ where

$$f(Fruits) = \{grape, apple, pear\} \text{ and } f(Eco\ logical\ Pr\ oducts) = \{milk, grape\}$$

$$K = \{grape, apple, pear, milk, beef\} \text{ then}$$

association rules are extracted based on rule schemas are as follows:

$$R_1 : grape, beef \rightarrow milk, pear$$

$$R_2 : apple \rightarrow beef$$

$$R_3 : apple, pear, milk \rightarrow grape$$

$$R_4 : grape, pear \rightarrow apple,$$

$$R_5 : beef \rightarrow grape,$$

$$R_6 : milk, beef \rightarrow grape$$

Based on schema relations i.e RS with different relations R1 and R3 in associated guided relatives R5 and R6 based on guidelines available on pruning process. Based on above rules with different attributes are filtered and association relation between different item sets with support and confidence shown below.

$$grape, pear \rightarrow milk(conf = 85\%)$$

$$grape \rightarrow milk(conf = 90\%)$$

$$pear \rightarrow milk(conf = 83\%)$$

In our strategy, we implement data relations based on customer transactions with different features with different domains.

IV. Experimental Study

This section describes the experimental study of our proposed hybrid data mining approach in large volumes of data based on relational attributes. For that, we use coding language as Java and implemented tool is Net beans for designing user interface based on our requirement. After constructing user interface like super market type business organization, using our proposed approach User Knowledge association mapping addressed between different item sets. Association mapping for different items based on their properties are discussed in section 5.1.

Associated Database Mapping

The client builds up a philosophy on database things. For our situation, beginning from the data relations, the philosophy was made by the master relations. Amid a few sessions, we examined with the master about the data relations ascribes and proposed approach with based parameter sequences. In addition, we discovered other fascinating data inquiring her to build up her desires and learning associated with database qualities with their mapping relations are shown in figure 4.

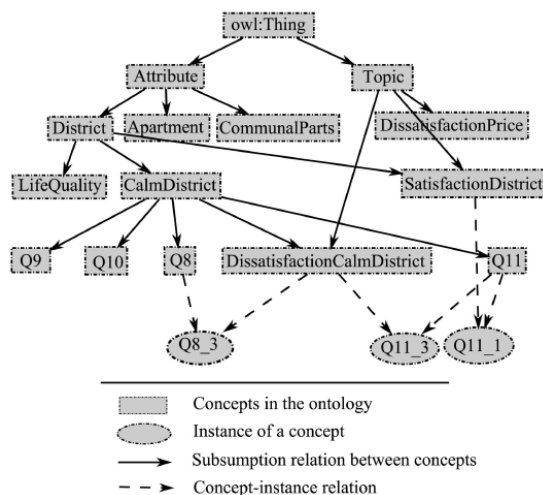


Figure 4. Association-instance mapping relations with User Knowledge visualization.

As a piece of lead mappings, metaphysics ideas are mapped to database things. Accordingly, a few associations amongst philosophy and database can be outlined. Because of execution prerequisites, the metaphysics and the data relations are connected through occasions. The philosophy of data relations association is made physically by the master. For our situation, with the 67 properties and four esteems, the master did not meet any issues to understand the association, however, we concur that for expressive data relations, a manual association could be extremely tedious. That is the reason coordinating a programmed metaphysics development module in our instrument is one of our important points of view.

Results

In this sections, defines the basic experimental results with different data relations and also define the effectiveness of the propose approach at different data relations

Table 1 and table 2 define set of relations. User several schemas during several conferences of testing the new oral appliance examining produced results.

Table 1. Notation meaning with descriptions

Nb	The <i>id</i> of the filter combination
MICF	Minimum improvement constraint filter
IRF	Item-relatedness filter
PRS	Pruning with Rule Schemas
Rule number	The number of rules remaining after filter application

Table 2. Pruning rule schemas for different association rules.

Rule Schema
$RS_5 : < EntryHall \rightarrow CloseSurrounding >$
$RS_6 : < Stairwell \rightarrow EntryHall >$
$RS_7 : < CloseSurrounding > \rightarrow < EntryHall >$
$RS_8 : < EntryHall \rightarrow Stairwell >$
$RS_9 : < CommonAreas \rightarrow GarbageRoom >$
$RS_{10} : < TechnicalMaintenance \rightarrow TechnicalMaintenance >$

Table 3. Filtering schemas for different relations.

Rule Schema	Operator
$RS_1 : < DissatisfactionPrice >$	$C(RS_1)$
$RS_2 : < DissatisfactionCalmDistrict >$	$C(RS_2)$
$RS_3 : < DissatisfactionPrice$ $DissatisfactionCommonAreas >$	C (RS_3)
$RS_4 : < DissatisfactionPrice \rightarrow$ $DissatisfactionCommonAreas >$	U_p/E (RS_4)

Based on above table's information, product information is shown in figure 5 with different item sets.

product_id	product_category	product_name	product_quantity	unit_price	bulk_price	bulk_quantity	barcode_id	rf_id
1	Cosmetics	Grey Flannel for Men Gents Perfume	100	1296	100	129600	12356	1235
2	Cosmetics	ARAMIS 900 EDT 110ML FOR MEN	100	3449	175000	100	12358	1235
3	Cosmetics	HugoDarkBlue For Men	100	1750	100	175000	12358	1235
4	Cosmetics	ROYAL MIRAGE BROWN PERFUME FOR MEN	100	594	59400	100	12359	1235
5	Groceries	Milk-Gold	1000	20	1000	20000	22365	2236
6	Groceries	Milk-Blue	1000	17	1000	17000	22366	2236
7	Groceries	Milk-Orange	1000	15	1000	15000	22363	2236
8	Groceries	Milk	1000	17	17000	1000	22369	2236
9	Groceries	Bread	1000	20	1000	20000	27560	2756
10	Groceries	Salt	1000	10	1000	10000	27561	2756
11	Groceries	Oil	1000	60	1000	60000	27562	2756
12	Groceries	Chaps	1000	15	1000	15000	27563	2756
13	Groceries	Nuts	1000	100	1000	100000	27564	2756
14	Vegetables	Onions	1000	30	1000	30000	77736	7773
15	Vegetables	Carrots	1000	25	1000	25000	12332	1233
16	Vegetables	BeetRoot	1000	40	1000	40000	12322	1232

Figure 5. Different product information based on items presented in business intelligence.

