# **MOVIE RECOMMENDER SYSTEM**

Project report submitted in partial fulfillment of the Requirements for the Award of the Degree of Bachelor of Technology

In

#### **Computer Science and Engineering**

BY

NAME: AMIT RAWAT

**ROLL NO. 2011302** 

NAME: UMANG NAITHANI

ROLL NO. 2011650

NAME: ADITYA NAUTIYAL

**ROLL NO. 2011604** 



**Department of Computer Science and Engineering** 

Graphic Era Deemed to be University, Dehradun, Uttarakhand(248001), June 2020



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Github Link-( <a href="https://github.com/BLink291/Movie recommendation">https://github.com/BLink291/Movie recommendation</a>)

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AMIT RAWAT Roll no.-2011302

UMANG NAITHANI Roll no.-2011650

ADITYA NAUTIYAL Roll no.-2011604

Under the guidance of MR. VIJAY SINGH PROFESSOR, GEU



**Department of Computer Science and Engineering** 

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## **CERTIFICATE**

This is to certify that the project report entitled "MOVIE RECOMMENDER SYSTEM" being submitted by

AMIT RAWAT	2011302
<b>UMANG NAITHANI</b>	2011650
ADITYA NAUTIYAL	2011604

in partial fulfillment for the award of the Degree of **Bachelor of Technology** in **Computer Science and Engineering** to the **Graphic Era Deemed to be University** is a record of bonafide work carried out under my guidance and supervision.

The results embodied in this project report have not been submitted to any other University or Institute for the award of any Degree or Diploma.

(Project Guide) Designation Date:

Head of the Department

## **ACKNOWLEDGEMENT**

I have taken efforts in this project. However, it would not have been possible without the kind support and help of **Graphic Era Deemed to be University**, **MR. Vijay Singh**, our mentor, **Umang Naithani and Aditya Nautiyal**, my project partners . I would like to extend my sincere thanks to all of them.

I am highly indebted to **MR. Vijay Singh** for his guidance and constant supervision as well as for providing necessary information regarding the project & also for his support in completing the project.

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My thanks and appreciations also go to my colleague in developing the project and people who have willingly helped me out with their abilities.

## **ABSTRACT**

In this project, we attempt to understand the different kinds of recommendation systems and compare their performance on the MovieLens dataset. We attempt to build a scalable model to perform this analysis. We start by preparing and comparing the various models on a smaller dataset of 100,000 ratings. Then, we try to scale the algorithm so that it is able to handle 20 million ratings by using Apache Spark. We find that for the smaller dataset, using user-based collaborative filtering results in the lowest Mean Squared Error on our dataset.

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#### **CHAPTER 1: INTRODUCTION**

#### 1.1 Introduction

Rapid development of mobile devices and the internet has made it possible for us to access different movie resources freely. People sometimes find it difficult to choose from millions of movies. Moreover, movie service providers need an efficient way to manage movies and help their customers to discover movies by giving quality recommendations. Thus, there is a strong need for a good recommendation system.

Currently, there are many movie streaming services, like Amazon, Netflix, etc. which are working on building high-precision commercial movie recommendation systems. These companies generate revenue by helping their customers discover relevant movies and charging them for the quality of their recommendation service. Thus, there is a strong thriving market for good movie recommendation systems. Movie recommender system is a system which learns from the users past watching history and recommends them movies which they would probably like to watch in future. We have implemented various algorithms to try to build an effective recommender system. We firstly implemented a popularity based model which was quite simple and intuitive. Collaborative filtering algorithms which predict (filtering) taste of a user by collecting preferences and tastes from many other users (collaborating) is also implemented. We have also done experiments on content based models, based on latent factors and metadata.

## **1.2 Project Overview**

We use machine learning to build a personalized movie scoring and recommendation system based on user's previous movie ratings. Different people have different tastes in movies, and this is not reflected in a single score that we see when we Google a movie. Our movie scoring system helps users instantly discover movies to their liking, regardless of how distinct their tastes may be.

## 1.3 Objectives

- System portal should have login for admins and users.
- All the users should be able to give reviews to movies.
- Admin can add movies to the database.
- Users can get recommendations of movies on popularity basis and can also filter the recommendations by genres.

