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Traditional Collaborative Filtering User Preferences in Big Data with Mapreduce

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Abstract: Recommendation system acts as a tool in providing most appropriate service to the user. Currently, information through online services increases. This leads to the overhead of data in online and there is a possibility of getting less accurate results. In previous approaches, recommendation of service is based on the feedbacks and ranking from the previous user. It doesn't consider the suggestion of the user at a time, who in need of searching for the particular service. The proposed system deals with the implementation of personalized recommendation to provide services for hotel reservation system. Preferences are collected from the active user about particular service for each application. Similar user's opinions are taken from the reviews using keyword extraction method and Supervised learning algorithms are used to identify sentiment orientation. It determines positive or negative opinion along with negation word near to each opinion word and then identifies the number of positive and negative opinions of reviews. Keywords with positive opinion are considered and similarity is calculated between user preferences with reviews of the previous user by jaccord and cosine measures. From this most similar keywords are provided to the user as recommended service. To provide more accurate prediction of the services needed by the active user the proposed system is implemented using MapReduce framework.

Keywords—*Preferences, Recommender system, Hadoop, MapReduce*

I. INTRODUCTION

In internet, amount of data increases day by day which leads to difficult analysis by data mining techniques. The sources of data can be a database, data warehouse, the web, other information repositories or data which are retrieved and stored in the system dynamically [1]. It causes inefficient and scalability problem. But when dataset are humongous in size, a wide distribution of data is needed and complexity arises which leads to the development of parallel and distributed data-intensive mining algorithms [2]. Big Data Analytics is the process of computing such large dataset in parallel using MapReduce environment [3].

A. Opinion Mining

Opinion Mining also refers to sentiment analysis is the process of analyzing the text in the document and provides the suggestions to the people by extracting opinion through online [4]. Users post their opinion about the services or products in the blogs, shopping sites, or review site Reviews about hotel, automobiles, movies, restaurants are available on the websites [18], [19], [20], [21]respectively.Text analysis in opinion mining is the process of getting high quality information from the text. Approximately, 90% of the world's data is available in unstructured format.By parsing this unstructured data, the patterns involved in it are identified and recommendations are provided.

B. Recommendation and Collaborative Filtering

Traditional system provides recommendation to particular application based upon the ranking given by the personalized user[5]. Now-a-days many application uses recommendation system which includes CDs, books, webpage, hotel reservation system and various[6], [7], [8]. In hotel reservation system, if one user is concerned about particular services and another user is looking for different services in the same hotel. But the same recommendation service is provided for both the user. It is not the good recommendation and the people will not satisfy to the recommendation. Moreover, in hotel reservation system the ratings of services and service recommendation list to the users are same does not consider user preferences [9]. Recommendation system is classified as content based, collaborative based and hybrid based recommendation system. Content based recommendation provides recommendation by taking the user preference from the previous user reviews. Collaborative Filtering (CF) recommends service based on the reviews of the previous user, by checking the similarity with the current user. Hybrid recommendation system combines recommendation of both content and CF.

C. Big Data Framework

Cloud computing is an effective platform to facilitate parallel computing in a collaborative way to tackle large-scale data. Big Data Analytics provides solution to these problems. Big data explains term of data sets which is large or complex so that traditional techniques failed to perform task [22].

The main characteristics of Big Data are volume, variety, veracity and velocity. In Big Data, the large dataset are partitioned into small data. Each data is further processed in parallel, by searching the patterns. The parallel process may interact with one another. The patterns from each partition are eventually merged and produce the result. Cloud computing tools are Hadoop, Mahout, MapReduce [10]. Hadoop is the open source tool for MapReduce and Google File System [11] which supports MapReduce programming framework written in Java, originally developed by Yahoo. Nowadays everything acts as a service, so creating and recommending the service using big data analytics in the social networking will be more efficient and accurate. The File System used for storing large data is Hadoop Distributed File System (HDFS) and simply adding the number of servers can achieve growth in storage capacity and computing power [12].

II. RELATED WORK

Recommendation is based on the people having similar preferences and interest (i.e. stable one) from past reviews [7]. It provides similarity computation using k-nearest neighbors. It uses user history profile as rows, their reviews as column and forms rating matrix. Cosine similarity used for calculating weight of rank matrix, which gives number of interaction between rows and columns. Finally, calculate the

item rating from weighted average rating of the neighbor user. It implements in MapReduce framework for overcoming scalability. It takes large computational time when dealing with huge amount of input data. So improvement must be done on Hadoop platform to reduce the computation time when dealing with these algorithms.

