



## EDUCATIONAL DATA MINING

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### ABSTRACT

Presently educational institutions compile and store huge volumes of data such as student enrolment and attendance records, as well as their examination results. Mining such data yields stimulating information that serves its handlers well. Rapid growth in educational data points to the fact that distilling massive amounts of data requires a more sophisticated set of algorithms. This issue led to the emergence of the field of Educational Data Mining (EDM). Traditional data mining algorithms cannot be directly applied to educational problems, as they may have a specific objective and function. This implies that a preprocessing algorithm has to be enforced first and only then some specific data mining methods can be applied to the problems. One such preprocessing algorithm in EDM is Clustering. Many studies on EDM have focused on the application of various data mining algorithms to educational attributes. Therefore, our project provides over three decades long (1983-2016) systematic literature review on clustering algorithm and its applicability and usability in the context of EDM. Our project classifies the student's record based on achievement. Their result is predicted through data mining with the help of the internal marks scored. Future insights are outlined based on the literature reviewed, and avenues for further research are identified.

**Keywords:** Data Mining, Extraction, Fuzzy Classification, Fuzzy Clustering

### I. INTRODUCTION

Data mining is a process used by companies to turn raw data into useful information. By using software to look for patterns in large batches of data, businesses can learn more about their customers to develop more effective marketing strategies, increase sales and decrease costs. The biggest challenge is to analyse the data to extract meaningful information that can be used to solve a problem or for the growth of the business. There are powerful tools and techniques available to mine data and find insights from it. The data mining process breaks down into five steps. First, organizations collect data and load it into their data warehouses. Next, they store and manage the data, either on in-house servers or the cloud. Business analysts, management teams and information technology professionals access the data and determine how they want to organize it. Then, application software sorts the data based on the user's results, and finally, the end user presents the data in an easy-to-share format,

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8<sup>th</sup> - 9<sup>th</sup> March 2019

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ISBN : 978-93-87793-75-0

such as a graph or table. Data mining serves the primary purpose of discovering patterns among large volumes of data and transforming data into more refined/actionable information

## II. BACKGROUND

In[1], *Pauziah Moud Arshad, Usamah Binmat, Norlida Buniyamin "Educational Data Mining for prediction and classification of student's achievement"*

This paper highlights the importance of using student data to drive improvement in education planning. It then presents techniques of how to obtain knowledge from databases such as large arrays of student data from academic Institution databases. Further, it describes the development of a tool that will enable faculty members to identify, predict and classify students based on academic performance measured using Cumulative Grade point average (CGPA) grades.

In [2], *Maizatul Akmar Ismail, Tutut Herawanand Ashish Dutt "A Systematic Review on Educational Data Mining"*

This paper has also outlined several future insights on educational data clustering based on the existing literatures reviewed, and further avenues for further research are identified. The key advantage of the application of clustering algorithm to data analysis is that it provides relatively an unambiguous schema of learning style of students given a number of variables like time spent on completing learning tasks, learning in groups, learner behavior in class, classroom decoration and student motivation towards learning. Clustering can provide pertinent insights to variables that are relevant in separating the clusters.

In[3], *Vandhana Dahiya's work survey on educational datamining*, the proposed solution explains the objective of educational data mining into 3 categories educational, administrative, commercial.

In[4], *Christobal Romero and Sebastian Ventura's data mining in education*, the proposed system introduces and reviews key milestones and the current state of affairs in the field of EDM, together with specific applications, tools, and future insights.

In[5], *Rajni Jindal and Malaya Dutta Borah's survey on educational data mining and research trends*, this survey work focuses on components, research trends (1998-2012) of Educational Data Mining highlighting its related tools, techniques and outcomes.

