matrix_factorization_svd

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In [1]: import pandas as pd
        import numpy as np
In [2]: training_files = ['ml-100k/u1.base', 'ml-100k/u2.base', 'ml-100k/u3.base', 'ml-100k/u4.base']
        validation_file = 'ml-100k/u5.base'
        out_file = 'mf-k.csv'
        k=5
In [3]: frames = []
        for training_file in training_files:
            frames.append(pd.read_csv(training_file, sep='\t', header=0,
                                names=['userId', 'movieId', 'rating', 'timestamp'], engine='pythe
        ratings = pd.concat(frames)
        ratings.head()
Out[3]:
           userId movieId rating timestamp
        0
                          2
                                    876893171
                1
        1
                          3
                1
                                  4 878542960
                          4
                                  3 876893119
        3
                1
                          5
                                     889751712
                          7
                                  4 875071561
In [4]: # R_df = ratings.pivot(index = 'userId', columns = 'movieId', values = 'rating').fillna
        R_df = pd.pivot_table(ratings, values='rating', index='userId', columns='movieId').fillna
        R_df.head()
Out[4]: movieId 1
                                          5
                                                                          10
        userId
                  0.0
                        3.0
                               4.0
                                     3.0
                                           3.0
                                                  5.0
                                                        4.0
                                                              1.0
                                                                    5.0
                                                                           3.0
                                                                                . . .
        2
                  4.0
                         0.0
                               0.0
                                     0.0
                                           0.0
                                                  0.0
                                                        0.0
                                                              0.0
                                                                    0.0
                                                                           2.0
        3
                  0.0
                         0.0
                               0.0
                                     0.0
                                           0.0
                                                  0.0
                                                        0.0
                                                              0.0
                                                                    0.0
                                                                           0.0
        4
                  0.0
                         0.0
                               0.0
                                     0.0
                                           0.0
                                                  0.0
                                                        0.0
                                                              0.0
                                                                    0.0
                                                                           0.0 ...
        5
                  4.0
                         3.0
                                                  0.0
                                                        0.0
                                                                    0.0
                                                                           0.0 ...
                               0.0
                                     0.0
                                           0.0
                                                              0.0
                1673 1674 1675
                                                 1678
                                                       1679
                                                             1680
                                                                   1681
                                                                          1682
        movieId
                                    1676
                                          1677
        userId
        1
                  0.0
                         0.0
                               0.0
                                     0.0
                                           0.0
                                                  0.0
                                                        0.0
                                                              0.0
                                                                    0.0
                                                                           0.0
                  0.0
                         0.0
                               0.0
                                     0.0
                                           0.0
                                                  0.0
                                                        0.0
                                                              0.0
                                                                    0.0
                                                                           0.0
```

```
3
                  0.0
                        0.0
                              0.0
                                    0.0
                                          0.0
                                                0.0
                                                      0.0
                                                            0.0
                                                                  0.0
                                                                        0.0
                                                                        0.0
        4
                  0.0
                        0.0
                              0.0
                                    0.0
                                          0.0
                                                0.0
                                                      0.0
                                                            0.0
                                                                  0.0
        5
                  0.0
                        0.0
                              0.0
                                    0.0
                                          0.0
                                                0.0
                                                      0.0
                                                            0.0
                                                                  0.0
                                                                        0.0
        [5 rows x 1682 columns]
In [5]: # normalized matrix
       R = R_df.as_matrix()
        user_ratings_mean = np.mean(R, axis = 1)
        R_demeaned = R - user_ratings_mean.reshape(-1, 1)
In [6]: \# SVD decomposition
        from scipy.sparse.linalg import svds
        U, sigma, Vt = svds(R_demeaned, k = 50)
        sigma = np.diag(sigma)
In [7]: # Prediction
        all_user_predicted_ratings = np.dot(np.dot(U, sigma), Vt) + user_ratings_mean.reshape(
        preds_df = pd.DataFrame(all_user_predicted_ratings, columns = R_df.columns)
In [8]: # top k recommendations for users
        top_k_list = {}
        for user in preds_df.index.unique():
            rated_movie = ratings[ratings['userId'] == (user)]['movieId']
            pred_rating = preds_df[[x for x in preds_df.columns.values if x not in rated_movie
            top_k_list[user] =pred_rating.index.values.tolist()[:k]
In [9]: top_k_df = pd.DataFrame.from_dict(top_k_list, orient='index')
       top_k_df.head()
Out[9]:
                  1
                       2
                            3
          100
        0
               176
                      89
                           12 135
        1 286
               302 313 275 285
                340 333 328
          268
                               327
           50
               748
                     301
                         269
                               289
          168
               181 228 222 208
In [10]: top_k_df.to_csv(out_file)
In [11]: # Validation, not sure how to do yet
         validation_rating = pd.read_csv(validation_file, sep='\t', header=0,
                                names=['userId', 'movieId', 'rating', 'timestamp'], engine='pyt
        validation_rating.head()
Out[11]:
            userId movieId rating timestamp
                                  3 876893171
        0
                 1
                          2
        1
                 1
                          4
                                  3 876893119
         2
                 1
                          5
                                  3 889751712
         3
                 1
                          6
                                  5 887431973
```

4 875071561

1

7

```
In [12]: def predict(row):
            return preds_df.loc[row['userId'] - 1, row['movieId']]
        validation_rating['pred rating'] = validation_rating.apply (lambda row: predict(row)
        validation_rating.head()
Out[12]:
           userId movieId rating timestamp pred rating
                         2
                                 3 876893171
                                                  2.911756
        1
                1
                         4
                                 3 876893119
                                                  3.180375
                                                  1.615407
        2
                1
                         5
                                 3 889751712
                1
                         6
                                 5 887431973
                                                  1.591248
                1
                         7
                                 4 875071561
                                                 3.495347
In [13]: from sklearn.metrics import mean_squared_error
        mean_squared_error(validation_rating['rating'], validation_rating['pred rating'])
Out[13]: 3.4196413821591585
```