

A Survey on Index Support for Item Set Mining

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Abstract

It is very difficult to handle the huge amount of information stored in modern databases. To manage with these databases association rule mining is currently used, which is a costly process that involves a significant amount of time and memory. Therefore, it is necessary to develop an approach to overcome these difficulties. A suitable data structures and algorithms must be developed to effectively perform the item set mining. An index includes all necessary characteristics potentially needed during the mining task; the extraction can be executed with the help of the index, without accessing the database. A database index is a data structure that enhances the speed of information retrieval operations on a database table at very low cost and increased storage space. The use index permits user interaction, in which the user can specify different attributes for item set extraction. Therefore, the extraction can be completed with the use index and without accessing the original database. Index also supports for reusing concept to mine item sets with the use of any support threshold. This paper also focuses on the survey of index support for item set mining which are proposed by various authors.

Index terms— Frequent Item Set, Index Support, Vertical Index, MTR (Modified Transaction Reduction).

To manage with these databases association rule mining is currently used, which is a costly process that involves a significant amount of time and memory. Therefore, it is necessary to develop an approach to overcome these difficulties. A suitable data structures and algorithms must be developed to effectively perform the item set mining. An index includes all necessary characteristics potentially needed during the mining task; the extraction can be executed with the help of the index, without accessing the database. A database index is a data structure that enhances the speed of information retrieval operations on a database table at very low cost and increased storage space. The use index permits user interaction, in which the user can specify different attributes for item set extraction. Therefore, the extraction can be completed with the use index and without accessing the original database. Index also supports for reusing concept to mine item sets with the use of any support threshold. This paper also focuses on the survey of index support for item set mining which are proposed by various authors.

1 INTRODUCTION

ANY real-life databases are updated through the blocks of occasionally received business information. In these databases, the content is occasionally updated through either insertion of new transaction blocks or removal of outdated transactions. Data can be portrayed as a sequence of incoming data blocks, where new blocks arrive periodically or old blocks are discarded. Examples of developing databases are transactional data from huge retail chains, web server logs, financial stock tickers, and call detail records [15]. Since the data develop overtime, algorithms must be developed to incrementally maintain itemset mining models.

Different categories of examination might be performed over such data like (a) mining all existing data (b) mining only the most current data (e.g., current week data), (c) mining periodical data (e.g., half-yearly data) and (d) mining preferred data blocks (e.g., data associated to the last month of this year and the 2nd month of last year). Consider for instance transactional data from huge retail chains, after shop closing on each day, a

44 set of transactions is inserted into the database. In this situation, market analysts are interested in investigating
45 different portions of the database to realize customer behaviors. For instance, they might be interested in
46 investigating purchases before New Year or during summer vacation.

47 Physical analysis of these huge amount of information stored in modern databases is very difficult. A recognized
48 data mining technique is association rule mining. It is able to discover all interesting relationships which are called
49 as associations in a database. Association rules are very efficient in revealing all the interesting relationships in
50 a relatively large database with huge amount of data. The large quantity of information collected through the
51 set of association rules can be used not only for illustrating the relationships in the database, but also used for
52 differentiating between different kinds of classes in a database. But the major difficulty in association rule mining
53 is its complexity.

54 Association rule mining is a costly method that involves a significant amount of time and memory. Hence,
55 suitable data structures and algorithms must be developed to efficiently carry out the task. Association rule
56 mining is a two-step method: Frequent itemset mining [1] and generation of association rule. Since the first stage
57 is the most computationally intensive knowledge mining task, research activity has been primarily focused on
58 significant and effective algorithms to achieve this extraction task. The data to be analyzed is perhaps extracted
59 from a database and stored into binary files. Numerous algorithms, both memory-based and disk-based are
60 concentrated on specialized data structures and buffer management approach to efficiently mine the desired type
61 of information from a flat dataset.

62 In this paper, itemset mining on developing databases has been focused. A database index [4] is a data
63 structure that enhances the speed of information retrieval operations on a database table at very low cost and
64 increased storage space. Indexes can be generated using one or more columns of a database table, offering the
65 foundation effective access of ordered records. The disk space needed for storing the index is normally less than
66 that needed by the table. Because indices typically include only the key-fields based on which the table is to be
67 organized, and eliminate all the other details in the table. Thus providing the chance to store indices in memory
68 for a table whose data is excessively large to store in memory.

