

# E-Commerce Review Web-based Application Using Text Mining

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*Abstract:* Recent year modern e-commerce is growing very rapidly. E-commerce websites deals with the online shopping and it's all about internet marketing buying and selling product. Retailers encouraging customers to write reviews about products to express their opinions on various aspects. Customer's common seek quality information from online reviews for shortcut their research and make decisions faster and with greater confidence ever before. In this project, by applying product aspect ranking to two real-world applications, first is document-level sentiment classification and second is extractive review summarize and achieve performance improvement. The aim of document-level sentiment classification is determine positive or negative opinion from consumer reviews and aim to extractive review summarize is to summarize consumer reviews by selecting informative review sentences.

*IndexTerms* – Text-Mining, Sentiment Analysis, Opinion words.

## I. INTRODUCTION

With the rapid expansion of ecommerce, more products are sold and more people are buying products on the Web. In order to grow customer choice of satisfaction and their shopping experiences, it has become a common practice for online salesman to validate their customers to review or to express opinions on the products that they buy. With more common users becoming comfortable with the Internet, literally increasing survey number of people are writing reviews. As a consequence, the number of reviews that a product receives, that particular product grows rapidly. Some popular products can get hundreds of reviews at some large salesmanship sites. This makes it very hard for a customer to read reviews and this helps him to make a decision on whether to buy the product.

It is a common practice that salesman selling products on the Web ask their customers to write review on the products and similar services. For a popular product, the number of reviews can be in thousands or hundreds. This makes it hard for a potential customer to read them in order to make a decision on whether to buy the product. In this paper, it is aimed to summarize all the customer reviews on a product. This summarization task is completely different from normal text summarization because it is focused are on specific features of the product that customers have opinions on and also whether the opinions are positive or negative. By selecting or rewriting the grouped reviews of shared subset it will not be summarized on the original sentences from the reviews to capture their main points from traditional text summarization.

In this paper, we only focus on opinion word target and mining opinion product features that the reviewers have commented on. A number of techniques are recommended to mime such features. Our Experimental results show that these techniques are highly cooperative.

## II. EXISTING SYSTEM

In previous methods, surfing the opinion inter-relations between defined opinion targets and defined opinion words was the key to collective extraction. To this end, the most recognized techniques is the nearest-neighbour rules and syntactic patterns. Nearest neighbor rules encourages the nearest adjective/verb to a noun/noun phrase in a limited window as its modifier. Syntactic information, are related to designated formation in which the opinion relations among words are decided according to their dependency relations in the parsing tree.

### 2.1 DRAWBACKS

1. Cannot obtain precise results.
2. Syntactic patterns are prone to errors. Usually Online reviews have informal writing styles such like grammatical errors, typographical errors, and punctuation errors. This makes the existing parsing tools, which are usually performed on formal texts such as news reports, prone to generating errors.
3. Time consuming because of manual process
4. Less efficient.
5. Opinion of sentiment word would be content sensitive.

## III. PROPOSED SYSTEM

We propose a product aspect ranking framework to automatically filter the important aspect of products from lots of consumer reviews. Developing probabilistic aspect ranking algorithm to clear the importance of many more aspect. By simultaneously exploiting aspect frequency and the impact of consumer opinions on the product.

### 3.1 ADVANTAGES

1. Precise results can be obtained.
2. Errors free.
3. Less Time consuming because of computerized process
4. More efficient.

### IV. MODULE DESCRIPTION

This paper contains following modules as represented in figure 4.1

#### I.Admin Module

1. Admin Login
2. Admin Dashboard
3. Add product/Delete product
4. Review Opinion words/Targets
5. Logout

#### II.User Module

1. User Registration/Login
2. View Products and Reviews
3. Buy Products
4. Logout

### I.ADMIN MODULE

#### 1. Admin Login

Admin must login using specified username and passwords which is already given.

#### 2. Admin Dashboard

Admin Dashboard displays the number of products and number of reviews given by the user and the classification is suggested by end user. In this module admin can view the number of reviews added to the product. Admin can also view the product classification whether product falls into positive Opinion words or negative Opinion words.

#### 3. Review Opinion Target

In this module admin can view the number of reviews, number of Positive Opinion, number of Negative Opinion added to the product. Opinion Targets are classified based on rule

**Positive Opinion Targets= (No. of Positive Opinion) / (Total no.of reviews)**

**Negative Opinion Targets= (No. of Negative Opinion) / (Total no.of reviews)**

#### PRODUCT SUGGESTION

**If Positive Opinion > Negative Opinion, the product is classified as Positive Opinion Target, This Product can be suggested to user. These product get higher rating and gets listed first in product listing.**

**If Negative Opinion > Positive Opinion, the product is classified as Negative Opinion Target, This Product cannot be suggested to user. These product get lower rating and gets listed last in product listing.**

#### 4. Add Product/Delete Product

New arrivals of products can be updated or added on web with respective descriptions. Product which is already added on web can be deleted if it is unavailable.

#### 5. Logout

After view Opinion Target reviews of the product, admin can now logout from the session.

