



Introduction of Opinion Mining and Sentiment Analysis Research

Varma Kajal S., Raval Jyoti A., Mr. Romil Patel

Department of Information Technology, Sigma Institute of Engineering, Vadodara, Gujarat, India.

varmakajal94@gmail.com

Department of Information Technology, Sigma Institute of Engineering, Vadodara, Gujarat, India.

ravaljyoti12@gmail.com

Assistant Professor, Department of IT, Sigma Institute of Engineering, Vadodara, Gujarat, India.

romil.patel1@gmail.com

Abstract— In our project important parts are opinion mining, spam detection, company rating and data cleaning. In data cleaning we had clean the data which are unwanted and also count the how many data are removed. Spam detection is based on the data cleaning process. Spam detection is identify in data which are spam or non-spam data based on same user, same ip address, same product id and same time. Opinion mining is based on the Non-spam data. In opinion mining process we can find the positive reviews, negative reviews and neutral reviews. For opinion mining we are used naive Bayesian algorithm. All data are stored in dataset. Company rating is also based on data cleaning process. In company rating we can find the highest rated company.

Keywords—Opinion mining and sentiment analysis, web mining, system flow, naïve Bayesian algorithm, Applications, Techniques.

I. INTRODUCTION

Opinions are one type of human activity or behaviour. People can make decisions on some things or thoughts. People can share own thoughts on web. In other words we can say sentiment, attitudes, emotions, evaluations of people. It is not only for individual. It is basically natural language process. [3] People can give own opinion for some tasks. This project we will make for the online shopping or various e-commerce websites. “What other people think” has always been an important piece of information for most of us during the decision-making process. In today’s world, there are so much data available on the internet.[2] It includes the customer reviews on different products also. It is general tendency that before purchasing any product, we go through the reviews written on the website of that product. [1] By reading those reviews customer takes decision. Reviews, the company can come to know that what is lacking in their product. This gives good incentives for review/opinion spam. In this system we are implement the data in Microsoft visual studio 2010/2013. We have already database in MS-access We had done the all steps of system flow. We had implemented the system on cleaning data and non-spam data. [1]

II. LITERATURE REVIEW

In recent year web mining is favourite area for many researchers. Peoples are used many sites on web. so User can give opinion on product and other items. Users share the thoughts. User interact with server, site, and people.[2] An important part of our information-gathering behaviour has always been to find out what other people think. In that project we will done spam detection, text mining, review rating opinion mining, data mining, emotion mining on opinion of the people. Our goal is rate to the product based on review mining. Opinion mining, spam detection and company rating is this project main concept.[1] We are take the permission for dataset to author. E-commerce website is main site is our project. People can give reviews on the different products. It is based on product reviews. Our activities are finding dataset, data cleaning, data transformation, data integration, data mining, spam detection and company rating. This project is based on different types of analysis.[3] Naive Bayesian algorithm is used for opinion mining and sentiment analysis. Naive Bayesian algorithm is apply on dataset. It is compare reviews with positive words and negative words. It is split to sentence and check one by one word. It can find positive reviews and negative reviews. [1]

III. SYSTEM FLOW

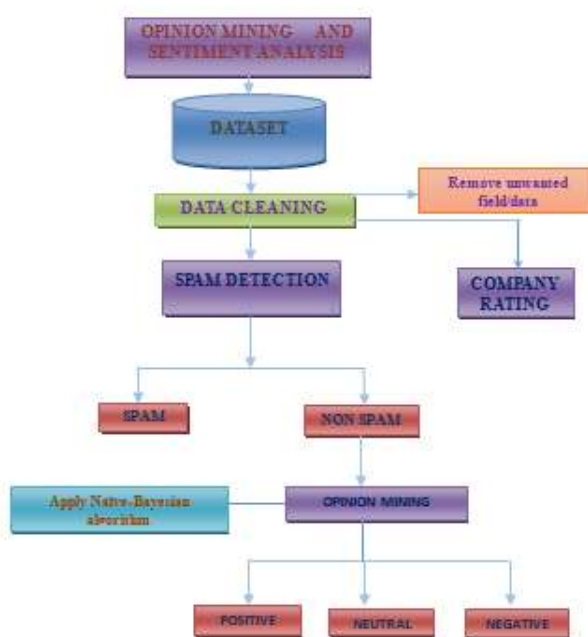


FIG.1 System work

IV. ALGORITHM

We are using Naive Bayesian algorithm for our project. Three conditions applied in opinion mining. It is classification theorem.