## III. PREDEFINED SYSTEM

In Existing system, the Clustering and Classification Algorithms in Education Data Mining Process are used. In Clustering process c-means, K-means and classification process SVM classification or naive bayes Algorithms are used for clustering and classification process. After the Clustering process is completed the clustered input is passed to the Classifier process. It is an unsupervised approach for analyzing data in statistics, machine learning, pattern recognition, DM, and bio informatics. It refers to collecting similar objects together to form a group or cluster. Each cluster contains objects that are similar to each other but dissimilar to the objects of other groups. This approach when applied to analyze the dataset derived from educational system is termed as Educational Data Clustering (EDC). An educational institution environment broadly involves three types of actors namely Teacher, student and the environment. Interaction between these three actors generates voluminous data that can systematically be clustered to mine invaluable

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information. Data clustering enables academicians to predict student performance, associate learning styles of different learner types and their behaviours and collectively improve upon Institutional performance. Researchers, in the past have conducted studies on educational datasets and have been able to cluster students based on academic performance in examination

## IV. PROPOSED SYSTEM

In proposed system, the Fuzzy Classification and Clustering Algorithm for Educational Data Mining Process is used. Fuzzy clustering (also referred to as soft clustering) is a form of clustering in which each data point can belong to more than one cluster.

In Fuzzy Classification process it based on Student all semester marks details, total counting and College overall result. It gives the Classification result. This Classification result based Fuzzy Graph generate process it based on Education Department Based

### 4.1 ADVANTAGES

- ✓ Mining process speed is high
- ✓ Educational datas can be easily accessed
- ✓ The preprocessing speed is high
- ✓ This process fully concentrates on educational purpose

## V. OVERVIEW OF THE SYSTEM

Input design is the process of connecting the user-originated inputs into a computer to used formats. The goal of the input design is to make data entry Logical and free from errors. Errors in the input database controlled by input design this application is being developed in a user-friendly manner. The forms are being designed in such a way that during the processing the cursor is placed in the position where the data must be entered. An option of selecting an appropriate input from the values of validation is made for each of the data entered. Concerning clients comfort the project is designed with perfect validation on each field and to display error messages with appropriate suggestions. Help managers are also provided whenever user entry to a new field he/she can understand what is to be entered. Whenever user enter a error data error manager displayed user can move to next field only after entering a correct data

Computer output is the most important and direct source of information to the user. Efficient intelligible output design should improve the system's relationship with the user and admin in decision-making. Output design generally refers to the results generated by the system. For many end users on the basis of the output the evaluate the usefulness of the application. Efficient software must be able to produce and efficient effective reports

## VI. MODULES

- ADMIN LOGIN
- LOAD DATA/ PREPROCESSING

- FUZZY CLASSIFICATION AND CLUSTERING
- CLASSIFICATION REPORT
- GRAPH REPORT

## 6.1 MODULE DESCRIPTION

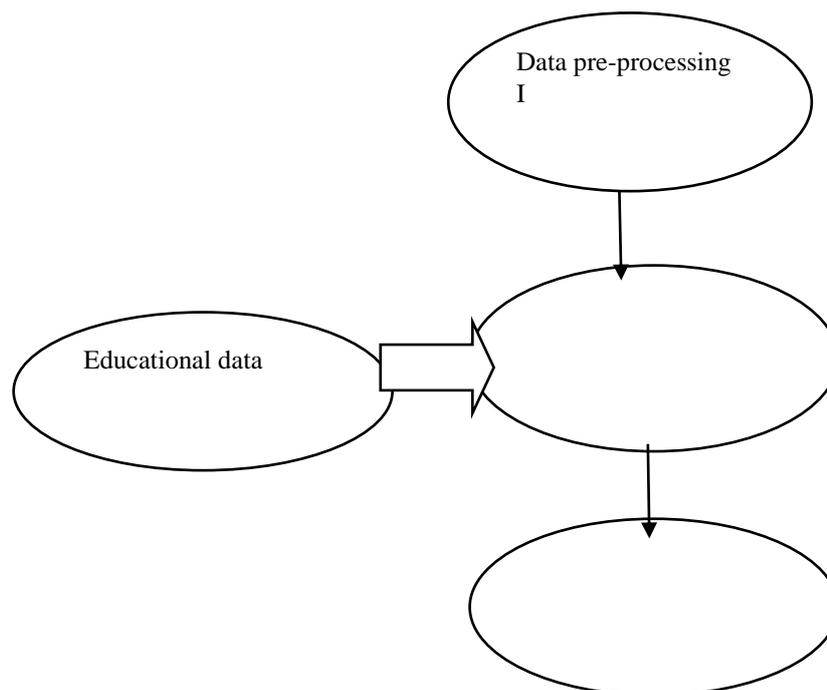
### 6.1.1 Admin Module

In this admin login page provide the Process security the admin is handle in overall process so we implement the admin login security it avoid the un authentication person.

