

Recommenders

```
In [1]: import difflib
import random
import pandas as pd
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
from surprise import Reader, Dataset
from surprise.prediction_algorithms import SVD, KNNWithZScore
from surprise.model_selection import train_test_split
from collections import defaultdict
import warnings
warnings.simplefilter(action='ignore', category=FutureWarning)
```

Importing and Splitting data

```
In [2]: full = pd.read_csv('../Data/filtered-cleaned')
full = full.drop(columns = 'Unnamed: 0')
full['title'] = full.title.str.extract('([a-zA-Z\s]+)')
min_cols = full[['userId', 'movieId', 'rating']]
min_cols = min_cols.sample(1000000)
smaller = min_cols.sample(50000)
```

```
In [3]: reader = Reader()
data = Dataset.load_from_df(min_cols, reader)
datasmall = Dataset.load_from_df(smaller, reader)
kdata = datasmall.build_full_trainset()
trainset, testset = train_test_split(data, test_size = 0.10)
```

```
In [4]: trainset = data.build_full_trainset()
print('Unique users: ', trainset.n_users, '\n')
print('Unique Movies: ', trainset.n_items)
```

Unique users: 99078

Unique Movies: 6044

```
In [5]: # Recommendations Based on Movie Title input
```

```
In [6]: knn_zscore = KNNWithZScore(sim_options={'name': 'pearson', 'user_based': False})
knn_zscore.fit(trainset)
```

Computing the pearson similarity matrix...
Done computing similarity matrix.

```
Out[6]: <surprise.prediction_algorithms.knns.KNNWithZScore at 0x7fe27d3221f0>
```

```
In [7]: def out_neighbors(m_id):
    tsr_inner_id = knn_zscore.trainset.to_inner_id(m_id)
    tsr_neighbors = knn_zscore.get_neighbors(tsr_inner_id, k=5)
    neighbors = full[full.movieId.isin([knn_zscore.trainset.to_raw_id(inner_id)
                                     for inner_id in tsr_neighbors])]
    print(list(neighbors.title.unique()))

def neighborer(df):
    movie_title = input("Enter Movie Title... Spelled correctly" )
    searchable = df.copy()
    searchable['search'] = searchable['title']
    searchable['search'] = searchable['search'].astype('string')
    searchable['search'] = searchable['search'].str.lower()
    searchable['search'] = searchable['search'].str.replace(' ', '')
    movie_title = movie_title.lower()
    movie_title = movie_title.replace(' ', '')
    for i, title in searchable.search.items():
        if movie_title == title:
            m_id = searchable.iloc[i]['movieId']
            name = full.loc[full['movieId'] == m_id, 'title'].iloc[0]
            out_neighbors(m_id)
            return m_id, name
```

```
In [8]: print('Sample Movies: ' , list(full.title.sample(5)))
```

Sample Movies: ['Finding Neverland ', 'Brother Bear ', 'Motorcycle Diaries', 'Love', 'Dumb and Dumber To ']

Enter Title Here -----
-

```
In [9]: neighborer(full)
```

Enter Movie Title... Spelled correctlyBrokeback Mountain
 ['Won', 'Joy Ride ', 'Icarus ', 'Who Am I ', 'Red State ']

```
Out[9]: (39183, 'Brokeback Mountain ')
```

Recommendations From Rating Input

```
In [10]: def movie_rater(movie_df,num, genre=None):
        userID = 1000
        rating_list = []
        while num > 0:
            if genre:
                movie = movie_df[movie_df['genres'].str.contains(genre)].sample(1)
            else:
                movie = movie_df.sample(1)
            print(movie.title)
            rating = input('How do you rate this movie on a scale of 1-5, press 1-5: ')
            if rating == 'n':
                continue
            else:
                rating_one_movie = {'userId':userID,'movieId':movie['movieId']}
                rating_list.append(rating_one_movie)
                num -= 1
        return rating_list
```

Rate Movies Here -----
-

Run Next 2 Cells

```
In [ ]: user_rating = movie_rater(full, 5)

1104586      A Space Odyssey
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
n
1857414      Snatch
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
n
243096       Thing
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
4
823367       Day After Tomorrow
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
4
13674        Burn After Reading
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
n
1899284       Whip It
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
n
1861565       Birthday Girl
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
n
204015        Avengers
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
5
364397        Kingsman
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
4
1909652       Oz the Great and Powerful
Name: title, dtype: object
How do you rate this movie on a scale of 1-5, press n if you have not see
n :
n
957452        X
Name: title, dtype: object
```

```

In [13]: print("Please Wait... looking for awesome movies \n \n")

new_ratings_df = min_cols.append(user_rating,ignore_index=True)
new_data = Dataset.load_from_df(new_ratings_df,reader)

svd_ = SVD(n_factors= 50, reg_all=0.05)
svd_.fit(new_data.build_full_trainset())

list_of_movies = []
for m_id in min_cols['movieId'].unique():
    list_of_movies.append( (m_id,svd_.predict(1000,m_id)[3]))

ranked_movies = sorted(list_of_movies, key=lambda x:x[1], reverse=True)

def recommended_movies(user_ratings,movie_title_df,n):
    for idx, rec in enumerate(user_ratings):
        title = movie_title_df.loc[movie_title_df['movieId'] == int(rec)]
        print('Recommendation # ', idx+1, ': ', title, '\n')
        n-= 1
    if n == 0:
        break

print('Movie Recs: \n')
recommended_movies(ranked_movies,full,5)

```

Please Wait... looking for awesome movies

Movie Recs:

Recommendation # 1 : HyperNormalisation

Recommendation # 2 : Planet Earth

Recommendation # 3 : Planet Earth II

Recommendation # 4 : 0

Recommendation # 5 : Blue Planet II

In []: