

BLM3021 Algorithm Analysis Semester Project Book Recommendation System

Group 2

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General definitions:

Some explanation about the variables used in the code:

```
#define BNUM 8
                        //book numbers
#define RNUM 20
                        //readers number
#define NRNUM 5
                        //new readers numbers
#define LEN 100
winBook (int):
                        will store the final recommended book id.
                        Stores reader's name the user will enter.
target (string):
targetIndex (int):
                        gets the readers id, in other words, the row it is in.
users (string):
                        Stores the users (readers) names.
nUsers (string):
                         Stores the new users' names.
                         Stores the books name.
books (string):
rates (int **):
                         Stores the rating given by x user to y book in a matrix.
nuRates (int **):
                         Stores the rating given by x new user to y book in a matrix.
```

Project aspects:

1.STEP

1. Bir okuyucunun diğer okuyuculara benzerliğinin hesaplanması: Bir okuyucunun diğer okuyuculara benzerliği pearson katsayısı(pearson coefficient) kullanılarak hesaplanacaktır. Hesaplama yapılırken iki kişinin de okuduğu kitaplar için işlem yapılacaktır. Pearson katsayısı kullanılarak a ve b kişilerinin benzerliği aşağıdaki şekilde hesaplanır:

rnum: Row number.

nrnum: New user row number.

This function is used to calculate the mean of read books ratings of a user.

```
181 - float meanSolo(int **rates,int rnum){
182
           int i,cnt=0,sum=0;
183
           float res=0;
184 🖃
           for(i=0;i<BNUM;i++){
185 🗕
               if(rates[rnum][i] != 0 ){
                   sum+=rates[rnum][i];
186
187
                   cnt++;
               }
188
189
           if(cnt!=0) //just in case if something went wrong and cnt still zero
190
191
               res = (float)sum/(float)cnt;
192
193
           return res;
194
```

Calculating the Pearson Coefficient between two users.

```
146  float pearsonCor(int **rates,int rnum,int **nuRates,int nrnum){
147
           int i;
148
           float ra=0,rb=0,sum=0,sdasum=0,sdbsum=0,result=0;
149
          meanCommon(rates,rnum,nuRates,nrnum,&ra,&rb);
150
           ra = meanSolo(rates, rnum);
151
           rb = meanSolo(nuRates,nrnum);
152 =
153 =
           for(i=0;i<BNUM;i++){
               if(rates[rnum][i] != 0 && nuRates[nrnum][i] != 0){
154
                        += ( rates[rnum][i] - ra )*( nuRates[nrnum][i] - rb );
                   sdasum += ( powf((rates[rnum][i] - ra) ,2) );
155
156
                   sdbsum += ( powf((nuRates[nrnum][i] - rb) ,2) );
               }
157
           }
158
159
160
           result = sum/(sqrt(sdasum)*sqrt(sdbsum));
161
           return result;
162
```

Combining these two functions we can achieve the first step of the recommendation process which is to calculate similarity.

2.STEP

 Bir okuyucunun en benzer olduğu k kişinin belirlenmesi: Verilen bir k değeri için bir önceki hesaplamanın sonucuna göre, okuyucuya en benzer k okuyucuyu bulunuz. En benzer k kişiyi bulmak için bütün elemanları sıralamaktan daha efektif bir çözüm planlayınız.

we calculate the similarity between new reader and each reader then inserts the similarity values with their row number in the correct cell, when a bigger similarity value comes, first it slides the table array (k length) in a way that deletes the last element from the table and insert the new value associated with its id in the right cell.

time complexity: worst case - O(k).

```
205 Trable * getKtable(KTable *table,int **rates,int **nuRates,int tar,int k){
           int i,j,l,flag;
206
207
           float similarity;
208
           for(i=0;i<k;i++){
209
                table[i].sim = -10;
210
211
           for(i=0;i<RNUM;i++){
212
213
                similarity=pearsonCor(rates,i,nuRates,tar);
214
                j=0;
215
                flag=0;
216 <del>|</del>
217 <del>|</del>
                while(j < k && flag == 0 ){
                    if(similarity < table[j].sim){</pre>
218
                        j++;
219
                    }else{
220
                        flag = 1;
221
222
223 📥
                for(l=k-1;l>j;l--){
224
                    table[l].sim = table[l-1].sim;
                    table[1].id = table[1-1].id;
225
226
227
                table[j].sim = similarity;
228
                table[j].id = i;
229
230
           return table;
```

3.STEP

3. Okuyucuya yeni kitap önerme : Okuyucuya kitap önermek için en benzer k kişi için aşağıdaki eşitlik ile hesaplama yapınız . Bu eşitlikte N değeri kullanıcının en çok benzediği k kişiyi göstermektedir Hesaplanan pred(a,p) değeri kişinin okumadığı bir kitaba vereceği puanın tahmini değeridir.

