*Research Article* 

# Prediction Of Consumer Review Analysis Using Naive Bayes And Bayes Net Algorithms

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**Abstract:** Datamining field that aims to bring out knowledge from three different form of Structured and unstructured, Semi structured forms. Classification techniques can be used to classify the large Volume of data and Variety of data. Classification supervised learning approach to data processing and WEKA tool is an effective and efficient tool with many inherent methods for extracting useful information. In this paper using the weka tool to analyze consumer data in the weka tool having the number of algorithm. In that algorithm we are used classification algorithm Bayes Network algorithm, Naive Bayes algorithm. Consumer behavior analysis is important of making decisions in the supermarket, consumer behavior prediction different data include in consumer behaviour analysis explains the all data and also use to identify the hidden relationships of data.

Index Terms: Consumer behaviour, Data Mining, Weka tool, Classification, Naïve Bayes, Bayes net.

#### **1.Introduction**

Data Mining is a group of process that the applies huge, complex databases. This is to eliminate random data and find the hidden data. These data mining methods are always computational concept. Data mining concepts are used to extract the hidden information or useful most informative data. Data mining concepts is most powerful technique in these techniques to implement the different data to analysis and get different result. In this result helps to taking the decision making process. We use data mining, procedures and concepts to reveal patterns in data. There is many of driving power at present. This is the reason was data mining processing has become such an important study.

WEKA is machine learning tool. In this developed by Java. It is having the set of instructions to implement the DM algorithm and machine learning algorithm. Directly put in to the data or java code using WEKA tool. WEKA is a group of tools for: Regression, Clustering, Associations, Data pre-processing, Classifications and Visualizations.

#### 1.1 Shopping Data set

Consumer actions is different based on action data are stored and use to analysis in this data information are categories in several criteria's age, income, budget, needs, products this are all in the dataset .in this paper all dataset how it is used and also how to implement the algorithm, which algorithm is give the best result .

Attributes	Description								
Personal detail	Consumer Name, Age, Gender, City								
Consumer income details	Categories to low level(income10000), middle level (income50000), high level( income100000 above)								
Budget	Categories to different budget levels								
Price	Describes the price level								
Brand	Categories to different brand like the consumer								
Items	What the items purchased consumer								
Payment mode	Payment mode on cash, card, net bank, or other payment mode								
Number of super market	Consumer like which super market								
Consumer satisfaction	Based on consumer satisfaction								

#### Table 1: List of Various Dataset Descriptions

Based on the data set classification to predict the consumer behaviours in the all attributes give the current scenario of the market. Research of analysis different category one is who buy the goods, why, what, where in this all category analysis used to market prediction, Most important process in analysis of consumer patterns estimates the market data[1].

## 1.2Classification

First introduce the sample data model to obtain the best limit conditions use the stimulate everyselected group of data. Only one limitedConstance has been determining, the coming upassignment is to measurement the selected groups of data. Total progress is called classification techniques. Two types of learning one is supervised

learning and second is unsupervised learning .Classification techniques one of the supervised learning. The classification technique used to classify the data by implementing tested data.

Selected group of data examples:

- Predicating the consumer data either he will purchase food item (selected group of data: Yes or No)
- Classify the food item from characteristic like colour, taste, size, weight (selected group of data: dhal, oil, milk, and coffee )
- categorize from gender (selected group of data : Male or Female)

The selected group of data classification involves following process:

- Implementing the testeddata file.
- Associate the categories labels and groups.
- Classified the functional elements to the located.
- Sample data and Tested data used by learning method.
- Used to identify the model and to categorize the hidden data set[2].



Fig 1: Classification process system

This paper summarizes the Consumer based analysis has different approach. In this survey section 2 specifies various related works. Section 3 gives the information for type of classification algorithms. Section 4 implies the Execution methods section 5 deal results and discussions Consumer behaviors and section 6 explain the conclusions.

