# A COLLABORATIVE FILTERING BASED RECOMMENDER SYSTEM ALLEVIATING THE MOST COMMON CHALLENGES

#### A DISSERTATION

Submitted in Partial fulfilment of the Requirement for the Award of the Degree of

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IN

#### **COMPUTER SCIENCE AND ENGINEERING**

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### DECLARATION

I hereby declare that the Major Project-II work entitled "A Collaborative Filtering based Recommender System Alleviating the most common Challenges" which is being submitted to Delhi Technological University, in partial fulfilment of requirements for the award of degree of Master of Technology (Computer Science and Engineering) is a bona fide report of Major Project-II carried out by me. The material contained in the report has not been submitted to any university or institution for the award of any degree.

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### CERTIFICATE

This is to certify that Project Report entitled "A Collaborative Filtering based Recommender System Alleviating the most common Challenges" submitted by Sahil Manchanda (Roll no. 2k16/CSE/12) for partial fulfilment of the requirement for the award of degree Master Of Technology (Computer Science and Engineering) is a record of the original work carried out by him under my supervision. To the best of my knowledge, this work has not been submitted in part or full for any degree or Diploma to this University or elsewhere.

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#### ABSTRACT

Recommender Systems (RS) is a subclass of information filtering system that seeks to predict the "rating" or "preference" that a user would give to an item. These are the Systems for recommending items (e.g. books, movies, web pages, newsgroup messages) to users based on examples of their preferences. With increasing information availability online, it has become extremely challenging to build accurate Recommender Systems. But at the same time, they are extremely important for businesses all around the globe. Hence it becomes vital to study these systems and improve their accuracy.

Recommender Systems are generally classified into Content based and Collaborative filtering(CF) based RS. Content based RS focus on the properties of the items while CF based RS estimates user/item similarity. Here, we discuss RS' with special emphasis on CF based techniques which are further classified as user/item based CF Algorithms.

The Issues and Challenges faced by CF based RS makes it an active area of research with many new variants and improvements being proposed in the past years. This work proposes a recommender system that alleviates three challenges simultaneously i.e. shilling attacks, sparsity problem and cold start problem that are most common in this domain. As a result, the proposed system proves to be more effective as compared to the conventional algorithm. Practically, this system can be utilized to produce quality recommendations to its users, as a result of which, the profitability can be maximized. The dataset used for doing the analysis and building the system is the popular MovieLens dataset. The work also presents a study of the Collaborative Filtering Improvements along with analysis of the important aspect of each improvement. This study can be utilized by any researcher working in this direction to have a complete idea of the past and the latest improvements in this area.

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## LIST OF ABBREVIATIONS

- 1. RS: Recommender/Recommendation System
- 2. CF: Collaborative Filtering
- 3. ACOS: Adjusted Cosine Similarity
- 4. PCC: Pearson Correlation Coefficient
- 5. UBCF: User based Collaborative Filtering
- 6. IBCF: Item based Collaborative Filtering
- 7. PCA: Principal Component Analysis
- 8. WS1: Weighted Slope One
- 9. HELF: Harmonic Mean of Entropy and Logarithm of Frequency
- 10. MAE: Mean Absolute Error