

Data Mining Approach for Children's Frequent Medicine Item Set using FP-Growth Algorithm

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Abstract— Know-how mining is a redesigning innovation for studying extraction in countless fields like restorative, instructive, today's, and many others. Extricate a critical understanding from large database is most crucial component. Know-how extraction process were finished through numerous methods like component extraction, forecast, characterization, etc. for our examination investigations expectation of understanding digging helps a first-rate deal for attending to useful knowledge. In this paper we targeting road site visitors dataset and we utilized fluffy expertise extraction for participation work by using FCM. For the understanding extraction method right here we actualized the connection and coefficient calculation for avenue visitors dataset and property decrease had been completed through utilising Genetic calculation ultimately with the help of A-Priori calculation we create the ordinary for the digging the associate object for spotlight lower.

Key words: Data Mining, Prediction, Feature Reduction, Fuzzy, Association Rule and Rule Generation

I. INTRODUCTION

In recent years the dimensions of info has multiplied speedily. This has crystal rectifier to a growing interest within the development of tools capable within the automatic extraction knowledge of information from data. The term data mining or knowledge discovery in database has been adopted for a field of research dealing with the automatic discovery of implicit information or knowledge within the databases. The implicit info among databases, in the main the attention-grabbing association relationships among sets of objects that cause association rules might disclose helpful patterns for call support, prognosis, selling policies, even diagnosis and lots of different applications.

The problem of mining frequent itemsets arose 1st as a sub-problem of mining association rules. Frequent itemsets play an important role in several data processing tasks that attempt to notice attention-grabbing patterns from databases like association rules, correlations, sequences, classifiers, clusters and plenty of additional of that the mining of association rules is one among the foremost common issues. the first motivation for looking association rules came from the requirement to research thus referred to as market group action knowledge, that is, to look at client behavior in terms of the purchased product. Association rules describe however typically things ar purchased along, as an example, associate association rule “beer, chips (80%)” states that four out of 5 customers that bought brewage conjointly bought chips. Such rules may be helpful for choices regarding product rating, promotions, store layout and plenty of others.

II. RELATIVE STUDY

A. Designing a Knowledge Discovery of Clustering Techniques in Pharmaceutical Compounds, V. Palanisamy and A. Kumarkombaiya

To develop data processing techniques to support higher cognitive process and discovery of useful cluster of the property atom for drug effects by analyzing substance information within the type of structured information. Existing studies in data processing largely specialize in ranked clump techniques applied in giant and little dataset of pharmaceutical compound and analyse its performance supported time accuracy. during this paper focuses to use cluster techniques of partition technique like increased K-means algorithmic rule and ranked technique like Birch and Chameleon algorithmic rule utilized in pharmaceutical compound specifically painted as atom range, atom name like carbon, hydrogen, nitrogen, gas with connected atoms. These dataset kind a useful cluster of atoms by functioning in 3 phases. The performance is experimented supported time taken to create the calculable cluster, additionally overall execution time is reduced by improvement of increased Kmeans algorithmic rule in comparison to chameleon and Birch algorithmic rule.

B. Data Mining Usage and Applications in Health Services, Mehmet Akif Cifci, Sadiq Hussain

Data Mining (DM), accustomed extract massive amounts of hidden, valuable, helpful data in massive quantities and to supply strategic call support, has created a brand new perspective on the utilization of health knowledge. it's become a apace growing technique of responding to problematic areas of information in massive quantities in most sections. though in health services it appears to be slow, a significant leap has return to the scene. The aim of this study is to supply a brand new perspective on decision-making processes by making associate infrastructure for the health knowledge and to supply examples for aid employees within the aid business victimisation DM techniques. Forasmuch as, the abstract framework of information discovery in databases, knowledge reposition, DM, Business Intelligence (BI) has been given. DM applications and usages square measure given as samples of priority problems and downside areas within the health sector.

C. Data Mining for Health Care Industry: A Practical Machine Learning Tool, Zeinab Sam Daliri

Healthcare generates mountains of body knowledge concerning patients, hospitals, claims, etc. Clinical trials, electronic patient records and laptop supported illness management can more and more manufacture mountains of clinical knowledge. data processing product ar designed to

require this one stage any. It brings the ability to find patterns and correlation hidden at intervals the information repository and assists professionals to uncover these patterns and place them to figure. Therefore, selections rest with health care professionals, not the data system specialists. The key to triple-crown data processing is to 1st outline the business or clinical downside to be resolved. So information will mechanically be obtained by the utilization of machine learning techniques within the hands of health care decision-makers. data processing applications will greatly advantages most areas concerned in health care business. the massive amounts of knowledge generated by health care transactions ar too complicated and enormous to be processed and analyzed by ancient strategies. data processing provides the methodology and technology to rework Brobdingnagian quantity data} into helpful information for deciding. This paper explores data processing applications in major areas like analysis of treatment effectiveness, and management of health care. KEYWORDS: Machine Learning, data processing, health care application, knowledge deposition.

