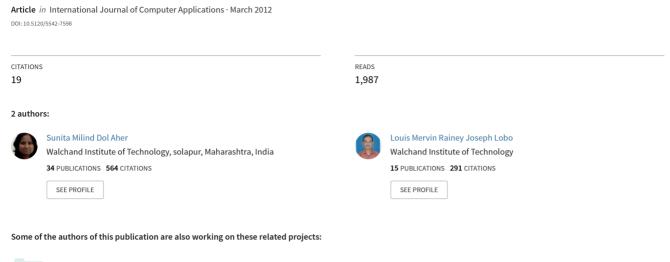
### Best Combination of Machine Learning Algorithms for Course Recommendation System in E-learning





 $T24S\ Modified\ TPS\ Activity\ for\ Mathematical\ Courses\ to\ Improve\ Students'\ Fundamental\ Knowledge\ View\ project$ 

# Best Combination of Machine Learning Algorithms for Course Recommendation System in E-learning

Sunita B Aher
M.E. (CSE) -II
Walchand Institute of Technology
Solapur University India

#### ABSTRACT

Data Mining is the extraction of hidden predictive information from large database which can be used in various commercial applications like bioinformatics, E-Association Rule, classification and clustering are three different algorithms in data mining. Course Recommender System plays an important role in identifying the behavior of students interested in particular set of courses. We collect the data regarding the course enrollment for specific set of data. For collecting this data, we use the learning management system like Moodle. After collecting the data, we apply the different combination of data mining algorithm like classification & association rule algorithm, clustering & association rule algorithm, association rule mining in classified & clustered data, combining clustering & classification algorithm in association rule algorithms or simply the association rule algorithm. Here in this paper we use ADTree classification algorithm, Simple K-means Algorithm & Apriori Association Rule algorithm as different machine learning algorithm. So we propose the five different methods to find the best combination of algorithm in recommending the courses to students in E-learning. We compare the result of this combined approach as well as only the association rule algorithm & present the best combination of algorithm for recommendation of courses in E-learning according to our simulation.

#### **Keywords**

Weka, Machine Learning Algorithm, Simple K-means Algorithm, ADTree Classification Algorithm, Apriori Association Rule Algorithm, Moodle.

#### 1. INTRODUCTION

Data mining also known as Knowledge Discovery in Database (KDD) is the extraction of interesting (nontrivial, implicit, previously unknown and potentially useful) patterns or knowledge from huge amount of data. Alternative names to data mining are knowledge extraction, data/pattern analysis, data archeology, data dredging, information harvesting, business intelligence, etc.

Data mining can also be used to extract the knowledge from E-learning system such as Moodle. The Course Recommendation System in E-Learning is a system that recommend to the student, the best combination of courses in which the students are interested.

The rest of the document is organized as follows. Section 1.1 provides a background of the related research fields covering a brief introduction about each. Section 2 describes the Literature Review. Section 3 discusses the

#### Lobo L.M.R.J.

Associate Professor, Head, Department of IT Walchand Institute of Technology Solapur University India

architecture of proposed system. Section 4 & 5 discusses methodology & implementation and best combination of data mining algorithm. The conclusion & future work is presented in Section 6 followed by references.

### 1.1 Background

This research integrates issues from the research field of data mining algorithms such as Classification, clustering & Association Algorithm), Moodle and open source data mining tool, Weka. The following subsections include a brief overview of these topics.

#### 1.1.1 Apriori Association Rule Algorithm

Association rules are used to show the relationship between data items. Association rule generation is usually split up into two separate steps: First, minimum support is applied to find all frequent itemsets in a database. Second, these frequent itemsets and the minimum confidence constraint are used to form rules.

Apriori Association rule is used to mine the frequent patterns in database. Support & confidence are the normal method used to measure the quality of association rule. Support for the association rule X->Y is the percentage of transaction in the database that contains XUY. Confidence for the association rule is X->Y is the ratio of the number of transaction that contains XUY to the number of transaction that contain X [9]. The Apriori association rule algorithm is given in below [9]:

### **Apriori Association Rule Algorithm**

Output : Association Rule satisfying Support & Confidence

### Method:

$$\begin{split} &C_1 = \text{Itemsets of size one in I;} \\ &\text{Determine all large itemsets of size 1, $L_{1;}$} \\ &i = 1; \\ &\text{Repeat} \\ &i = i+1; \\ &C_i = \text{Apriori-Gen}(L_{i-1}); \end{split}$$

Apriori-Gen( $L_{i-1}$ )

- Generate candidates of size i+1 from large itemsets of size i.
- 2. Join large itemsets of size i if they agree on i1.
- Prune candidates who have subsets that are not large.

