Package 'CFF'

October 12, 2022

Title Simple Similarity for User-Based Collaborative Filtering Systems

Version 1.0
Date 2020-02-25
Maintainer Farimah Houshmand Nanehkaran <hoshmandcomputer@gmail.com></hoshmandcomputer@gmail.com>
Description A simple, fast algorithm to find the neighbors and similarities of users in user-based filtering systems, to break free from the complex computation of existing similarity formulas and the ability to solve big data.
License GPL (>= 2)
Encoding UTF-8
RoxygenNote 7.0.2
NeedsCompilation no
Author Farimah Houshmand Nanehkaran [aut, cre] (https://orcid.org/0000-0003-1687-1719 , University=Islamic Azad University of Kashan,Kashan,Iran), Seyed Mohammad Reza Lajevardi [ctb] (https://orcid.org/0000-0002-4744-2784 , University=Islamic Azad University of Kashan,Kashan,Iran), Mahmoud Mahlouji Bidgholi [ctb] (https://orcid.org/0000-0001-8895-8501 , University=Islamic Azad University of Kashan,Kashan,Iran) Repository CRAN Date/Publication 2020-03-04 10:00:05 UTC
R topics documented:
CFF-package

2 CFF-package

CFF-package Simple Similarity for User-Based Collaborative Filtering Systems

Description

A simple, fast algorithm to find the neighbors and similarities of users in user-based filtering systems, to break free from the complex computation of existing similarity formulas and the ability to solve big data.

Details

The DESCRIPTION file:

CFF Package:

Title: Simple Similarity for User-Based Collaborative Filtering Systems

Version:

Date: 2020-02-25

Authors@R: c(person(given="Farimah", family="Houshmand Nanehkaran", role = c("aut", "cre"), email="hoshmandcom

Farimah Houshmand Nanehkaran <hoshmandcomputer@gmail.com> Maintainer:

Description: A simple, fast algorithm to find the neighbors and similarities of users in user-based filtering systems, to brea

License: GPL (>= 2)UTF-8 Encoding: RoxygenNote: 7.0.2

Author: Farimah Houshmand Nanehkaran [aut, cre] (https://orcid.org/0000-0003-1687-1719, University=Islamic

Index of help topics:

CFF-package Simple Similarity for User-Based Collaborative

Filtering Systems

Score_replace Replacing of Neighbor Users' Ratings on

Non-Rated Items By The Active User

simple_predict Prediction Unseen Items For The Active User

simple_similarity Finding Neighbor Users And Their Similarity

Values

User-Based Collaborative Filtering Systems

Author(s)

NA

Maintainer: Farimah Houshmand Nanehkaran <hoshmandcomputer@gmail.com>

References

Kumar, P., Kumar, V., & Thakur, R. S. (2019). A new approach for rating prediction system using collaborative filtering. Iran Journal of Computer Science, vol.2, no. 2, pp. 81-87.

CFF-package 3

Zhang, P., Zhang, Z., Tian, T., & Wang, Y. (2019). *Collaborative filtering recommendation algorithm integrating time windows and rating predictions*. Applied Intelligence, vol. 49, no. 8, pp. 3146-3157.

Gadekula, S. K., Rao, U. P., Vyas, R. K., Dontula, A. L., & Gaikwad, S. V. (2019). *Improved Pearson Similarity for Collaborative Filtering Recommendation System*. In 2019 6th International Conference on Computing for Sustainable Global Development (INDIACom), pp. 1047-1054, IEEE.

Examples

```
ratings <- matrix(c( 2,</pre>
                                                    4,
                         5, NaN, NaN, NaN,
                                                  5,
                    NaN, NaN, NaN, 1, NaN,
                         4,
                   NaN,
                               5, NaN,
                                            4,
                                                 NaN,
                     4, NaN, NaN, 5, NaN,
                                                 NaN,
                      5, NaN,
                               2, NaN, NaN,
                                                 NaN,
                                       4, 2, NaN), nrow=6, byrow=TRUE)
                   NaN,
                          1, NaN,
active_users <- c(1:dim(ratings)[2])</pre>
time_all <- c(rep(NaN, length(active_users)))</pre>
ratings3 <- ratings
for (ac in 1:length(active_users))
 cat("====== user",active_users[ac], "========", "\n","\n")
 ##1
 T1_start <- Sys.time()
 sim <- simple_similarity(ratings, max_score=5, min_score=1, ac)</pre>
 T1_end <- Sys.time()
                                                                 "\n","\n")
          Similar Users =", sim$sim_index,
 cat("Similarity Values =", sim$sim_x,
                                                                 "\n","\n")
 ##2
 T2_start <- Sys.time()
 ratings2 <- Score_replace(ratings, sim_index= sim$sim_index, ac)</pre>
 T2_end <- Sys.time()
                                                                 "\n","\n")
 cat(" Predicted Scores =", ratings2[,ac],
 ##3
 T3_start <- Sys.time()
 predictedItems <- simple_predict(ratings, ratings2, ac)</pre>
 T3_end <- Sys.time()
                                                                "\n","\n")
 cat(" Predicted Items =", predictedItems,
 ##4
 time_all[ac] \leftarrow (T1\_end - T1\_start) + (T2\_end - T2\_start) + (T3\_end - T3\_start)
                                                                "\n","\n")
 cat("
                  Time =", time_all[ac],
 ##5
```

