# Comparative Study of various association rule-mining algorithms applied with special reference to Educational Data Mining

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## Abstract:

Data is important property for everyone. Large amount of data is available in the world. Educational Data Mining helps in understanding the trends, inclination of students as well as their performance. This analysis can help in adapting and designing student cerbic teaching-learning methods. In this paper, we have studied what is Educational data mining, various techniques of EDM and methods used for association-rule mining algorithm. Algorithms are compared based on some performance factors like accuracy, data support, execution speed, etc.

*Keywords* — Data Mining (DM), Educational data mining (EDM), Educational systems, Association rulebased algorithm.

### **I. INTRODUCTION**

Data mining is the computing process of discovering patterns in large data set involving methods at the intersection of machine learning, statistics. and database system. It is an interdisciplinary subfield of computer science. The overall goal of the data mining process is to extract information from a data set and transform it into an understandable structure for further use. Aside from the raw analysis step, it involves database and data management aspects, data pre-processing, model and inference considerations, interestingness metrics, complexity considerations, post-processing of discovered structures, visualization, and online updating. Data mining is the analysis step of the "knowledge discovery in databases" process, or KDD.

Educational data mining (EDM) is the latest advancement in education field, dealing with the development of techniques for exploring and analysing the huge data that come from the educational database. This involves analysing the results in depth in order to monitor the student's

academic activities closely. Competition between the institutions is an all-time high forcing management to focus on increasing enrolment and registration whilst controlling costs. Data mining, which is defined as the process of extracting previously unknown knowledge, and detecting the interesting patterns from a massive set of data, offers a way of dealing with this problem in educational institutions.

In most of the colleges several assessments are prepared to evaluate student performance. Types of assessments that are common to all the colleges and institutions are assignments, sessional, and practical. With the help of sessional marks, assignments and practical the professor assesses the performance of students but this become the very tedious task and need to maintain the records. However, it also takes time to do these tasks because they need to formulate their own calculation for evaluation of result. These tasks will become much easier if all calculations for class position (ranking) can be automatically done when the teacher enter assignments, practical and exams marks. Students result is the criteria for the

measurement of the student's capability in terms of academic performance in the any academics. It is used to measure a student's capability in each subject offered by the student, because the student result is very important to the student and his/her parents, there is need to avoid common mistakes made during the result processing, this can easily be achieved by an automated result processing system, The system is an effective, efficient and error free results processing system designed and implemented for any college or institution.

### **II.LITRATURE SURVY**

Nur Hani Zulkifli, Jamalahyahaya, Aziz Deraman has proposed the empirical study which has been done to discover requirement analysis process in Business Intelligence and Analytics implementation. The approach involves interviewing business intelligence experts in various domains. The proposed model can be used as a guideline for practitioners in analyzing requirements for BIAPM[1].

Mohiuddin Ali Khan, WajeGharibi, Sateesh Kumar Pradhan has proposed the analysis of real data collected from several institutions, a sample study has been conducted to show how the Apriori Algorithm can be used in educational field and the results have been observed. Data mining can be used effectively in educational institutes for leading education activities in an effective way, for watching students' performances continuously and directing students in course and profession choosing. Thus, the level of student's success can be raised, and we can concentrate on students weakness at selected courses thus improving the standard of students[2].

Leena Khanna,Dr.Shailendra NarayanSingh, Dr. MansafAlam has proposed the systematic review on EducationalData Mining technique which helps in predicting thefuture and changing the future. It containsapplicationareas of EDM,various methods of EDMand the factors affecting the Student's academicperformance and the teaching learning process[3].

Oswaldo Moscoso-Zea1, Andres-Sampedro, and Sergio Luján-Mora has proposed the differences between existing methodologies for DW design.

Looking at differences between a business DW design and an educational DW design we found out that the value chain is the most important characteristic that changes the vision of the implementation project. Kimball and Inmon are the two most popular and recommended methodologies in the literature[4].

Lida Shams, Hassan Rashidi has proposed the BI architecture based on the service oriented concept (i.e. the flexible services that can be accessed on demand) is used for analyzing the current situation of educational courses and predicting their near future in order to support decision making. Thus the proposed SOA for BI considerably improves the decision making process[5].

