

nmf-implementation

January 23, 2018

1 Fundamentals of Machine Learning - Exercise 9

Dominique Cheray & Manuel Krämer

```
In [1]: import math
import matplotlib.pyplot as plt
import numpy as np
from sklearn.datasets import load_digits
from sklearn import decomposition
from sklearn import preprocessing

digits = load_digits()
X = digits["data"] / 255.
Y = digits["target"]
```

1.1 4 Implementation - Non-negative matrix factorization

```
In [2]: def non_negative(data, num_components):

    X = data
    p = num_components
    m,n = data.shape

    #initialize H and W with positive normal random distribution
    H = np.absolute(np.random.standard_normal(size=(p,n)))
    W = np.absolute(np.random.standard_normal(size=(m,p)))

    # Iteration limit
    limit = 1000

    for i in range(limit):
        H = H * (np.dot(W.T,X)) / np.clip((np.dot(W.T,np.dot(W,H))), a_min = 0.01, a_max = 10)
        W = W * (np.dot(X,H.T)) / np.clip((np.dot(W, np.dot(H,H.T))), a_min = 0.01, a_max = 10)

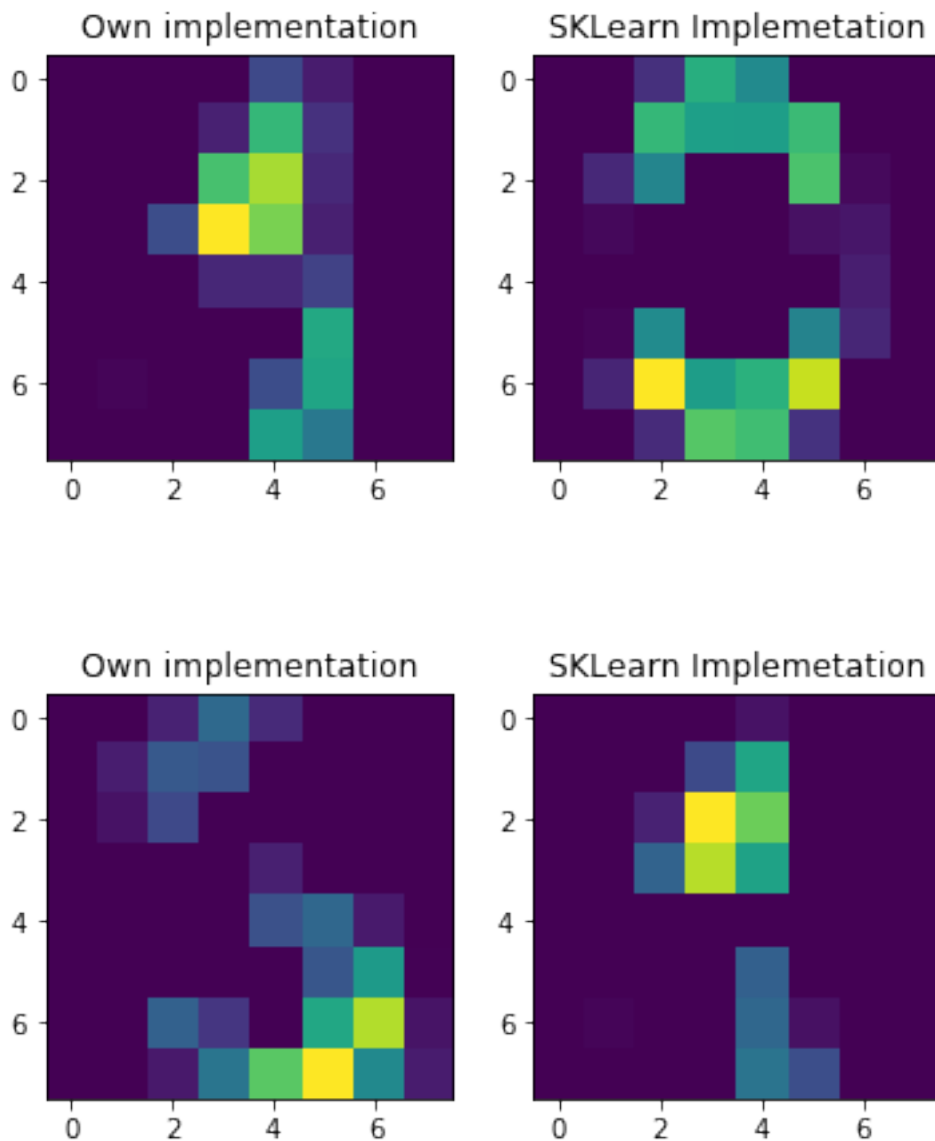
    return W, H

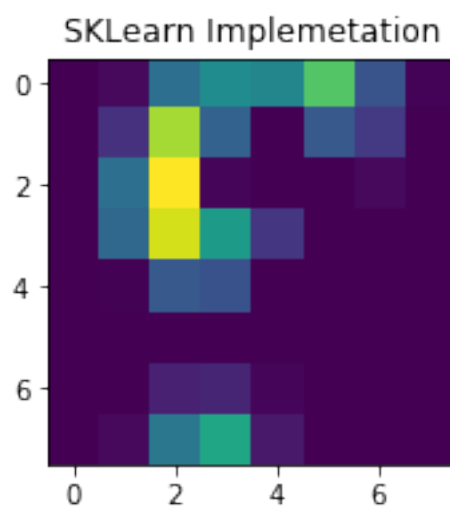
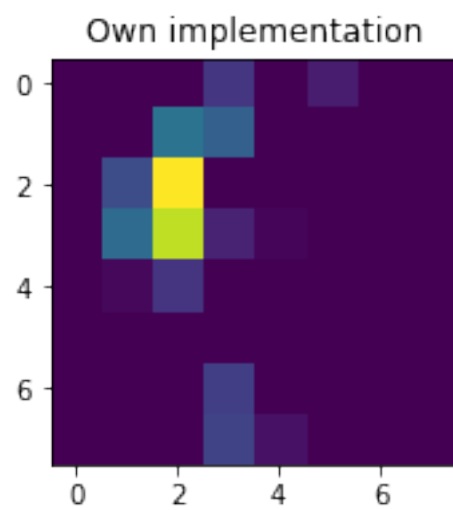
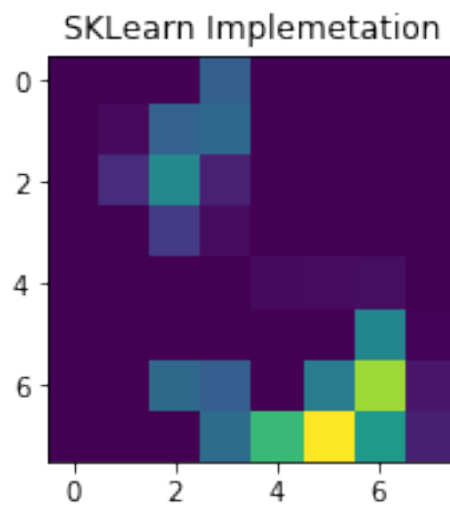
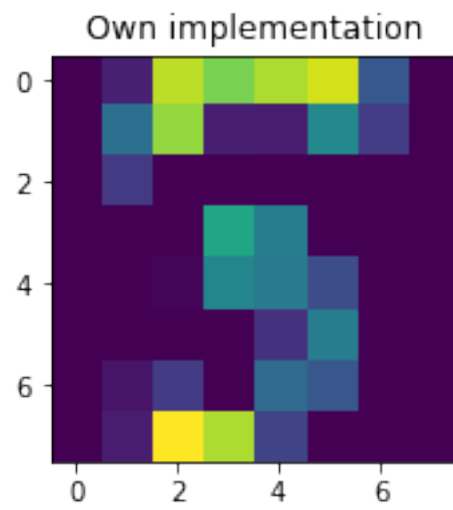
In [3]: W, H = non_negative(X, 10)
print(W.shape,H.shape)
```

