Intelligent Recommendation System Using Big Data and Hadoop

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ABSTRACT

More E-business Websites give items distinctive costs which made it hard for customers to discover the items and administrations they need. Keeping in mind the end goal to defeat this information over-burden, clever proposal motors are utilized to recommend items and to furnish customers with applicable information to help them choose which items to buy. Suggestion motors are exceedingly computational and subsequently perfect for the Hadoop Platform. Result shows approximately 75 to 80 percent is achieved for register users and 65 to 70 percent is achieved for unregister users. For register users accuracy is comparatively high for propose system then the existing ones E-commerce product to the register as well as the unregister user with increased accuracy by analyzing the interest of the users. This framework goes for building a Web Recommendation motor which utilizes thing or client based suggestion for prescribing Items. It will investigate the information and give recommendations taking into account what comparable clients did and on the past exchange history of the client.

Keyword: - E-Commerce, Recommender System, User-based CF, Item-based CF, Hadoop

I. INTRODUCTION

1. Recommendation System

The items can be prescribed taking into account the top general venders on a site, or on an examination of the past purchasing conduct of the buyer as a forecast for future purchasing conduct. The types of proposal incorporate recommending items to the customer, giving customized item data, outlining group sentiment, and giving group evaluates. Comprehensively, these proposal systems are a piece of personalization on a site since they help the site adjust to every client. Content outline is a chain of a packing a given archive into a condensed variation by removing the most basic data from it.

1.1 Collaborative filtering

These methodologies constructing a model from a client's past conduct (things beforehand acquired or chose and/or numerical appraisals given to those things) and also comparable choices made by different clients. This model is then used to anticipate things (or evaluations for things) that the client may have an enthusiasm for. Communitarian separating approaches regularly experience the ill effects of three issues: chilly begin, adaptability, and sparsity.

• There are two major approaches for collaborative filtering algorithms:

- User-based approaches: It certain majority of the customer have same taste then they join into the one group. Recommendation is given to user based on evolution of items by other users from the same group.
- Item-based approaches: This method tries to predict user's opinion on different items and then recommends an item based on the users transaction history as well as a numerical value that expresses the predicted likeliness of an item about which the user has not given his opinion.

1.2 Content-based filtering

This approach uses a set of discrete characteristics of an item to recommend more items with similar properties. It is based on item description and user preference.

1.3 Demographic Recommender System

Demographic recommender systems aim to classify the user based on personal attributes and make recommendations based on demographic classes. A demographic recommender system uses the personal information of the user to make predictions regarding their likes. The benefit of a demographic approach is that it may not require a history of user ratings of the type needed by collaborative and content-based techniques.

1.4 Knowledge-Based Recommender System

Knowledge-based recommender systems are a specific type of recommender_system that is based on explicit knowledge about the item assortment, user preferences, and ultimately, how the item is useful for the user. Knowledge-based systems tend to work better than others at the beginning of their deployment but if they are not equipped with learning components they may be surpassed by other shallow methods that can exploit the logs of the human/computer interaction.

1.5 Hybrid Recommendation

In some cases, combining the results of any two recommendation technique like collaborative and content-based filtering proves to be more efficient. It can be done by making collaborative and content based recommendations separately and then merging them or by adding content based to collaborative approach or vice versa.

2. Big Data

Big data analysis is one of the upcoming disciplines in data mining where the large unstructured data that is very difficult to store and retrieve in an efficient manner. The three perspectives of big data are volume, velocity and variety. Volume refers to the amount of data that is being processed. Velocity refers to the speed at which the data can be processed with minimal error rate. Variety refers to all types of data starting from unstructured raw data to semi-structured and structured data.

3. Hadoop

Hadoop is an Open Source Framework given by Apache Software Foundation Founded by Doug Cutting. In Big Data large amount (Volume) of data is there may be in form of GB, TB or PB. So, Hadoop is used to manage that massive amount of data. Main aim of Hadoop is the partition a large amount of data set into data fragments. Because efficiency is the big problem for Big Data.