Mohammed I.Al. Twijri, Amin Y. Noaman has proposed a new Data Mining model to be applied in higher education institutions. The Suggested model assists in decision modeling process in the strategic levels of higher institutions as well as regulates the disciplines of student's admission[6].

Ms.TismyDevasia ,Ms.Vinushree T P, Mr.VinayakHegde has proposed the web based application which makes use of the Navie Bayesian mining technique for the extraction of useful information. The system aims at increasing the success graph of students using Naive Bayesian and the system which maintains all student admission details, course details, subject details, student marks details, attendance details, etc[7].

| Name      | Source    | year | Description       |
|-----------|-----------|------|-------------------|
|           | of        |      |                   |
|           | publicati |      |                   |
|           | on        |      |                   |
| А         | Conferen  | 2015 | Study distillates |
| Review    | ce        |      | the gaps in       |
| on        |           |      | existing          |
| Predictin |           |      | prediction        |
| g         |           |      | methods,          |
| Student's |           |      | variables that    |
| Performa  |           |      | define the        |
| nce using |           |      | academic          |
| Data      |           |      | performance and   |
| Mining    |           |      | the prediction    |
| Techniqu  |           |      | methods that can  |
| es.       |           |      | be used to        |

III. EDM SURVEY OF PAPERS

|           |          |      | determine the      | and       |         |      | EDM from 2002      |
|-----------|----------|------|--------------------|-----------|---------|------|--------------------|
|           |          |      | students'          | Predictio |         |      | to 2014 with main  |
|           |          |      | performance.       | n in      |         |      | focus on papers    |
|           |          |      | Techniques         | Educatio  |         |      | published in       |
|           |          |      | utilized in        | nal Data  |         |      | Educational Data   |
|           |          |      | previous           | Mining:   |         |      | Mining.            |
|           |          |      | researches :       | A         |         |      | Predicting         |
|           |          |      | Neural Network.    | Research  |         |      | Academic           |
|           |          |      | decision Tree.     | Travelog  |         |      | Performance with   |
|           |          |      | SVM. K.NN.         | ue        |         |      | Pre/Post           |
|           |          |      | Naïve Baves with   |           |         |      | Enrolment          |
|           |          |      | Neural Network     |           |         |      | Factors. DM        |
|           |          |      | having the highest |           |         |      | techniques to      |
|           |          |      | accuracy and       |           |         |      | predict learning   |
|           |          |      | Classification     |           |         |      | performance,       |
|           |          |      | method the most    |           |         |      | association        |
|           |          |      | frequent in EDM.   |           |         |      | between Pre/Post   |
| Modellin  | Conferen | 2015 | In this study the  |           |         |      | Enrolment          |
| g Key     | ce       |      | author developed   |           |         |      | Factors and        |
| paramete  |          |      | an effective       |           |         |      | Employability,     |
| rs in     |          |      | model to predict   |           |         |      | DM applications    |
| Higher    |          |      | the effect of      |           |         |      | in supplementary   |
| Educatio  |          |      | various identified |           |         |      | areas of education |
| n using   |          |      | parameters in      |           |         |      | like Faculty       |
| Logistic  |          |      | CGPA of the        |           |         |      | Evaluation,        |
| Regressio |          |      | students using     |           |         |      | Analysis of the    |
| n: an     |          |      | Logistic           |           |         |      | Curriculum         |
| Indian    |          |      | Regression.        |           |         |      | Chosen by          |
| case      |          |      | Parameters like    |           |         |      | Students etc.      |
| based     |          |      | Academic           |           |         |      | Need of            |
| Data      |          |      | background,        |           |         |      | Integrated         |
| Analysis  |          |      | Family Closeness   |           |         |      | Models/Framewo     |
| •         |          |      | and Freedom to     |           |         |      | rks for all the    |
|           |          |      | make Choices etc   |           |         |      | stakeholders to    |
|           |          |      | does have an       |           |         |      | ensure             |
|           |          |      | impact on the      |           |         |      | Sustainable        |
|           |          |      | students overall   |           |         |      | Growth             |
|           |          |      | CGPA. The          | Educatio  | Journal | 2014 | This paper         |
|           |          |      | model could not    | nal Data  |         |      | highlights         |
|           |          |      | make a distinction | Mining    |         |      | Evaluation of      |
|           |          |      | between the Low    | and its   |         |      | Students and       |
|           |          |      | and Medium         | role in   |         |      | Teachers           |
|           |          |      | CGPA and High      | Educatio  |         |      | performances       |
|           |          |      | and medium         | nal field |         |      | using different    |
|           |          |      | CGPA.              |           |         |      | Data mining        |
| Performa  | Journal  | 2015 | Comprehensive      |           |         |      | methods.           |
| nce       |          |      | literature review  |           |         |      | Attributes         |
| Analysis  |          |      | of researches in   |           |         |      | considered in the  |