Count C<sub>i</sub> to determine L<sub>i</sub>;

Until no more large itemsets found;

#### 1.1.2 Simple K-means Clustering Algorithms

Clustering is finding groups of objects such that the objects in one group will be similar to one another and different from the objects in another group [8]. Clustering can be considered the most important unsupervised learning technique.

Simple K-means algorithm is a type of unsupervised algorithm in which items are moved among the set of cluster until required set is reached. This algorithm is used to classify the data set, provided the number of cluster is given in prior. This algorithm is iterative in nature.

### Algorithm: Simple K-means clustering algorithm Input:

Set of Elements or Database of transaction  $D = \{t_1, \, t_2, \, t_3, \, \dots, \, t_n\}$  Number of required Cluster k

#### **Output:**

Set of Cluster K

#### Method:

Make initial guesses for the means  $m_1$ ,  $m_2$ , ...,  $m_k$ ; Repeat

Assign each element t<sub>i</sub> to the cluster having the closest mean.

Calculate the new mean for each cluster. Until there are no changes in any mean

### 1.1.3 ADTree Classification Algorithm

Classification is a data mining task that maps the data into predefined groups & classes. It is also called as supervised learning.

An alternating decision tree (ADTree) is a machine learning method for classification which generalizes decision trees. An alternating decision tree consists of two nodes. Decision nodes specify a predicate condition. Prediction nodes contain a single number. ADTree always have prediction nodes as both root and leaves. An instance is classified by an ADTree by following all paths for which all decision nodes are true and summing any prediction nodes that are traversed.

### 1.1.4 Learning Management System Moodle

Moodle is an open-source course management learning system to help educators create effective online learning communities [15]. It is also possible to modify the source code of any file of Moodle. It is very easy to add the course in the system. Here are adding 13 course category & near about 82 subjects which are related to Computer Science & Engineering and Information Technology Department. Here we are using the Moodle as it is very user friendly. Also it maintains detailed logs of all activities of students. It stores the record of every click that students make. We can use these logs to find courses in which student are interested. It stores the logs in relational database MYSQL. Moodle is shown in figure 1.

### 1.1.5 Open Source Data Mining Tool Weka

Weka is a collection of machine learning algorithms for data mining tasks. The algorithms can either be applied directly to a dataset or called from your own Java code [12]. The Weka workbench contains a collection of visualization tools and algorithms for data analysis and predictive modeling, together with graphical user interfaces for easy access to this functionality [11]. It is freely available software. Weka has several standard data mining tasks, data preprocessing, clustering, classification,

association, visualization, and feature selection. Figure 2 shows Weka 3.5.3 with Explorer window.



Figure 1: Moodle

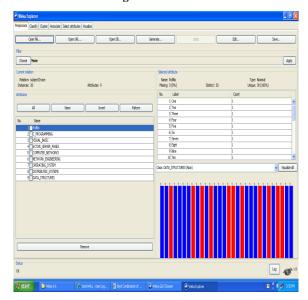


Figure 2: Weka 3.5.3 with Explorer window

### 2. LITERATURE REVIEW

C. Carmona [1], they propose proposed how to use adaptive machine learning algorithms to learn about the student's preferences over time. First they use all the background knowledge available about a particular student to build an initial decision model based on learning styles. This model can then be fine-tuned with the data generated by the student's interactions with the system in order to reflect more accurately his/her current preferences.

Castro [2] provides an up-to-date snapshot of the current state of research and applications of Data mining methods in e-learning. It provides the taxonomy of e-learning problems to which the Data Mining techniques have been applied including, for instance: Students' classification based on their learning performance; detection of irregular learning behaviors; e-learning system navigation and interaction optimization; clustering according to similar e-learning system usage; and systems' adaptability to

students' requirements and capacities.

Lili He & Hongtao Bai [3] explained Aspect oriented programming which offers a unique module, an aspect, to encapsulate scattered and tangled code, which made it hopeful to solve the problem of crosscutting concerns. Identification and encapsulation of crosscutting concerns was the key problem in the migration from OO system to AO system. A novel aspect mining method which combined clustering and association rule technology is provided in this article. Clustering analysis based on the execution traces was provided to find out candidate aspects; while association rule mining based on the execution traces with ordered call was used to find out the crosscuts. Both the aspect code (advice body) and the crosscuts (point cuts) were gotten after the above two processes, which constituted the aspect mining process.

