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Raj Kamal Yadav, Rajat Bansal, Manish Mahalwal **Problem statement** Introduction/ Use Case

In today's world of Netflix and Chill, recommendation systems have become an essential part of our life, especially due to

commodity like movies, suggestions are made to users by

We shall be using both Content-Based recommendations with

features such as actors, director, budget, crew etc. as well as

Collaborative filtering for which the input to our algorithm

the increasing # of choices available to us. For a media

finding user profiles of individuals with similar tastes.

will be observed user ratings in the past. The output of the framework will be such that foy any given user, it would be able to tell us the top-N movies based on the metric used for calculating similarity. **Data Set** Dataset used is available at Kaggle containing 9,000 different

movies, 26 million ratings over 700 users. The data is provided to us in the form of: Credits.csv: About movie's cast and crew information

- B. Keywords.csv: Keywords associated to each movie.

 - Links.csv: IMDB and TMDB IDs' for all movies.

Rating.csv: Contains 100 ratings from all users for all movies. Movies Metadata.csv: Contains features related to each movie like Budget, Genre, Language, Title etc. We will split our data randomly into 75% and 25% for Training and Testing Set respectively. We will also perform Stratified K-Fold Cross Validation for different models.

Building a personalised movie recommendation system for users based on their past ratings.

Project objectives 1. We shall be using TF-IDF metric as part of the

Personalised Movie Recommendation System

content based recommendation as the Baseline for our project. 2. We shall also be using Matrix Factorization techniques such as SVD and KNN algorithm as part of the Collaborative Filtering approach. 3. For our final model, we shall be using Bayesian

Personalized Ranking along with Matrix Factorization. 4.Root Mean Squared Error is used to evaluate the performance of the models.

Literature Review BPR: Bavesian Personalized Ranking from Implicit

Feedback: This paper presents a generic optimization criterion (BPR-Opt) for item recommendation from implicit feedback for personalized ranking. Reference Blogs:

Recommender System using Bayesian Personalized Ranking Introduction to Recommender System

Week 11-12 Training and testing

Stratified Cross Validation

Timeline

Week 5-6

Understanding problem

and reading relevant literature.

Week 7-8

Data Cleaning and Baseline

Model

Week 9-11

Implementing KNN, Matrix

Factorization as well as Bayesian

Personalized Ranking

analysis.

Week 13 Compilation of results and its