|  |         |      | paperare:Attendance,Assignment,GPA,GPA,Sessionalmarksand finalgrade.The mainlimitation of thisstudyisthatlimitedattributes   |   |                |      | sequencing, and<br>teachers support<br>etc Analysis and<br>discovery of<br>useful patterns<br>from these<br>approaches<br>identified.  |
|--|---------|------|--|---|----------------|------|--|
| Learning<br>Analytics<br>and<br>Educatio<br>nal Data<br>Mining in<br>Practice:<br>A<br>Systemati<br>c<br>Literature<br>Review<br>of<br>Empirical<br>Evidence | Journal | 2014 | are studied.<br>Primary research<br>objective<br>included<br>identification of<br>factors such as<br>grades,<br>demographic<br>characteristics,<br>students'<br>portfolios,<br>students'<br>participation and<br>enrolment in<br>various activities,<br>multimodal skills, | Educatio<br>nal data<br>mining<br>applicatio<br>ns  | Journal        | 2014 | Broad areas of<br>application in<br>which EDM can<br>be applied to e-<br>learning are<br>discussed. Major<br>application areas<br>include: User<br>Modelling, User<br>Knowledge<br>Modelling, User<br>Behaviour<br>Modelling, User<br>Profiling, Domain<br>Modelling, Trend<br>Analysis etc. |
|  |         |      | students' mood<br>etc. and<br>evaluating the<br>affect of these<br>parameters on<br>his/her<br>performance.  | Educatio<br>nal data<br>mining:<br>A review   | Conferen<br>ce | 2013 | Targeted 9 latest<br>studies to which<br>the data mining<br>methods are<br>applied in<br>educational<br>setting ranging<br>from the year   |
| Educatio<br>nal data<br>mining:<br>A survey<br>and a<br>data<br>mining<br>based<br>analysis<br>of recent<br>works  | Journal | 2014 | Main areas of<br>EDM focussed in<br>this paper is<br>Students<br>modelling<br>approaches which<br>includes<br>behaviour,<br>performance,<br>assessment,<br>student support<br>and feedback,<br>domain<br>knowledge,  | Techniqu<br>es<br>focused :<br>Classifica<br>tion,<br>Predictio<br>n,<br>Associati<br>on Rule<br>Analysis,<br>Sequentia<br>1 Pattern<br>Increasin | Journal        | 2013 | 2004 until 2012.Data mining isnecessary inorganizations toenhancecompetitiveadvantage anddecision making.Educational dataMining helps toanalyze theeducational dataand developingmodel for   |