69 Moreover, an index structure for extracting item set as sequence of data blocks. The index supports M
70 Keywords : Frequent Item Set, Index Support, Vertical Index, MTR (Modified Transaction Reduction). user
71 communication, where the user specifies many constraints for itemset extraction. It permits the mining of
72 the complete set of itemsets which satisfy (a) time constraints and (b) support constraints. Since the index
73 contains all feature potentially required during the mining task, the extraction can be carried out by means
74 of the index, without accessing the database. The data representation is absolute, i.e., no support threshold
75 is enforced throughout the index construction stage, to permit reusing the index for mining itemsets with any
76 support threshold.

77 Constraints like support and confidence is not enforced throughout the index creation stage. Therefore, the
78 extraction can be carried out using the index alone, without accessing the original database. As the databases
79 are necessary in almost all the retail stores, super markets, etc., it is necessary to develop an approach for item
80 set mining with the help of index support. There are many approaches available in the literature based on item
81 set mining and index support.

82 2 II.

83 3 LITERATURE SURVEY

84 S. Sahaphong [1] suggested frequent itemsets mining using vertical index list. In this paper, the author proposes
85 a new technique to mine all frequent itemsets [12] that executes database scanning only once to create data
86 structure. This arrangement uses the conceptual of vertical data outline to include transaction data. The
87 altering of minimal support is not effected by the data structure and rescan of database is not needed. This
88 technique has the capability of discovering frequent itemsets without creation of candidate itemsets. It achieves
89 absolute and accurate frequent itemsets. The experimental observation illustrates that this technique provides
90 all definitions and accuracy of frequent itemsets.

91 Mining frequent itemsets from secondary memory was put forth by G. Grahne et al., [2]. Mining frequent
92 itemsets is main function of mining association rules, and it is understood algorithmically for main memory
93 databases. In this approach, the authors examine techniques for mining frequent itemsets when the database or
94 the data structures utilized in the mining are excessively large to fit in main memory. Experimental observations
95 show that this technique reduces the required disk access by order of magnitude, and allow actual scalable data
96 mining.

97 Yin-Ling Cheung et al., [3] suggested mining frequent itemsets without support threshold: with and without
98 item constraints. In traditional association rules mining, a minimum support threshold is considered to be
99 available for mining frequent itemsets. But, fixing such a threshold is characteristically tough. This makes an
100 additional practical difficulty; it is to mine N kitemsets with the maximum supports for k equal to a certain
101 kmax value. The final output is the N-most interesting itemsets. Normally, it is very simple for users to conclude
102 N and kmax value. The author proposed two new approaches, namely LOOPBACK and BOMO. Experimental
103 observation proves that this technique provides better result than the existing Itemset-Loop algorithm, and the
104 output of BOMO can be an order of magnitude enhanced than the original FP-tree algorithm [24], still with

105 the supposition of an optimally chosen support threshold. The author also proposed the mining of "N-most
106 interesting k-itemsets with item constraints." This permits the user to denote different degrees of interestingness
107 for dissimilar itemsets. Experimental observations show that this proposed Double FP-trees algorithm, is very
108 effective in solving this problem which depends on BOMO.

109 E. Baralis et al., [4] recommended itemset mining on indexed data blocks. Numerous attempts have been
110 offered to combine data mining activities with relational DBMSs, but a correct incorporation into the relational
111 DBMS kernel has been infrequently achieved. This paper suggested an innovative indexing method [16], which
112 denotes the transactions in a succinct form, suitable for tightly incorporating frequent itemset mining in a
113 relational DBMS. The data illustration is complete, i.e., no support threshold is imposed, with the intention to
114 permit reusing the index for mining itemsets with any support threshold. In addition, a suitable structure of
115 the stored information has been developed, in order to permit a selective access of the index blocks essential
116 for the current extraction stage. The index has been executed into the PostgreSQL open source DBMS and
117 utilizes its physical level access techniques. Many experiments have been done on several datasets, characterized
118 by dissimilar data distributions. The implementation time of the frequent itemset [12] mining task exploiting
119 the index is constantly similar with and sometime quicker than a C++ execution of the FPgrowth technique
120 accessing data stored on a flat file. A fast algorithm for frequent itemset mining using FP-trees was proposed by
121 G. Grahne et al., [5].