# 2.Related works

Shaffy goyal, Namisha modi using different classification algorithm and compare the Naïve Base algorithm identified probability of two class label[3].Dr.anil sharma and Balrajpreet kaur to compare the different classification algorithm like KNN, Naïve Base, decision tree in that three algorithm give the best result naïve base only.[4] Nagaraju Orsu Gopala Krishna Murthy Nookala , Suresh B. Mudunuri ,Bharath Kumar Pottumuthu, using different classification algorithm and different data all classification algorithm and data is give the different result[5]. Abdul Hamid M. Ragab, Amin Y. Noaman, Abdullah S. AL-Ghamdi, Ayman I. Madbouly, in the all authors to compare the different likeC4.5 algorithms, Random Forest algorithm , IBK-E algorithms, LibSupport Vector machine algorithms, MLP algorithms, Multilayer Perceptron algorithms, Naïve Bayes algorithms, and PART algorithms. Comparative study and analysis related to classification predict included Recall, Precision, F-Measure, Matthews's Correlation Coefficient (MCC), Precision-Recall Curve (PRC), ROC curve, FP-rate, and TP-rate have been analysis by simulation using Weka Toolkit. Experimental Results given that C4.5 gives the best performance and accuracy and lowest absolute errors, then PART, Random Forest, Multilayer Perceptron, and Naïve Bayes, respectively[6]. Yugal kumar and G. Sahoo compare the all algorithms BayesNet, Naïve Bayes, Naïve Bayes Up table, Multilaverperceptron, Voted perceptron and J48 But, it is not easy to say which one is the best mean absolute value is false rate of J48 is small cancer data for breast i.e. small data and large data from the table one, seen give the action of the J48 classification is best compare to another classification technique[7]. PrudhviRaj G,Arka Haldar , Lakshmi S.V.S.S, Arka Halder ,Prudhvi Raj G, Lakshmi S.V.S.S, all authors make to compare this three algorithm decision tree ,naïve bayes, zero R using weka tool so that final result is naivebayes algorithm is best one running time is very low and accurate value is given[8]. Md. Faisal Kabir, Alamgir Hossain, Keshav Dahal in these three authors to identify the classification algorithm is the best one of the algorithm[9].

# **3.**Classification Algorithms

A classification algorithm in common, is a role that weighs the input value so that the output separates one class into positive data values and the other into negative data values. Linear Classification algorithms, Logistic regression classification algorithms, Naive Bayes classification algorithm, Fisher's linear discriminate classification algorithms, Support vector machines classification algorithm, Least squares support vector machines classification algorithms, Quadratic classification algorithm, Kernel estimation classification algorithm, k-nearest neighbor classification algorithm, Decision trees classification algorithm, Random forests classification algorithm, Neural networks, Learning vector quantization different classification algorithms available in data mining. Here are many Bayes classification algorithms available inWeka toolkit in this article explain and compare the data only these two algorithm Bayes Net, Naïve Bayes[10].

## 3.1 Bayesian classification

Bayes' Theorem construct through the Bayesian classification. Bayesian classification are statistics classification. Bayesian classification are predicting the data labels. Belonging to the probability that a given attributes held by the specific data label. Bayes classification algorithm is more efficient and correct one. Probability having different types :

Probability of posterior [P(S/X)]

Prior Probability of prior [P(S)]

Hear X - data columns/rows, S -hypothesis. as stated by to Bayesian Theorem,

Posterior probability possibility



#### 3.1.1 Bayes net algorithm

Bayesian belief networks refer to collective conduction probability. That is called trust networks, Bayesian network graph or probability network graph

- The Bayes Networks graph primate the group of condition probability to the describe between subgroup of attributes.
- It display a visual method connection graph
- It is used to developing the Bayesian classification Networks

Bayes network having two types one is Directed acyclic graph, Second one is condition probability tables.

DAG is directly connected the eggs random variable, that variable may be is a different value or continues value, that variables may be same to the correct value given in the dataset[11].

The below figure shows a DAG for having six values



Fig2: Bayesian Network Directed Acyclic Graph

In the above flow chart show the information of cancer disease. For the example, cancer disease is affacted by a reason to human first one family history, and either or not the human is smoking. So that the value comes PositiveXray is not dependent variables of either the human has a family history and also smoking or that the human is cancer disease, Now we know the human has cancer.

• Condition Probability

In this table for the given values and variable Cancer show the every possibility of combination in the values of present nodes, Family History and Smoking is given in the table.