III. PROPOSED ALGORITHM

A. FP Growth algorithm

FP Growth is one in all the fundamental algorithmic rule use for generate association rules. FP growth is AN approach supported divide and conquers technique. the most purpose of this method is to provide frequent item sets by victimization the mix of knowledge attributes. It essentially works on to come up with frequent item set while not candidate set generation. Another sister technique of FP growth is understood as apriori algorithmic rule additionally use in association data processing. Apriori may be a classic algorithmic rule for learning association rules. Apriori is meant to control on databases containing transactions (for example, collections of things bought by customers, or details of a web site frequentation). a method to construct a less complicated model computed from information, easier to grasp and with a lot of prognostic power is to form a collection of simplified rules.

The major steps of FP growth ar include the subsequent steps:

- 1) "Step1- 1st condenses the information showing frequent item set in to FP-tree.
- 2) Step2: It divides the FP-tree in to a collection of conditional information and mines every information severally, so extract frequent item sets from FP-tree directly. It include one root labeled as null, a collection of item prefix sub trees because the kids of the basis, and a frequent .item header table. every node within the item prefix sub tree consists of 3 fields: item-name, count and node link where---itemname registers that item the node represents; count registers the quantity of transactions diagrammatic by the portion of path reaching this node, node link links to successive node within the FP- tree. every item within the header table consists of 2 fields--- item name and head of node link, that points to the primary node within the FP-tree carrying the item name".

B. FP Growth Pseudo Code

Input: constructed FP-tree

Output: complete set of frequent patterns

Method: Call FP-growth (FP-tree, null).

procedure FP-growth (Tree, α)

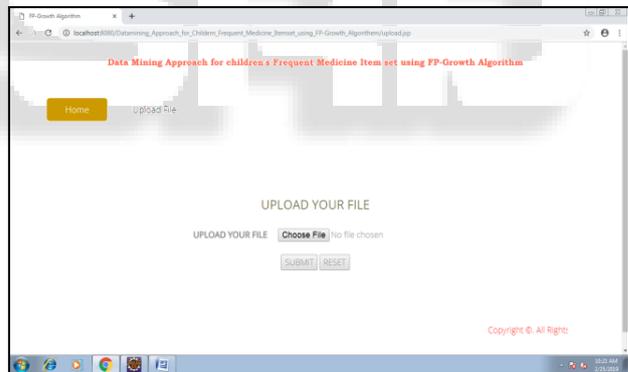
- ```
{
1) if Tree contains a single path P then
2) for each combination do generate pattern β α with support = minimum support of nodes in β .
3) Else For each header ai in the header of Tree do {
4) Generate pattern $\beta = ai$ α with support = ai.support;
5) Construct β .s conditional pattern base and then β .s conditional FP-tree Tree β
6) If Tree $\beta =$ null
7) Then call FP-growth (Tree β , β) }
```

### IV. SCREEN SHOTS

#### A. Index Page



#### B. Upload Page



#### C. Home Page



D. View Page

| S.No | Childrens Names          | Age | Diseases            | Medicines                       | Months  |
|------|--------------------------|-----|---------------------|---------------------------------|---------|
| 1    | Chitrani Abhinav         | 10  | Fever               | Hamdard Khamira Marwareed Khas  | January |
| 2    | Immedi Divya Sai         | 9   | Chickenpox          | Potassium                       | January |
| 3    | K Sai Satya Setu Mohan   | 1   | Diarrhea            | Amicolon Capsule                | January |
| 4    | Alkella V S Vinay Kartik | 6   | Diarrhea            | Amicolon Capsule                | January |
| 5    | Borithu Adha Kiran       | 7   | Diarrhea            | Amicolon Capsule                | January |
| 6    | Pensubaru Velmurugan     | 9   | Fever               | Cough & Runny Nose              | January |
| 7    | P Harish Kumar           | 8   | 5th disease         | SBL Varicellum Dilution 1000 CH | January |
| 8    | Nareddy Chandra Shekar   | 5   | Common cold         | Abratarone                      | January |
| 9    | Alula Gruhith Sai        | 1   | Fever               | Himalaya Septilin               | January |
| 10   | Mamidi Sumikendar        | 5   | Hematologic disease | Cough & Runny Nose              | January |
| 11   | Guggilla Nayya           | 10  | 5th disease         | Abratarone                      | January |
| 12   | Immedi Divya Sai         | 6   | Hematologic disease | Himalaya Septilin               | January |

