

DATA MINING IN RETAIL SECTOR TO PREDICT SALES BEHAVIOUR ANALYSIS

A Project Report of Capstone Project - 2

Submitted by

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in partial fulfillment for the award of the degree

of

Bachelor of Technology

IN

Computer Science and Engineering

SCHOOL OF COMPUTING SCIENCE AND ENGINEERING

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APRIL / MAY- 2020



SCHOOL OF COMPUTING AND SCIENCE AND ENGINEERING

BONAFIDE CERTIFICATE

Certified that this project report <u>**"DATA MINING IN RETAIL SECTOR TO PREDICT**</u> <u>SALES BEHAVIOUR ANALYSIS</u> is the bonafide work of <u>MANISHA (1613101374)</u>" who carried out the project work under my supervision.

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TABLE OF CONTENTS

- Abstract
- Introduction
 - Overall description
 - Purpose
 - Motivations and scope
- Proposed model
- Existing System
- Implementation or Architecture Diagrams
- Output/Results
- Conclusion
- References

[1] ABSTRACT

Data mining is proved to be one of the important tools for identifying useful information from very large amount of data bases in almost all the industries. Industries are using data mining to increase revenues and reduce costs. This article begins the concept of data mining that has emerged as a technique of discovering patterns to make better strategies and decisions. It also discusses standard tasks involved in data mining, reviews various data mining applications in different sectors. This paper attempts, how data mining can be applied in retail industry to improve market campaign. It is an inter disciplinary promising field that focuses on access of information useful for high level decisions and also include machine learning to help online shopping stores to indentify online customer behavior the recommend for him the appropriate products he/she is interesting to them because the growing popularity and acceptance of e-commerce platforms, users face an ever increasing burden in actually choosing the right product from the large number of online offers. Thus, techniques for personalization and shopping guides are needed by users. For a pleasant and successful shopping experience, users need to know easily which products to buy with high confidence.

In this paper eleven data mining classification techniques will be comparatively tested to find the best classifier fit for consumer online shopping attitudes and behavior according to obtained dataset for big agency of online shopping ,the results shows that decision table classifier and filtered classifier gives the highest accuracy and the lowest accuracy is achieved by classification via clustering and simple cart, also this paper will provide a recommender system based on decision table classifier helping the customer to find the products he/she is searching for in some ecommerce web sites .Recommender system learns from the information about customers and products and provides appropriate personalized recommendations to customers to find the

desired products..It is also known as Knowledge Discovery in Databases (KDD) and enables data exploration, data analysis, and data visualization of huge databases at a high level of abstraction, without a specific hypothesis in mind. The working of data mining is understood by using a method called modeling with it to make predictions. Data mining techniques are results of long process of research and product development and include artificial neural networks, decision

trees and genetic algorithms. This paper surveys the data mining technology, its definition, motivation, its process and architecture, kind of data mined, functionalities and classification of data mining, major issues, applications and directions for further research of data mining technology. The main aim of the data mining process is to extract the useful information from the dossier of data and mold it into an understandable structure for future use. There are different process and techniques used to carry out data mining successfully.

[2]. INTRODUCTION

The enormous usage of computers has provided a huge amount of data for one disposal. Because of the spiraling amount of data, experts have been facing challenges in extracting useful and meaningful information from it. This has lead to data mining. Data mining is a non trivial process of extraction of information which is hidden, previously unknown and is potentially useful, from large databases. Data mining can also be explained as finding the correlations in a large relational database based on the different depth of angles we analyze it. It is a powerful tool with high potential that helps the organizations or companies to increase their sales and gain more profit from the information about the dealings of their customers. Data mining provides us with the useful information that queries and reports are not able to provide us efficiently. The information that is extracted by the data mining etiquette is not explicitly available in the database, whereas database application only projects the information that is available in the info bank with a restricted manipulation capacity. So data mining is best described as knowledge unearthing in databases.