Score_replace

```
ratings3[,ac] <- ratings2[,ac]
}
Mean_Time <- mean(time_all)

cat("========= Mean Time ========", "\n","\n")

cat(" Mean Time =", Mean_Time, "\n","\n")

cat(" Full Matrix =", "\n","\n")

print(ratings3)</pre>
```

Score_replace

Replacing of Neighbor Users' Ratings on Non-Rated Items By The Active User

Description

The ratings of each user that has more similar to the active user are directly replaced in his unseen items.

Usage

```
Score_replace(ratings, sim_index, ac)
```

Arguments

ratings A rating matrix whose rows are items and columns are users.

sim_index Descending sorted indexes based on similarity to the active user who is a vector

of integers.

ac The id of an active user as an integer $(1 \le ac \le length of users)$.

Details

The unseen items of the active user are filled by the ratings of the similar users, respectively. Each element remains unchanged after one placement.

Value

ratings2 A matrix the size of the original user-item matrix in which the active user's

empty elements are filled.

Author(s)

Farimah Houshmand Nanehkaran

Maintainer: Farimah Houshmand Nanehkaran <hoshmandcomputer@gmail.com>

simple_predict 5

References

Gadekula, S. K., Rao, U. P., Vyas, R. K., Dontula, A. L., & Gaikwad, S. V. (2019). *Improved Pearson Similarity for Collaborative Filtering Recommendation System*. In 2019 6th International Conference on Computing for Sustainable Global Development (INDIACom), pp. 1047-1054, IEEE.

Examples

```
ratings <- matrix(c( 2,</pre>
                                        NaN,
                                                       4,
                             5, NaN,
                                               NaN.
                     NaN,
                           NaN,
                                 NaN,
                                               NaN,
                                                       5,
                                         1,
                     NaN,
                            4,
                                    5,
                                        NaN,
                                                     NaN,
                           NaN,
                                 NaN,
                                          5,
                                               NaN,
                                                     NaN,
                          NaN,
                                    2,
                                        NaN,
                                              NaN,
                                                     NaN,
                     NaN,
                             1,
                                 NaN,
                                          4,
                                                 2,
                                                     NaN), nrow=6, byrow=TRUE)
sim <- simple_similarity(ratings, max_score=5, min_score=1, ac=1)</pre>
ratings2 <- Score_replace(ratings, sim_index= sim$sim_index, ac=1)
```

simple_predict

Prediction Unseen Items For The Active User

Description

In the predicted items list, items with more scores replace in top of the list.

Usage

```
simple_predict(ratings, ratings2, ac)
```

Arguments

ratings A rating matrix whose rows are items and columns are users.

ratings2 A matrix the size of the original user-item matrix in which the active user's

empty elements are filled.

ac The id of an active user as an integer $(1 \le ac \le lengthofusers)$.

Details

Collaborative filtering is a recommender system for predicting the missing ratings that an active user might have given to an item. These ratings have been calculated and accumulate in a vector by this function.

Value

predictedItems A sorted vector of predicted items based on the scores.

6 simple_similarity

Author(s)

Farimah Houshmand Nanehkaran

Maintainer: Farimah Houshmand Nanehkaran <hoshmandcomputer@gmail.com>

References

Song, B., Gao, Y., & Li, X. M. (2020, January). *Research on Collaborative Filtering Recommendation Algorithm Based on Mahout and User Model*. In Journal of Physics: Conference Series, Vol. 1437, no. 1, p. 012095, IOP Publishing.

Ramakrishnan, G., Saicharan, V., Chandrasekaran, K., Rathnamma, M. V., & Ramana, V. V. (2020). *Collaborative Filtering for Book Recommendation System*. In Soft Computing for Problem Solving, pp. 325-338, Springer, Singapore.