If the correct user was the login it send the successful message else it show the alert message.

### 6.1.2 Load Data/ Preprocessing

In this module, the admin may select the any Education Dataset (2013,2014,2015) in this dataset many type of attributes are there for Ex(Student department, Semester result ,Register/Roll



**Fig 6.1 preprocessing**

### 6.1.3 Fuzzy Classification and Clustering

Fuzzy clustering (also referred to as soft clustering) is a form of clustering in which each data point can belong to more than one cluster. Clustering or cluster analysis involves assigning data points to clusters such that items in the same cluster are as similar as possible, while items belonging to different clusters are as dissimilar as possible. Clusters are identified via similarity measures these similarity measures include distance, connectivity, and intensity Different similarity measures may be chosen based on the that

Fuzzy classification is the process of grouping elements into a fuzzy set whose membership function is defined by the truth value of a fuzzy propositional function. After the Classification process is completed the data was stored in department, class ,total ,average wise.This data's are stored in databases we calculate the over all class count for each department and total year college student education level report.a or the application.

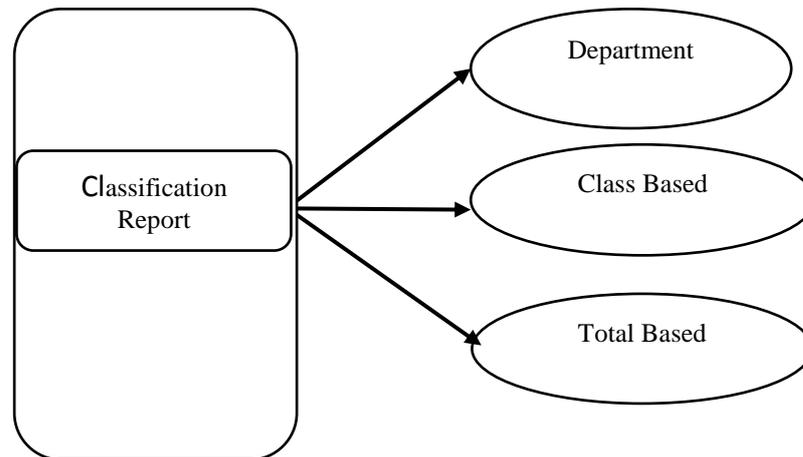


Fig 6.2 fuzzy classification

#### 6.1.4 GRAPH REPORT

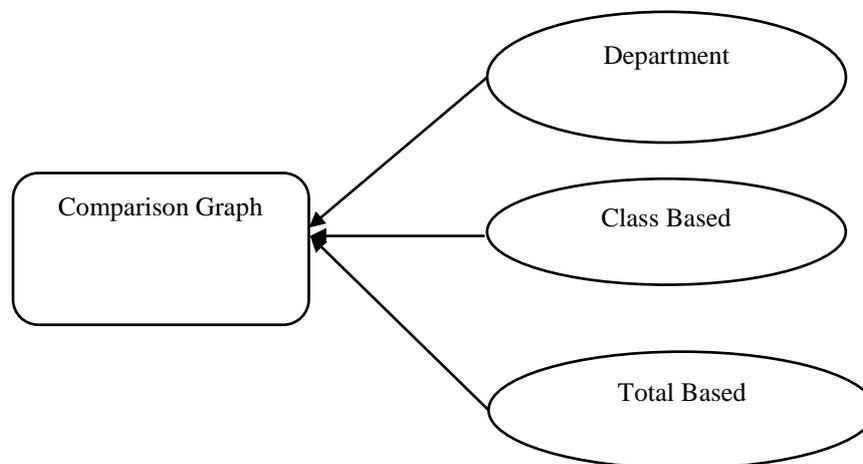


Fig 6.3 Graph Report

Finally the classification report is completed it generate the Graph for based on Student department, class, total average, semester result

