

Final project Step 2

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In this step, I've started to import libraries and data. I'll be looking at the data with `str()` function and also will be seeing the first few rows by `head()` function. Later on, I'll be creating the matrices of genres of movies so that the recommenderlab could work with the data.

Importing libraries

Importing the dataset

```
setwd("~/MSDS/DSC520/dsc520/Final_project")
movie_data <- read.csv("IMDB-Dataset/movies.csv", stringsAsFactors=FALSE)
rating_data <- read.csv("IMDB-Dataset/ratings.csv")
str(movie_data)
```

```
## 'data.frame': 10329 obs. of 3 variables:
## $ movieId: int 1 2 3 4 5 6 7 8 9 10 ...
## $ title : chr "Toy Story (1995)" "Jumanji (1995)" "Grumpier Old Men (1995)" "Waiting to Exhale (1995)"
## $ genres : chr "Adventure|Animation|Children|Comedy|Fantasy" "Adventure|Children|Fantasy" "Comedy|Drama|Romance"
```

Glimpse of the data

```
# Movies data
head(movie_data)
```

```
##   movieId      title
## 1      1  Toy Story (1995)
## 2      2    Jumanji (1995)
## 3      3  Grumpier Old Men (1995)
## 4      4  Waiting to Exhale (1995)
## 5      5  Father of the Bride Part II (1995)
## 6      6      Heat (1995)
##              genres
## 1 Adventure|Animation|Children|Comedy|Fantasy
## 2      Adventure|Children|Fantasy
## 3      Comedy|Romance
```

```
## 4          Comedy|Drama|Romance
## 5                      Comedy
## 6          Action|Crime|Thriller
```

```
# Ratings data
head(rating_data)
```

```
##   userId movieId rating timestamp
## 1      1      16   4.0 1217897793
## 2      1      24   1.5 1217895807
## 3      1      32   4.0 1217896246
## 4      1      47   4.0 1217896556
## 5      1      50   4.0 1217896523
## 6      1     110   4.0 1217896150
```

```
# extracting the genres as a dataframe
movie_genre <- as.data.frame(movie_data$genres, stringsAsFactors=FALSE)
```

```
# Splitting the collective genres into individual ones
movie_genre2 <- as.data.frame(tstrsplit(movie_genre[,1], '[|]', type.convert=TRUE), stringsAsFactors=FALSE)
```

```
# Assigning column names as serial numbers assuming each movie may have maximum of 10 genres
colnames(movie_genre2) <- c(1:10)
```

```
# List of all the genres
```

```
list_genre <- unique(movie_genre2[c("1")])[1:18,] # there was 19 values but last one was without genre
```

```
#list_genre <- c("Action", "Adventure", "Animation", "Children", "Comedy",
#              "Crime", "Documentary", "Drama", "Fantasy", "Film-Noir",
#              "Horror", "Musical", "Mystery", "Romance", "Sci-Fi", "Thriller", "War", "Western")
```

```
# Initializing a matrix
genre_mat1 <- matrix(0,10330,18)
```

```
genre_mat1[1,] <- list_genre
```

```
# Assigning genres as column names
colnames(genre_mat1) <- list_genre
```

```
for (index in 1:nrow(movie_genre2)) {
  for (col in 1:ncol(movie_genre2)) {
    gen_col = which(genre_mat1[1,] == movie_genre2[index,col])
    genre_mat1[index+1,gen_col] <- 1 }}
```

```
genre_mat2 <- as.data.frame(genre_mat1[-1,], stringsAsFactors=FALSE) #removing first row, which was the
```

```
for (col in 1:ncol(genre_mat2)) {
  genre_mat2[,col] <- as.integer(genre_mat2[,col]) #convert from characters to integers
}
```

```
str(genre_mat2)
```

```
## 'data.frame': 10329 obs. of 18 variables:
## $ Adventure : int 1 1 0 0 0 0 0 1 0 1 ...
## $ Comedy : int 1 0 1 1 1 0 1 0 0 0 ...
## $ Action : int 0 0 0 0 0 1 0 0 1 1 ...
## $ Drama : int 0 0 0 1 0 0 0 0 0 0 ...
## $ Crime : int 0 0 0 0 0 1 0 0 0 0 ...
## $ Children : int 1 1 0 0 0 0 0 1 0 0 ...
## $ Mystery : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Animation : int 1 0 0 0 0 0 0 0 0 0 ...
## $ Documentary: int 0 0 0 0 0 0 0 0 0 0 ...
## $ Thriller : int 0 0 0 0 0 1 0 0 0 1 ...
## $ Horror : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Fantasy : int 1 1 0 0 0 0 0 0 0 0 ...
## $ Western : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Film-Noir : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Romance : int 0 0 1 1 0 0 1 0 0 0 ...
## $ Sci-Fi : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Musical : int 0 0 0 0 0 0 0 0 0 0 ...
## $ War : int 0 0 0 0 0 0 0 0 0 0 ...
```