## 1.4 Genesis of problem

The number of movies available exceeds the watching capacity of a single individual. In the past, people used to shop in a physical store, in which the items available are limited. For instance, the number of movies that can be placed in a Blockbuster store depends on the size of that store. By contrast, nowadays, the Internet allows people to access abundant resources online. Netflix, for example, has an enormous collection of movies. Although the amount of available information increased, a new problem arose as people had a hard time selecting the items they actually wanted to see. This is where the recommender system comes in. Therefore many movie streaming services, like Amazon, Netflix, etc. are working on building high-precision commercial movie recommendation systems. These companies generate revenue by helping their customers discover relevant movies and charging them for the quality of their recommendation service. Thus, there is a strong thriving market for good movie recommendation systems.

#### **CHAPTER 2: LITERATURE SURVEY**

#### 2.1 ALGORITHMS

For our project, we focused on two main algorithms for recommendations:

- Content-based filtering
- Collaborative filtering

#### 2.1.1 Content Based Recommendations:

Content Based Recommendation algorithm takes into account the likes and dislikes of the user and generates a User Profile. For generating a user profile, we take into account the item profiles( vector describing an item) and their corresponding user rating. The user profile is the weighted sum of the item profiles with weights being the ratings user rated.

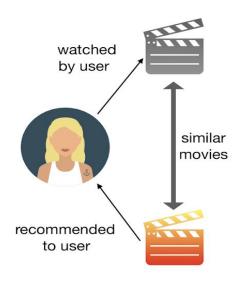
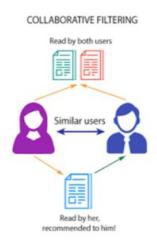


Figure 1. Content based filtering

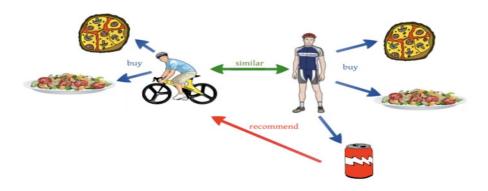
## 2.1.2 Collaborative Filtering:

Collaborative Filtering techniques make recommendations for a user based on ratings and preferences data of many users. The main underlying idea is that if two users have both liked certain common items, then the items that one user has liked that the other user has not yet tried can be recommended to him. We see collaborative filtering techniques in action on various Internet platforms such as Amazon.com, Netflix, Facebook.

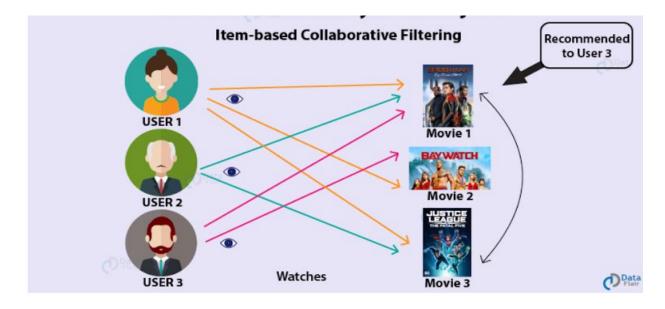


#### Collaborative Filtering is of two types:

**User-based:** For a user U, with a set of similar users determined based on rating vectors consisting of given item ratings, the rating for an item I, which hasn't been rated, is found by picking out N users from the similarity list who have rated the item I and calculating the rating based on these N ratings



**Item-based:** For an item I, with a set of similar items determined based on rating vectors consisting of received user ratings, the rating by a user U, who hasn't rated it, is found by picking out N items from the similarity list that have been rated by U and calculating the rating based on these N ratings.



#### 2.2 Python

Python is an interpreted, high-level, general-purpose programming language. Created by Guido van Rossum and first released in 1991, Python's design philosophy emphasizes code readability with its notable use of significant whitespace. Its language constructs and object-oriented approach aim to help programmers write clear, logical code for small and large-scale projects. Python is dynamically typed and garbage-collected. It supports multiple programming paradigms, including structured (particularly, procedural), object-oriented, and functional programming. Python is often described as a "batteries included" language due to its comprehensive standard library.