STEP:-1

If the probability of positive label is greater than the negative, then the sentence or opinion is positive.

$$P(X/Pos_Count)P(Pos_Count) > P(X/Neg_Count)P(Neg_Count) \text{ (Positive)}$$

STEP:-2

If the probability of negative label is greater than the positive, then the sentence or opinion is negative.

$$P(X/Neg_Count)P(Neg_Count) > P(X/Pos_Count)P(Pos_Count) \text{ (Negative)}$$

STEP:-3

If the probability of positive minus probability of negative is zero, then it is neutral.

$$P(X/Neg_Count)P(Neg_Count) - P(X/Pos_Count)P(Pos_Count) = 0 \text{ (Neutral)}$$

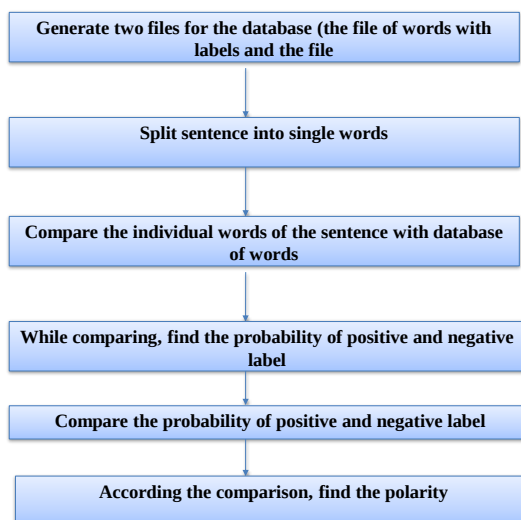


FIG.2 Steps of algorithm

V. RESULT

- 1) Here display the original dataset.

Camera										
IpAddress	Rating	UserId	Productid	Date	Reviewtime	Reviews	Price	Company	Reviewid	
134.56.78.90	5		631GFK43DF	07-Sep-13	4:26:13.14586	i wish the can	\$20.55	Samsung	R2EQFVBTBYRTUUN	
34.56.13.34	5		34YURYH4N3YR	02-Mar-13	2:10:41.1478956	no instruction	\$20.55	Samsung	R2QP9Z5F2Q4KER	
135.87.90.90	5	CKLWY487R3ESUN	635Y84URHF7	25-Mar-13	10:58:08.1472394	after a lot of n	\$20.55	Canon	R2C3HQMXLYZK30	
135.87.90.90	5	CKLWY487R3ESUN	635Y84URHF7	25-Mar-13	10:58:08.1472394	i have not ye	\$20.55	Canon	R2C3HQMXLYZK30	
135.87.90.90	5	CKLWY487R3ESUN	735843UHF7	25-Mar-13	10:58:08.1472394	and the body	\$20.55	Canon	R130Q01GD3X0U	
135.87.90.90	5	CKLWY487R3ESUN	635Y84URHF7	25-Mar-13	10:58:08.1472394	everything el	\$20.55	Canon	R2C3HQMXLYZK30	
135.87.90.90	5	CKLWY487R3ESUN	635Y84URHF7	25-Mar-13	10:58:08.1472394	these things n	\$20.55	Canon	R2C3HQMXLYZK30	
135.87.90.90	5	CKLWY487R3ESUN	635Y84URHF7	25-Mar-13	10:58:08.1472394	the moveabl	\$20.55	Canon	R2C3HQMXLYZK30	
135.87.90.90	5	CKLWY487R3ESUN	635Y84URHF7	25-Mar-13	10:58:08.1472394	i love photog	\$20.55	Nikon	R2C3HQMXLYZK30	
135.87.90.90	5	CKLWY487R3ESUN	635Y84URHF7	25-Mar-13	10:58:08.1472394	i love photog	\$20.55	Nikon	R2C3HQMXLYZK30	
135.87.90.90	3	CKLWY487R3ESUN	735843UHF7	25-Mar-13	10:58:08.1472394	i love photog	\$20.55	Samsung	R2C3HQMXLYZK30	