Today's online shopping is becoming another shopping channel or pattern for doing shopping because the internet has provided consumers with a platform where they can shop smartly. Consumers are free to explore various alternatives from a wide range and choose the best one. Many companies use Internet with the purpose to cut costs and hence reduce the price of the products. It also helps them to reach a larger audience who would require their product. Nowadays, customers use the internet not only to buy the product but also to compare products, prices and find out benefits of buying the product from a particular store.

Online shopping is the process whereby goods and services are bought by consumers from a seller, over the Internet with no intermediary service. It is a form of electronic commerce in which online shop or virtual store evokes the physical analogy of purchasing products or services in a shopping centre.

There are many advantages of online shopping. There are no time and location limits in online shopping.

DATA MINING AND ONLINE SHOPPING

Data mining is crucial for extracting and identifying useful information from a large amount of data that is why retailing companies operate purchase databases in a long way, such that all transactions are stored in arranged order.

A record-of-transaction database typically contains the transaction date and the products bought in the course of a given transaction. Usually, each record also contains eshopper ID, particularly when the purchase was made using a credit card or a frequent buyer card. Therefore, the purchasing sequence of an e-shopper in the database that has made repeated purchase can easily be determined.

This purchase sequence provides a description of the changes in an e-shopper's preferences over time ,because a purchase sequence can reveal the changes of e-shopper's preferences over time.

(i) Overall Description :

<u>KNOWLEDGE DISCOVERY IN DATABASE :</u>

Data mining is composed of seven phases, the first four phases are used for data preprocessing that is data is prepared in a format for further use and the rest three are used to work on the data so formed to retrieve the hidden information. Data cleaning is use to remove all the noise and other inconsistent data from the input database. Data integration is used to integrate the data as data can be inputted from various sources. Data warehouse is a place where all this cleaned and integrated data is kept. Data selection phase selects the data which is best suited for data mining task. Data transformation transforms the data into a format suitable for data mining. Data mining phase use to employ intelligent methods on the data to generate the knowledge or patterns. These patterns are evaluated in the succeeding phase that is the patterns evaluation phase and in the last phase the knowledge is presented in a user Knowledge discovery is the most valued output of computing. Finding new phenomenon or generating new patterns involves Data Mining and which includes a large set of data basically known as Big Data. Big Data comes into role because previous technologies were not able to handle that much amount of data efficiently.

(ii) Purpose :

The Amount of data generated and published over internet is drastically increasing day by day, the collection of Big Data is a set of large data which is too large and complicates to handle in a traditional data processing ways. Big data is the ability of processing data with the following properties velocity, variety and volume.Knowledge Discovery is the most desirable end Finding new phenomena or product of computing enhancing our knowledge about them has a greater value than optimizing production processes or and is second only to task that preserve ourinventories world and our environment

(iii) Motivational and Scope :

Retail industry is also realizing that it is possible to gain a competitive advantage utilizing data mining. Retailers have been collecting enormous amounts of data throughout the years, just like the banking industry, and now have the tool needed to sort through this data and find useful pieces of information. For retailers, data mining can be used to provide information on product sales trends, customer buying habits and preferences, supplier lead times and delivery performance, seasonal variations, customer peak traffic periods, and similar predictive data for making proactive decisions. Here are some examples of how the retail industry has been utilizing

data mining effectively. Marketing: One of the most widely used areas of data mining for the retail industry, as in the banking industry, is marketing. 'Market basket analysis' is a marketing method used by many retailers to determine optimal locations to promote products. Simply stated, it is the study of retail stock movement data recorded at a Point of Sale (PoS)—to support decisions on shelf space allocation, store layout, product location and promotion effectiveness.