		0 0		
Disease	Family Family		-Family	-Family
	History,	History,	History,	History,
	Smoking	-Smoking	Smoking	-Smoking
+CA	.07	.05	.06	.02
-CA	.03	.05	.04	.03

Table2: Condition Probability

#### 3.2 Naïve Bayes algorithm

One of the "supervised learning "algorithms is naïve bayes algorithm .it is give the both input data and output classified data's (i.e., the data having the right input and output for every point). Naïve Bayes classification algorithms based on Bayes' theorem which is powerful to the predicted the variables . Naïve Bayes algorithm is to is classified the group of data items to efficient and, currect, fast. It is more accept in different group of data prediction analysis .When we assume of non dependence data variables is handle, a Naïve Bayes algorithm to perform the good compare to other model like regression analysis. It is good performinggive to the different input data compare the numeric value of variable in data, for numericvalue of distribution is predicted. In the assumption is they have now appear of character isindependent to the current of any other character in the adjusted class. For example, a food item may be considered as different quality of food if it is having quality of food and no quality of food . all thoseproperty are independent or dependent of each other, they contribution of independent to thevariable probability distribution of that the food is a quality or no quality. Bayesian classification, the main aim is to find the posterior probabilities i.e. the probability of a label given some observed features, P (B | characteristic). With the help of Bayes theorem, we want to express this in unit form as follows [12].

# $P(B|char) = \frac{P(B)P(characteristic|B)}{P(characteristic)}$

P(B|char)= (P(B)P(characteristic|B))/(P(characteristic))

Here, (B | characteristic)- posterior probability.

P(B) -prior probability.

 $P(characteristic \ | B) \ - \ probability \ prediction \ class.$ 

P(characteristic) - prior probability of prediction.

In the given example of the data is climate and equal of given class variable possibility is playing. In there we want to identify the players playing or not playing based on climate condition. See following pace to performing it. fits transfer data into transaction table second Create table to calculate the probability like cloudy probability = 0.29 and playing probability is 0.64 and next we use Naive Bayes theorem equation to the posterior probability for everyclass. The class with the top level posterior probability is the output come of prediction

Problem: Players will play if climate in sunny. In this statement is true?

Table.3: Weather						
climate	play					
Sunny	No					
Cloudy	Yes					
Rain	Yes					
Sunny	Yes					
Sunny	Yes					
Cloudy	Yes					
Rain	No					
Rain	No					
Sunny	Yes					
Rain	Yes					
Sunny	No					
Cloudy	Yes					
Cloudy	Yes					
Rain	No					

Transaction table								
climate	Yes	No						
Sunny	3	2						
Rain	2	3						
Cloudy	4	-						
Total	9	5						

Potentiality table									
Climate	Yes	No							
Sunny	3	2							
Rain	2	3							
Cloudy	4	-							
Total	9	5							
Total no yes& no	=9/14	=5/14							

Table.4: Transaction

Table.5: Potentiality

0.64

0.34

Total no of Sunny = 5/14 = 0.36Total no of rain = 5/14 = 0.36Total no of cloudy = 4/14 = 0.26

We apply the formula of posterior probability.

Total no yes& no

P(Yes OF Sunny) = P(Sunny OF Yes) \* P(Yes) / P(Sunny), Here P(Sunny OF Yes) = 3/9 = 0.33, P(Sunny) = 5/14 = 0.36, P(Yes) = 9/14 = 0.64, Now (Yes OF Sunny) = 0.33 \* 0.64 / 0.36 = 0.60, which has higher one probability ratio.

Naïve Bayes use the equal method of predicting the probability ratio of different class an entity. In that algorithm frequentlyused in classification of data [13].

### **4.Experimental Setup**

In this experimental setup, using the weka tool differentanalysis done bytheclassification algorithm .model of WEKA has been implemented to the executing classification algorithms using the consumer Data for csv file format and arff file format also used. When startup Weka software, Explorer and experiment is the first and second display the various menu display Weka. First select Explorer menu the aim of analysis the Consumer data in this is analysis in two process one Preprocess and second one is Classify the data

#### 4.1. Preprocess

Starting Weka Explorer, first we selected and loaded the data after that filtering the data by using supervised attribute selection option. Before the filtering, 16 attributes have selected for classification and after the filtering identify the 8 attributed only selected. In this step is basic step for all classification algorithms. How the data set to select and loading, filter into weka tool. Select the Classifiyoption in the first step of Explorer by choosing the Bayesian method of classification algorithm Bayes Network are selected the ten times multiple Confirmation of experiment method setup the classified. If you want change the fold cross change it using classification for different data to showing in explanation of execution method selecting and set up the classify data has deen show the Fig.3,4 For(egGender, review)[14].