Al¶³pio Jorge [4] proposed a method for grouping and summarizing large sets of association rules according to the items contained in each rule. They used hierarchical clustering to partition the initial rule set into thematically coherent subsets. This enabled the summarization of the rule set by adequately choosing a representative rule for each subset, and helped in the interactive exploration of the rule model by the user. Rule clusters can also be used to infer novel interest measures for the rules.

B.Ramasubbareddy, A. Govardhan & A.Ramamohanreddy [13] proposed Associative classification which was a classification of a new tuple using association rules. It was a combination of association rule mining and classification. They searched for strong associations between frequent patterns and class labels. The main aim

of this paper was to improve accuracy of a classifier. The accuracy can be achieved by producing all types of negative class association rules.

The main hypothesis discussed in the paper [4] was that Web content analysis can be used to improve Web usage mining results. They proposed a system that integrated Web page clustering into log file association mining and used the cluster labels as Web page content indicators. They experimented with several approaches to content clustering, relying on keyword and character n-gram based clustering with different distance measures and parameter settings.

### 3. ARCHITECTURE OF COURSE RECOMMENDER SYSTEM

In Course Recommendation System, we are consider the 13 course category which is shown in following table 1. Under each category there will courses. So there are about 82 subjects.

We have created the student login & gave the access to the student. We have considered the student of two courses Computer Science & Engineering and Information Technology for collecting the data. Student will enroll for those subjects in which they are interested. This data is stored in the moodle database which we use to find out the best combination. The framework for this recommendation system is shown in the figure 3.

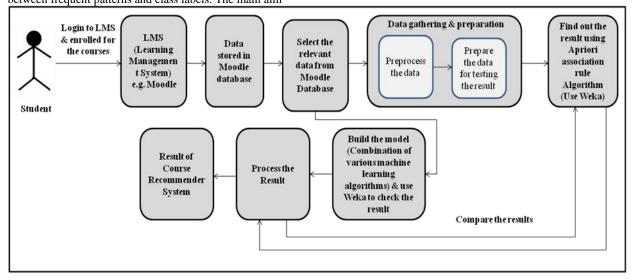


Figure 3: Course Recommender System in E-learning

In this architecture , student first logs in the Learning Management System e.g. Moodle. The system verifies the username & password. After verifying the username & password, student will search the course category & courses (subjects). Students will enroll for subject in which they are interested. This enrollment information is stored in database as Moodle store each & every action of student.

After collecting the data from student which is stored in Moodle database, the next stage is to gather & prepare the data. In this step, first we select the data from database which is relevant. We check the result using the Apriori association rule algorithm. After that we clean & transformed into the format for testing. To test the result

we use the open source data mining tool WEKA. Since our project is to find the best combination of subject, we use the Apriori association rule algorithm. In the step, build the model; we try to develop the algorithm which may be combination of various data mining algorithm. We find out the result using this model & compare the result with result obtained using already existing algorithm in WEKA. Finally we deploy the result.

### 4.METHODOLOGY& MPLEMENTATION RESULT

Here we are considering the sample data extracted from Moodle database. Here we consider data of 15 courses. We consider the subjects like C-programming (C), Visual

Basic (VB), Active Server Pages (ASP), Computer Network (CN), Microprocessor (MP), Computer Organization (CO), Database Engineering (DBE), Advanced Database System (ADS), Operating System (OS), Distributed System (DS), Finite Automata System (FSA), Data Structure (DS-I), Software Engineering (SE), and Software Testing & Quality assurance (STQA).