## **2.3.1** NumPy

NumPy stands for 'Numerical Python' or 'Numeric Python'. It is an open source module of Python which provides fast mathematical computation on arrays and matrices. Since arrays and matrices are an essential part of the Machine Learning ecosystem, NumPy along with Machine Learning modules like Scikit-learn, Pandas, Matplotlib, TensorFlow, etc. complete the Python Machine Learning Ecosystem.

#### 2.3.2 Pandas

Similar to NumPy, Pandas is one of the most widely used python libraries in data science. It provides high-performance, easy to use structures and data analysis tools. Unlike NumPy library which provides objects for multi-dimensional arrays, Pandas provides in-memory 2d table objects called Dataframe. It is like a spreadsheet with column names and row labels.

## 2.4 MongoDB

MongoDB is a cross-platform document-oriented database program. Classified as a NoSQL database program, MongoDB uses JSON-like documents with optional schemas. MongoDB is developed by MongoDB Inc. and licensed under the Server Side Public License (SSPL).

#### **2.5 HTML**

Hypertext Markup Language (HTML) is the standard markup language for documents designed to be displayed in a web browser. It can be assisted by technologies such as Cascading Style Sheets (CSS) and scripting languages such as JavaScript. Web browsers receive HTML documents from a web server or from local storage and render the documents into multimedia web pages. HTML describes the structure of a web page semantically and originally included cues for the appearance of the document.

#### 2.6 CSS

Cascading Style Sheets (CSS) is a style sheet language used for describing the presentation of a document written in a markup language like HTML. CSS is a cornerstone technology of the World Wide Web, alongside HTML and JavaScript.CSS is designed to enable the separation of presentation and content, including layout, colors, and fonts. This separation can improve content accessibility, provide more flexibility and control in the specification of presentation characteristics, enable multiple web pages to share formatting by specifying the relevant CSS in a separate .css file, and reduce complexity and repetition in the structural content.

## 2.7 JavaScript

JavaScript is a cross-platform, object-oriented scripting language used to make webpages interactive (e.g., having complex animations, clickable buttons, popup menus, etc.). There are also more advanced server side versions of JavaScript such as Node.js, which allow you to add more functionality to a website than simply downloading files (such as real time collaboration between multiple computers). Inside a host environment (for example, a web browser), JavaScript can be connected to the objects of its environment to provide programmatic control over them.

#### 2.7 Visual Studio Code

Visual Studio Code is a free source-code editor made by Microsoft for Windows, Linux and macOS. Features include support for debugging, syntax highlighting, intelligent code completion, snippets, code refactoring, and embedded Git. Users can change the theme, keyboard shortcuts, preferences, and install extensions that add additional functionality. The source code is free and open-source, released under the permissive MIT License. The compiled binaries are freeware for any use.

## 2.8 Application Programming Interface (API)

An application programming interface (API) is a computing interface which defines interactions between multiple software intermediaries. It defines the kinds of calls or requests that can be made, how to make them, the data formats that should be used, the conventions to follow, etc. It can also provide extension mechanisms so that users can extend existing functionality in various ways and to varying degrees. An API can be entirely custom, specific to a component, or it can be designed based on an industry standard to ensure interoperability. Some APIs have to be documented, others are designed so that they can be "interrogated" to determine supported functionality. Since other components/systems rely only on the API, the system that provides the API can (ideally) change its internal details "behind" that API without affecting its users.