Calculates rating prediction for the unread book with (bookIndex) id.

Instead of recalculating the similarities I am just using the K similarity table.

```
236 ☐ | float prediction(KTable *table,int **rates,int **nuRates,int nrnum,int bookIndex,int k){
237
           int i;
238
           float predict,ra,rb,pay=0,payda=0;
239 🖃
           for(i=0; i<k ;i++){
240
               rb=meanSolo(rates,table[i].id);
241
               pay += (table[i].sim) * (rates[table[i].id][bookIndex] - rb);
242
               payda += table[i].sim;
243
244
          ra = meanSolo(nuRates,nrnum);
245
          if (payda!= 0)
246
               predict = ra + pay/payda;
247
248
               exit(1);
249
250
          return predict;
251
```

And In this function, we calculate the predicted rate for each unread book of the new user and returns the winning book id.

```
253 ☐ int getRecommendedBookIndex(KTable *table,int **rates,int **nuRates,int nrnum,int k,char **books){{
254
           float pred, max=-20.0;
           int i,maxInd;
255
256
           for(i=0;i<BNUM;i++) {
257 -
258 -
               if(nuRates[nrnum][i] == 0){
                   pred=prediction(table,rates,nuRates,nrnum,i,k);
259
260 =
                   if(pred>max){
261
                       max = pred;
262
                       maxInd = i;
263
264
                   printf("\t%s\t\t%f\n",books[i],pred);
265
266
           }
267
268
           return maxInd;
269
```

Screenshots:

```
Enter the user name: (EX: NU1)

->NU2

Enter the similarity table capacity (k): 3

The (3) most similar persons are:

1. U2 user's similarity rate with NU2 is: 0.961

2. U1 user's similarity rate with NU2 is: 0.952

3. U11 user's similarity rate with NU2 is: 0.821

The prediction rate of the unread books:

TRUE BELIEVER 2.303307

THE KITE RUNNER 2.010123

HARRY POTTER 1.955190

According to our statistics..

NU2 user should read next: TRUE BELIEVER
```

```
Enter the user name: (EX: NU1)
->NU3

Enter the similarity table capacity (k): 3

The (3) most similar persons are:
1. U16 user's similarity rate with NU3 is: 0.511
2. U14 user's similarity rate with NU3 is: 0.445
3. U15 user's similarity rate with NU3 is: 0.309

The prediction rate of the unread books:

THE WORLD IS FLAT 0.822540
MY LIFE SO FAR 0.770752

According to our statistics..
NU3 user should read next: THE WORLD IS FLAT
```

```
Enter the user name: (EX: NU1)

->NU4

Enter the similarity table capacity (k): 3

The (3) most similar persons are:

1. U2 user's similarity rate with NU4 is: 0.990

2. U13 user's similarity rate with NU4 is: 0.988

3. U16 user's similarity rate with NU4 is: 0.841

The prediction rate of the unread books:

THE TAKING
2.150650
RUNNY BABBIT
2.343484

According to our statistics..

NU4 user should read next: RUNNY BABBIT
```

```
Enter the similarity table capacity (k) : 3
 The (3) most similar persons are:
             U9 user's similarity rate with NU5 is: 0.953
U18 user's similarity rate with NU5 is: 0.741
U6 user's similarity rate with NU5 is: 0.643
The prediction rate of the unread books:
              TRUE BELIEVER
              THE KITE RUNNER
              HARRY POTTER
                                                           3.039378
            user should read next: HARRY POTTER
inter the user name: (EX: NU1)
            U2 user's similarity rate with NU4 is: 0.990
U13 user's similarity rate with NU4 is: 0.988
U16 user's similarity rate with NU4 is: 0.841
U10 user's similarity rate with NU4 is: 0.817
U3 user's similarity rate with NU4 is: 0.766
U12 user's similarity rate with NU4 is: 0.765
U20 user's similarity rate with NU4 is: 0.651
U18 user's similarity rate with NU4 is: 0.623
              THE TAKING
According to our statistics..
NU4 user should read next: RUNNY BABBIT
```