Table 1: Sample data extracted from Moodle database [11]

Courses→	С	VB	ASP	CN	NE	MP	CO	DBE	ADS	OS	DS	FSA	DS-	SE	STQA
Doll No													I		
Roll No↓															
1	yes	yes	yes	yes	yes	no	no	no	no	no	no	no	yes	no	no
2	no	no	no	no	no	no	no	no	no	no	no	no	no	no	no
3	yes	yes	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	yes	yes
4	no	no	no	yes	yes	no	yes	no	no	no	no	no	no	no	no
5	yes	yes	yes	yes	yes	no	no	yes	no	yes	yes	no	yes	no	no
6	yes	yes	yes	no	no	no	no	no	no	yes	no	no	yes	no	no
7	no	no	no	yes	yes	yes	yes	no	no	no	no	no	no	yes	no
8	no	no	no	no	no	no	no	yes	yes	yes	yes	no	yes	no	no
9	no	no	no	yes	yes	yes	yes	no	no	no	no	yes	no	no	no
10	yes	no	no	no	no	no	no	no	no	no	no	no	no	no	no
11	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
12	yes	yes	yes	yes	yes	no	no	no	no	no	no	no	no	no	no
13	no	no	no	no	no	no	no	yes	yes	yes	yes	no	yes	yes	yes
14	yes	yes	yes	yes	yes	no	no	no	no	yes	yes	no	no	no	no
15	yes	yes	yes	no	no	no	no	no	no	no	no	no	yes	no	no
16	no	no	no	yes	yes	no	no	yes	yes	yes	yes	no	yes	no	no
17	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	yes	yes
18	yes	yes	yes	no	no	no	no	no	no	no	no	no	no	no	no
19	no	no	no	yes	yes	yes	yes	yes	yes	no	no	no	no	no	no
20	yes	no	no	no	no	no	no	no	no	yes	yes	no	yes	yes	yes
21	yes	no	yes	no	no	yes	yes	no	no	yes	yes	yes	no	no	no
22	no	no	no	no	no	no	no	yes	yes	yes	yes	no	yes	no	no
23	yes	yes	yes	yes	yes	yes	yes	no	no	yes	yes	no	yes	no	no
24	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
25	no	yes	yes	no	no	yes	yes	yes	yes	yes	yes	no	no	no	no
26	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
27 28	yes	yes	yes	yes	yes	no	no	no	no	no	no	no	no	no	no
29	no no	no	no	yes	yes	no	no	no	no voc	yes	yes	no	yes	no	no
30		no	no	no	no	yes	yes	yes	yes	no	no	no	no	no	no
31	yes no	yes no	yes no	yes no	yes no	no no	no no	no no	no no	no no	no no	no no	no no	yes no	yes no
32	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	no	yes	no	no
33	no	no	no	yes	yes	no	no	no	no	yes	yes	no	yes	no	no
34	yes	yes	yes	no	no	no	no	no	no	no	no	no	no	no	no
35	no	no	no	no	no	no	no	no	no	yes	yes	no	no	no	no
36	no	no	no	yes	yes	no	no	no	no	no	no	no	yes	no	no
37	yes	yes	yes	yes	yes	yes	yes	yes	yes	no	no	no	no	no	no
38	no	no	no	no	no	no	no	no	no	yes	yes	yes	yes	yes	yes
39	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
40	no	no	no	no	no	no	no	no	no	no	no	no	no	yes	yes
41	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
42	no	no	no	yes	yes	no	no	no	no	no	no	no	no	no	no
43	no	no	no	no	no	no	no	no	no	yes	yes	no	yes	no	no
44	no	no	no	no	no	no	no	no	no	no	no	no	no	no	yes
45	no	no	no	no	no	no	no	no	no	no	no	no	no	no	no

### 4.1 Association Rule Mining

If we apply Apriori association rule mining to the data extracted from Moodle database using data mining tool Weka then we get all association rule containing "no"

only. This result is shown in first row of table 7. As we are recommending the course to the student, we need all

association rules containing "yes" only. So in order to get the correct association rule, we prepare the data.

### **4.1.1 Data Preparation Strategy**

Data preparation strategy is explained in [10]. For data preparation, we delete those rows having very less student count as well as that course having very less course count.

Subject	С	VB	AS	CN	NE	OS	DS	DS
→ ···			P					-I
Roll								
No of								
Student								
<b>1</b>								
1	yes	yes	yes	yes	yes	no	no	yes
3	yes							
4	no	no	no	yes	yes	no	no	no
5	yes							
6	yes	yes	yes	no	no	yes	no	yes
7	no	no	no	yes	yes	no	no	no
8	no	no	no	no	no	yes	yes	yes
9	no	no	no	yes	yes	no	no	no
11	yes	yes	yes	no	no	yes	yes	yes
12	yes	yes	yes	yes	yes	no	no	no
13	no	no	no	no	no	yes	yes	yes
14	yes	no						
15	yes	yes	yes	no	no	no	no	yes
16	no	no	no	yes	yes	yes	yes	yes
17	yes	yes	yes	no	no	yes	yes	yes
18	yes	yes	yes	no	no	no	no	no
19	no	no	no	yes	yes	no	no	no
20	yes	no	no	no	no	yes	yes	yes
21	yes	no	yes	no	no	yes	yes	no
22	no	no	no	no	no	yes	yes	yes
23	yes							
24	yes							
25	no	yes	yes	no	no	yes	yes	no
26	yes	yes	yes	no	no	yes	yes	yes
27	yes	yes	yes	yes	yes	no	no	no
28	no	no	no	yes	yes	yes	yes	yes
30	yes	yes	yes	yes	yes	no	no	no
32	yes	yes	yes	no	no	yes	yes	yes
33	no	no	no	yes	yes	yes	yes	yes
34	yes	yes	yes	no	no	no	no	no
35	no	no	no	no	no	yes	yes	no
36	no	no	no	yes	yes	no	no	yes

37	yes	yes	yes	yes	yes	no	no	no
38	no	no	no	no	no	yes	yes	yes
39	yes							
41	yes	yes	yes	no	no	yes	yes	yes
42	no	no	no	yes	yes	no	no	no
43	no	no	no	no	no	yes	yes	yes

After data preparation strategy, we get table 2. After preparation of data, we check the result using Weka. These results are shown in second row of table 7. Now, we get all association rules containing "yes" only. As we increase the support, we get the refined rule. The meaning of the rule (DS=yes →OS=yes) is: if student is interested in Distributed system then he/she is interested in Operating System & vice versa. Due to above rule, we can recommend to new student who has recently enrolled for DS course, the Operating System as a course to be opted. Association rule obtained here also match with the in general real world interdependencies among the course.

### 4.2 Combination of Classification & Association Rule algorithm

Here we consider the ADTree algorithm as classification algorithm & Apriori Rule as association rule algorithm. First we apply the ADTree classification algorithm on sample data. We consider only those courses for which the value of ADTree is negative (negative value represent "yes"). Then we select those rows from table having the negative course value. The result after application of ADTree classification algorithm in table 1 is shown in table 3. Now we apply the Apriori association rule algorithm to this table. The result of application of this algorithm is shown in third row of table 7. Using this combination of classification & association rule algorithm, there is no need to prepare the data obtained from Moodle courses.

Table 3: After application of ADTree classification algorithm on table 2 [17]

Courses→	С	VB	ASP	CN	NE	MP	CO	DBE	ADS	OS	DS	FSA	DS-I	SE	STQA
Roll No ↓															
3	yes	yes	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	yes	yes
5	yes	yes	yes	yes	yes	no	no	yes	no	yes	yes	no	yes	no	no
11	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
17	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	yes	yes
20	yes	no	yes	yes	no	yes	yes	yes							
23	yes	no	no	yes	yes	no	yes	no	no						
24	yes	yes	yes												
26	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
32	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	no	yes	no	no
39	yes	yes	yes												
41	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no

### 4.3 Combination of Clustering & Association Rule Algorithm

In this combination of algorithms i.e. Simple K-means clustering algorithm & Apriori association rule algorithm, first we apply the Simple K-means clustering algorithm to sample data extracted from Moodle course. After

application of this algorithm we get some clusters. So we divide the table into sub tables depending on the number of clusters. Then we check the result of these sub-tables. So we consider only that sub-table of cluster which gives the correct result. The cluster which gives the correct result is shown in table 4. As this is Course Recommending System, we consider only that association rule containing "yes" only. After clustering the data using Simple K-means clustering algorithm, we apply the Apriori

association rule algorithm to the sub-table whose result are correct. The result of application of Apriori to clustered

data is shown in forth row of table 7.

Table 4: After application of Simple K-means clustering algorithm on table 2 [19]

Courses→	С	VB	ASP	CN	NE	MP	CO	DBE	ADS	OS	DS	FSA	DS-I	SE	STQA
Roll No↓															
1	yes	yes	yes	yes	yes	no	yes	no	no						
3	yes	yes	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	yes	yes
5	yes	yes	yes	yes	yes	no	no	yes	no	yes	yes	no	yes	no	no
6	yes	yes	yes	no	no	no	no	no	no	yes	no	no	yes	no	no
11	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
12	yes	yes	yes	yes	yes	no	no	no							
14	yes	yes	yes	yes	yes	no	no	no	no	yes	yes	no	no	no	no
16	no	no	no	yes	yes	no	no	yes	yes	yes	yes	no	yes	no	no
17	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	yes	yes
23	yes	no	no	yes	yes	no	yes	no	no						
24	yes	yes	yes												
26	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
27	yes	yes	yes	yes	yes	no	no	no							
28	no	no	no	yes	yes	no	no	no	no	yes	yes	no	yes	no	no
30	yes	yes	yes	yes	yes	no	yes	yes							
32	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	no	yes	no	no
33	no	no	no	yes	yes	no	no	no	no	yes	yes	no	yes	no	no
37	yes	no	no	no	no	no	no								
39	yes	yes	yes												
41	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no

### 4.4 Association Rule Mining of classified & clustered data

In this case we consider three algorithms i.e. ADTree classification algorithm, Simple K-means clustering algorithm & Apriori Association Rule algorithm. The order of application of these three algorithms is:

ADTree classification algorithm- Simple K-means clustering algorithm- Apriori association rule algorithm

So first we apply the ADTree classification algorithm on sample table of moodle course. Then we apply Simple K-means clustering algorithm on classified data. Again we consider that cluster which gives the correct association rules. Sample table after application of Simple K-means clustering algorithm on classified data which is shown in table 3 is given in table 5. Result of application of Apriori association rule on correct cluster of classified data is shown in fifth row of table 7.

Table 5: After application of Simple K-means clustering algorithm on table 3 [16]

Courses→	C	VB	ASP	CN	NE	MP	CO	DBE	ADS	OS	DS	FSA	DS-	SE	ST
Roll_No↓													Ι		QA
5	yes	yes	yes	yes	yes	no	no	yes	no	yes	yes	no	yes	no	no
11	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
23	yes	no	no	yes	yes	no	yes	no	no						
26	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no
32	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	no	yes	no	no
41	yes	yes	yes	no	no	no	no	no	no	yes	yes	no	yes	no	no

## 4.5 Combining Clustering & Classification algorithm with Association Rule algorithm

Here we apply three data mining algorithm in following order:

Simple K-means clustering algorithm- ADTree classification algorithm- ADTree classification algorithm

First we apply Simple K-means clustering algorithm on

sample data. After clustering the data, we apply the ADTree classification algorithm on clustered data. Sample table after application of ADTree classification algorithm on clustered data which is shown in table 4 is given in table 6. After application of this classification algorithm, we apply Apriori association rule algorithm to classified data whose results are shown in sixth row of table 7.

Courses→	С	VB	ASP	CN	NE	MP	CO	DBE	ADS	OS	DS	FSA	DS-I	SE	STQA
Courses		, 2	1101	OI (	1,2	1,11		222	1120	0.0	2.5	1011	201	22	21411
Roll No↓															
•															
3	yes	yes	yes	yes	yes	no	no	no	no	yes	yes	yes	yes	yes	yes
5	yes	yes	yes	yes	yes	no	no	yes	no	yes	yes	no	yes	no	no
23	yes	yes	yes	yes	yes	yes	yes	no	no	yes	yes	no	yes	no	no
24	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
39	ves	ves	ves	ves	ves	ves	ves	ves	ves	ves	ves	ves	ves	ves	ves

Table 6: After application of ADTree classification algorithm on table 2 [18]

### 5. BEST COMBINATION OF DATA MINING ALGORITHM

According to our simulation, the best combination of algorithm is the combination of clustering, classification & association rule mining. Here we preprocess the data to remove the noise or to add the missing value but there is no need to prepare the data as explained subsection 4.1.1. If we are using the Apriori association rule algorithm &

increase the support then we are getting the refined association rule but the number of rules, we get, are less & we need to prepare the data also. Association rule obtained using this combination i.e. clustering, classification & association rule algorithm also match with the in general real world interdependencies among the courses.