Combining the genre matrix with the movies data resulting in a search matrix

```
SearchMatrix <- cbind(movie_data[,1:2], genre_mat2[])
head(SearchMatrix)
```

```
##   movieId      title Adventure Comedy Action Drama
## 1      1      Toy Story (1995)      1      1      0      0
## 2      2      Jumanji (1995)      1      0      0      0
## 3      3      Grumpier Old Men (1995)      0      1      0      0
## 4      4      Waiting to Exhale (1995)      0      1      0      1
## 5      5      Father of the Bride Part II (1995)      0      1      0      0
## 6      6      Heat (1995)      0      0      1      0
##   Crime Children Mystery Animation Documentary Thriller Horror Fantasy Western
## 1      0      1      0      1      0      0      0      1      0
## 2      0      1      0      0      0      0      0      1      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      1      0      0      0      0      1      0      0      0
##   Film-Noir Romance Sci-Fi Musical War
## 1      0      0      0      0      0
## 2      0      0      0      0      0
## 3      0      1      0      0      0
## 4      0      1      0      0      0
## 5      0      0      0      0      0
## 6      0      0      0      0      0
```

For the movie recommendation system to make sense of the ratings through recommenderlabs, we have to convert our matrix into a sparse matrix one. This new matrix is of the class 'realRatingMatrix'.

```
ratingMatrix <- reshape2::dcast(rating_data, userId~movieId, value.var = "rating", na.rm=FALSE)
ratingMatrix <- as.matrix(ratingMatrix[,-1]) #remove userIds

#Convert rating matrix into a recommenderlab sparse matrix
ratingMatrix <- as(ratingMatrix, "realRatingMatrix")
ratingMatrix
```

```
## 668 x 10325 rating matrix of class 'realRatingMatrix' with 105339 ratings.
```

Exploring recommendation model options

```
recommendation_model <- recommenderRegistry$get_entries(dataType = "realRatingMatrix")
names(recommendation_model)
```

```
## [1] "HYBRID_realRatingMatrix"      "ALS_realRatingMatrix"
## [3] "ALS_implicit_realRatingMatrix" "IBCF_realRatingMatrix"
## [5] "LIBMF_realRatingMatrix"      "POPULAR_realRatingMatrix"
## [7] "RANDOM_realRatingMatrix"      "RERECOMMEND_realRatingMatrix"
## [9] "SVD_realRatingMatrix"        "SVDF_realRatingMatrix"
## [11] "UBCF_realRatingMatrix"
```

Since we're interested to create the model based on IBCF algorithm or Item Based Collaborative filtering, let's look at the parameters for that.

```
recommendation_model$IBCF_realRatingMatrix$parameters
```

```
## $k
## [1] 30
##
## $method
## [1] "Cosine"
##
## $normalize
## [1] "center"
##
## $normalize_sim_matrix
## [1] FALSE
##
## $alpha
## [1] 0.5
##
## $na_as_zero
## [1] FALSE
```

So far, I have reached at this step, Next steps will be: - Creating visualizations of ratings and most watched movies etc.. I'll be exploring data as much as possible. - Then, I'll work on the recommender system based on Collaborative Filtering System. For that, I'll split the data in 80:20 ratio for - training and testing purposes. - Train the model __ Make some predictions/recommendations - Validate the model