

OPINION MINING AND SENTIMENT ANALYSIS FOR ONLINE REVIEW

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Abstract - In today's world, online shopping has a tremendous growth. According to human nature, people buy products based on the reviews. When bad reviews are made by fake people, the rating of the product gradually falls. To overcome this, our project aims to generate a random key when the purchase is made. Only with the key provided, the person intended to make a review can enter into the review system. Once the review is made, the good words and the bad words are differentiated using clustering algorithm called sentimental analysis. Based on this differentiation, an overall rating is made as well as a graph is generated which indicates the quality of the product.

KeyWords: review, rating, sentiment analysis, product, clustering

1. INTRODUCTION

The important source of information in our day-to-day life constitute the opinions and experiences of other people. For example, we get suggestion from our friends which hospital, hotel or laptop they would recommend to us. Nowadays, online user reviews plays a major role to answer such questions. Online reviews provide a great value to vendors as they represent unsolicited and genuine customer feedback that is conveniently available at virtually no costs. However, for products in major demand there often exist several thousands of reviews so that manual analysis is not an option. In this paper, we provide a comprehensive study of how to model and automatically analyze the opinion-rich information contained in customer reviews. In specific, we consider the task of aspect oriented sentiment analysis. Given a collection of review texts, the task's goal is to detect the individual product aspects reviewers have commented on and to decide whether the comments are rather positive or negative. Text analysis systems development involves the tedious and costly work of creating appropriate resources — for instance, labelling training corpora for machine learning methods or constructing special-purpose knowledge bases. As an overarching topic of the thesis, we examine the utility of distant supervision techniques to reduce the amount of human supervision.

1.1 LITERATURE SURVEY:

1. Deep Learning for Aspect-Based Sentiment. This paper was submitted by Bo Wang, Min Liu. In 2014. Sentiment analysis plays a vital role in natural language understanding and has a wide range of real-world applications. The typical

sentiment analysis focuses on predicting both the sides of polarity of the given sentence(s). This task works in the specifying that the given text has only one aspect and polarity. A more general and complicated task would be to predict the aspects mentioned in a sentence and the sentiments associated with each one of them. This generalized task is known as aspect-based sentiment analysis (ABSA). In the annual SemEval competition, an ABSA task has been included since 2014. Among submissions of the past two years, most of the winning models use support vector machines (SVM). In the recent trends of deep learning, this work applies deep neural nets to solve this task. We design a model with aspect prediction and sentiment prediction. For both positive and negative predictions, we achieve better than or close to state-of-the-art performance using deep learning models. We propose a new method to combine the syntactic structure and convolutional neural nets which directly match aspects and corresponding polarities.

2. Approaches, Tools and Applications for Sentiment Analysis Implementation. This paper was submitted by Alessia D'Andrea in 2015. The paper gives an complete overview of the different sentiment classification approaches and tools used for sentiment analysis. Starting from this overview the paper provides a classification of (i) approaches with respect to features/techniques and advantages/limitations and (ii) tools with respect to the different techniques used for sentiment analysis. Different application fields of sentiment analysis are: business, politic, public actions and finance are also discussed in the paper.

3. Distribution-Based Cluster Structure Selection. This paper was submitted by Ankit Singh in 2016. The aim of cluster structure ensemble is to find a unified cluster structure from multiple cluster structures obtained from different datasets. Unfortunately, not all the cluster structures contribute to the monofied cluster structure. This paper investigates the problem of how to select the appropriate cluster structures in the ensemble which will be summarized to a representative cluster structure. Specifically, the cluster structure is first represented by a mixture of Gaussian distributions, the parameters which are estimated using the expectation-maximization algorithm. Finally, several distribution-based distance functions are designed to evaluate the similarity between two cluster structures.

D. Ensemble clustering selection by optimization of accuracy-diversity trade off. The main aim of clustering is to cluster objects which have common features within the group but

have dissimilarities with other groups. Clustering algorithms involve finding a common structure without using any label, similarly like other unsupervised methods. Recent studies show that the decision of the ensemble clusters gives more accurate results than any other single clustering solution. Besides that, the accuracy and diversity of the ensemble are one of the important factors which effect the overall success of the algorithm. There is a trade off between accuracy and diversity, in other words, you sacrifice one while you increase the performance of the other. On the other hand, the optimum number of clustering solutions is one of the parameters that effect the final result. Recently, finding the best subset of the ensemble clustering solutions by eliminating the redundant solutions has become one of the most challenging problems in the literature. The proposed study here aims to find a best model which optimizes the accuracy and diversity trade off by selecting the best subset of cluster ensemble.

2. PROPOSED SYSTEM

In proposed, we propose a simple yet efficient model, called dual sentiment analysis (DSA), to address the polarity shift problem in sentiment classification. We first propose a data expansion technique by creating sentiment reversed reviews. The original and reversed online reviews are constructed in a one-to-one correspondence. Polarity shift is a kind of linguistic phenomenon which will reverse the sentiment polarity of the text. Negation is the important type of polarity shift. For example, by adding a negation word "wouldnot" to a positive text "I would do this work" in front of the word "do", the sentiment of the text will be reversed from positive to negative.