| g Ouality           |          |      | improving           |          |         |      | models and           |
|---------------------|----------|------|---------------------|----------|---------|------|----------------------|
| of                  |          |      | institutional       |          |         |      | distillation of data |
| Educatio            |          |      | effectiveness       |          |         |      | for human            |
| n Using             |          |      | Educational Data    |          |         |      | iudoment are the     |
| n Using<br>Educatio |          |      | Mining is a kay     |          |         |      | judgment are the     |
|                     |          |      | Willing is a key    |          |         |      | main techniques      |
| hai Data            |          |      | area in mining      |          |         |      | uses in EDM with     |
| Mining              |          |      | students            |          |         |      | the fourth and       |
|                     |          |      | performance and     |          |         |      | fifth kind having    |
|                     |          |      | helpful in          |          |         |      | prominence           |
|                     |          |      | predicting          |          |         |      | within               |
|                     |          |      | educational         |          |         |      | educational data     |
|                     |          |      | institutions        |          |         |      | mining. Author       |
|                     |          |      | performance         |          |         |      | discussed the        |
|                     |          |      | taking parameters   |          |         |      | main application     |
|                     |          |      | like Teaching       |          |         |      | areas of EDM         |
|                     |          |      | Skills, Course      |          |         |      | which includes       |
|                     |          |      | content.            |          |         |      | enhancing the        |
|                     |          |      | Infrastructure etc. |          |         |      | existing student     |
| A Survey            | Conferen | 2012 | This paper          |          |         |      | models that          |
| and                 | ce       | 2012 | focuses on          |          |         |      | present              |
| Future              | 66       |      | importance of       |          |         |      | comprehensive        |
| Vision of           |          |      | need of data        |          |         |      | information about    |
| Vision of           |          |      | mining in various   |          |         |      | a student's          |
| Data<br>mining in   |          |      | Educational         |          |         |      | individuality        |
| Educatio            |          |      | Sustam which        |          |         |      | discovering or       |
| Educatio            |          |      | System which        |          |         |      | improving the        |
| nal Field           |          |      |                     |          |         |      | domain modele        |
|                     |          |      | traditional, web-   |          |         |      | domain models,       |
|                     |          |      | based, e-learning   |          |         |      | studying the         |
|                     |          |      | and intelligent     |          |         |      | pedagogical          |
|                     |          |      | tutoring system     |          |         |      | provision            |
|                     |          |      | etc. This paper     |          |         |      | stipulated by        |
|                     |          |      | describes the use   |          |         |      | education            |
|                     |          |      | of various          |          |         |      | software,            |
|                     |          |      | techniques such     |          |         |      | systematic           |
|                     |          |      | as prediction and   |          |         |      | investigation of     |
|                     |          |      | classification,     |          |         |      | learners and their   |
|                     |          |      | association rule    |          |         |      | learning.            |
|                     |          |      | mining,             | Educatio | Journal | 2010 | Peer review of 15    |
|                     |          |      | clustering, and     | nal Data |         |      | years from 1993      |
|                     |          |      | social area         | Mining:  |         |      | to 2009 which        |
|                     |          |      | networking to       | A        |         |      | focuses on EDM       |
|                     |          |      | educational data.   | Review   |         |      | an upcoming          |
| Data                | Journal  | 2012 | Classification and  | of the   |         |      | research area        |
| Mining              |          |      | prediction.         | State of |         |      | which take           |
| for                 |          |      | relationship        | the Art  |         |      | account of           |
| Educatio            |          |      | mining              |          |         |      | various fields like  |
| n                   |          |      | clustering          |          |         |      | data mining          |
| 11                  |          |      | discovery with      |          |         |      | learning analytics   |
|                     |          |      | uiscovery with      |          |         |      | icarning analytics,  |

|  | e-learning,       |
|--|-------------------|
|  | adaptive          |
|  | hypermedia,       |
|  | intelligent       |
|  | tutoring systems, |
|  | web mining etc.   |

V.

Apriori algorithm is one of most common and widely used methods in Association Mining. Association rules are if/then statements that help to

#### IV. TECHNIQUES USED IN EDUCATIONAL ncover relationships between unrelated data in a **DATA MINING** database, relational database or other information

Some of the important techniques used in EDM are pository. Association rules are used to find the discussed below: relationships between the objects which are

a) Linear Regression: this Predication techniquarequently Applications used together. of predicts a number. Various attributes like incomessociation rules are basket data analysis, sales, age, weight etc. can be predicted using assification , cross-marketing, clustering, catalog Regression. For example in Educational Dataesign, and loss-leader analysis etc.