Figure 4 shows the graph for courses after application of data mining techniques

Table 7: Result of applying the different data mining technique

Results  iori Association Rule before preprocessing & application of combination of Clustering & Association Rule  Minimum support: 0.7  Minimum metric <confidence>: 0.9  Best rules found:</confidence>
Rule  Minimum support: 0.7  Minimum metric <confidence>: 0.9</confidence>
Minimum metric <confidence>: 0.9</confidence>
1. Computer_Organization=no → Microprocessor=no conf:(1) 2. Database_Engineering=no → Advanced_Database_System=no conf:(1) 3. Computer_Organization=no Finite_State_Automata=no 32 → Microprocessor=no conf:(1) 4. Microprocessor=no → Computer_Organization=no conf:(0.97) 5. Softare_Tesing_And_Quality_Assurance=no 35 → Software_Engineering=no conf:(0.97) 6. Software_Engineering=no → Softare_Tesing_And_Quality_Assurance=no conf:(0.97) 7. Advanced_Database_System=no → Database_Engineering=no conf:(0.97) 8. Microprocessor=no Finite_State_Automata=no → Computer_Organization=no conf:(0.97) 9. Finite_State_Automata=no Softare_Tesing_And_Quality_Assurance=no → Software_Engineering=no conf:(0.97) 0. Finite_State_Automata=no Software_Engineering=no → Softare_Tesing_And_Quality_Assurance=no conf:(0.97)
Apriori Association Rule after preprocessing & before application of combination of Clustering & Association Rule
Minimum support: 0.5 Minimum metric <confidence>: 0.9  Best rules found:  1. Distributed_Systems=yes → Operating_System=yes conf:(1)  2. Visual_Basic=yes → Active_Server_Pages=yes conf:(1)  3. Network_Engineering=yes → Computer_Networks=yes conf:(1)  4. Computer_Networks=yes → Network_Engineering=yes conf:(1)  5. C_Programming=yes Visual_Basic=yes → Active_Server_Pages=yes conf:(1)  6. Distributed_Systems=yes Data_Structures=yes → Operating_System=yes conf:(1)  7. Operating_System=yes → Distributed_Systems=yes conf:(0.96)  8. Active_Server_Pages=yes → C_Programming=yes conf:(0.96)</confidence>
2. 2. 3. 4. 4. 5. 6. 7. 7. 6. 6. 6. 7. 6. 6. 7. 7. 6. 6. 7. 7. 7. 6. 6. 7. 7. 7. 6. 6. 7. 7. 7. 6. 6. 7. 7. 7. 7. 7. 7. 7. 7. 7. 7. 7. 7. 7.

```
10. Active Server Pages=yes → Visual Basic=yes
                                                                conf:(0.96)
              Minimum support: 0.6
              Minimum metric <confidence>: 0.9
              Best rules found:
              1. Distributed Systems=yes → Operating System=yes
                                                                     conf:(1)
              2. Operating System=ves → Distributed Systems=ves
                                                                     conf:(0.96)
        After Application of Classification Algorithm- ADTree & Association Rule-Apriori Association Rule
C, VB,
              Minimum support: 0.95
ASP, CN,
              Minimum metric <confidence>: 0.9
NE, MP,
CO, DBE,
              Best rules found:
ADS, OS,
DS, FSA,
              1. Operating System=yes \rightarrow C Programming=yes conf:(1)
DS-I, SE,
              2. C Programming=yes → Operating_System=yes conf:(1)
STQA
              3. Distributed System=yes \rightarrow C Programming=yes
                                                                  conf:(1)
              4. C_Programming=yes → Distributed_System=yes
                                                                  conf:(1)
              5. Data_Structure-I=yes \rightarrow C Programming=yes conf:(1)
              6. C Programming=yes \rightarrow Data Structure-I=yes conf:(1)
              7. Distributed System=yes → Operating System=yes
                                                                    conf:(1)
              8. Operating_System=yes → Distributed_System=yes
                                                                    conf:(1)
              9. Data_Structure-I=yes → Operating System=yes conf:(1)
              10. Operating System=yes \rightarrow Data Structure-I=yes
      After Application of Clustering algorithm-Simple K-means & Association Rule-Apriori Association Rule
C. VB.
              Minimum support: 0.85
ASP, CN,
              Minimum metric <confidence>: 0.9
NE, MP,
CO, DBE,
              Best rules found:
ADS, OS,
DS, FSA,
              1. Visual_Basic=yes → C_Programming=yes conf:(1)
DS-I, SE,
              2. C Programming=yes \rightarrow Visual Basic=yes
STQA
              3. Active_Server_Pages=yes → C_Programming=yes
              4. C_Programming=yes → Active_Server_Pages=yes
                                                                   conf:(1)
              5. Active_Server_Pages=yes → Visual_Basic=yes conf:(1)
6. Visual_Basic=yes → Active_Server_Pages=yes conf:(1)
              7. Visual Basic=yes Active Server Pages=yes → C Programming=yes
                                                                                     conf:(1)
              8. C Programming=yes Active Server Pages=yes → Visual Basic=yes
                                                                                     conf:(1)
              9. C Programming=yes Visual Basic=yes → Active Server Pages=yes
                                                                                     conf:(1)
              10. Active Server Pages=yes → C Programming=yes Visual Basic=yes
                                                                                      conf:(1)
    After Application of Classification algorithm-ADTree, Clustering algorithm-Simple K-means Algorithm &
                                    Association Rule-Apriori Association Rule
C, VB,
              Minimum support: 0.95
ASP, CN,
              Minimum metric <confidence>: 0.9
NE, MP,
CO. DBE.
              Best rules found:
ADS, OS,
DS, FSA,
              1. Visual_Basic=yes \rightarrow C_Programming=yes conf:(1)
              2. C_Programming=yes → Visual_Basic=yes conf:(1)
DS-I, SE,
STQA
              3. Active_Server_Pages=yes → C_Programming=yes
              4. C Programming=yes → Active Server Pages=yes
```