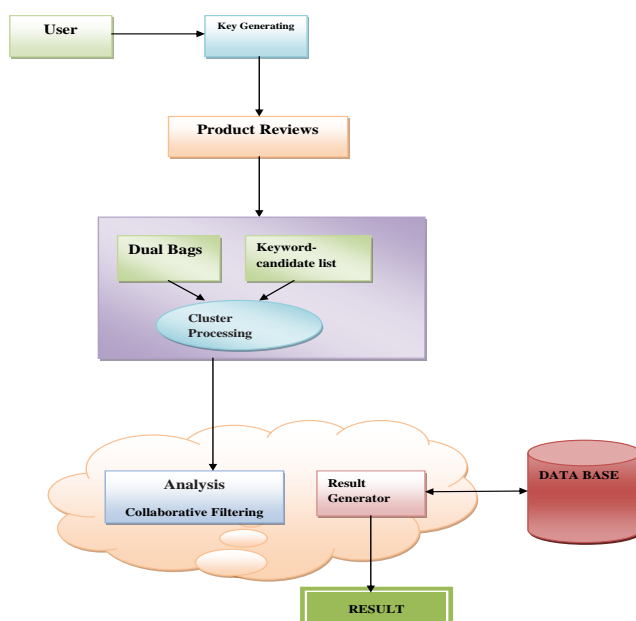


Figure 1: system architecture

The proposed system consists of the following parts:

- 1) User interface Design
- 2) Admin maintain the products
- 3) User transaction
- 4) Key generate and Review sharing
- 5) Dual Sentiment Analysis
- 6) Performance Evaluation

User interface:

This is the first module of our project. The important role for the user is to move login window to user window. This module has created for the security purpose. In this login page we have to enter login credentials. It checks whether the username and password matches or not (valid user id and valid password). If we enter any invalid username or password we can't enter into login window to user window it will shows error message. So we are preventing unauthorized user entering into the login window to user window. It provides good security for our project. So server contains user id and password. Server also checks the authentication of the user. It improves the security and prevents unauthorized user entering into the network. In our project we are using JSP for designing. Here we validate the user who has logged in and authentication of the server.

Admin maintain the products:

This is the second module of our project. The important role for the Product owners is to move login window to Product owner window. This module has created for the security purpose. In this page we have to enter login user id and password. It checks the credentials for correct match (valid user id and valid password). If we enter any invalid username or password we can't enter into login window to Product owner window it will shows error message. So we are preventing from unauthorized product owner entering into the login window to product owner window. It provides good security for our project. So server contains user id and password server also check the authentication of the user. Here Product Owner is Updated their Products. It well improves the security and preventing from unauthorized data owner enters into the network. In our project we are using JSP for designing. Here we validate the user who has logged in and authentication of the server.

User transaction:

This is the third module in our project, here symbolizes a unit of work performed within a database management system (or similar system) against a database, and treated in a coherent and reliable way independent of other transactions of the user. A transaction generally represents any change in database. User will transfer the amount to provider.

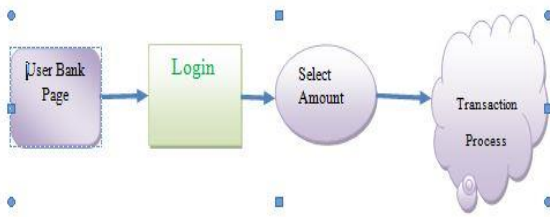


Figure 2: User transaction

Key generate and Review sharing:

This module is used to help the Buyer to share their Opinions and check their content is in safe also providing protection. A random-number generator (RNG) is a computational or physical device designed to generate a unique sequence of numbers or symbols that cannot be reasonably predicted better than by a random chance. Key Generation is the process of generating keys to our files. That key will have to be unique for every member while at the time of receives.

Dual Sentiment Analysis:

In this module all of the original training samples are reversed to their opposites. These are referred as “original training set” and “reversed training set” respectively. In our data expansion technique, there is a one-to-one correspondence between the original and reversed reviews. The classifier is trained by maximizing a combination of the likelihoods of the original and reversed training samples which is called dual training (DT).

Performance Evaluation:

In this project what we are going to perform means, the evaluation of the entire datasets, keywords, tables, requests, requestors and how much we are going to perform efficiently without getting any late to the end user or requestor. This module also performs the functionalities like user requested queries and what the information getting to them and what the cluster we need to join to the table to the efficient handling of queries and performing operations like analysis, evaluation of the application.

3. CONCLUSION

In this paper we have presented two methods for detecting aspect categories that is useful for online review this article has been accepted for inclusion in a future issue of this journal. The first, unsupervised method uses spreading activation over a graph built from word co-occurrence data, enabling the use of both direct and indirect relations between words. This results in every word having an activation value for each category that represents how likely

it is to imply that category. While other approaches need labelled training data to operate, this method works unsupervised. The major drawback of this method is that a few parameters need to be set beforehand, and especially the category firing thresholds need to be carefully set to gain a good performance. We have given heuristics on how these parameters can be set.

4. FUTURE ENHANCEMENT

In the future, we can generalize the DPA(Double propagation algorithm) to a wider range of sentiment analysis tasks. We also plan to consider more complex polarity shift patterns such as transitional, subjunctive and sentiment-inconsistent sentences in creating reversed reviews.

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