## **2020 Massive Data Analysis Term Project**

### **Recommendation System**

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A movie recommendation system which describes the details as following. In the first part, I will build an "Item(movie)-Based Collaborative Filtering" to find the similarity of movie using "Cosine Similarity", and in the second part I will build a simple recommender system for "Rating Predictions" to predict the movie rating for each user by the top 10 similar movies' rating from first part.

The whole program is implement by MapReduce on PySpark.

#### I. Dataset

From MovieLens: https://grouplens.org/datasets/movielens/

This dataset (*ml-latest-small*) describes 5-star rating from MovieLens, a movie recommendation service.

It contains 100836 ratings across 9742 movies which created by 610 users.

All ratings are contained in the file *ratings.csv*. Each line of this file after the header row represents one rating of one movie by one user, and has the following format:

userId,movieId,rating,timestamp

### II. Item-item Collaborative Filtering

Calculate the similarity for each movie pairs

Cosine similarity:

$$sim(x,y) = cos(r_x, r_y) = \frac{r_x \cdot r_y}{\|r_x\| \cdot \|r_y\|}$$

\* The mapper and reducer implement detail please see the ipynb file

Output Result: <a href="https://www.dropbox.com/s/ao53hl6rfkbwl4p/similarity.out?dl=0">https://www.dropbox.com/s/ao53hl6rfkbwl4p/similarity.out?dl=0</a>

# **III. Rating Predictions**

Select top 10 similarity to calculate the movie rating for each user

- Rating Predictions:

$$r_{xi} = \frac{\sum_{j \in N(i;x)} s_{ij} \cdot r_{xj}}{\sum_{j \in N(i;x)} s_{ij}}$$

\* The mapper and reducer implement detail please see the ipynb file

Output Result: https://www.dropbox.com/s/wtcogl0lv1fxx7n/predict.out?dl=0