

## DESIGN DOCUMENTATION

This assignment evaluates the speed and accuracy of different recommendation systems given below

1. Collaborative filtering
2. Collaborative along with Baseline approach
3. SVD
4. SVD with 90% retained energy
5. CUR
6. CUR with 90% retained energy

This assignment is divided into 7 subcodes.

- svd.py
- collaborative\_filtering.py
- cur.py
- data\_handling.py
- error\_funcs.py
- similarity\_funcs.py
- final\_run.py

### **svd.py**

In this method a data matrix(A) is divided into three sub matrices, using singular value decomposition.

$$A = U * Sig * V'$$

A = Original Data Matrix ( Users \* Items)

U = Users to Concept matrix

V = Items to Concept matrix

Sig = Concept Strength matrix containing Eigen values in decreasing order

Here this part of the code takes the A matrix with few values removed and it tries predicting it. Now the error between the predicted values and the original values will be compared.

### **collaborative\_filtering.py**

This method implements an approach where we take ratings of users similar to the one we need to predict, from A matrix and estimate a value based on weighted similarity of their ratings. Here we take two similar users but it may vary as per the requirement. In baseline approach we add a new term b, while estimating the weighted similarity, where b is the sum of average of all the ratings, deviation of the user and deviation of the item.

### **cur.py**

Here A matrix is split into C, U, R matrices where C contains of few columns randomly picked from matrix A. U is a matrix which is generated through a particular algorithm and R contains few rows which are randomly picked from A.

$$A = C * U * R$$

Where, C =column matrix,

U = pseudo inverse of intersection of C and R,

R = row matrix.

### **data\_handling.py**

It will retrieve the data from the dataset which is of the form user-id, movie-id, rating, time-stamp as four columns in it. From this four columns user item matrix is created which is used in the recommendation system

### **error\_funcs.py**

This part of the code contains three different measures to find the accuracy of recommendation systems:

1. RMSE - Root Mean Square Error
2. Precision in Top K
3. Spearman Correlation

RMSE - Root Mean Square Error:

Formula:  $(\text{sum}((\text{predicted} - \text{actual}) ** 2) / n) ^ 0.5$

Precision in Top K:

Gives an estimate of how many of the predicted ratings are present in the top K ratings of the user since only the good one's count in the error measure.

Spearman Correlation:

Formula:  $1 - [\text{sum}(\text{diff}(\text{predicted} - \text{actual})^2) / n((n^2)-1)]$

### **similarity\_funcs.py**

In this part of the code pearson similarity for given two matrices will be calculated

$$\text{sim}(x, y) = \frac{\sum_{s \in S_{xy}} (r_{xs} - \bar{r}_x)(r_{ys} - \bar{r}_y)}{\sqrt{\sum_{s \in S_{xy}} (r_{xs} - \bar{r}_x)^2} \sqrt{\sum_{s \in S_{xy}} (r_{ys} - \bar{r}_y)^2}}$$

### **final\_run.py**

This is a control unit for the whole code, which combines all the functions written above in an appropriate way and gives output.

### **Data\_set**

All the data here is stored in the folder dataset which contains a file namely u.data. In this file, data of  $10^5$  ratings are stored which is collected from internet. In this file data is stored in four columns, where 1st column contains user id's, 2nd column contains movie id, 3<sup>rd</sup> column contains rating given by the specific user and 4<sup>th</sup> column contains the time stamp.