Mining to predict the CGPA of Student various V.COMPARSION OF ASSOCIATION RULE parameters like Age, Gender, Attendance, Family Income, Occupation, Family Qualification can be used as predictors.

b) Clustering: Clustering is the method that helps to group similar records together. This is unsupervised learning approach, which mainly focuses on highdimensional data. Clustering can be Hierarchical or Non-Hierarchical. K-Means is the most common method of Clustering. For example in Educational Data Mining, clustering can be used to group students based on their Learning styles as Visual, Aural, and Kinesthetic. Cluster model has a disadvantage that there are no clear rules to characterize each cluster.

c) Classification: Classification helps in classifying data based on the Training set and then uses that pattern to classify the new data which is also known as the training set. This is a supervisedlearning technique because the classes are predefined before extracting patterns on the target data. Some popular Classification methods used in EDM are Decision Tree, Neural Network, Naive Bayes Classifier, SVM etc.

d) Association Rule Mining: Association rules in Educational Data Mining are used to determine remarkable and strong association rules from Educational databases using support and confidence as the predefined measures. Association rules further identify which parameters are closely related to each other and have relations among them. MINING ALGORITHM

Apriori algorithm requires large amount of memory space due to large number of candidates are generated. It need multiple scans for generating candidate sets. In this algorithm execution time is more as time is wasted in producing candidates set every time. Hence, we are using FP-Growth algorithm which is particularly more efficient than apriori algorithm when it comes to long patterns.

### FP (Frequent Pattern) Growth Algorithm

The FP-Growth algorithm, proposed by Han. In his study, Han proved that his method outperforms other popular methods for mining frequent patterns, e.g. the apriori algorithm and the tree projection.

The FP-Growth algorithm is an alternative way to find frequent itemsets without using candidate generations, thus improving performance. For so much it uses a divide-and-conquer strategy. The core of this methods is the usage of a special data structure named frequent-pattern tree (FP tree), which retains the itemset association information.

The algorithm extracts frequent item sets that can be extract association rules. This is done using the support of an item set.

By using divide-and-conquer strategy, it compress the database which provides the frequent sets; then divide this compressed database into a set of conditional databases, each associated with a frequent set and apply data mining on each database. This algorithm performs mining on FP-tree recursively. There is a problem of finding frequent itemsets which is converted to searching and constructing trees recursively.

The key points of FP-Growth algorithm are:

- 1. FP growth improves Apriority to a big extent
- 2. Frequent Item set Mining is possible without candidate generation
- Two step approach used in this algorithm:

1: Build a compact data structure called the FP-tree.

Built using 2 passes over the data set.

2: Extracts frequent itemsets directly from the FP-tree

FP-Tree is constructed using 2 passes over the datatheir support.

Pass 1

1.Scan data and find support for each item.

2. Discard infrequent items.

| Cha<br>racte<br>ristic<br>s              | AIS          | SET<br>M | Apri<br>ori | Apri<br>oriti<br>d                              | Apri<br>ori<br>hybr<br>id                              | FP-<br>Gro<br>wth        |
|--|--------------|----------|-------------|---|--|--------------------------|
| Data<br>Supp<br>ort                      | Less         | Less     | Limi<br>ted | Ofte<br>n<br>supp<br>ose<br>large               | Very<br>large  | Very<br>large            |
| Spee<br>d in<br>initia<br>l<br>phas<br>e | Slow         | Slow     | High        | Slo<br>w  | High   | High                     |
| Spee<br>d in<br>later<br>phas<br>e       | Slow         | Slow     | Slow        | High  | High   | High                     |
| Accu<br>racy                             | Very<br>less | Less     | Less        | Mor<br>e<br>accu<br>rate<br>than<br>Apri<br>ori | Mor<br>e<br>accu<br>rate<br>than<br>Apri<br>oriti<br>d | Mor<br>e<br>accu<br>rate |

3. Sort frequent items in decreasing order based on  $\frac{a}{b}$  their support.

4. Use this order when building the FP-Tree, so common prefixes can be shared.

Pass 2:

Nodes correspond to items and have a counter

- 1. 1. FP-Growth reads 1 transaction at a time Mining", International Symposium on Educational Technology (ISET),2015 IEEE. maps it to a path
- 2. 2. Fixed order is used, so paths can overlap when performance Prediction", International Journal of Advanced transactions share items (when they have the samputer Science and Applications, Vol.7, No.5, 2016. [9] S. M. Merchan and J. A. Duarte, "Analysis of Data Mining prefix). Techniques for Constructing a Predictive Model for Academic
- -In this case, counters are incremented
- 3. Pointers are maintained between containing the same item, creating singly linked Trupti A. Kumbhare, Prof. Santosh V. Chobe, "An Overview of lists
- 930. -The more paths that overlap, the highet 1the. Romero and S. Ventura, "Educational data mining: A review of the state of the art," IEEE Transactions on Systems, Man, and compression. FP-tree may fit in memory.

nndes016.