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## VII. UPSHOTS

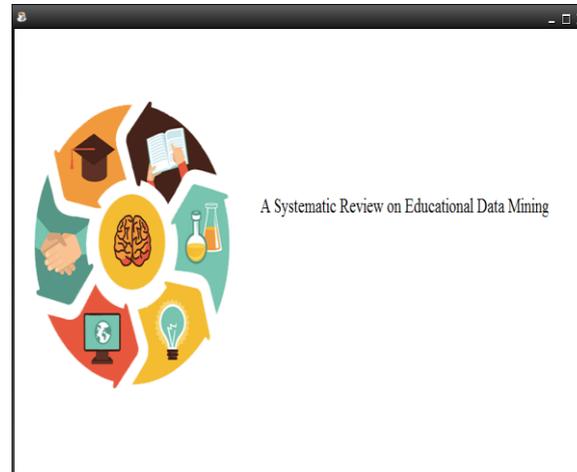


Fig : 7.1 Home page

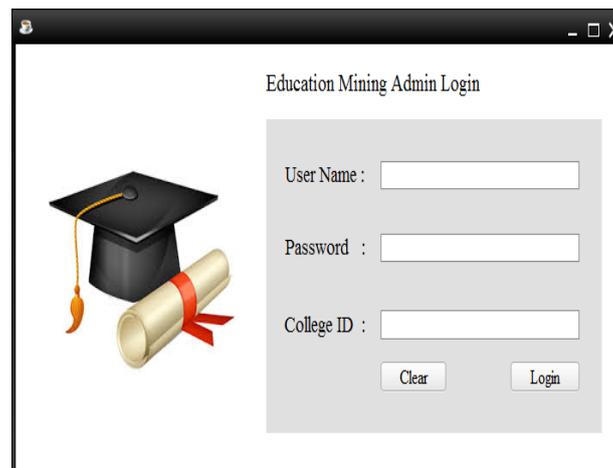


Fig : 7.2 Login module

Fuzzy Classification-1

EDM Department Based Classification: Classification Student Counting: 157

Student-Department: Department MCA

firstname	lastname	dept	college	regulation	gender	address	city	sem1	sem2	sem3	sem4	sem5	sem6	total
Britt	Britt Ma...	MCA	55111-697	68428-102	Male	9 Dwigh...	Krasnos...	53	37	68	97	65	74	394
Ike	Ike Bold...	MCA	76310-001	45281-286	Male	73 Unio...	Fenglu	50	83	91	70	61	65	420
Bertie	Bertie D...	MCA	68016-460	0603-0283	Female	860 Lin...	Beaug...	84	84	81	44	43	85	421
Lane	Lane Sk...	MCA	65841-658	37808-291	Male	69013 ...	Markaz...	82	86	61	84	38	52	403
Kimmi	Kimmi ...	MCA	55316-822	65841-666	Female	158 Dix...	Slantsy	37	36	62	82	39	89	345
Elisabet	Elisabet...	MCA	0395-4202	76053-001	Female	2 Cresc...	San Raf...	97	91	95	68	54	71	476
Padraic	Padraic...	MCA	54458-920	60793-608	Male	60 Ler...	Fort Wo...	70	79	59	87	47	46	388
Sorcha	Sorcha ...	MCA	50845-0...	54569-0...	Female	4636 De...	Fort Wo...	53	48	80	83	60	95	419
Rod	Rod Ric...	MCA	0093-7442	36987-3...	Male	70 Bash...	Grevená	44	50	87	52	59	86	378
Flossi	Flossi D...	MCA	55154-6...	13668-029	Female	0304 le...	Nanling	60	43	45	92	79	56	375
Shenlyn	Shenlyn...	MCA	52125-809	59779-291	Female	2 Schm...	Dortmund	73	77	88	80	37	45	400
Tibold	Tibold B...	MCA	49349-975	53144-1...	Male	93762 ...	Longmen	46	70	62	89	50	46	363
Cullie	Cullie H...	MCA	50436-9...	61919-165	Male	0678 Kn...	Strawbe...	88	64	58	62	65	81	418
Rancell	Rancell ...	MCA	49349-480	33261-732	Male	0 Prent...	Novo-Ni...	35	60	92	70	74	61	392
Helge	Helge M...	MCA	69153-060	0378-0073	Female	00 Moul...	Werang	65	44	85	70	50	43	357
Dabi	Dabi Py...	MCA	0884-2592	0173-0823	Female	051 Con...	Humpolec	53	82	46	77	77	45	380
Sibley	Sibley ...	MCA	13668-158	57237-006	Female	95 Chiv...	Tambob...	68	55	46	84	42	92	387