- Fraud Detection
- Customer Acquisition and Retention

[3]. PROPOSED MODEL

Data mining can be implemented in many fields depending on the aim of the business. In today' s business world, many industries can apply data mining technology such as Manufacturing, Banking, Healthcare, Insurance, Telecom, Medicine, and Retail. Telecommunication can apply data mining for customer retention, fraud analysis, and defect management. It helps to identify telecommunication patterns, catch fraudulent activities, intrusion detection, make better use if resources and improve service quality. Data mining can help healthcare insurers to detect fraud and abuse, healthcare organizations make customer relationship management decisions, physicians identify effective treatments and best practices, and patients receive better and more affordable healthcare services. Computer Science and Engineering Data mining in computer science and engineering can be used to help monitor system status, improve system performance, isolate software bugs, detect software plagiarism, analyze computer system faults, uncover network intrusions and recognize system malfunctions. Bank and financial sector Data mining in bank or any financial sector for loan payment analysis, customer credit policy analysis, target marketing, detection of money laundering and other financial crimes can be applied.

Nowadays Big data is playing a very big role in industrialization and sales industry and we can use that Data using Data Mining to extract more information and patterns to make it more valuable and useful.And key components of making it possible are described below.

1. Market Basket Analysis Market :

Market basket analysis is a <u>data mining</u> technique used by retailers to increase sales by better understanding customer purchasing patterns. It involves analyzing large data sets, such as purchase history, to reveal product groupings, as well as products that are likely to be purchased together. The adoption of market basket analysis was aided by the advent of electronic point-of-sale (<u>POS</u>) systems. Compared to handwritten records kept by store owners, the digital records generated by POS systems made it easier for applications to process and analyze large volumes of purchase data.

2. Consumer behavior :

Consumer behavior means the study of individuals, groups or organizations about their process of selecting, securing, using and disposing the products, services, experiences or ideas to satisfy needs and the impact of these process on the consumer and the society.

Behavior concerns either with the individual or the group (e.g. In college friends influence what kind of clothes a person should wants to wears) or a firm (peoples working in firm make decision as to which products the firm should use.) The use of product is often so important to the marketer because this may influence how a product is best positioned or how we can encourage increased consumption.

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Consumer Behaviour Analysis :

Before diving into the deep end of consumer behaviour analysis, it is important to understand its meaning. Consumer behaviour analysis is the study of how people make purchase decisions with regard to a product, service or organisation. Studying consumer behaviour would allow you to answer several questions, such as:

- How consumers feel about alternatives to their preferred brands;
- How consumers choose between the alternatives;
- How consumers behave while shopping;
- How consumer behaviour is swayed by their surrounding environment;
- How marketing campaigns can be improved to more effectively influence customer behaviour.

In order to understand this concept better, let's take a look at the factors that affect consumer

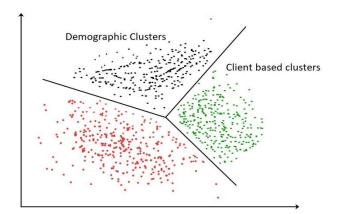
behaviour:

- **Psychological:** This is considered to be the most important factor that affects consumer behaviour. Traits like perception, motivation, personality, beliefs and attitude are important to decide why a consumer would buy a product.
- **Personal:** These are characteristics that are applicable to individuals and may not relate to other people in a group. These factors can include age, occupation, financial situation and lifestyle.
- Social: Social characteristics play an important role in consumer behaviour, and it can include family, communities and social interaction. These factors are difficult to assess while preparing marketing plans.
- **Geographical:** The location of consumers also play a role in how they purchase products. For example, a person living in warmer weather would be less likely to purchase winter clothing compared to someone living in temperate climate

3. Algorithms Used :

• <u>Clustering :</u>

Clustering is a Machine Learning technique that involves the grouping of data points. Given a set of data points, we can use a clustering algorithm to classify each data point into a specific group. In theory, data points that are in the same group should have similar properties and/or features, while data points in different groups should have highly dissimilar properties and/or features. Clustering is a method of unsupervised learning and is a common technique for statistical data analysis used in many fields.