3. 4. Frequent itemsets extracted from the FP-Trope-618, Nov. 2010.

#### VI. CONCLUSION

[12] Thakar, Pooja. "Performance Analysis and Prediction in Educational Data Mining: A ResearchTravelogue." arXiv preprint

Performance", IEEE Latin America Transactions, Vol. 14, No. 6,

Association RuleMining Algorithms", IJCSIT, Vol. 5(1), 2014, 927-

association rule-mining B. Kumar and S. Pal, "Mining educational data to analyse There various are algorithms. In this paper, we have discussed tubeats' performance," International Journal of Advanced Computer association rule mining algorithms used Science and Applications, Vol.7, No.5, 2016.

education system. Educational Data Mining [14] A. M. Shahiri, W. Husain, and N. A. Rashid, "A review on Education system. Educational Data Mining pedatiting student's performance using data mining techniques," be used effectively in institutes for leading dia Computer Science, vol. 72, pp. 414-422, 2015.

education activities in an effective way, for<sup>[15]</sup> P. Baepler and C. Murdoch, "Academic Analytics and data monitoring student's performances continuously. Scholarship of Teaching and Learning, vol. 4, no. 2, Jul. 2010.

Comparison is done based on the above[16] S. Harikumar, "A study on educational data mining," performance criteria. Each algorithm has some<sup>International Journal of Computer Trends and Technology, vol. 8,</sup> advantages and disadvantages. From the above<sup>[17]</sup> M. Berland, R. S. Baker, and P. Blikstein, "Educational data

comparison, we can conclude that, FP-growth ismining and learning Analytics: Applications to Constructionist more suitable than all other algorithms discussedresearch," Technology, Knowledge and Learning, vol. 19, no. 1-2, pp. 205–220, May 2014. [18] M. S. Bhullar and A. Kaur. "Use of data mining in education here in terms of large data set support and speed.

sector". Lecture Notes in Engineering and Computer Science 2200(1), pp. 513-516. 2012.

#### REFERENCES

[19] M. Mayilvaganan, D. Kalpanadevi ," Comparison of

[1] Nur Hani Zulkifli, Jamalahyahaya, Aziz Deraman (Article Students Academic Environment," in International Conference on August2016), "Business Intelligence and Analytics in Managing Communication and Network Technologies (ICCNT), 2014. Organizational Performance: The Requirement Analysis", Journal [20] BipinBihariJayasingh, "A Data Mining Approach to Inquiry

of Advances in Information Technology Vol. 7, No.3, August 2016. Based Inductive Learning Practice In Engineering Education", in [2] Mohiuddin Ali Khan, WajebGharibi, Sateesh Kumar Pradhan IEEE 6<sup>th</sup>International Conference on Advanced Computing 2016. (2014) "Data Mining Techniques for Business Intelligence in Educational System: A CaseMining"978-14799-3351

8/14/\$31.00,2014 IEEE.

Leena Khanna Dr.ShailendraNarayan [3] Singh, Dr. Mansaf Alam, "Educational Data Mining and its Role in Determining Factors Affecting Student AcademicPerformance: A Systematic Review, 978 1-4673-6984/8/16/\$31.00 © 2016 IEEE.[4]Oswaldo Moscoso-Zea1, Andres Sampedro, and Sergio Luján-Mora (2016),"Data warehouse design for Educational Data Mining", International Conference on Enterprise Information System (ICEIS)2016, pp. p. 244–249.

[5] Lida Shams, Hassan Rashidi (2016), "Factors affecting the Increase and Decrease student achievement in Primary School with Business Intelligence Approach", International Journal of Computer & Information Technologies(IJOCIT).

[6] MohammedI .Al.Twijri, Amin Y. Noaman, "A New Data Mining Model Adopted for Higher Institutions", Procedia Computer Science 65 (2015).

[7] Ms.TismyDevasia, Ms.Vinushree T P, Mr.VinayakHegde, "Prediction of Students Performance using Educational Data