Fig 7.3 Department wise classification

Fuzzy Classification-2

EDM Department Based Classification :  Fist Class Student Count :

Student Department :  Second Class Student Count :  >>>>>

First Class Student :

firstname	lastname	dept	college	regulation	gender	address	city	sem1	sem2	sem3	sem4	sem5	sem6	total	status
Britt	Britt Mavin	MCA	55111-697	68428-102	Male	9 Dwight...	Krasnos...	53	37	68	97	65	74	394	First Class
Ike	Ike Bold...	MCA	76310-001	49281-286	Male	73 Union...	Fenglu	50	83	91	70	61	65	420	First Class
Bertie	Bertie D...	MCA	68016-460	0603-0283	Female	860 Lind...	Beaug...	84	84	81	44	43	85	421	First Class
Lane	Lane Skaid	MCA	65841-658	37808-291	Male	69013 W...	Markaz...	82	86	61	84	38	52	403	First Class
Elisabet	Elisabet ...	MCA	0395-4202	76053-001	Female	2 Cresce...	San Rafael	97	91	95	68	54	71	476	First Class
Sorcha	Sorcha ...	MCA	50845-0...	54569-01...	Female	4636 De...	Fort Worth	53	48	80	83	60	95	419	First Class
Shenlyn	Shenlyn ...	MCA	52125-809	59779-291	Female	2 Schme...	Dortmund	73	77	88	80	37	45	400	First Class

Second Class Student :

firstname	lastname	dept	college	regulation	gender	address	city	sem1	sem2	sem3	sem4	sem5	sem6	total	status
Kimmi	Kimmi G...	MCA	55316-822	65841-666	Female	158 Dixo...	Slantsy	37	36	62	82	39	89	345	SECON...
Padraic	Padraic ...	MCA	54458-920	60793-608	Male	60 Leroy...	Fort Worth	70	79	59	87	47	46	388	SECON...
Rod	Rod Ricc...	MCA	0093-7442	36987-30...	Male	70 Bashf...	Grevena	44	50	87	52	59	86	378	SECON...
Flossi	Flossi Di...	MCA	55154-6...	13668-029	Female	0304 Ilen...	Nanling	60	43	45	92	79	56	375	SECON...
Tibold	Tibold Bli...	MCA	49349-975	53144-10...	Male	93762 O...	Longmen	46	70	62	89	50	46	363	SECON...
Helge	Helge Mi...	MCA	69153-060	0378-0073	Female	00 Moul...	Werang	65	44	85	70	50	43	357	SECON...
Debi	Debi Pyner	MCA	0884-2592	0173-0823	Female	051 Cont...	Humpolec	53	82	46	77	77	45	380	SECON...
Sibley	Sibley B...	MCA	13668-158	57237-006	Female	95 Chive ...	Tambob...	68	55	46	84	42	92	387	SECON...