5. Operating\_System=yes → C\_Programming=yes conf:(1)
6. C\_Programming=yes → Operating\_System=yes conf:(1)
7. Distributed\_System=yes → C\_Programming=yes conf:(1)
8. C\_Programming=yes → Distributed\_System=yes conf:(1)
9. Finite\_State\_Automata=no → C\_Programming=yes conf:(1)

	10. C_Programming=yes → Finite_State_Automata=no conf:(1)
After Applie	cation of Clustering algorithm-Simple K-means , Classification Algorithm- ADTree & Association Rule- Apriori Association Rule
C, VB, ASP, CN, NE, MP, CO, DBE, ADS, OS, DS, FSA, DS-I, SE, STQA	Minimum support: 0.95 Minimum metric <confidence>: 0.9  Best rules found:  1. Visual_Basic=yes → C_Programming=yes conf:(1) 2. C_Programming=yes → Visual_Basic=yes conf:(1) 3. Active_Server_Pages=yes → C_Programming=yes conf:(1) 4. C_Programming=yes → Active_Server_Pages=yes conf:(1) 5. Computer_Network=yes → C_Programming=yes conf:(1) 6. C_Programming=yes → Computer_Network=yes conf:(1) 7. Network_Engineering=yes → C_Programming=yes conf:(1) 8. C_Programming=yes → Network_Engineering=yes conf:(1) 9. Operating_System=yes → C_Programming=yes conf:(1) 10. C_Programming=yes → Operating_System=yes conf:(1)</confidence>

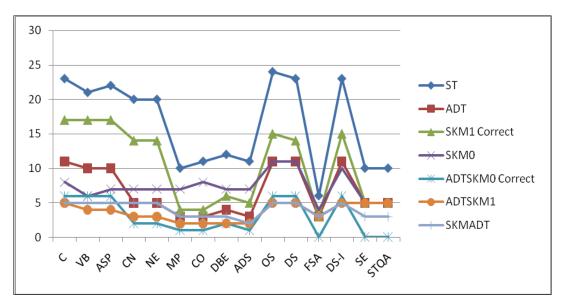


Figure 4 Graph for courses after application of various data mining algorithms

ADT- Courses after application ADTree classification Algorithm to Sample Table
SKM1 Correct – Courses after application of Simple K-means clustering algorithm (cluster 1 correct result)
SKM0 – Courses after application of Simple K-means clustering algorithm (cluster 0 incorrect result)
ADTSKM0 Correct – Courses after application of Simple K-means algorithm to classified data (Cluster 0 correct result)
ADTSKM0 Correct – Courses after application of Simple K-means algorithm to classified data (Cluster 0 correct result)
SKMADT- courses after application of ADTree classification algorithm to correct (SKM1 Correct) clustered data.

### 6. CONCLUSION AND FUTURE WORK

In this paper, we compare different combination of data mining algorithms i.e. Classification & association rule algorithm, clustering & association rule algorithm, Association Rule Mining of classified & clustered data, combining Clustering & Classification algorithm into Association Rule algorithm & only association rule algorithms. We consider ADTree classification algorithm, Simple K-means clustering algorithm & Apriori association rule algorithm. We compare the result & found that the combination clustering, classification &

association rule algorithm is the best combination. Future work includes the atomization of this combination algorithm i.e. clustering, classification & association rule algorithm on huge amount of data obtained from Moodle course of a college.

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