Veka Explorer									- C	1
Preprocess Classify Cluster Associate	e Select attributes V	isualize								
assifier										
Choose BayesNet -D -Q weka.classifier	s.bayes.net.search.loca		BAYES-E	weka.classifie	rs.bayes.n	et.estimate.Sin	npleEstima	atorA 0.5		
et ontions	Classifier output									
	Clubbiner output									
O Use training set	Time solves as b									
O Supplied test set Set	lime taken to b	ulla model	L: 0.03 Se	conas						
Cross-validation Folds 10	Stratified	cross-vali	dation ==	-						
	Summary									
Percentage split % 66										
More options	Correctly Class	Correctly Classified Instances			615 85.4167 %					
	Kappa statistic	ssified if	is callees	0.5147						
	Mean absolute e	rror		0.15	93					
lom) Gender	Root mean squar	ed error		0.33	7					
	Relative absolu	te error		55.04	99 %					
Start Stop	Root relative s	quared ern	ror	88.70	08 %					
sult list (right-click for options)	Total Number of	instances	5	720						
18:45:26 - bayes.BayesNet	=== Detailed Ac	curacy By	Class ===							
		TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Cl
		0.901	0.365	0.921	0.901	0.911	0.516	0.909	0.981	Ma
		0.635	0.099	0.576	0.635	0.604	0.516	0.909	0.689	Fe
	Weighted Avg.	0.854	0.319	0.860	0.854	0.857	0.516	0.909	0.929	
	=== Confusion M	atrix								
		-1								
	535 59   a	- Male	1 43							
	46 80 j b	- Female								
										7 1

Fig.3: Bayean Network Classifier algorithm and Result for Gender

BayesNet -D -G Weka.classife	rs.bayes.net.search.loca	1.K2 P 1 -8 BA	TES -E Weka.classm	ers.bayes.h	et.estimate.sin	pieestima	norA 0.5		
Test options	Classifier output								
<ul> <li>Use training set</li> </ul>	=== Summary ===								- F
O Supplied test set Set	Correctly Class	ified Testan	512		71 1111				- 1
Cross-validation Folds 10	Incorrectly Cla	ssified Inst.	ances 208		28.8889				
O Bereentage collit	Kappa statistic		0.5	821					
C Percentage spint 40 00	Mean absolute e	error	0.1	.809					
More options	Relative absolu	te error	52.1	62.1049 %					
	Root relative s	quared error	73.1	794 %					
(Nom) Review	Total Number of	Instances	720						
	Detailed Ac	curacy By Cla							
Start Stop									
Result list (right-click for options)		0.634 0	.178 0.644	0.634	0.639	0.457	0.852	0.711	Ser
18:45:25 - bayes BayesNet		0.713 0	.211 0.682	0.713	0.697	0.498	0.862	0.809	Qua
18:47:05 - bayes.BayesNet		0.711 0	.039 0.807	0.711	0.756	0.706	0.959	0.865	Pri
	Weighted Avg.	0.711 0	.150 0.712	0.711	0.710	0.562	0.889	0.803	Och
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	Confusion M	latrix ===							
	a b c	d < class	sified as						
	154 73 9	7 I a = Se:	rvices						
	66 199 14	0   b = Qui	ality						
	0 0 0 6	3   d = 0t)	hers						

Fig.4: Bayean Network Classifier algorithm and Result for review

The weka Tool, Select the Classifiyoption in the first step of Explorer by choosing the Bayesian method of classification and Bayes Network algorithm and selected the ten times multiple Confirmation of experiment menu and set up the classifyed data. if you want change the fold cross change it .using classification for different data to showing in explanation of execution method selecting and set up the classify data has deen show the Fig.5,6 For (eg Gender, review)[15].

🜍 Weka Explorer										ı ×
Preprocess Classify Cluster Associate	Select attributes Vi	sualize								
Classifier										-
Choose NaiveBayes										
Test options	Classifier output									
Use training set Use t	Time taken to bu === Stratified of === Summary === Correctly Classi Incorrectly Classi Roppa statistic Mean absolute or Root mean square Rot relative ac Total Humber of === Detailed Acc	nild model pross-vali sified Inst sified In ror derror pared err Instances puracy By	: 0 secon dation == ances stances cor Class ===	603 117 0.42 0.17 61.84 94.95 720	203 79 509 131 %	83.75 16.25	8			
	Weighted Avg. Confusion Ma a b < c 538 56   a = 61 65   b =	TP Rate 0.906 0.516 0.838 atrix === classified = Male = Female	FP Rate 0.484 0.094 0.416	Precision 0.090 0.537 0.835	Recall 0.906 0.516 0.838	F-Measure 0.902 0.526 0.836	MCC 0.428 0.428 0.428	ROC Area 0.834 0.834 0.834	PRC Area 0.958 0.542 0.885	Cla Mal Fem
Status										