### 11. USER MODULE

#### 1. User Registration

The System requests for user name and password. User must enter the correct values and submit. Now the system checks for user name validity sub code format and existence, then it will store the user's information. The acknowledgement will be received from the system to the user.

#### 2. User Login

User need to login into the system If user not yet having username and password to login, they need to register by giving their basic details and get username and password.

#### 3. View Products and Review

1) After logging into the system, user can now view the shopping products online. Now user can add reviews to the product. There are two types of Opinion words dataset classified, one is positive Opinion words and another one is negative Opinion words.

2) User reviews are compared with the Opinion words dataset and classified whether it is Positive Opinion or negative Opinion. User can review the individual product only once, if he review same product twice a 'Review already added to this product' error alert will be shown to user.

#### 4. Buy Products

User can buy products and can give reviews to that particular products.

#### 5. Logout

After adding reviews to the product, user can now logout from the session.

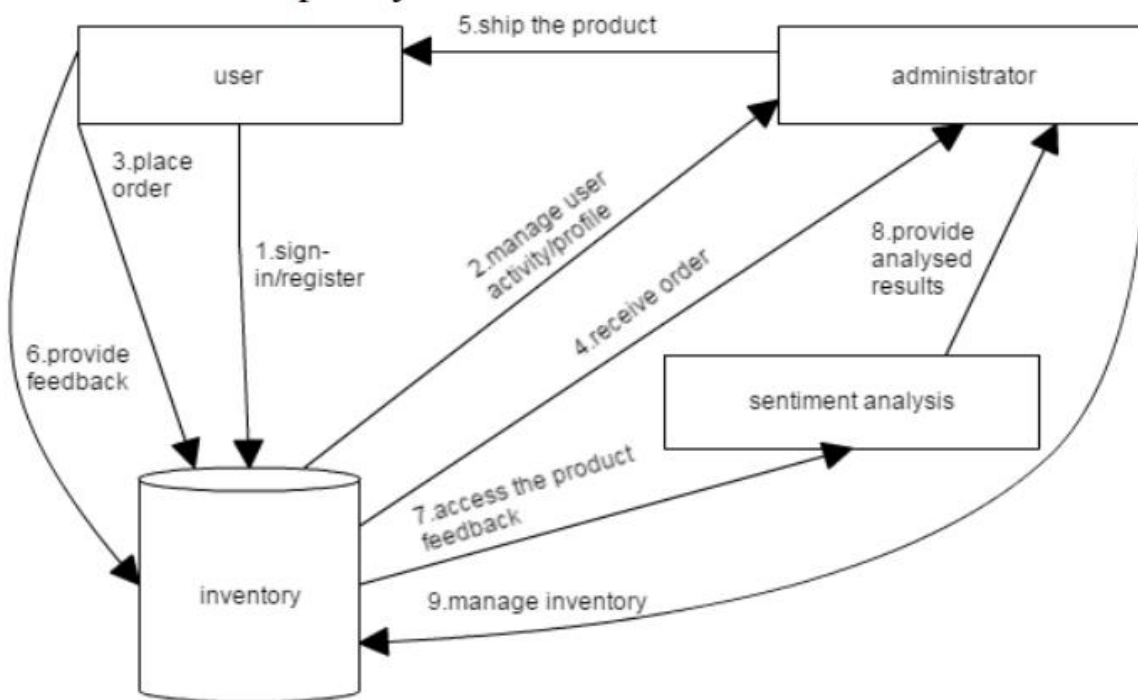
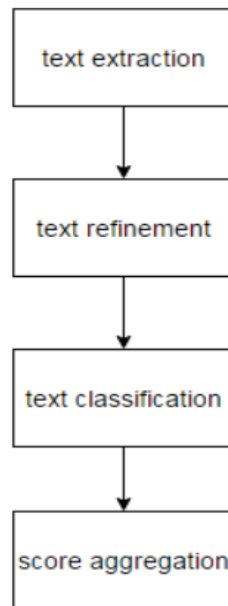


Fig 4.1 SYSTEM ARCHITECTURE OF WEB-BASED E-COMMERCE

## V. SENTIMENT ANALYSIS

Sentiment analysis refers to the process of mining the texts in order to identify the tone of the review written by the user. Positive, negative and neutral are passage of tones written by user. Sentiment analysis is actually used for the purpose of humans as we read the passage we can easily identify the tone of the passage using the context and the words used but for computers it is totally a different scenario to sort out, they mainly use the contexts to identify the sentiments of the passage. Here comes the complicated words, words with opposite meanings such as good and bad are actually used in the same context but they represent different sentiments, this can't be recognized by the computers. Consequently this method is opted to the utilization of polarity scores for certain words in the passage in order to identify the sentiment of the passage. Hence the utilization of sentiment analysis in e-commerce websites in order to analyse the customer's feedback on the products, customer's feedback and their ratings are encountered by manual reading which proves to be the best alternative to the process. Details will be provided by user to administrator where the overall score is calculated for each product from the feedback given by user. This will constitute in detail the amount of positive score and negative score for the product clearly to the administrator, based on which the important and regular decisions on managing the record directory can be made by the administrator.