## **CHAPTER 3: SYSTEM DEVELOPMENT**

## 3.1 Software Requirements

Client End: Internet Explorer, Google Chrome Operating System Windows/linux

Developer End: Visual studio Code, Python3, Operating System Windows/ linux

#### 3.2 Hardware Requirements

Processor: dual core CPU

RAM: 1GB or higher

Database Server: Mongoengine for Mongodb

## 3.3 Functional Requirements

#### **Register Users**

Register new user Login for existing user Register admin

#### **Add Movies**

Admin can add movies to the database

#### **Find Movies**

Users can find any movie by their name

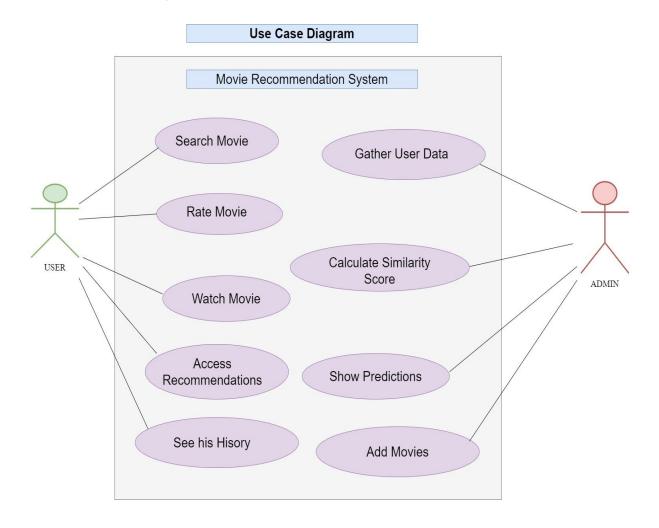
#### **Review Movies**

Users can give their reviews on movies

#### **Get Recommendations**

User can get personalized recommendations
User can filter the recommendations on basis of genres
User can see Top charts of movies

## 3.4 Use Case Diagram



## 3.4 User Interface (UI)

Should be user friendly Should adapt to different screen sizes

## **CHAPTER 4: PROJECT DEVELOPMENT**

## 4.1 Database Implementation

We use the MovieLens dataset available on Kaggle, covering over 45,000 movies, 26 million ratings from over 270,000 users. The data is separated into two sets: the first set consists of a list of movies with their overall ratings and features such as budget, revenue, cast, etc. After removing duplicates in the data, we have 45,433 different movies. Table 1 is the top 10 most popular movies by their weighted score, calculated using the IMDB weighting.

#### 4.2 Tables

#### 4.2.1 Ratings

Sl.No	Field Name	Field Description	Field Data	Constraints
			Type	
1.	movie_id	Unique movie Id	INTEGER	PK, NOT NULL
2.	user_id	Unique user Id	INTEGER	NOT NULL
3.	rating	Rating given by user	FLOAT	NOT NULL
4.	timestamp	Date and time of rating	DATETIME	NOT NULL

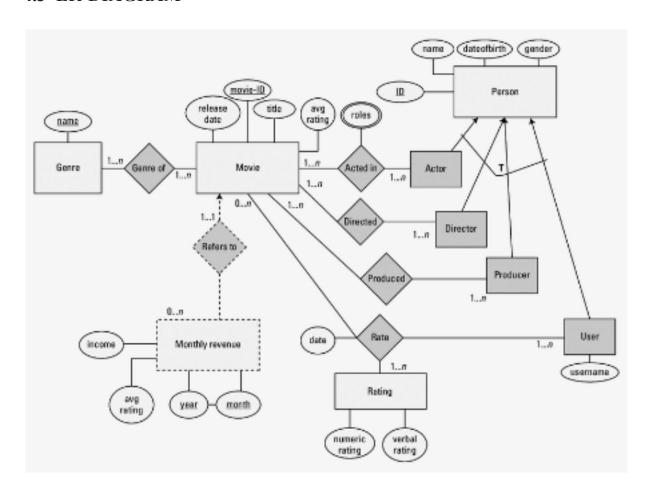
#### 4.2.1 User

Sl.No.	Field Name	Field Description	Field Data	Constraints
			Type	
1.	id	User Id	INTEGER	PK, NOT NULL,
				AUTO
				INCREMENT(100)
2	name	Name of user	STRING	NOT NULL
2.	email	Login Id	EMAIL FIELD	NOT NULL
3.	pswd	User Login Password	STRING	NOT NULL
4.	admin	User Creation Time	BOOLEAN	NOT NULL
5.	age	User's age	INTEGER	
6.	gender	User's Gender	STRING	
7.	country	User's country	STRING	
8.	spoken_lang	Languages User know	List of STRING	
	uages			