Fig.5: Naïve Bayes Classifier algorithm and Result for Gender

Weka Explorer									- c	1
Preprocess Classify Cluster Associat	e │ Select attributes │	Visualize								
Classifier										
Choose NaiveBayes										
est options	Classifier output									
<ul> <li>Use training set</li> </ul>	Summary -									_
Supplied test set     Set.     Set.     Cross-validation Folds 10     Percentage split % 66     More options	Correctly Cla Incorrectly C. Kappa statist Mean absolute Root mean squ Relative abso Root relative Total Number	ssified Inst lassified In ic error ared error lute error squared err of Instances	ances stances or	457 263 0.46 0.21 0.33 61.66 80.37 720	79 39 46 14 % 2 %	63.4722 36.5278	*			
Nom) Review	Detailed a	Accuracy By	Class							
Start Stop		TP Pate	FD Pate	Presiden	Recell	F-Measure	MCC	BOC Area	PPC Area	<b>C1</b> -
esult list (right-click for options) 18:36:43 - bayes.NalveBayes 18:41:16 - bayes.NalveBayes	Weighted Avg.	0.572 0.670 0.570 0.857 0.635	0.218 0.268 0.048 0.020 0.188	0.572 0.613 0.733 0.806 0.639	0.572 0.670 0.570 0.857 0.635	0.572 0.640 0.642 0.831 0.634	0.354 0.397 0.578 0.814 0.453	0.817 0.811 0.926 0.984 0.850	0.663 0.762 0.763 0.781 0.730	Qua Pri Oth
	Confusion	Matrix ===								
	a b c 139 89 15 66 187 13 35 23 77 3 6 0	d < cl 0   a = 13   b = 0   c = 54   d =	assified Services Quality Price Others	as						
										7 F
tatus										

Fig.6: Naïve Bayes Classifier algorithm and Result for Review

# 5. Result Analysis and Discussion

In this paper we used consumer data set like Consumer Name, Age, Gender, City, Categories to low level(income10000), middle level (income50000), high level( income100000 above), Categories to different budget levels Describes the price level, Categories to different brand like the consumer, What the items purchased consumer, Payment mode on cash, card, net bank, or other payment mode Consumer like which super market ,Based on consumer satisfaction based on review in mainly two data is classify and compare Naïve Bayes and Bayean Network discuss in details [16].

## 5.1 Method for Evaluation

The conduction is applied in the method of evaluation:

- 1. ConfusionMatrix: it is the used to analysis how predict to your data classifier can recognize attribute of different attribute class member. In this matrix an M\*M, where M is the number of attribute class memberbeing calculated. For the issue in our hand, we have M equal to 2 and M equal to 4, and hence we get a 2\*2 and 4\*4 matrix.
- 2. Kappa: Measurements the connection link together categorizes in the data elements and true attribute. In the value intermediate of (ZERO,ONE). The mean value of ONE is correct value and ZERO means dynamic value.
- 3. True Positive: in this measurement that give the correct classification attribute.
- 4. False Positive: in this detail data of inaccurate instances labels as correct instances.
- 5. Recall Measurement: Measurement of recall in the rate of all relevant datasets provided by the classifier. Highest –level of recall is a model of a data set that provides relevant data.
- 6. Precision Calculation: Measurement of the precision calculation exacting the applicable knowledge data is gathering. High level precision model return the more applicable attribute and inapplicable attribute [17].

Name of algorithms	Used for classified data	Correctly c	lassified data	Incorrectly da	classified ta
		Number of data	Percentage (%)	Numbers of data	Percentage (%)
Bayes Net	Gender	615	85.4167	105	14.5833
Naïve Bayes	Gender	603	83.75	117	16.25
Bayes Net	Review	512	71.1111	208	28.8889
Naïve Bayes	Review	457	63.4722	263	36.5278