Hadoop includes:

HDFS: Hadoop Distributed File System

Map Reduce: Offline computing engine.

II. Problem with Existing System

This system aims to building a book recommendation engine which uses item or user based recommendation from Mahout for recommending books. After surveying various recommendation systems on

e-commerce web-site we found that most of the recommendation system recommends the product to the register users only it can't recommend the product to the un-register users of e-commerce web-site. Most of the recommendation system have cold star problem that's way accuracy of the system is less.

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Fig.2 Structured Data

III. Propose System

Propose system is accurately recommend the product to the registered as well as the unregister users with help of web-logs files.

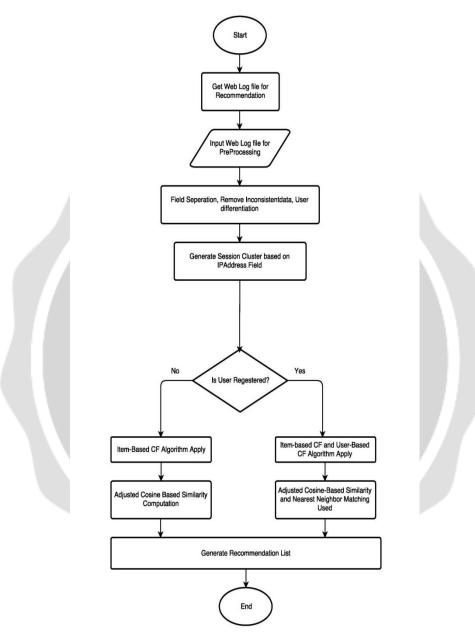


Fig. 1 Flow chart of propose system

1. Data acquisition

In this stage the whole navigational information which incorporates all the website pages went by is gathered and put away. The propose work makes utilization of regular log record configuration to keep up the information. This information got is very unstructured and conflicting in nature so that must be preprocessed for further examination.

2. Data Preprocessing

In this stage conflicting, excess information is wiped out utilizing taking after stride. Field division stage concentrates on recognizing one trait from another by making utilization of separator character, for example, comma. In information wiping stage we sift through exceptions information. We check for URL postfixes. Log sections having filename postfixes, for example, gif, jpeg, jpg are disposed of. All records having fizzled http status code are evacuated i.e status code more noteworthy than 200 and under 299 are killed. In client separation stage we dole out one of a kind ID, to every IP deliver and enlisted clients to separate one client structure other. At long last we build session in session ID stage. In this stage we aggregate together session having a place with special client.

3. Recommendation system

Two different recommendation systems have been proposed. One is for register user which is work based on user profile and another is for unregister user which is based on recent transaction history of the users.

3.1 For Register Users

If users are register then recommend the item based on user profile and transaction history of the users. In this section Item-based CF and User-based CF both approaches are apply so that this system use adjusted cosine algorithm for finding item similarity and nearest neighbor algorithm for finding similar kind of user. After finding the similarity of users and items system will generate the recommendation list for register users.

3.2 For Unregister Users

If users are unregister then recommend the item based on current transaction history of the user based on IP address of the users. In this section Item-based CF approach is apply so that this system use adjusted cosine algorithm for finding item similarity. Based on item similarity system will generate the recommendation list for unregister users.

IV. Result



Fig. 2 Result of Propose System

IV. CONCLUSION

In this research work we focus on providing good quality recommendation to all the users especially unregister users of e-commerce site. The idea of propose system is to develop a recommendation engine that can recommend Result Shows that accuracy of propose system is high then the existing system. Result shows approximately 75 to 80 percent is achieved for register users and 65 to 70 percent is achieved for unregister users. For register users accuracy is comparatively high for propose system then the existing ones E-commerce product to the register as well as the unregister user with increased accuracy by analyzing the interest of the users.

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