Using this information to construct data relations on supermarket business shown in figure 6 with suitable support and confidence in different sequences.

```

Association Rules
localhost:8084/WebRuleMining/rulemining.jsp
[HugoDarkBlue] => [ROYAL] (conf:100.0%)
[Grey, Milk-Gold] => [ROYAL] (conf:100.0%)
[ARAMIS, Grey, Milk-Gold] => [ROYAL] (conf:100.0%)
[ARAMIS, HugoDarkBlue] => [ROYAL] (conf:100.0%)
[HugoDarkBlue, Milk-Gold] => [ROYAL] (conf:100.0%)
[Milk-Gold] => [ROYAL] (conf:100.0%)
[ARAMIS] => [ROYAL] (conf:33.33333333333336%)
[ARAMIS, Grey, HugoDarkBlue] => [ROYAL] (conf:100.0%)
[ARAMIS, Milk-Gold] => [ROYAL] (conf:100.0%)
[Grey, HugoDarkBlue] => [ROYAL] (conf:100.0%)
[ARAMIS, Grey, HugoDarkBlue, Milk-Gold] => [ROYAL] (conf:100.0%)
[ARAMIS, HugoDarkBlue, ROYAL] => [Milk-Gold] (conf:100.0%)
[ARAMIS] => [Milk-Gold] (conf:33.33333333333336%)
[Grey, HugoDarkBlue, ROYAL] => [Milk-Gold] (conf:100.0%)
[Grey, ROYAL] => [Milk-Gold] (conf:100.0%)
[ARAMIS, Grey, ROYAL] => [Milk-Gold] (conf:100.0%)
[ARAMIS, Grey] => [Milk-Gold] (conf:50.0%)
[HugoDarkBlue] => [Milk-Gold] (conf:100.0%)
[Grey, HugoDarkBlue] => [Milk-Gold] (conf:100.0%)
[Grey] => [Milk-Gold] (conf:33.33333333333336%)
[HugoDarkBlue, ROYAL] => [Milk-Gold] (conf:100.0%)
[ROYAL] => [Milk-Gold] (conf:100.0%)
[ARAMIS, Grey, HugoDarkBlue] => [Milk-Gold] (conf:100.0%)
[ARAMIS, HugoDarkBlue] => [Milk-Gold] (conf:100.0%)
[ARAMIS, ROYAL] => [Milk-Gold] (conf:100.0%)
[ARAMIS, Grey, HugoDarkBlue, ROYAL] => [Milk-Gold] (conf:100.0%)
[ARAMIS, Grey] => [Onions] (conf:50.0%)
[ARAMIS, Milk-Blue, Grey] => [Onions] (conf:100.0%)
[ARAMIS] => [Onions] (conf:33.33333333333336%)
[Bread, Milk-Blue] => [Onions] (conf:100.0%)
[Grey] => [Onions] (conf:33.33333333333336%)
[Milk-Blue] => [Onions] (conf:66.66666666666667%)
[ARAMIS, Milk-Blue] => [Onions] (conf:100.0%)
[Bread] => [Onions] (conf:100.0%)
[Milk-Blue, Grey] => [Onions] (conf:50.0%)
    
```

Total Time Taken to Process : 77431.412510435 seconds.

Figure 6. Association rules based on item set description with their values.

Figure 6 shows the hybrid data mining approach results, after applying classification on each data item associated rule mining sequences shown in table 2,3. Time efficiency to classify different items with relationships performance results are shown in figure 7.

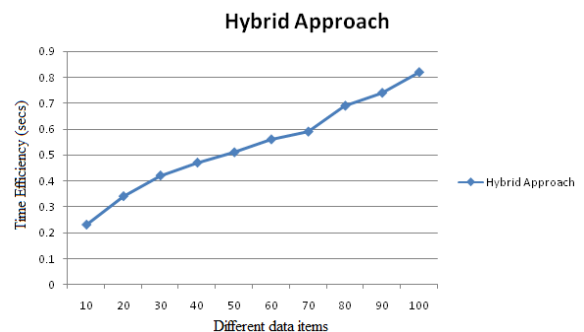


Figure 7. Time efficiency results of hybrid approach with different attributes

As shown in above figure and tables, the proposed approach gives better sequential results for different attributes. Finally, these results give best and effective associated attributes from different business oriented applications effectively.

V. Summery

In this paper, we propose Hybrid data mining approach (which consists association, classification). Data visualization and analysis are the basic relations from different data items based on user knowledge

throughout large volumes of data. The main contribution of this paper as follows: First, we propose rule mining schemas; ontologies are two basic parameters to operate and integrate user knowledge with respect to association rule mining. Second, set of data relations are applicable over different rule schemas in real time aspect in-order to utilize the usage of user in post processing based on several operators like pruning, confirming and exception and others. We apply our proposed approach to integrate user knowledge to reduce number of rule processing in relational databases. Furthermore, based on interesting interactive process to in-depth analysis of data with respect to user knowledge. In future work, we implement different pattern growth evaluation calculations are implemented to elaborate data analysis on transactional data bases with different relations.

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