

# Non-negative Matrix Factorization

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In this project, non-negative matrix factorization is used to train the recommendation algorithm.

**Dataset** : Movie Recommendation Data Sets

**Source** : <http://files.grouplens.org/datasets/movielens/ml-latest-small.zip>

## 0.1 Preliminary works

### 1. Import libraries

```
[ ]: pip install scikit_surprise
```

```
[ ]: import pandas as pd
from surprise import NMF, Dataset, Reader
from surprise.model_selection import train_test_split, cross_validate
```

### 2. Get a dataset

```
[ ]: from google.colab import files
data_to_load = files.upload()
```

<IPython.core.display.HTML object>

Saving ratings.csv to ratings.csv

```
[ ]: df = pd.read_csv('ratings.csv')
del df['timestamp']
df.head()
```

```
[ ]: 
```

	userId	movieId	rating
0	1	1	4.0
1	1	3	4.0
2	1	6	4.0
3	1	47	5.0
4	1	50	5.0

## 0.2 Non-negative Matrix Factorization algorithm

```
[ ]: reader = Reader(rating_scale=(1.0, 5.0))
data = Dataset.load_from_df(df, reader)
```

```
[ ]: # Fit the algorithm to the full dataset
train = data.build_full_trainset()

algo = NMF()
algo.fit(train)
```

```
[ ]: <surprise.prediction_algorithms.matrix_factorization.NMF at 0x7f74c8cf6518>
```

```
[ ]: #Run 5 fold CV
cross_validate(algo, data, measures=['RMSE', 'MAE'], cv=5, verbose=True)
```

Evaluating RMSE, MAE of algorithm NMF on 5 split(s).

	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	Mean	Std
RMSE (testset)	0.9240	0.9248	0.9175	0.9235	0.9097	0.9199	0.0057
MAE (testset)	0.7094	0.7083	0.7031	0.7074	0.6970	0.7050	0.0045
Fit time	6.29	6.54	6.67	6.62	6.30	6.48	0.16
Test time	0.23	0.26	0.11	0.11	0.25	0.19	0.07

```
[ ]: {'fit_time': (6.285089015960693,
 6.536697149276733,
 6.674468755722046,
 6.62369966506958,
 6.298094272613525),
'test_mae': array([0.70939679, 0.70830663, 0.70307781, 0.7074144 ,
0.69702027]),
'test_rmse': array([0.92404159, 0.92479641, 0.91750888, 0.92347584,
0.90965239]),
'test_time': (0.2251298427581787,
0.2599964141845703,
0.11418485641479492,
0.11135745048522949,
0.24715781211853027)}
```

```
[ ]: # Recommendation for a user or two as testing

pred_1 = algo.predict(378, 78499, r_ui=4, verbose=True)
pred_2 = algo.predict(453, 2324, r_ui=5, verbose=True)
```

```
user: 378      item: 78499      r_ui = 4.00    est = 3.75    {'was_impossible':
False}
user: 453      item: 2324      r_ui = 5.00    est = 4.19    {'was_impossible':
False}
```