FIG.3 Original dataset

- 2) Here select the table of dataset.



FIG.4 Select the database

- 3) Here select the columns of which we need.

Columns

☒ IPaddress

☒ Userid

☒ Productid

☒ Rating

☒ Date

☒ Reviewtime

☒ Reviews

☐ Price

☒ Company

☒ Reviewid

FIG.5 Select Column

- 4) Display the original dataset in Grid view.

No.	Problem	Class	Productid	Rating	Date	Reviewtime	Reviews	Company	Reviewid
1	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
2	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
3	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
4	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
5	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
6	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000

FIG.6 Display Original dataset

- 5) Data cleaning process is removed the unwanted data and display in Grid view.

No.	Problem	Class	Productid	Rating	Date	Reviewtime	Reviews	Company	Reviewid
1	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
2	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
3	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
4	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000
5	1000000000	1000000000	1000000000	1	1000000000	1000000000	1000000000	1000000000	1000000000

FIG.7 Display Data cleaning data

- 6) Spam detection process is comparing some fields and if match so it is in spam data and if not so it is in non-spam data. Here display the Spam data in Grid view.
- 7) Here display the spam data in grid view and also stored in database.

Product	Brand	ProductID	Rating	Date	ReviewText	Reviewer	ReviewerID
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	after a lot of research... i decided on the gl	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee

FIG.8 Display Spam data

- 8) Here display the non-spam data and also stored in database.

Product	Brand	ProductID	Rating	Date	ReviewText	Reviewer	ReviewerID
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	after a lot of research... i decided on the gl	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee

FIG.9 Display Non-spam data

- 9) Now start opinion mining process and it's based on the non-spam data. And it's also based on positive-negative word directory. All data stored in direct database. Below figure in display the data of positive reviews and stored in direct data table.

Product	Brand	ProductID	Rating	Date	ReviewText	Reviewer	ReviewerID
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	after a lot of research... i decided on the gl	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee
126.76.81.80	CHLNTW4700000000	64750430007	2	12-03-2011	10-03-06-1472000	i have not use any of the usual products with digital picture quality... it's not as good as the other	Casee

FIG.10 Display and stored Positive reviews data

- 10) Here stored the Negative reviews in data table.

Review ID	Review Text	Rating	Date
101-101-101	the manual is very hard to understand... and it is really difficult to use...	1	2016-03-10
102-102-102	the camera is very blurry... and it is really difficult to use...	1	2016-03-11
103-103-103	the battery is not good... and it is really difficult to use...	1	2016-03-12
104-104-104	the camera is very blurry... and it is really difficult to use...	1	2016-03-13
105-105-105	the battery is not good... and it is really difficult to use...	1	2016-03-14

FIG.11 Display and stored Negative reviews data

11) Here stored the Neutral reviews data in data table.

