## baseline model

## November 6, 2019

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In [2]: import pandas as pd
        import numpy as np
        import random
        import matplotlib.pyplot as plt
        from surprise import Dataset
        from surprise import accuracy
        from surprise import Reader
        import time
        import sklearn
        from sklearn.model_selection import train_test_split
        from random import shuffle
        import seaborn as sns
        from surprise.prediction_algorithms.baseline_only import BaselineOnly
In [3]: train_ratings = pd.read_csv("train_data.csv")
        test_ratings = pd.read_csv("test_data.csv")
In [4]: print('train_minimum_rating',min(train_ratings['rating']))
        print('train_maximum_rating', max(train_ratings['rating']))
        print('test_minimum_rating',min(train_ratings['rating']))
        print('test_maximum_rating', max(train_ratings['rating']))
('train_minimum_rating', 0.5)
('train_maximum_rating', 5.0)
('test minimum rating', 0.5)
('test_maximum_rating', 5.0)
In [18]: def coverage(threshold1, threshold2,prediction):
             start time=time.time()
             predictions = pd.DataFrame(prediction)
             pred = predictions.groupby('uid')
             df1= pred.apply(lambda x: x.sort_values(by=["est"],ascending=False))
             df2=df1.reset_index(drop=True)
             df3 = df2.groupby('uid').head(10)
             s1=df3[df3['r_ui'] > threshold1].groupby('uid')['r_ui'].count().reset_index()
             s2 = df3.pivot_table(index=['uid'],aggfunc='size').reset_index()
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s2.columns = ['uid', 'counts']
             df=pd.merge(s1, s2, on='uid')
             # #number of high true rating(larger than 4) devided by top N predictions
             df['rate']=df['r_ui']/df['counts']
             user_coverage=float(sum(df['rate']> threshold2))/df3['uid'].nunique()
             item=df3.groupby('iid').apply(lambda x: x.sort_values(by=["est"],ascending=False)
             s=item[item['r_ui'] > threshold1].groupby('iid')['r_ui'].count().reset_index()
             ss = item.pivot_table(index=['iid'],aggfunc='size').reset_index()
             ss.columns = ['iid','counts']
             dff=pd.merge(s, ss, on='iid')
             dff['rate'] = dff['r_ui'] / dff['counts']
             item_coverage=float(sum(dff['rate']> threshold2))/df3['iid'].nunique()
             catalog_coverage = float(df3['iid'].nunique())/predictions['iid'].nunique()
             end time=time.time()
             duration=end_time-start_time
             return user coverage, item coverage, catalog coverage, duration
In [5]: reader = Reader(rating scale=(0.5, 5))
        ratings = Dataset.load_from_df(train_ratings, reader)
In [6]: raw_ratings = ratings.raw_ratings
        random.seed(42)
        shuffle(raw_ratings)
In [7]: ratings.raw_ratings = raw_ratings
In [8]: copy_ratings=ratings
In [11]: test_rating = Dataset.load_from_df(test_ratings, reader)
         test_raw_rating=test_rating.raw_ratings
Fit a baseline model
In [14]: print('Using SGD')
         bsl_options = {'method': 'sgd',
                        'learning_rate': .00005,
         algo = BaselineOnly(bsl_options=bsl_options)
         algo.fit(ratings.build_full_trainset())
Using SGD
Estimating biases using sgd...
```

## user-coverage, item-coverage, catalog, running time on the test set