

A53271697
Bosi Cheng

1.

```
businessCount_train,businessCount_valid = defaultdict(int),defaultdict(int)

negative_valid=defaultdict(list)

train,positive_valid=defaultdict(list),defaultdict(list)

user_set,item_set=set(),set()

totalPurchases = 0

for l in readGz("train.json.gz"):

    if totalPurchases<100000:

        user,business = l['reviewerID'],l['itemID']

        user_set.add(user)

        item_set.add(business)

        train[user].append(business)

        businessCount_train[business] += 1

        totalPurchases += 1

    else:

        user,business = l['reviewerID'],l['itemID']

        user_set.add(user)

        item_set.add(business)
```

```

        positive_valid[user].append(business)

        businessCount_valid[business] += 1

        totalPurchases += 1

valid_if_purchase=0

while valid_if_purchase<100000:

    random_user=user_set.pop()

    user_set.add(random_user)

    random_item=item_set.pop()

    item_set.add(random_item)

    if random_item not in train[random_user] :

        if random_item not in negative_valid[random_user]:

            negative_valid[random_user].append(random_item)

            valid_if_purchase+=1

mostPopular = [(businessCount_train[x], x) for x in businessCount_train]

mostPopular.sort()

mostPopular.reverse()

return1 = set()

```

```
count = 0

for ic, i in mostPopular:

    count += ic

    return1.add(i)

    if count > totalPurchases/2: break


correction=0

for user in positive_valid.keys():

    for item in positive_valid[user]:

        if item in return1:

            correction+=1


for user in negative_valid.keys():

    for item in negative_valid[user]:

        if item not in return1:

            correction+=1


print(len(positive_valid))

print(len(negative_valid))

print(correction)
```

```
print(float(correction/200000))
```

```
36402  
39239  
100472  
0.50236
```

2.

No, it is not the best choice. This model has nothing to do with the training set. We should include the non-purchase pair only in the validation sets.

```
businessCount_train,businessCount_valid = defaultdict(int),defaultdict(int)
```

```
negative_valid=defaultdict(list)
```

```
train,positive_valid=defaultdict(list),defaultdict(list)
```

```
user_set,item_set=set(),set()
```

```
train_user_set=set()
```

```
totalPurchases = 0
```

```
valid_user_set,valid_item_set=set(),set()
```

```
for l in readGz("train.json.gz"):
```

```
    if totalPurchases<100000:
```

```
        user,business = l['reviewerID'],l['itemID']
```

```
        if user not in user_set:
```

```
            user_set.add(user)
```

```
        if business not in item_set:
```

```
            item_set.add(business)
```

```

train[user].append(business)

businessCount_train[business] += 1

totalPurchases += 1

else:

    user,business = I["reviewerID"],I["itemID"]

    if user not in valid_user_set:

        valid_user_set.add(user)

    if business not in valid_item_set:

        valid_item_set.add(business)

    positive_valid[user].append(business)

    businessCount_valid[business] += 1

    totalPurchases += 1


mostPopular = [(businessCount_train[x], x) for x in businessCount_train]

mostPopular.sort()

mostPopular.reverse()


return1 = set()

count = 0

for ic, i in mostPopular:

```

```

count += ic

return1.add(i)

if count > totalPurchases/2: break


valid_if_purchase=0

while valid_if_purchase<100000:

    random_user=valid_user_set.pop()

    valid_user_set.add(random_user)

    random_item=valid_item_set.pop()

    valid_item_set.add(random_item)

    if random_item not in train[random_user] :

        if random_item not in negative_valid[random_user]:

            negative_valid[random_user].append(random_item)

            valid_if_purchase+=1


correction=0

for user in positive_valid.keys():

    for item in positive_valid[user]:

        if item in return1:

```

```

        correction+=1

print(correction)

for user in negative_valid.keys():

    for item in negative_valid[user]:

        if item not in return1:

            correction+=1

print(len(positive_valid))

print(len(negative_valid))

print(correction)

print(float(correction/200000))

```

```

115673
0.578365

```

3.
while loading the json, create a dict that matches every item and its categories item_categories[item].

```

repeat_purchase=defaultdict()

for user in user_set:

    temp=set()

    for item in purchase[user]:

        for category in item_categories[item]:

```

```

if category in temp:

    repeat_purchase[user]=True

    temp.add(category)


if user not in repeat_purchase.keys():

    repeat_purchase[user]=False

```


4.

My username is bosicheng. But I can't participate because I'm not invited. (why?) I swear it's my UCSD email.



bosicheng

Joined a day ago · last seen in the past day



Competitions
Novice

[Home](#)
[Competitions](#)
[Kernels](#)
[Discussion](#)
[Datasets](#)
[...](#)
[Edit Profile](#)

Account (User ID 2476516) [Edit](#)

User Name

bosicheng

Your username cannot be changed.

Display Name

bosicheng

The name users will see publicly. We recommend using your full name.

Email Address

bocheng@eng.ucsd.edu

Here's the issue:

[Overview](#)
[Data](#)
[Kernels](#)
[Discussion](#)
[Leaderboard](#)
[Rules](#)

i This is a limited-participation competition. Only invited users may participate.