• <u>Demographic Clusters</u> :

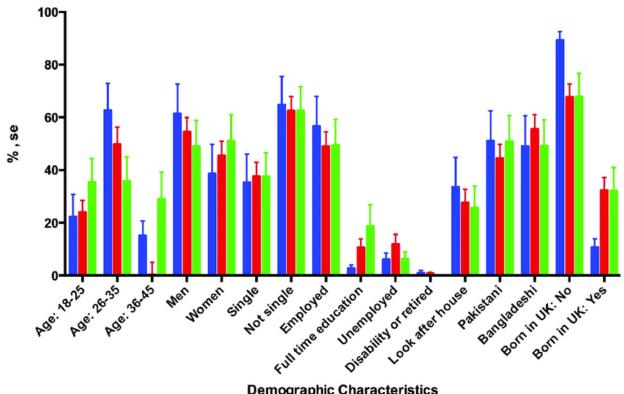
In <u>demographics</u>, **clustering** is the gathering of various <u>populations</u> based on <u>ethnicity</u>, <u>economics</u>, or <u>religion</u>.

Demographic clustering is distribution-based. It provides fast and natural clustering of very large databases. Clusters are characterized by the value distributions of their members. It automatically determines the number of clusters to be generated.

Typically, demographic data contains many categorical variables. The mining function works well with data sets that consist of this type of variables.

You can also use numerical variables. The Demographic Clustering algorithm treats numerical variables by assigning similarities according to the numeric difference of the values. Demographic Clustering is an iterative process over the input data. Each input record is read in succession. The similarity of each record with each of the currently existing clusters is calculated. If the biggest calculated similarity is above a given threshold, the record is added to the relevant cluster. This cluster's characteristics change accordingly.

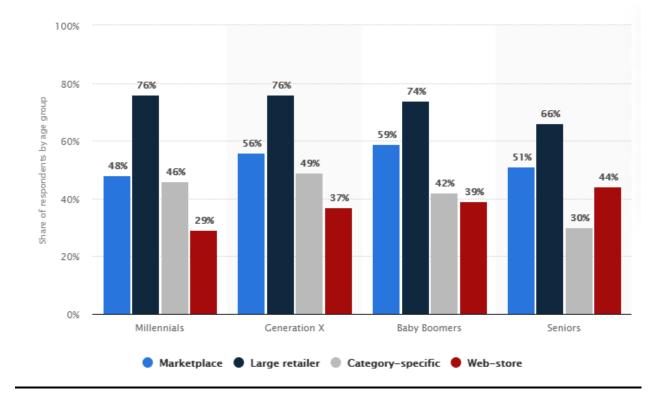




Demographic Characteristics (univariate, weighted prevalence estimates)

• Age Group Clusters :

This statistic presents the shopping preference regarding retail types in the United States as of 2017, by age group. According to the findings, the strongest preferred online shopping channel across all age groups were large retailers, with 76 percent of Millennials having recorded this shopping method. Additionally, 74 percent of Baby Boomers held the same preference as Millennials.



• <u>Client Based Clusters :</u>

In the context of <u>customer segmentation</u>, cluster analysis is the use of a mathematical model to discover groups of similar customers based on finding the smallest variations among customers

within each group. These homogeneous groups are known as "customer archetypes" or "personas".

The goal of cluster analysis in marketing is to <u>accurately segment customers</u> in order to achieve more effective customer marketing via personalization. A common cluster analysis method is a mathematical algorithm known as *k-means cluster analysis*, sometimes referred to as scientific segmentation. The clusters that result assist in better customer modeling and <u>predictive analytics</u>, and are also are used to target customers with offers and incentives personalized to their wants, needs and preferences.

		R		F		Μ		
Customers		Days : Purch	Since Last ase	Number of (Past 12 Mo		Net Reven (Past 12 M		
High Spenders	922		9		4			\$154
Mid Spenders	581		54		3			\$121
Risk of Churn	807		192		2			\$70
Low Spenders	1,361		192		2			\$4
	3,671		447		3			\$87

- Language Bases Clusters
- Location Based Clusters
- Sales Based Clusters

As compared with Knowledge Database Discovery segmentation, the three main advantages of the analytical segmentation approach represented by cluster analysis are:

- Practicality It would be practically impossible to use predetermined rules to accurately segment customers over many dimensions
- Homogeneity Variances within each resulting group are very small in cluster analysis, whereas
 rule-based segmentation typically groups customers who are actually very different from one
 another.
- Dynamic clustering The clusters definitions change every time the clustering algorithm runs, ensuring that the groups always accurately reflect the current state of the data.