## **4.2.1** Movie

Sl.No.	Field Name	Field Description	Field Data Type	Constraints
1.	id	Movie Id	INTEGER	PK, NOT NULL, AUTO INCREMENT(10 0)
2	name	Name of Movie	STRING	NOT NULL
2.	genres	List of genres of movie	List of STRING	NOT NULL
3.	release _date	Release Date of the movie	STRING	NOT NULL
4.	release _year	Release Year of the movie	STRING	NOT NULL
5.	vote_count	Number of Reviews of movie	INTEGER	DEFAULT =0
6.	vote_average	Average review of movie	FLOAT	DEFAULT =0
7.	crew	List of crew involved in movie	List of STRING	
8.	cast	List of Cast of the movie	List of STRING	
9.	director	Director of the movie	STRING	
10.	keywords	Keywords of that movie	List of STRING	
11.	overview	Overview of the movie	STRING	
12.	spoken_languag es	Languages In which movie is available	List of STRING	
13.	adult	If movie is Adult rated	BOOLEAN	DEFAULT =FALSE

## 4.3 ER-DIAGRAM



### **CHAPTER 5: RESULTS AND SCREENSHOTS**

## 5.1 Login window

```
Are you a [a]dmin or [g]uest? g
 ************** Welcome guest *********
What action would you like to take:
[C]reate an account
[L]ogin to your account
e[X]it app
 *********** LOGIN **********
What is your email? amit@gmail.com
What is your password? aammittt
Logged in successfully.
What action would you like to take:
What action would you like to take:
[T]op Movies
[W]atch a new movie
[R]eview a movie
View [y]our watched movies
e[X]it app
ami@gmail.com>
```

## 5.2 Register window

```
*************** REGISTER ***********
What is your name? Umang Naithani
What is your email? umang@gmail.com
What is your password? umang
Created new account with id 11093.

What action would you like to take:

What action would you like to take:

[T]op Movies
[W]atch a new movie
[R]eview a movie
View [y]our watched movies
e[X]it app

Umang Naithani>
```

## 5.3 See Top Charts

#### 5.4 Get Recommendation

```
[W]atch a new movie
[R]eview a movie
View [y]our watched movies
e[X]it app
ami@gmail.com> w
              Recommended movies for you ***********
                          name vote_count vote_average
                                                         score
                House of Games
3911
                                                    7 6.794167
          The Spanish Prisoner
                                      74
1726
                                                    7 6.772874
4758
                        Heist
                                     138
                                                     6 5.955232
7194
                       Spartan
                                     104
                                                    6 5.942222
                       Redbelt
12587
                                                     6 5.924889
               State and Main
                                                    6 5.900588
               The Winslow Boy
                                       22
                                                    6 5.806857
                 Things Change
                                       13
                                                    6 5.740000
14101 If These Walls Could Talk
                                       13
                                                    5 5.240000
                      Homicide
8971
                                       32
                                                    5 5.138667
What action would you like to take:
What action would you like to take:
[T]op Movies
[W]atch a new movie
[R]eview a movie
View [y]our watched movies
e[X]it app
ami@gmail.com>
```

#### 5.6 Review a Movie

#### **5.7** See Users History

### **CHAPTER 6: CONCLUSION AND FUTURE SCOPE**

#### 6.1 Conclusion

In this project, a recommendation system has been implemented based on a hybrid approach of collaborative filtering engine and context based engine. The system can be highly improved by making use of caching mechanisms, user clustering which will definitely boost the speed of the system. Further enhancements include storing users past history of results, contexts for future predictions. In our project, collaborative filtering and content based filtering algorithms are used to predict a user's movie rating.

#### 6.1 Goals Achieved

We were able to successfully build a recommendation system using a hybrid approach of both collaborative filtering and content based filtering algorithms using the MovieLens dataset available on Kaggle, covering over 45,000 movies, 26 million ratings from over 270,000 users.

## **6.1 Future Scope**

Run the algorithms on a distributed system, like Hadoop or Condor, to parallelize the computation, decrease the runtime and leverage distributed memory to run the complete MSD.

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