Examples

```
ratings <- matrix(c( 2,</pre>
                              5, NaN,
                     NaN,
                           NaN,
                                  NaN,
                                         1,
                                               NaN,
                                                        5,
                                    5,
                     NaN,
                             4,
                                        NaN,
                                                4,
                                                     NaN,
                       4,
                           NaN,
                                  NaN,
                                         5,
                                              NaN,
                                                     NaN.
                                    2,
                       5,
                           NaN,
                                        NaN,
                                               NaN,
                                                     NaN,
                     NaN,
                                                     NaN), nrow=6, byrow=TRUE)
                                  NaN,
                                                 2,
sim <- simple_similarity(ratings, max_score=5, min_score=1, ac=1)</pre>
ratings2 <- Score_replace(ratings, sim_index= sim$sim_index, ac=1)</pre>
predictedItems <- simple_predict(ratings, ratings2, ac=1)</pre>
```

simple_similarity

Finding Neighbor Users And Their Similarity Values

Description

Steps of calculating the similarity of one user to an active user:

- 1- Calculating the difference between the desired user ratings with the active user in common items.
- 2- Calculating the similarity value for each common item.
- 3- Calculating the mean value of similarities.

Usage

```
simple_similarity(ratings, max_score=5, min_score=1, ac)
```

Arguments

ratings	A rating matrix whose rows are items and columns are users.
max_score	The maximum range of ratings.
min_score	The minimum range of ratings.
ac	The id of an active user as an integer $(1 \le ac \le length of users)$.

simple_similarity 7

Details

The similarity of the active user with other users is obtained by the following formulas:

$$dif_{(u_i,j)} = |r_{(u_a,j)} - r_{(u_i,j)}|$$

$$sim_{dif(u_i,j)} = \frac{-dif_{(u_i,j)}}{max_score - min_score} + 1$$

$$sim_{(u_a,u_j)} = \frac{\sum_{j=1}^{N_j} sim_{(dif_{(u_i,j)})}}{N_j}$$

j is the row number for the items and i is the column number for the users in the ratings matrix.

 u_i is a ith column user and u_a is an active user.

 $r_{(u_a,j)}$ is the rating of active user in the jth row and $r_{(u_i,j)}$ is the rating of the ith user in the jth row.

 $dif_{(u_i,j)}$ is the difference of the rating for the ith user with the active user in the jth row.

 $sim_{dif_{(u,j)}}$ is the similarity of the ith user with the active user in the jth row.

 $sim_{(u_a,u_i)}$ is the similarity of the user i, with the active user.

 N_j is the number of common items.

For example, suppose active user ratings are: {2, nan, 3, nan, 5} and one user ratings are: {3, 4, nan, nan, 1} then for ratings between 1 and 5:

dif={1, nan, nan, nan, 4} and

 $sim(dif) = {\frac{-1}{5-1} + 1, nan, nan, nan, \frac{-4}{5-1} + 1} = {0.75, nan, nan, nan, 0}$

and mean of sim(dif) is sim=0.375.

Value

An object of class "simple_similarity", a list with components:

call The call used.

sim_x Neighboring user similarity values in descending order.

sim_index Number of columns for neighboring users in descending order of similarity.

Author(s)

Farimah Houshmand Nanehkaran

Maintainer: Farimah Houshmand Nanehkaran <hoshmandcomputer@gmail.com>

References

Mongia, A., & Majumdar, A. (2019). *Matrix completion on multiple graphs: Application in collaborative filtering*. Signal Processing, vol. 165, pp. 144-148.

Hong, B., & Yu, M. (2019). *A collaborative filtering algorithm based on correlation coefficient*. Neural Computing and Applications, vol. 31, no. 12, pp. 8317-8326.

8 simple_similarity

Examples

```
ratings <- matrix(c( 2, 5, NaN, NaN, NaN,
                 NaN, NaN, NaN, 1, NaN,
                 NaN,
                      4,
                            5, NaN,
                                      4, NaN,
                   4, NaN, NaN,
                                 5, NaN, NaN,
                   5, NaN,
                            2, NaN, NaN,
                                           NaN,
                 NaN,
                       1, NaN,
                                 4,
                                      2,
                                           NaN),nrow=6,byrow=TRUE)#items*users
sim <- simple_similarity(ratings, max_score=5, min_score=1, ac=1)</pre>
```

Index

```
* Collaborative Filtering

    CFF-package, 2

* Prediction

    simple_predict, 5

* Similarity

    simple_similarity, 6

* replacement

    Score_replace, 4

CFF-package, 2

Score_replace, 4

simple_predict, 5

simple_similarity, 6
```