The system with most predicted rating by same user for similar items [6]. User-item matrix is formed by finding relationship between different items and to provide recommendation to the user. Consider reviews of similar item and identify similarity computation for item-item based approach. It computes using cosine based similarity, correlation based similarity and adjusted cosine based similarity. Finally, predicted rating for the target user is provided. Some other method is used in order to overcome the scalability issue.

Keyword based service recommendation system [13] which takes the preferences from the previous user keyword set and finds the similarity with the active user keyword set. Using CF, personalized rating for each service is considered and lists the top recommended services. Drawbacks of this system, it does not consider the positive and negative preferences. In order to make more accurate the bigrams of words is taken.

III. PROPOSED SYSTEM

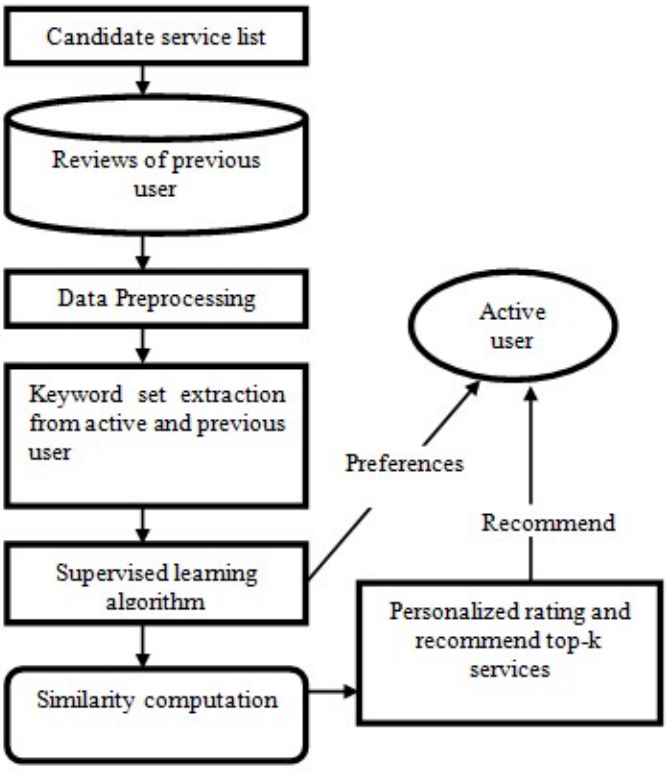

Figure 1.Architectural Diagram

The proposed system uses previous user reviews to find similarity with the active user and provide recommendation of service based on the active user needs [14]. First step is toform candidate service list for the application along with domain thesaurus i.e. semantic words [15]. Then collect the previous user post in the form of reviews, which includes their opinion about the application. After the collection of reviews, a review sentence is given to data preprocessing. Data preprocessing consist of stop word removal and Part-Of-Speech (POS) tagging. The keywords obtained are taken as keyword set of previous user. Meanwhile active user needs to provide the service as keywords. The system extracts opinion words in reviews and classified as n-level orientation scale [16]. Opinion orientation is an intended interpretation of the user satisfaction in terms of numerical values. It used to identify the number of positive and negative opinions of each keyword by using supervised learning algorithm [17]. Next, the similarity between the active and previous user's preference keyword set is calculated. The similarity computation is done by jaccord and cosine similarity method [13]. Finally, personalized rating for each service of the active user is calculated as shown in Fig. 1 and recommend top-k rating is provided to the active user [5].

The main steps of semantic based service recommendation system are described as follows:

A. Data Preprocessing

Stop word removal involves removing of unwanted and low priority words in each review sentence. Reviews are stored in HDFS which is given as input to stop word removal. Then each word is tagged using POS tagger.

B. Keyword Extraction

Active user gives their preferences of service as keywords by selecting from the candidate service list. From the active user preference services, keyword set is formed as Active Preference Keyword (APK). Then correspondingly previous reviews will be transformed as Previous Preference Keyword (PPK) set along with semantic words. Keywords tagged as noun by tagger is extracted [16] and check for the most frequent keyword using Apriori algorithm with minimum support count. The algorithm for keyword extraction is shown as follows:

```
keyword extraction (pos tagged input reviews)
    if word is in noun then
        extract (word)
    endif
count numbers of each word
set a minimum support count
    if count is  greater than minimum support count
        display (word)
    else
        remove (word)
    endif
```

C. Keyword Orientation

Bayes theorem calculates probability using supervised term counting based approach. It is used to identify keyword orientation by determining whether a given review is a positive, negative or neutral using opinion word [17]. In this algorithm, the probabilities of the labels are found according to the words. Steps are as follows:

- The positive and negative opinion words and review sentences are stored in text file.
- Split the sentence into the combination of words. It means first combination of two words and then single words.
- First compare the combination of two words, if matched then delete that combination from the opinion. Again start comparing for the single words.
- Initially, the probabilities of all the labels are zero [positive=0, negative=0]. Based on opinion, the probabilities of positive and negative labels get incremented

Similarly, the negation rule algorithm is applied for opinion orientation is as follows:

if opinion _word is near a negation word then

orientation←Apply Negation Rules(orientation)

Negation rules have a negation word or phrase which usually reverses the opinion expressed in a sentence. Three rules must be applied:
Negation Negative->Positive e.g., "no problem"
Negation Positive ->Negative e.g., "not clean" and
Negation Neutral-> Negative e.g.," does not suite", where "suite" is a neutral verb.

D. Jaccord Similarity Method

Jaccord similarity is an approximation method used for finding similarity between APK and PPK. It is does not consider the repetition of keywords in the keyword set. It takes the extracted keyword set of different previous users and compares the similarity with the preference keyword set of active user. The jaccord similarity method is given in algorithm as follows:

To calculate the similarity between APK and PPK,

$$sim\ (APK,PPK) = \frac{|\ APK \bigcap PPK\ |}{|\ APK \bigcup PPK\ |}$$

return sim (APK,PPK)

E. Cosine Similarity Method

It is an exact similarity method to find the most similarity between active preference keyword set and previous preference keyword set. The number of times the particular keywords is repeated in the APK and PPK is taken as weight of the keyword. If the keyword is not available in the preference keyword set, then the weight of the keyword will be taken as zero (i.e. wij = 0). Cosine similarity is also known as vector space model, in which the weight of the keywords in keyword set will be transformed as vector. Then the Term Frequency and Inverse Document Frequency (TF-IDF) is used for finding the number of times the particular term occurs in the document i.e. the frequency of the keywords. The frequency of the keyword will be taken as weight of the keyword in the keyword set. TF-IDF is calculated for both active preference keyword set and previous preference keyword set [5], [13].

TF-IDF in which Term Frequency(TF) takes the distinct keywords and number of times the particular keywords appears in the reviews and in the active keyword set in the following function:

$$TF = \frac{N_{pk_i}}{\sum_g N_{pk_i}} \qquad (2)$$

where, Npki number of times particular keyword appears in the keyword set, g is the number of keywords in the preference keyword set. The Inverse Document Frequency (IDF) is computed by number of documents containing the keywords divided by the number of keywords present in that document. It is given by following function:

$$IDF = 1 + \log_e\left(\frac{N}{n_i}\right) (3)$$

where, N is the total number of reviews posted by the user, ni is the number of occurrence of the keywords in all reviews. TF-IDF scores for each keywords is calculated as weight by the function:

$$w_{pk_i} = TF * IDF\ (4)$$

The weight of APK and PPK is used to calculate the cosine similarity is defined as follows,

$$\text{sim(APK,PPK)} = \cos\left(\vec{W}_{AP}, \vec{W}_{PP}\right)$$

$$= \frac{\vec{W}_{AP} * \vec{W}_{PP}}{\| \vec{W}_{AP} \|_2 * \| \vec{W}_{PP} \|_2} (5)$$

where, $\vec{W}_{AP}$ and $\vec{W}_{PP}$ be the weight of the keyword in the keyword set of the active user and previous user

F. Personalized Rating

Using CF algorithm [5], rating of each service is provided based on the cosine similarity value. The previous keyword set which is most similar to the active keyword set is filtered out from cosine similarity. Rating of each keyword using cosine similarity is calculated and provides the top-k rated service to the active user. The personalized rating for each service of the active user is calculated as follows:

$$pr = \bar{r} + \sum_{PPK_j \in R} sim(APK, PPK_j) * (r_j - \bar{r}) (6)$$

$$k = \frac{1}{\sum_{PPK_J \in R} sim(APK, PPK_J)} \qquad (7)$$

where, $\bar{r}$ be the average rating of service, rj be the corresponding rating of the different previous user,

sim (APK,PPKj) be the similarity of APK and PPK of cosine similarity value. K is the normalizing factor and R is used to store the previous user after each filtration.

III. IMPLEMENTATION ON MAPREDUCE

MapReduce [5], [7], [13] used to execute data in parallel manner. MapReduce used for implementing keyword and opinion extraction, similarity method, raking of services in parallel. It reduces time in running the algorithm.