This module extracts the feedback from the record directory and performs sentiment analysis on it. It extracts the regular important sentiment words from the feedbacks evaluated the score and finally aggregates and parses the scores of all feedbacks of the product to provide its overall score. It basically splits the words into positive and negative space embedding. Here positive words will be scored positive and negative words will be scored negative. Now a product quality is determined by the overall sentiment feedback conversion. If the product has positive feedback then the positive reviews outweigh and rating hits for 5 stars the negative reviews, consequently the product is considered to be good. If the product has negative feedback then the product is bad in quality or in its shipping. Other than these the product can also have a neutral score i.e., positive and the negative reviews are equal then product is considered to be a neutral form.

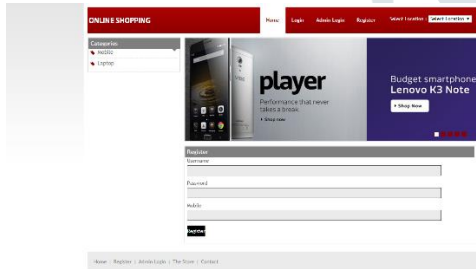


1. **TEXT EXTRACTION:** This is the first step in which the sentiment analysis module extracts the words from the feedback that could potentially affect the outcome.
2. **TEXT REFINEMENT:** This step involves the refinement of the extracted words from irrelevant words and phrases.
3. **TEXT CLASSIFICATION:** Here the extracted and the refined text are classified into positive and negative classes based upon their polarity.
4. **SCORE AGGREGATION:** This step computes the total score from the classifier and then aggregates it to produce the final sentiment score. Finally the positive and the negative scores of the product are shown in the form of bar graph to the user.

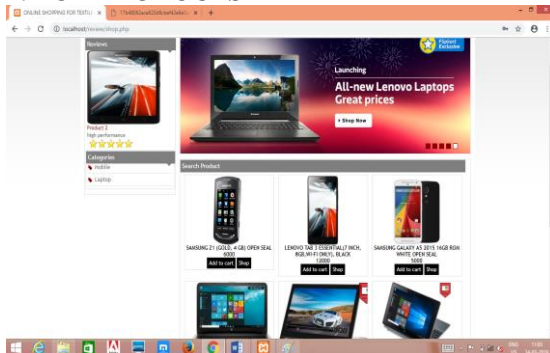
## VI. SCREENSHOTS

### I. USER

#### 1. USER REGISTRATION



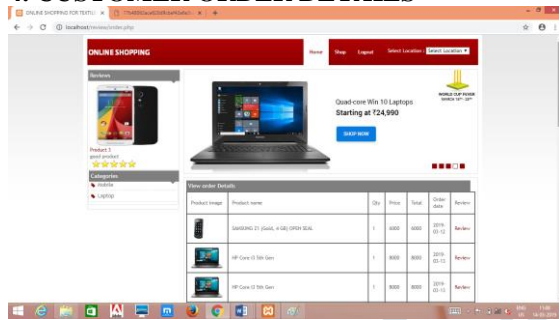
#### 2. BUY PRODUCTS



### 3. REVIEW PAGE

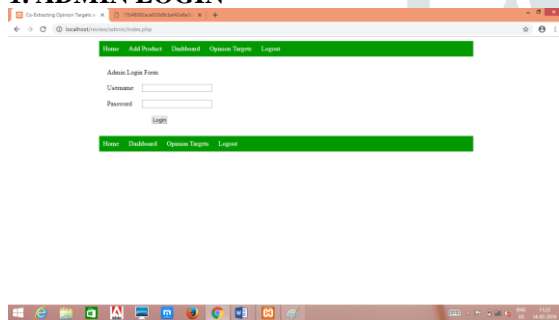


### 4. CUSTOMER ORDER DETAILS

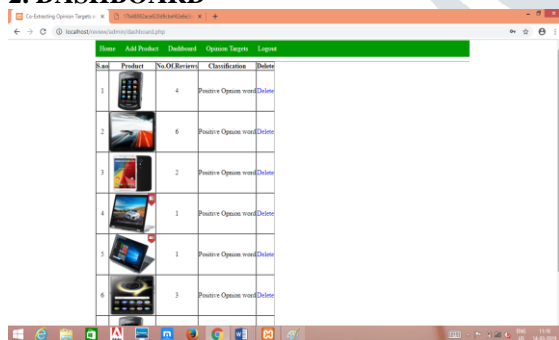


## II. ADMIN

### 1. ADMIN LOGIN



### 2. DASHBOARD



### 3. OPINION TARGETS



#### 4. ADD PRODUCT

#### VII. CONCLUSION

A marketer's quality in E-Commerce websites completely depends on the reviews given by the user who have purchased and used it. Since the urge star ratings aren't entirely confirmative, by using the concept of Text mining side to side with Sentiment analysis brings out the required information for a user who desires to evolve others opinion on that product. Sentiment data of texts in conjunction with its actual context is encoded in sentiment embedding to support sentiment analysis algorithm. As a result the written feedback of users is taken into the work out the particular quality of the marketer's from which it will be either promoted to next or to be removed by the administrator.

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