Table.6: comparison of Accuracy for Bayes classification Algorithms

# **5.2.** Considerations of study

The consumer dataset categorized using classification techniques. It is implemented that Bayes Networks appear in the sequence of right categorized occurrence in the instances based on gender 615 and based on review 512 of with the accuracy of percentage 85.45% and 71.11%. The number of incorrectly classified instances based on gender 105 and based on review 208 that is 14.58% and 28.89%. Naïve Bayesian classified indicate the number of right class instances based on gender 603 and based on review 457 with the accuracy of

gender 83.75%, and based on review accuracy 63.47% and the number of incorrectly classified instances based on gender 117 and based on review263 of that is incorrect percentage based on gander and review16.25%, 36.53% .Table one indicate in correct the different BayesianClassification .Table two display the last calculation like False Positive Rate calculation, True Positive Rate measurements , calculation ofPrecision, measurement ofRecall, calculation of F-Measure, measurement of MCC area, measurementROC Area of all the Bayes Net algorithm and the Naïve Bayes algorithm in the terms of two classes of based on gender is male and female, based on reviewservices, quality, price, others. Table three indicates the Comparison based on gender and review of calculation in the Weighted Average different BayesClassification. Table four calculations given the Confusion Matrix based on gender and reviewpair of the Bayesian Network algorithm and the Naïve Bayes algorithm. [18].

÷											
	Name of algorithms	Used for classified data	TP Rate	FP Rate	Precision	Recall	F- Measure	MCC	ROC	PCR	CLASSES
		Gender	0.901	0.365	0.921	0.901	0.911	0.516	0.909	0.981	Male
	<u>Bayes</u> Net algorithms		0.635	0.099	0.576	0.635	0.604	0.516	0.909	0.689	Female
		Review	0.634	0.178	0.644	0.634	0.639	0.457	0.852	0.711	Services
			0.713	0.211	0.682	0.713	0.697	0.498	0.862	0.809	Quality
			0.711	0.039	0.807	0.711	0.756	0.706	0.959	0.865	Price
			1.000	0.011	0.900	1.000	0.947	0.944	1.000	0.998	Others
		Gender	0.906	0.484	0.898	0.906	0.428	0.428	0.834	0.958	Male
			0.516	0.094	0.537	0.516	0.428	0.428	0.834	0.542	Female
	Naive		0.572	0.218	0.572	0.572	0.572	0.354	0.817	0.633	Services
	Bayes		0.670	0.268	0.613	0.670	0.670	0.397	0.811	0.762	Quality
	argorithms	Review	0.570	0.048	0.733	0.570	0.570	0.578	0.926	0.763	Price
			0.857	0.020	0.806	0.857	0.857	0.814	0.984	0.781	Others

Table.7: Bayes classification AlgorithmsFinal Statistics

Different	Name of classification algorithms							
Measurements	Bayes Ne	t algorithms	Naïve Baye	<u>s</u> algorithms				
	Gender	Review	Gender	Review				
TP Rate	0.854	0.711	0.838	0.635				
FP Rate	0.319	0.150	0.416	0.188				
Precision	0.860	0.712	0.835	0.639				
Recall	0.854	0.711	0.838	0.635				
F-Measure	0.857	0.710	0.836	0.634				
MCC	0.516	0.562	0.428	0.453				
ROC	0.909	0.889	0.834	0.850				
PCR	0.929	0.803	0.885	0.730				

Table.8: Comparison of Weighted Average.

Name of classification algorithms	Ge	ender	attribute	Classified data
	a	b	variables	
Bayes Net algorithms	535	59	а	Male
	46	80	b	Female
Naïve <u>Bayes</u> algorithms	a	b	variables	Classified data
	538	58	а	Male
	61	65	b	Female

**Table .9: Confusion Matrix for Gender** 

Name of classification algorithms	Gender attribute					Classified data
	а	ь	c	d	variables	
	154	73	9	7	а	Services
	66	119	14	0	ь	Quality
Bayes Net algorithms	19	20	96	0	c	Price
	0	0	0	63	d	Others
	а	ь	c	d	variables	Classified data
			_			Classifieu uata
	139	89	15	0	a	Services
	139 66	89 187	15 13	0	a	Services Quality
Naïve <u>Bayes</u> algorithms	139 66 35	89 187 23	15 13 77	0 13 0	a b c	Services Quality Price



**Table.10: Confusion Matrix for Review** 







6. Conclusion

Classification approach is data mining approach it is supervised learning algorithm. It identified the categorized data, so that the classification of the pre-determined classes. Data classification problem is one of the most important researches in data mining , in which the classification of data of interest is the minimum classification, and has very low level sample data compared to the major classes. In this leads classifier prediction to be based on towards majority class, so solutions needs to be found out to handle this problem. Here we have evaluated solutions to class imbalance problem on consumer behaviour dataset using WEKA.In this paper compare the two classification algorithm. In this analysis to identify that Bayesian Network algorithm create the hugeamount of correct classified data compared to the Naïve Bayes algorithm.

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