### Data Mining on Reverse Set Enumeration Tree and DFS for Sequential Pattern Generation

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### **ABSTRACT**

Data mining consist of their may storing of large number of data that generate useful pattern. In database consist of there are large number of group item that group shows set of item. Item may shows value that storing in database. Item may have list that store in database. Data Mining can extract the information that store in database to understand data. Pattern shows the relationship between item set in database.

Keywords: Data mining, Database, Item Set, pattern, Reverse Set

### **I.INTRODUCTION**

Data mining is process of analysis that may large number of quantities of data in order to discover meaningful patterns and rules. Mining may be useful for generating the patterns. Enumeration Tree are the connected graph that consist the search technique can be associated backtrack search that associated with graph. Graph search are similar to Depth First Search(DFS) That Depth first search first that identified the unvisited nodes means DFS can find which nodes not visited on the current path. When current path is completely traversed then selected the next path. DFS consist of nodes can arrange in tree structure.

Tree can fist start in root node and this root have child node. Root have parent node under the root node have child node. DFS can explore the vertex that suspended as soon as new vertex will reach. When new vertex will be explored then vertex vs continue. The search terminates when all reached vertices will be fully explored. Pattern growth approach

### 1.1 Motivation:

The mining process first finds frequent items set based on user-defined support thresholds and then generates association rules from the frequent item set based on the user-defined confidence threshold. Approach to discover frequent item set can divide in two categories first is level-wise and second pattern growth. Data mining can derive the association rule from transaction database such that the presence of some other item. The database is first scanned to find all items with their count. The database is scanned again to construct the tree according to the sorted order. The construction process is executed tuple by tuple from the first transaction to the last one. After all transaction is processed the tree is completely constructed.

### **II.OBJECTIVES**

- Generating sequential pattern in single transaction
- Minimum time to required for pattern generation

### III.PROPOSED WORK

### 3.1 Scope

The scope of the system that patterns will be generate in single transaction. may complete within single scanning that scalability and efficiency of a system may improved. Sequential pattern consist number of item may generate in item set in sequential order transaction complete within single transaction no more time take to complete the transaction. To developed Reverse Set Enumeration Tree input is transactional dataset with name online retail. It is product cell database from that dataset stock code is used as a node in Reverse Set Enumeration Tree. The pattern growth approach can be viewed as searching a reverse set enumeration tree in depth-first manner. The construction of the reverse set enumeration tree follows an imposed ordering of item.

### IV.IMPLEMENTATION MODULE

- Read Dataset
- Reverse Set Enumeration Tree
- DFS Pattern
- 4.1 Read Dataset

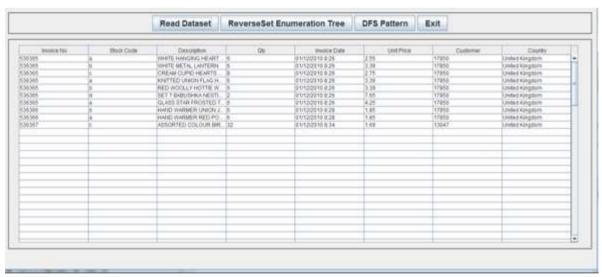


Fig 1: Read Dataset

Data can be collected together and making their set. A standard method to mine high utility patterns is to enumerate each subset of I and test those subset have a utility over the threshold. However such an enumeration is infeasible due to the large number of subsets of I. Fig 1 shows that there are eight coloum and this coloum name is Invoice No, Stock Code,Description,Qty,Invoice Date,Unite price, Customer ,Country. Invoice No show that the number. Stock code show that item define stock number.description show that description of

item.Qty show quantity mean how many quantity are required.Invoice Date shows shows date of that item that may transactions. Unite price that show price of that item.Customer show that the id of that customer and Country shows that which country may implement transaction.

### 4.2 Reverse Set Enumeration Tree

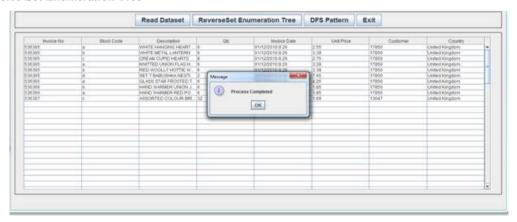


Fig 2: Reverse Set Enumeration Tree

The module for given database D the first reverse set enumeration tree will be generated. It is a tree that includes reverse lexicographic order. Here pattern will be search before its super set. In this tree root will not be contain any item and each node N will be filled by pattern. The child nodes of N will be labeled by its sequential transaction item for sequential pattern mining. Fig 2 shows that there are eight coloum and this coloum name is Invoice No, Stock Code, Description, Qty, Invoice Date, Unite price, Customer, Country. Invoice No show that the number. Stock code show that item define stock number.description show that description of item.Qty show quantity mean how many quantity are required.Invoice Date shows shows date of that item that may transactions. Unite price that show price of that item.Customer show that the id of that customer and Country shows that which country may implement transaction. After completing this transaction the process may completed.