Existing System

Data is raw material of information that can be understood as any facts, numbers, or text which can be processed by machines. Information is the data that has been given some meaning nu way of relational connections. For ex data collected from sales transaction can be used to analyze sales trends of particular years. Knowledge is application of data and information.it can be considered as general awareness of information, facts, ideas, truth or principle.

Across a wide variety of fields, data are being collected and accumulated at a dramatic pace. There is an urgent need for a new generation of computational theories and tools to assist humans in extracting useful information (knowledge) from the rapidly growing volumes of digital data. These theories and tools are the subject of the emerging field of knowledge discovery in databases. At an abstract level, the KDD field is concerned with the development of methods and techniques for making sense of data. The basic problem addressed by the KDD process is one of mapping low-level data (which are typically too voluminous to understand and digest easily) into other forms that might be more compact (for example, a short report), more abstract (for example, a descriptive approximation or model of the process that generated the data), or more useful (for example, a predictive model for estimating the value of future cases). At the core of the process is the application of specific data-mining methods for pattern discovery and extraction.

Data Mining is basically used today by most of the companies with a very strong consumer focus retail, financial, communication, and marketing organizations, to drill down into their transactional data and determine pricing, client preferences and product related information, impact on the sales, client satisfaction and corporate

profits. With the help of data mining, a retailer can use point-of-sale records of client purchases to develop products and promotions to appeal to specific client segments.

Knowledge discovery and data mining have become areas of growing significance because of the recent increasing demand for KDD techniques, including those used in knowledge acquisition, machine learning, databases, statistics, data visualization, and high performance computing. Knowledge discovery and data mining can be very useful for the field of Artificial Intelligence in many areas, for example education, industry, commerce, government, and so on. The relation between Knowledge and Data Mining, and Knowledge Discovery in Database (KDD) process are presented in the paper. Data mining theory, Data mining tasks, Data Mining technology and Data Mining challenges are also proposed.

Disadvantages of Existing Model :

- Excessive work intensity may require investment in high performance teams and staff training.
- The difficulty of collecting data.Depending on the type of data that you want to collect can be a lot of work.
- The Existing Model is totally based upon the customers past behavior, and determine the future event totally based upon that.
- Now the main problem in this Model is that if Customer changes its taste and not buying the things from the past event in that case it won't be able to give accurate results.

Implimentation or Architecture Diagram

Case Study of Wal Mart :

Wal-Mart is often described as a pioneering leader in data mining and data management: Wal-Mart captures point-of-sale transactions from over 2,900 stores in six countries and continuously transmits this data to its massive 7.5 terabyte data warehouse.

Wal-Mart allows more than 3,500 suppliers to access data on their products and perform data analyses. These suppliers use this data to identify customer-buying patterns at the store display level. They use this information to manage local store inventory and identify new merchandising opportunities.

Other companies supplement their customers' transactional information with external data such as postal codes to do a market basket analysis. Practically every retailer now records all the details of each POS (Point of Sale) transaction for stock keeping purposes. Sometimes these are supplemented by customer information. *Home Depot*, for example supplements the data with ZIP or postal code of the purchaser. Sometimes the cashier may also enter the sex and appropriate age of the customer into the cash register. Affinity cards and credit card numbers can be used to track repeat customers.

DATA SETS

ONLINE SHOPPING DATA SET :

The Dataset used is obtained from highly reputational online shopping agency which sells only online .The dataset is composed of online ordering log file for three months. The dataset consists of 304 instances and 26 attributes.