Fig 7.4 Mark wise Classification

Education Result

Department Result

Department	First Class Student Count	Second Class Student Count	Over All Student Count	Over All Pass Result
BSC :	<input type="text" value="110"/>	<input type="text" value="56"/>	<input type="text" value="1414"/>	<input type="text" value="55%"/>
MCA :	<input type="text" value="88"/>	<input type="text" value="69"/>		
MSC :	<input type="text" value="77"/>	<input type="text" value="69"/>		
MPhil :	<input type="text" value="77"/>	<input type="text" value="69"/>	First Class Student Count : <input type="text" value="780"/>	Second Class Student Count : <input type="text" value="613"/>
MCom :	<input type="text" value="96"/>	<input type="text" value="77"/>	Education Result Graph	
BCom :	<input type="text" value="75"/>	<input type="text" value="73"/>	<input type="button" value="serching"/>	
MA :	<input type="text" value="78"/>	<input type="text" value="66"/>		
BA :	<input type="text" value="87"/>	<input type="text" value="62"/>		
MBA :	<input type="text" value="92"/>	<input type="text" value="72"/>		

Fig 7.5 Result

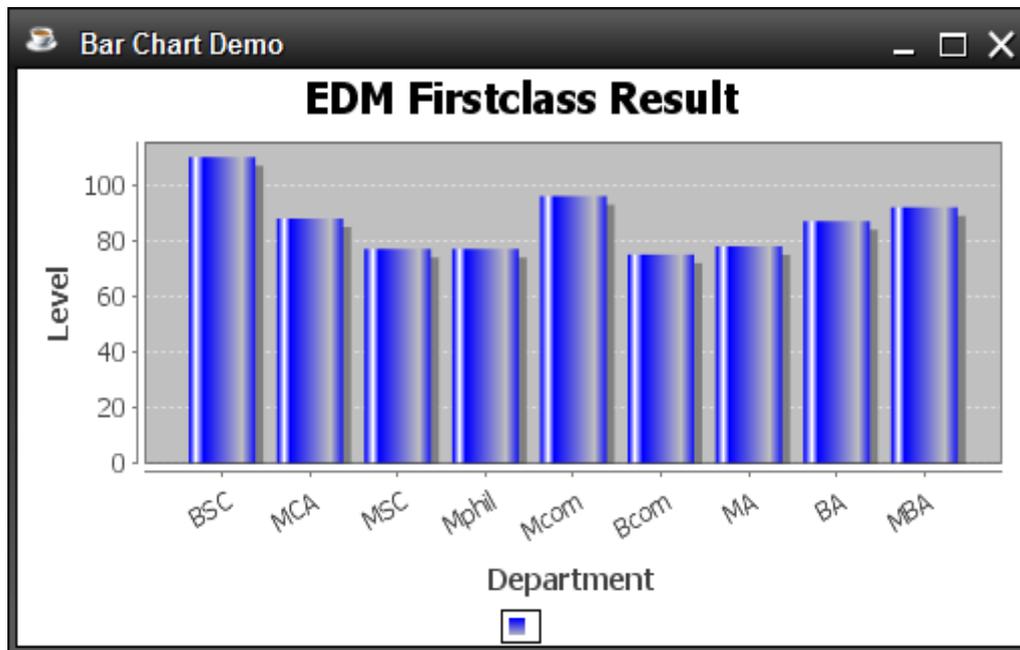


Fig 7.6 First Class Graph

## VIII. CONCLUSION

In our Existing system following the Clustering and Classification Algorithms in Education Data Mining Process In Clustering process c-means , K-means and classification process SVM classification or naive bayes Algorithms are used for clustering and classification Process After the Clustering process is completed the clustered input passing the Classifier process It is an unsupervised approach for analyzing data in statistics, machine learning, pattern recognition, DM, and bio informatics. It refers to collecting similar objects together to form a group or cluster. Each cluster contains objects that are similar to each other but dissimilar to the objects of other groups. This approach when applied to analyze the dataset derived from educational system is termed as Educational Data Clustering (EDC). An educational institution environment broadly involves three types of actors namely Teacher, student and the environment. Interaction between these three actors generates voluminous data that can systematically be clustered to mine invaluable information. Data clustering enables academicians to predict student performance, associate learning styles of different learner types and their behaviors and collectively improve upon Institutional performance. Researchers, in the past have conducted studies on educational datasets and have been able to cluster students based on academic performance in examinations. So it was one of the drawback of to accessing educational data's. To accessing educational data's the mining process may slow. The basic concepts and basic algorithms may only using the educational mining process

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