E. Extract Page

| S.No | Childrens Names          | Age | Diseases            | Medicines                       | Months  |
|------|--------------------------|-----|---------------------|---------------------------------|---------|
| 1    | Chitrani Abhinav         | 10  | Fever               | Hamdard Khamira Marwareed Khas  | January |
| 2    | Immedi Divya Sai         | 9   | Chickenpox          | Potassium                       | January |
| 3    | K Sai Satya Setu Mohan   | 1   | Diarrhea            | Amicolon Capsule                | January |
| 4    | Alkella V S Vinay Kartik | 6   | Diarrhea            | Amicolon Capsule                | January |
| 5    | Borithu Adha Kiran       | 7   | Diarrhea            | Amicolon Capsule                | January |
| 6    | Pensubaru Velmurugan     | 9   | Fever               | Cough & Runny Nose              | January |
| 7    | P Harish Kumar           | 8   | 5th disease         | SBL Varicellum Dilution 1000 CH | January |
| 8    | Nareddy Chandra Shekar   | 5   | Common cold         | Abratarone                      | January |
| 9    | Alula Gruhith Sai        | 1   | Fever               | Himalaya Septilin               | January |
| 10   | Mamidi Sumikendar        | 5   | Hematologic disease | Cough & Runny Nose              | January |
| 11   | Guggilla Nayya           | 10  | 5th disease         | Abratarone                      | January |
| 12   | Immedi Divya Sai         | 6   | Hematologic disease | Himalaya Septilin               | January |

F. Search File

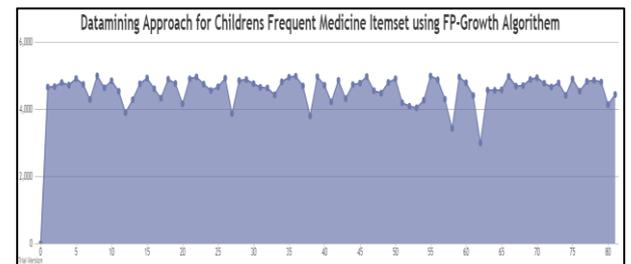
G. Search File Out Put

| S.No | Childrens Names          | Age | Diseases            | Medicines                       | Months  |
|------|--------------------------|-----|---------------------|---------------------------------|---------|
| 1    | Chitrani Abhinav         | 10  | Fever               | Hamdard Khamira Marwareed Khas  | January |
| 2    | Immedi Divya Sai         | 9   | Chickenpox          | Potassium                       | January |
| 3    | K Sai Satya Setu Mohan   | 1   | Diarrhea            | Amicolon Capsule                | January |
| 4    | Alkella V S Vinay Kartik | 6   | Diarrhea            | Amicolon Capsule                | January |
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| 9    | Alula Gruhith Sai        | 1   | Fever               | Himalaya Septilin               | January |
| 10   | Mamidi Sumikendar        | 5   | Hematologic disease | Cough & Runny Nose              | January |
| 11   | Guggilla Nayya           | 10  | 5th disease         | Abratarone                      | January |
| 12   | Immedi Divya Sai         | 6   | Hematologic disease | Himalaya Septilin               | January |
| 13   | Kani Sai Kiran           | 11  | Diarrhea            | Hamdard Khamira Marwareed Khas  | January |
| 14   | Nagbor Saikesh           | 6   | Fever               | Potassium                       | January |

H. FP-Growth Algorithm

| Diseases, Medicines                                          | Count |
|--------------------------------------------------------------|-------|
| Fever Hamdard Khamira Marwareed Khas                         | 4665  |
| Pneumonia Amicolon Capsule                                   | 4680  |
| Attention deficit hyperactivity disorder, Cough & Runny Nose | 4800  |
| 5th disease Abratarone                                       | 4725  |
| Chickenpox, Potassium                                        | 4920  |
| Diarrhea Amicolon Capsule                                    | 4745  |
| Pneumonia SBL Varicellum Dilution 1000 CH                    | 4295  |
| Pneumonia, Cough & Runny Nose                                | 5000  |
| Attention deficit hyperactivity disorder, Potassium          | 4645  |
| Mumps Hamdard Khamira Marwareed Khas                         | 4855  |
| Pneumonia Himalaya Septilin                                  | 4540  |
| Pneumonia Hamdard Khamira Marwareed Khas                     | 3910  |
| Hematologic disease Hamdard Khamira Marwareed Khas           | 4290  |

I. Graph Page



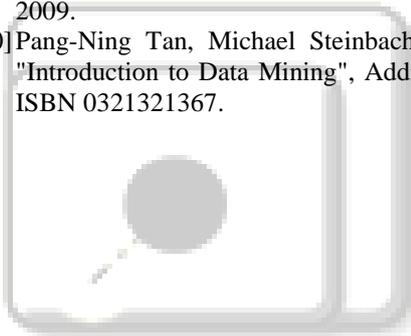
V. CONCLUSION

these days world the massive quantity of knowledge has been saving frequently within the enterprises. there's a necessity of special treatment for the simplest use of those information. we have a tendency to applied In there during this study data processing techniques that think about the simplest thanks to extract new info from the info. Started with the massive quantity of knowledge during this analysis, once choice some attributes we have a tendency to bestowed the simplest use of knowledge by making some association rules for the medical students and doctors facilitate. data management is providing the power to seek out these rules any time once would like. Additionally update version of association rules can modify the previous rules consistent with new information collected from many sources.

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