All data sets are given in the Table Below ------

Attribute Name	Description
Personal information	Include serial number, buyer name ,gender ,age
Educational level	Describes buyer educational level and it is
	classified into categories from (1-10)
	(1-3) Graduated,(4-6) Master,(7-10) PHD.
Brand	Describes product brand name.
Product name	Describes the product name.
Item description	Describes the product specification.
Category	Describes product Category.
Quantity	Describes product ordered quantity per order.
Price	Describes product price.
Item Type	Describes the product different types.
Payment Method	Describes order payment method which is
	classified here into three methods (COD):cash
	on delivery ,credit card, buyer web site
	account.
Number of visits	Describes buyer visit number for the web site
	page.
duration of visit	Describes buyer duration visit and it is
	measured by minutes.
Rating	Describes product rating from the buyer and
	it's measured by scale from(1-5)
	(1) represent poor and (5) represent excellent.
User Satisfaction of	Describes user satisfaction from the product
the product	and it is rated from(1-100) .100 represent
	highly satisfied and 1 represent not satisfied.
Best deal	Describes the best offer for the product 1

The test procedure is repeated 10 times .The final accuracy of an algorithm will be the average of the 10 trials.

<u>Algorithms sampling –</u>

Cluster Analysis

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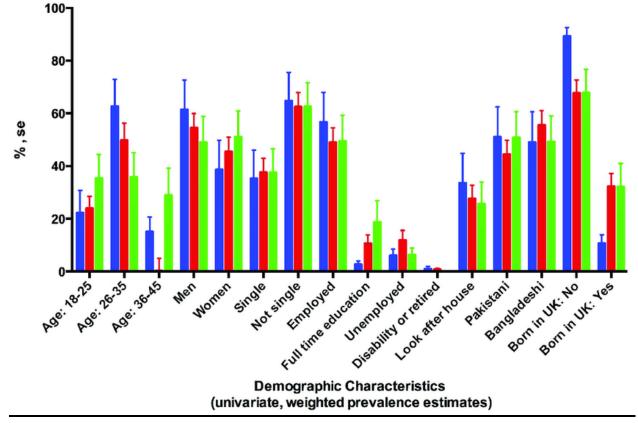
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The process is not based on any predetermined thresholds or rules. Rather, the data itself reveals the customer prototypes that inherently exist within the population of customers.

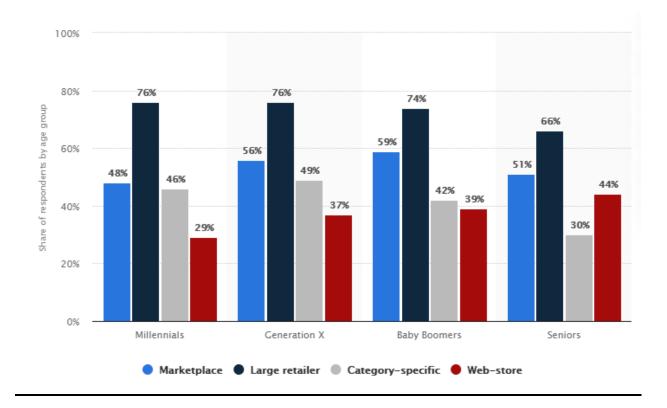


Demographic Based Clusters Sample :





Age Based Clusters Sample Data :



<u>Client Based Clusters :</u>

		R		F		Μ		
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Output / Results

Types of Outputs

Generally, applications of data mining can generate outputs such as:

- Buying patterns of customers; associations among customer demographic characteristics;
- Predictions on which customers will respond to which mailings;
- Patterns of fraudulent credit card usage;
- identities of "loyal" customers; credit card
- spending by customer groups;
- Predictions of customers who are likely to change their credit card affiliation;
- Predictions on which customers will buy new insurance policies; behavior patterns of risky customers;
- Expectations of fraudulent behavior;
- Characterizations of patient behavior to predict frequency of office visits.

The Result investigate the performance of selected classification algorithms that was previously described by using WEKA. Figure 4 and figure 5 represent snap in for the result is composed of two sectors **WEKA explorer** interface that include classification algorithms analysis and results and WEKA knowledge flow interface that include the knowledge flow charts analysis for each algorithm respectively.