### 4.2.1 Reverse Set Enumeration Tree Structure

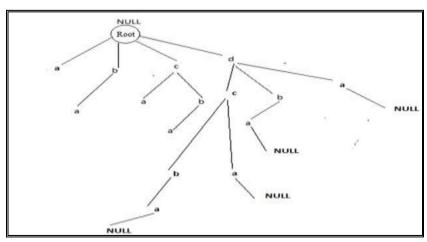


Fig 3: Reverse Set Enumeration Tree Structure

Fig 3 shows that structure of Reverse Set Enumeration Tree. Generating tree in the base of root node and child node.

Root node consists of parent node that place in higher level. This root node can divided in child node. This node can arrange in alphabetic order. There are four node that is a to d. the root have no value that it can denote by null. The method is to approximate an upper bound on utility of all visible patterns represented by nodes in the sub tree rooted at the node currently being explored when growing the reverse set enumeration tree.

### 4.3 DFS Pattern

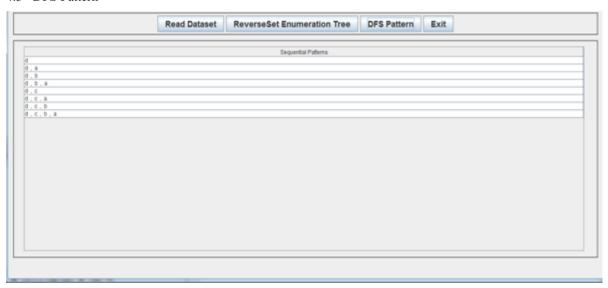


Fig 4: DFS Pattern

Depth first search is another way of traversing graphs which is closely related to reorder traversal of tree. That preorder traversal simply visits each node before its children. The depth first search algorithm then simply initializes a set of markers that tell that which vertex has been visited. Depth first search similar to the breadth first search that finding an unvisited node on the current path and when current path is completely traversed then select the next path. Depth first search is an algorithm for searching a tree data structure. Fig 4 shows generating DFS pattern. Depth first search is another way of traversing graphs which is closely related to reorder traversal of tree. That preorder traversal simply visits each node before its children. The depth first search algorithm then simply initializes a set of markers that tell that which vertex has been visited. Depth first search similar to the breadth first search that finding an unvisited node on the current path and when current path is completely traversed then select the next path. Depth first search is an algorithm for searching a tree data structure. The algorithm starts at the root node of a tree and goes down a given branch. And backtracks until it finds an unexplored path, and then explored it. Depth First Search of a graph differs from exploration of vertex v is suspended as soon as new vertex will be reached, When new vertex will be explored then vertexes of v continue, The search terminates when all reached vertices will be fully explored. Reverse Set Enumeration Tree and traversed with Depth First Search algorithm as specified above and generate sequential pattern

### Algorithm DFS (Node root)

Input: Root
Output: Pattern

Begin

Pattern=[]

If root<> Null Then

Pattern ←root, Data

Childnodelist = root. Children

For Node c in Childnodelist

DFS(c)

Next

End

### V.STEPS OF DATA MINING FOR SEQUENTIAL PATTERN

- Data Cleaning
- Data Integration
- Data Selection
- Data Transformation
- Pattern Evaluation
- Knowledge Presentation
- Data Mining

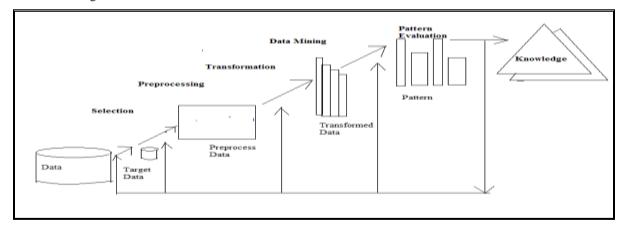


Fig 5: Step of Data Mining For Sequential Pattern

- 5.1 Data Cleaning: This handles noisy, erroneous, missing, or irrelevant data that can clean the entire error.
- 5.2 Data Integration: This handles multiple and heterogeneous data source may be integrated into one.
- 5.3 Data Selection: This handles data relevant to analysis of task that can be generated from database.
- 5.4 Data Transformation: This handle data are transformed into appropriate form for data mining perform summary.
- 5.5 Pattern Evaluation: This identifies the truly interesting patterns representing knowledge based for interesting measure.

- 5.6 Knowledge Presentation: This handle visualization and knowledge representation technique in mined knowledge.
- 5.7 Data Mining: This handle large number of quantities of data in order to discover meaningful patterns

### VI. CONCLUSION

Mining may be useful for generating the patterns. Frequent pattern is important for data mining that all the item in transaction that may represent in values. Finding the item set in data mining it is important. In high utilities patterns of item set may be consist of the group of item in transaction it is called as item set. High utility refers to the set of item that may refer to the database system. High utility item set may mined to the list may generate from mined database. High utility of pattern may identify the item set that utility may satisfy. Mining of high utility of item set is efficiently most of challenging task may consist of cost, quantity, profit that measure the utility.

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