IV. EXPERIMENTAL EVALUATION

The dataset used in the experiment is real dataset [18] which consist of 400mb of different hotels with overall rating of each hotel. The accuracy is measured by the parameters of precision, recall and F-measure as shown below,

$$\text{Precision} = \frac{|ExtractedValues \bigcap TrueValues|}{|ExtractedValues|} (8)$$

$$\text{Recall} = \frac{|ExtractedValues \cap TrueValues|}{|TrueValues|} \quad (9)$$

$$F-measure = \frac{2*\text{Recall}*\text{Precision}}{(\text{Recall}+\text{Precision})} \quad (10)$$

Keyword extraction by apriori algorithm shows the accuracy in Figure.2,

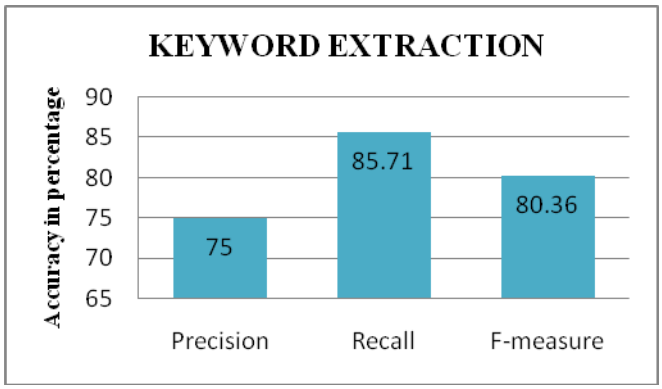

Figure2. Keyword Extraction

By naïve baysen, opinion orientation is analyzed and accuracy is measured for the precision, recall, F-measure is shown in Figure.3,

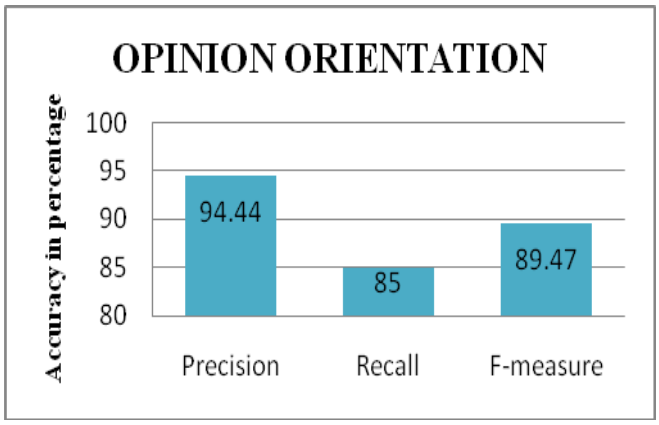

Figure 3: Opinion Orientation

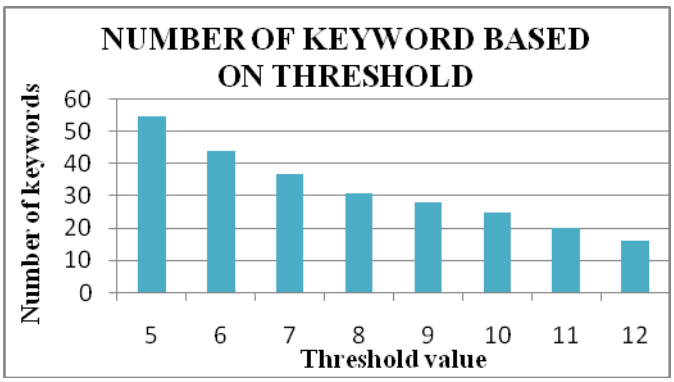

Figure 4. Number of keywords based on threshold

Keyword extraction gives accuracy of 80.36% using frequent itemset mining. Opinion orientation provides 89.47% of accuracy using naïve bayes for hotel dataset.Fig. 4, shows the outcome of number of keywords based on threshold count in terms of number of keywords threshold. Keywords are extracted for the threshold of 1, 2, 3 and 4, in which some of the keywords are not related to hotel keywords. If the threshold count is greater than 12, there is a chance to ignore some of the keywords. So, the threshold count is set from 5 to 12.