5.


```

allRatings_train = []

userRatings = defaultdict(list)

totalPurchase=0

sum_mse=0

for l in readGz("train.json.gz"):

    user,business = l['reviewerID'],l['itemID']

    if totalPurchase<100000:

        allRatings_train.append(l['rating'])

        userRatings[user].append(l['rating'])

        totalPurchase+=1

    else:

        if totalPurchase==100000:

            globalAverage = sum(allRatings_train) / len(allRatings_train)

            sum_mse+=abs(l['rating']-globalAverage)*abs(l['rating']-globalAverage)

            totalPurchase+=1

globalAverage = sum(allRatings_train) / len(allRatings_train)

userAverage = {}

for u in userRatings:

    userAverage[u] = sum(userRatings[u]) / len(userRatings[u])

```

```
print(globalAverage)
```

```
print(sum_mse/100000)
```

```
4.232
```

```
1.222481119999121
```

6.

```
import gzip
```

```
import numpy as np
```

```
from collections import defaultdict
```

```
#get the data
```

```
def readGz(f):
```

```
    for l in gzip.open(f):
```

```
        yield eval(l)
```

```
allRatings = []
```

```
data=[]
```

```
userRatings = defaultdict(list)
```

```
for l in readGz("train.json.gz"):
```

```
    user,business = l['reviewerID'],l['itemID']
```

```

allRatings.append(l['rating'])

userRatings[user].append(l['rating'])

data.append(l)


#shuffle the data

np.random.shuffle(data)


def getPrediction2(alpha,uB,iB,i,j,y_u,y_i,uMap,iMap):

    rating = alpha + (uB[i] if i in uB else 0) + (iB[j] if j in iB else 0)

    if i in uMap and j in iMap:

        rating +=np.inner(y_u[uMap[i]],y_i[iMap[j]])

    return rating


#Method to Train The Latent Factor Model. This method doesn't use any Machine Learning library.

def trainLFModel(lam,tData,vData,trials):

    uTrainDict = defaultdict(lambda: defaultdict(int))

    iTrainDict = defaultdict(lambda: defaultdict(int))

    uValidDict = defaultdict(lambda: defaultdict(int))

    # iValidDict = defaultdict(lambda: defaultdict(int))

    uB = defaultdict(float)

```

```
iB = defaultdict(float)

uMap = defaultdict(int)

uCount=0

iMap = defaultdict(int)

iCount=0

for i in tData:

    user, item, rating = i['reviewerID'], i['itemID'], i['rating']

    uTrainDict[user][item] = rating

    iTrainDict[item][user] = rating

    if user not in uMap:

        uMap[user]=uCount

        uCount+=1

    if item not in iMap:

        iMap[item]=iCount

        iCount+=1

for i in vData:

    user, item, rating = i['reviewerID'], i['itemID'], i['rating']

    uValidDict[user][item] = rating


y_u=np.random.normal(scale=1,size=(len(uTrainDict),1))
```

```

y_i=np.random.normal(scale=1,size=(len(iTrainDict),1))

alpha = 0

totalTrials=trials

for _ in range(totalTrials):

    alpha=0

    for i in uTrainDict:

        for j in uTrainDict[i]:

            alpha += uTrainDict[i][j] - uB[i] - iB[j] - np.inner(y_u[uMap[i]],y_i[iMap[j]])

    alpha /= len(tData)

    print(alpha)

    for i in uTrainDict:

        uB[i] = 0

        for j in uTrainDict[i]:

            uB[i] += uTrainDict[i][j] - alpha - iB[j] - np.inner(y_u[uMap[i]],y_i[iMap[j]])

        uB[i] /= (lam + len(uTrainDict[i]))

    for j in iTrainDict:

        iB[j] = 0

        for i in iTrainDict[j]:

            iB[j] += iTrainDict[j][i] - alpha - uB[i] - np.inner(y_u[uMap[i]],y_i[iMap[j]])

```

```

iB[j] /= (lam + len(iTrainDict[j]))

for i in uTrainDict:

    for lf in range(1):

        y_u[uMap[i]][lf] = 0

        for j in uTrainDict[i]:

            y_u[uMap[i]][lf] += y_i[iMap[j]][lf]*(uTrainDict[i][j] - alpha - iB[j]
+y_i[iMap[j]][lf]*y_i[iMap[j]][lf]-np.inner(y_u[uMap[i]],y_i[iMap[j]]))

            y_u[uMap[i]][lf] /= (lam + y_i[iMap[j]][lf]*y_i[iMap[j]][lf])

        for j in iTrainDict:

            for lf in range(1):

                y_i[iMap[j]][lf] = 0

                for i in iTrainDict[j]:

                    y_i[iMap[j]][lf] += y_u[uMap[i]][lf]*(uTrainDict[i][j] - alpha - uB[i] -
np.inner(y_u[uMap[i]],y_i[iMap[j]])) +y_u[uMap[i]][lf]*y_u[uMap[i]][lf] )

                    y_i[iMap[j]][lf] /= (lam + y_u[uMap[i]][lf]*y_u[uMap[i]][lf])

vMSE = 0

for i in uValidDict:

    for j in uValidDict[i]:

#         vMSE += ((alpha + (uB[i] if i in uB else 0) + (iB[j] if j in iB else 0) - uValidDict[i][j]) **2)

```

```

        vMSE += ((getPrediction2(alpha,uB,iB,i,j,y_u,y_i,uMap,iMap) - uValidDict[i][j]) **2)

    vMSE /= len(vData)

    print (vMSE)

    return vMSE,alpha,uB,iB,uMap,iMap

print("done")

tData=data[:100000]

vData=data[100000:]

lamdas=[1,2,3,5]

trials=[2]

for i in lamdas:

    tempvMSE=1

    for t in trials:

        tempvMSE,alpha,uB,iB,uMap,iMap=trainLFModel(i,tData,vData,t)

    print ("MSE:"+str(tempvMSE))

```

```
MSE: 1.2608578988522912
```

```
7. biggest: 'U536579649', 'U910050838', 'l471768594'
```

```
smallest: 'l011994385', 'U926666668'
```

8.

when $\lambda=5$:

MSE: 1.1453497407928057