122 Well-organized algorithms for mining frequent itemsets are vital for mining association rules, also for many
123 additional data mining tasks. Techniques for extracting frequent itemsets have been employed by means of
124 a prefix-tree structure called as FP-tree, implemented for storing compressed information regarding frequent
125 itemsets. Many experimental observations have illustrated that these techniques execute very well. In this
126 technique, the author proposed an innovative FP-array method that significantly decreases the necessity of
127 traversing FP-trees, thus acquiring enhanced performance for FP-tree-based algorithms. The proposed approach
128 works particularly well for sparse data sets. Additionally, the author proposed a novel technique for mining
129 all, maximal, and closed frequent itemsets [21]. This approach uses the FP-tree data structure along with the
130 FP-array method effectively and integrates several optimization techniques. Experimental result proves that this
131 technique is the best for many cases. Although this approach takes much memory when the data sets are sparse
132 but it is the fastest technique when the minimum support is low. Furthermore, this technique is the fastest
133 techniques and uses less memory than previous techniques when the data sets are dense.

134 Xuegang Hu et al., [6] suggested mining frequent itemsets using a pruned concept lattice. Extracting frequent
135 itemsets is a critical step in association rule mining. On the other hand, most of the approaches which mine
136 frequent itemsets examine databases numerous times, which reduces the efficiency. In this technique, the
137 association among the concept lattice and frequent itemsets is used, and the method of pruned concept lattice
138 (PCL) is established to characterize frequent itemsets in a specified database, and the scale of frequent itemsets is
139 compressed efficiently. A technique for extracting frequent itemsets based on PCL is implemented, which prunes
140 infrequent concepts appropriately and dynamically throughout the PCL's construction based on the Apriori
141 property. The effectiveness of the approach is illustrated with experiments.

142 Improved paralleled algorithm for mining frequent item-set used in HRM was presented by XuePing Zhang
143 et al., [7]. This approach established the technique of multi-thread processing and a Multi-Threaded Paralleled
144 frequent item-set extraction Algorithm -MTPA was implemented depending on FPtree algorithm [24]. It has been
145 implemented in an enterprise human resources management organization. Based on the experiments of paralleled
146 mining by utilizing increasing multi-thread processing, it is confirmed that MTPA which on the circumstance
147 of multi-core processors can enhance the efficiency of frequent item-set mining successfully. Dong Liyan et al.,
148 [8] proposed a novel method of mining frequent item sets. The goal of mining association rules is to determine
149 the association relationship among the item sets from mass data. In a number of practical applications, its
150 responsibility is mostly to support in decision-making. In this paper, the author proposed an association rule
151 algorithm of mining frequent item sets, which establishes a new data structure and adopts compressed storage
152 tree to develop the run performance of this algorithm. At last, the experiment indicates that the algorithm
153 proposed in this paper has much more advantages in load balance and run time compared with most existing
154 algorithms. Guo Yi-ming et al., [9] presented a vertical format algorithm for mining frequent item sets. Apriori
155 is a traditional algorithm for association rules. With the purpose of obtaining the support degree of candidate
156 sets, Apriori requires to scan the database for several times. This author proposed a new algorithm, which mine
157 frequent item sets by means of vertical format. The proposed technique only needs to scan database one time.
158 And in the follow-up data mining procedure, it can obtain new frequent item sets through 'and operation' among
159 item sets. This technique requires less storage space, and can enhance the efficiency of data mining.

160 Frequent closed informative itemset mining was suggested by Fu et al., [10]. In modern years, cluster analysis
161 and association investigation have attracted a lot of attention for large data analysis such as biomedical data
162 analysis. In this paper, the author developed a novel algorithm of frequent closed itemset mining [21]. The
163 algorithm deals with two challenges of data mining which are mining huge and elevated dimensional data and
164 interpreting the outputs of data mining. Frequent itemset extraction is the main job of association analysis. The
165 approach depends on concept lattice structure with the intention that frequent closed itemsets can be produced
166 to decrease the complicity of extracting all common itemsets and each recurrent closed itemset has additional

167 knowledge to make easy understanding of mining results. From this aspect, the paper also deals with the extension
168 of the approach for cluster analysis. The experimental observation shows the efficiency of this algorithm.

