## Plagiarism Scan Report

Summary	
Report Genrated Date	27 Oct, 2017
Plagiarism Status	100% Unique
Total Words	623
Total Characters	4061
Any Ignore Url Used	

## **Content Checked For Plagiarism:**

Last part o $\Box$  data preprocessing help to provide deviation  $\Box$ rom predicted rating with the extraction o $\Box$  adjectives  $\Box$ rom the review which are scored on a scale o $\Box$  1 to 5. An overall expected rating associated with the review is calculated. This calculated rating is then compared to the actual rating. The weight o $\Box$  the review is adjusted on the basis o $\Box$  the deviation o $\Box$  the calculated rating  $\Box$ rom the actual rating.

The preprocessed data  $\$ rom data preprocessor is  $\$ ded as input to the opinion parser, which is the most important component o $\$ system provides output which contain recommendations  $\$ or users and overall product sentiment ( $\$ or business). Opinion parsing includes bootstrapping process  $\$ or aspect extraction. Bootstrapping is an opinion words and opinion target (aspect) extraction process. In this process a set o $\$ opinion words like "good", "bad", "amazing", called as Opinion lexicon, is given as an input to the bootstrapper. This Opinion lexicon is used by the bootstrapper to identi $\$ y opinion words  $\$ rom the reviews.

## Fig 3:Opinion Parser

It then extracts corresponding aspects and  $\square$ orms pairs. Known Opinion Lexicon and extracted opinion words and target (aspect) are then used together to  $\square$ urther extract opinion words and targets. Subtasks included in this process are extracting targets using opinion words, extracted targets and extracting opinion words using extracted targets and using both the given and the extracted opinion words. This process goes on till no opinion words and target are  $\square$ 0 to be extracted.

Fig 4: Bootstrapping Using Double Propagation

Other part o $\square$  opinion parser is sentiment analysis, a block o $\square$  opinion parser assigns sentiment weights to the aspects. It uses Sentiment Lexicon to identi $\square$ y sentiment weights to the aspects. It uses Sentiment Lexicon to identi $\square$ y sentiment intensity associated with the particular opinion word. Sentiment lexicon is a database o $\square$  lexical units o $\square$  a language along with their sentiment orientations. Strengths o $\square$  these sentiments are averaged to get the overall sentiment  $\square$  or the given aspect

## V. Conclusion

In particular, we are aiming to measure how much the ratings have changed <code>or</code> those users who "<code>ollow</code>" our recommendations on the test set by mentioning the recommended aspect in the review. In addition to the average rating on the test set, we can compare our results with three strong baseline approaches. These baselines basically indicate the strengths and weaknesses oo the establishments based on their user reviews: Trending Similarly, those users who will not <code>ollow</code> our negative recommendations (and experience the negative aspect oo an item against our advice) might give higher/lower ratings to the items than the average rating oo the items given by all users in the application and those users who did not <code>ollow</code> the recommendations provided with the baseline approach. The results are likely to show that managers who "<code>ollowed</code>" our positive recommendations, obtained higher or lower ratings <code>or</code> the user experiences than the managers who <code>ollowed</code> recommendations provided with the baseline approaches. The results shall hold not only <code>or</code> the restaurants but also across hotels and beauty domains.

The state-o[]-the-art approach to enhance the []unctionality o[] recommender systems by recommending not only the product itsel[] but also some positive aspects o[] the product to []urther enhance user experiences with the it. Recommendations o[] a set o[] valuable aspects are expected to work well as those users who []ollow our recommendations must rate their experiences signi[]icantly higher than those who []ollow the baseline recommendations.

The way in which users on the system behave is nundamental in judging the pernormance on the system. Considering the user proniles and product proniles or recommendation would really assist retaining the aspects on the user-product pairs and being highly relevant in making recommendations and provide even better insights to businesses with the help on product-user association.

Report generated by smallseotools.com