In Figure. 5, provides the results of keyword orientation in terms of number of opinions for 5keywordin reviews is shown below,

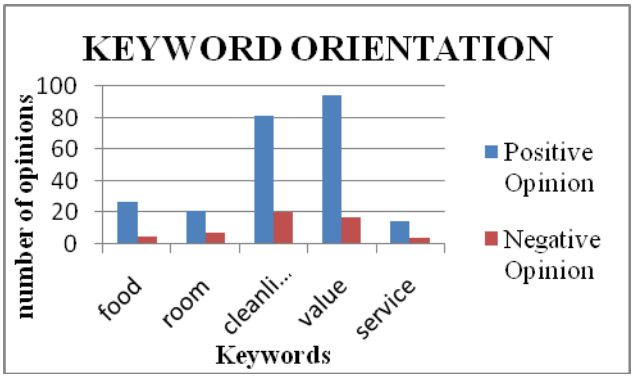

Figure 5. Number of Positive and Negative opinions of each keywords

The result is taken for similarity computation of APK with PPK keyword set using jaccord and cosine similarity method. The APK consist of 3 keywords (cleanliness, food, value). From the computation, cosine similarity provides the highest value for keyword between APK and PPK is shown in Fig.6.

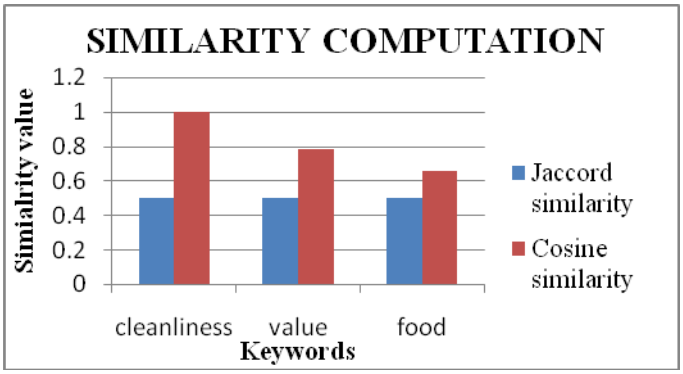

Figure 6. Similarity Computation of jaccord and cosine similarity

Rating of keyword for the most similar is rated between (0-5), where the highest value gives most needed keyword to the user. Semantic based service recommendation provide mostaccurate rating than Keyword Aware Service Recommendation (KASR) [13] as shown in Fig. 7,

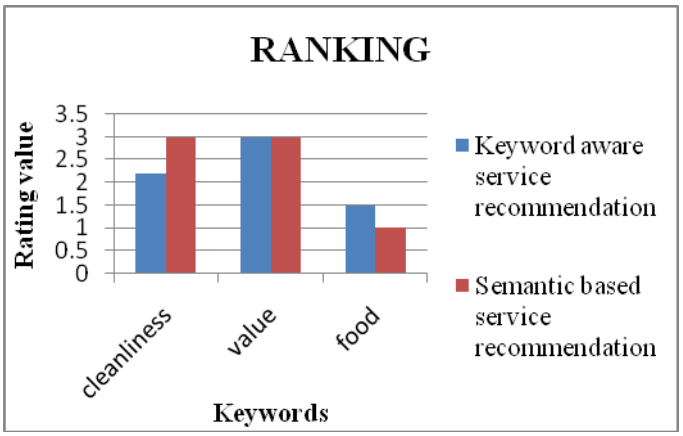

Figure 7. Ranking of keywords

Execution time for a single mapper is higher for both similarity methods. By increasing the number of mapper, execution time is decreased as shown in Figure. 8,

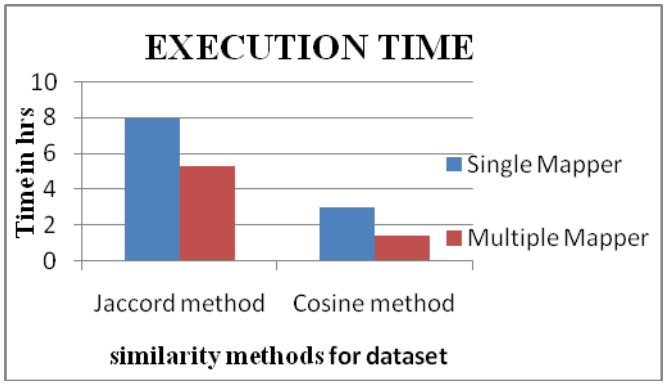

Figure 8. Execution time of similarity methods

CONCLUSION

The proposed system extracts keyword from customer reviews with minimum support threshold. The opinion words are extracted in reviews. Bayes theorem based on probabilities using supervised term counting based approach is used to identify sentence and keyword orientation. The number of positive and negative opinions in review sentences is estimated. And count the number of positive and negative opinion for each keyword in online customer reviews. To validate the performance of the system the rating of each keyword is calculated. The proposed system gives keyword rating but tripadvisor website gives overall rating of the hotel without analyzing the opinions on each keyword. This would make the proposed technique more complete and effective. In future, further implementation is done by increasing the number of node to make the system more efficient and reduces the time in execution.

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