Review ID	Review Text	Rating	Date
106-106-106	the camera is good... and it is really difficult to use...	2	2016-03-15
107-107-107	the battery is good... and it is really difficult to use...	2	2016-03-16
108-108-108	the manual is good... and it is really difficult to use...	2	2016-03-17
109-109-109	the camera is good... and it is really difficult to use...	2	2016-03-18
110-110-110	the battery is good... and it is really difficult to use...	2	2016-03-19

FIG.12 Display and stored Neutral reviews data

VI. GRAPH

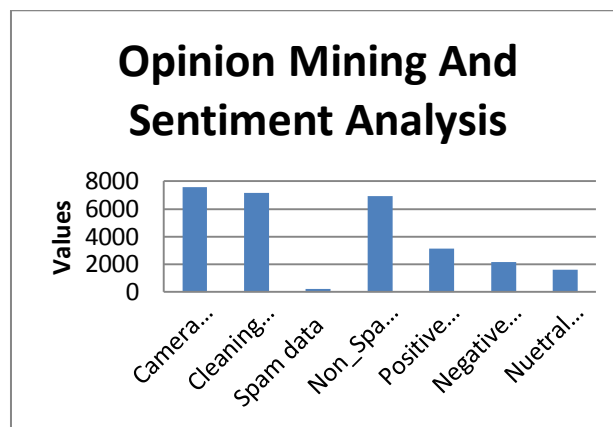


FIG.13 System chart

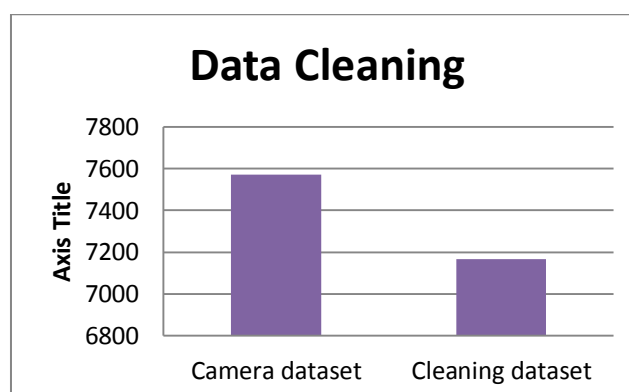


FIG.14 Data cleaning

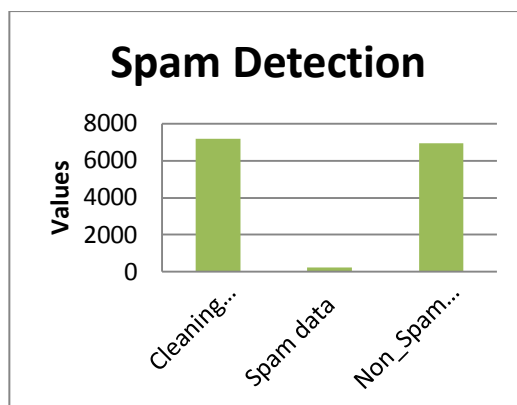


FIG.15 Spam detection

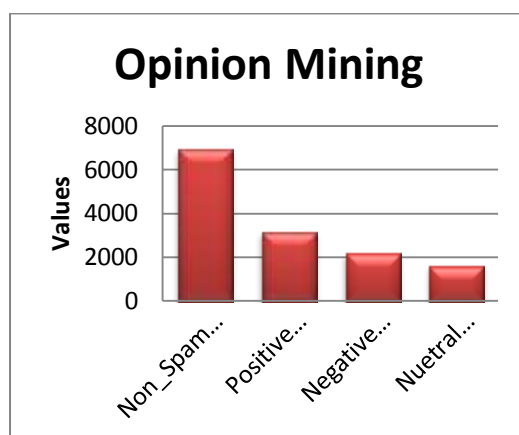


FIG.16 Opinion mining

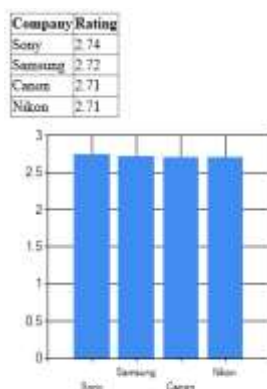


FIG.17 Company rating

VII. CONCLUSION

In this paper we had used naive Bayesian algorithm. We had done three activities opinion mining, spam detection and company rating. We will improve the system on large dataset.

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