Test options	Classifier output						
 Use training set 	Time taken to I	ouild mode	el: O seco	nda			
O Supplied test set Set	=== Stratified	cross-val	lidation ==				
Cross-validation Folds 10	=== Summary ===						
Percentage split % 66		and we want				- Alexandra	2
	Correctly Class			154		50.8251 %	
More options	Incorrectly Cla Kappa statistic		Instances	149		49.1749	-
	Mean absolute			0.23			
(Nom) class -		a start and a st		0.46			
		Relative absolute error					
Start Stop	Relative absolute error 74,9141 % Root relative squared error 115.0949 %						
Result list (right-click for options)	Total Number of	f Instance	es	303			
17:06:08 - lazy.KStar							
	=== Detailed A	ccuracy B	y Class ==	•			
		TP Rate	FP Rate	Precision	Recall	F-Measure	
		0.29	0.149	0.333	0.29	0.31	
		0.386		0.466			
		0.719	The second s			0.667	
		0.16		0.182			
	Weighted Avg.	0.508	0.263	0.49	0.508	0.496	
	=== Confusion 1	Matrix ===					
				11			

WEKA Explorer Tools results are given below -----

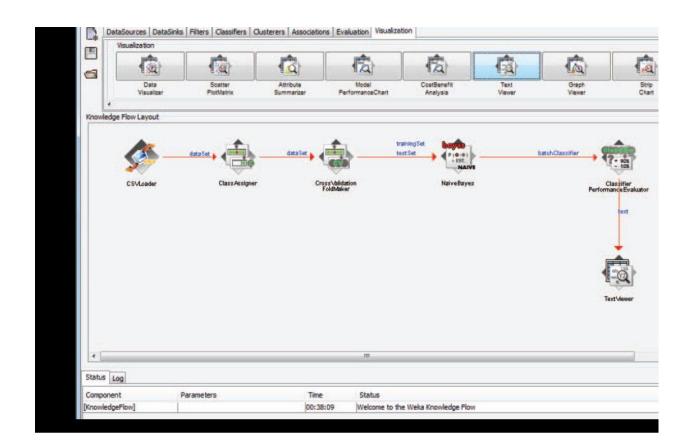
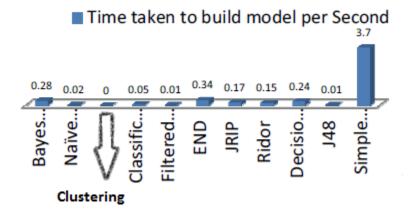


TABLE VI. CLASSIFIERS EXECUTION TIME

Classifier Name	Time taken to build model per
Clustering	0 Second
Filtered Classifier	0.01 Second
J48	0.01 Second
Naïve Bayes	0.02 Second
Classification via Clustering	0.05 Second
Ridor	0.15 Second
JRIP	0.17 Second
Decision Table	0.24 Second
Bayes Net	0.28 Second
END	0.34 Second
Simple Cart	3.7 Second

Classifiers Time taken to build model



Conclusion

Data mining is neither a magic bullet nor a simple process and as such presents challenges that go well beyond the technical. Data mining should be used with carefully outlined criteria, which should not be based on any kind of prejudice or stereotype assumptions. Otherwise the outcome of datamining will not give the desired result. Many data management challenges remain, both technical and societal. Large online databases raise serious societal issues. To cite a few of the societal issues: Electronic data interchange and data mining software make it relatively easy for a large organization to track all of your financial transactions.

By doing that, someone can build a very detailed profile of your interests, travel, and finances. Is this an invasion of your privacy? Indeed, it is possible to do this for almost anyone in the developed world.

Data mining is a very powerful tool that should be used with utmost care for increasing customer satisfaction, providing best, safe and useful products at reasonable and economical prices. This should be used for making the business more competitive and profitable. Data mining should be used in any way that affects the privacy of common man, so that the confidentiality and individuality of human being is preserved. It should not be used in any way that may cause undue hardship, financial loss or emotional setback.

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