169 A new approach of modified transaction reduction algorithm for mining frequent itemset was proposed by
170 R.E. Thevar et al., [13]. Association rule mining is to take out the interesting association and relation among
171 the huge volumes of transactions. This procedure is segmented into two sub problem: first problem is to discover
172 the frequent itemsets from the transaction and then the second problem is to build the rule from the mined
173 frequent itemset. Frequent itemsets creation is the necessary and most time huge procedure for association rule
174 mining. Currently, most wellorganized apriori-like algorithms rely deeply on the minimum support constraints to
175 prune the enormous amount of non-candidate itemsets. These algorithms store numerous unnecessary itemsets
176 and transactions. In this paper, the authors proposed an innovative frequent itemsets creation algorithm called
177 MTR-FMA (modified transaction reduction depends on frequent itemset mining algorithm) that sustains its
178 performance even at relative low supports. The experimental output also proves that proposed MTR-FMA
179 algorithm on an outset is quicker than high efficient AprioriTid and other algorithms.

180 Lei Wen et al., [14] developed an efficient algorithm for mining frequent closed itemset. Association rule mining
181 was a significant field of data mining investigation. Determining the potential frequent itemset was a vital step.
182 The existed frequent itemset discovery algorithms could find out all the frequent itemset or maximal frequent
183 itemset. N. Pasquier developed an innovative job of mining frequent closed itemset. The size of frequent closed
184 itemset was much lesser than all the frequent itemsets and did not lose any information. In this paper, a new
185 frequent closed itemset approach depends on the directed specified itemset graph. This approach can discover
186 all the frequent closed itemset powerfully by using depth first search method. The experiment report confirms
187 that it is effective for mining frequent closed itemsets.

188 A. Omari [17] developed a new temporal measure for interesting frequent itemset mining. Frequent itemset
189 mining supports the data miner in searching for strongly associated items and transactions in huge transaction
190 databases. Since the number of frequent itemsets is typically very large and uncontrollable for a human user,
191 techniques for mining interesting rules have been developed to identify meaningful and summarized representation.
192 Furthermore, many measures have been implemented in the literature to find out the interestingness of the rule.
193 In this paper, the author establishes a novel temporal measure for interesting frequent itemset mining. This
194 measure depends on the suggestion that interesting frequent itemsets are generally covered by many recent
195 transactions. This measure diminishes the cost of searching for frequent itemsets by reducing the search interval.
196 Furthermore, this measure can be used to improve the search strategy implemented by the Apriori algorithm.

197 A novel parallel frequent itemset mining algorithm was recommended by Chen Xiaoyun et al., [18]. Frequent
198 itemset mining is an essential and important matter in data mining field and can be utilized in many data mining
199 tasks. The majority of these mining tasks need multiple passes over the database and if the database size is huge,
200 which is typically the case; scalable elevated performance solutions involving multiple processors are needed. In
201 this paper, the authors proposed a new equivalent frequent itemset mining approach which is called HPFP-Miner.
202 The proposed technique depends on FP-Growth and establishes modest communication overheads by efficiently
203 partitioning the list of frequent elements list over processors. The reports of experiment prove that HPFP-Miner
204 has good scalability and performance.

205 E. Baralis et al., [19] suggested itemset mining on indexed data blocks. This paper proposes a novel index,
206 called I-Forest, to maintain data mining activities on developing databases, whose content is occasionally updated
207 through insertion or deletion of data blocks. I-Forest permits the mining of itemsets from transactional databases
208 such as transactional data from huge trade chains. Item, support and time constraints may be imposed throughout
209 the extraction phase. The proposed index is a covering index that corresponds to transactional blocks in a succinct
210 form and permits different kinds of examination (e.g., analyze quarterly data). Throughout the creation stage no
211 support constraint is needed. Thus, the index offers an entire illustration of the developing data. The I-Forest
212 index has been executed into the Post-greSQL open source DBMS and develop its physical level access methods.
213 Experiments conducted on both sparse and dense data distributions. The implementation time of the frequent
214 itemset mining task exploiting the index is always comparable with and for least support threshold quicker than
215 the Prefix-Tree algorithm accessing static data on at file.

216 Jian Chen et al., [20] provided a method for new items based on indexing techniques. The amount of
217 information in the World Wide Web is growing very rapidly than our skill to process. Recommender systems
218 provide knowledge innovation approaches to assist people find what they actually want. These approaches consist
219 of collaborative filtering (CF), association rules detection and Bayesian networks, etc. But all of these methods
220 have a significant disadvantage: items or pages which being uploaded to a site in recent times cannot be found.
221 This is commonly called as the new item problem. The authors established a common structure for solving this
222 difficulty and present a single index arrangement x-features-tree for using heuristic information retrieval approach
223 to find the right items for the right users.

224 Mining association rules from XML data with index table was suggested by Xin-Ye Li et al., [22].

225 Mining XML association rule is tackled with extra challenge because of the inherent flexibilities of XML in
226 both arrangement and semantics. With the purpose of making mining XML association rule very efficient, this
227 paper provides a new definition of transaction and item in XML environment, then construct transaction database
228 depending on an index table. Based on the definition and the index table utilized for XML searching, it is easy
229 to check the relation among the transaction and retrieve an item quickly. A high adaptive mining approach is

230 also illustrated. By using this approach, mining rules can be processed with no assistance of interest associations
231 specified by users and mining unknown rules. The effectiveness of these approaches is proved with the help of
232 experiments on real-life data.

233 L. Golab et al., [23] proposed indexing time method for evolving data with variable lifetimes. Numerous
234 applications store data items for a predetermined, fixed duration of time. Examples consist of sliding windows
235 over online data streams, in which old data are thrown out as the window slides forward. Earlier researches
236 on management of data with limited lifetimes have emphasized online query processed in main memory. In
237 this approach, the authors concentrate on the difficulty of indexing time-developing data on disk for offline
238 investigation [16]. With the intention of decreasing the I/O costs of index updates, existing work separates the
239 data chronologically. Thus, only the previous separation is examined for expirations, only the youngest separations
240 acquire insertions, and the remaining partitions in the middle are not processed. On the other hand, this result is
241 based upon the hypothesis that the order in which the data are introduced is equivalent to the termination order,
242 which means that the lifetime of each data item is the similar. In order to break this hypothesis, the authors
243 reveal that the existing solutions no longer be relevant, and suggested a new index partitioning strategies that
244 provide low update costs and quick access times. E.J. Keogh et al., [25] proposed an indexing scheme [11] for fast
245 similarity search in large time series databases. This paper addresses the trouble of similarity searching in huge
246 time-series databases. The authors proposed an innovative indexing approach that permits quicker retrieval. The
247 index is produced by generating bins that include time series subsequences of roughly the similar shape. For
248 every bin, this proposed approach can rapidly compute a lower bound on the distance among a given query and
249 the most similar element of the bin. This bound permits to search the bins in greatest-first order, and to prune
250 some bins from the search space without verifying the contents. Further speedup can be achieved by optimizing
251 the data inside the bins in such a way that ignores the process of comparing the query to every item in the bin.
252 This technique could be called as STB (Shape To Bit-vector) indexing, and experimentally confirm it on space
253 telemetry, medical and synthetic data, demonstrating roughly an order-of-magnitude speedup.

254 III.

255 4 PROBLEMS AND DIRECTIONS

256 Many problems are faced in the above discussed existing techniques. To solve these problems many researches has
257 to be done. The research should be focussed on the following areas.

258 5 a) Problems involved in Static Datasets

259 Most of the data mining algorithms available in the literature concentrate on the static datasets. These are the
260 kind of the datasets which do not change over time. So, there is a need for some researches to focus on this static
261 dataset problem and to develop some technique which supports the developing database based on index support
262 to mine the item sets. Incremental update of the index will solve this difficulty. A technique should be developed
263 to update the index whenever a new data is inserted.

264 6 b) Use of Correlation Information

265 It takes more time to read all the physical data blocks. Therefore, it is necessary to reduce the number of physical
266 data blocks read throughout the mining process. For this purpose, correlated information can be stored in the
267 same block. Also the system can be adopted for the tree based rule mining schemes and the system can be
268 improved to detect positive and negative rule mining process.

269 IV.

270 7 CONCLUSION

271 In recent business trends, it is necessary to transform the data available in a database into an informational
272 advantage. Data mining is the process of extracting relatively useful information from a large data base. Since
273 the usage of database increased in all the fields of research and also in the retail shop, there is a need for some
274 techniques to make these available databases into a sequence of valuable information. For the quick retrieval of
275 information indexes can be used in item set mining. It reduces the cost of storage and provides quick access
276 of information. The available information in databases can be arranged based on some order, in such a way
277 that should provide strong association among the data. Many algorithms and techniques have been discussed
278 in the literature, which exists in the current approaches. The problems faced in the existing techniques and the
279 directions will form the basis for the innovation of the new approaches. This survey will help the researchers to
280 develop an efficient technique based on index support for item set mining. ^{1 2}

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7 CONCLUSION

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