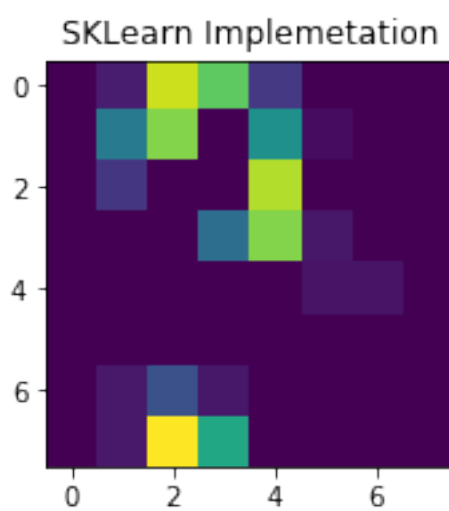
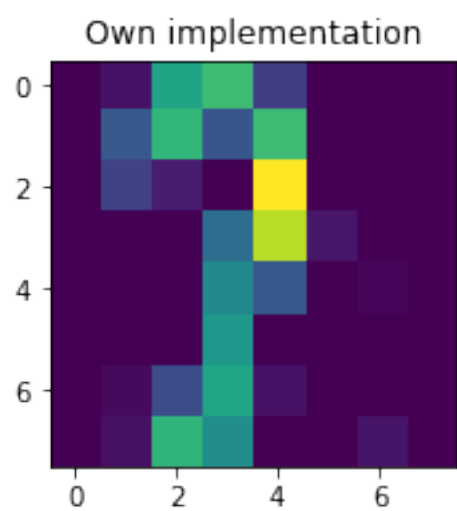
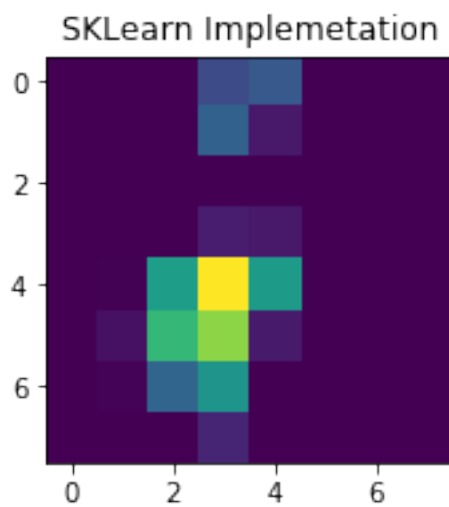
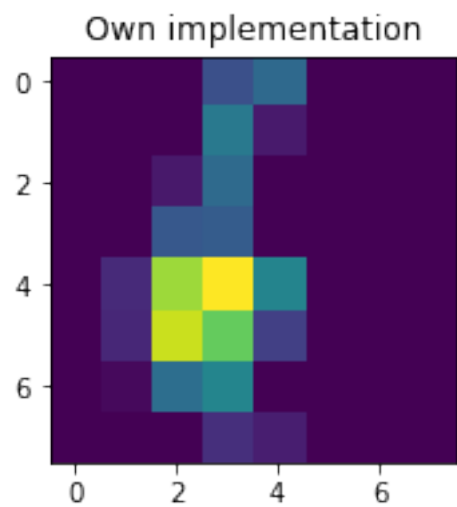
(1797, 10) (10, 64)

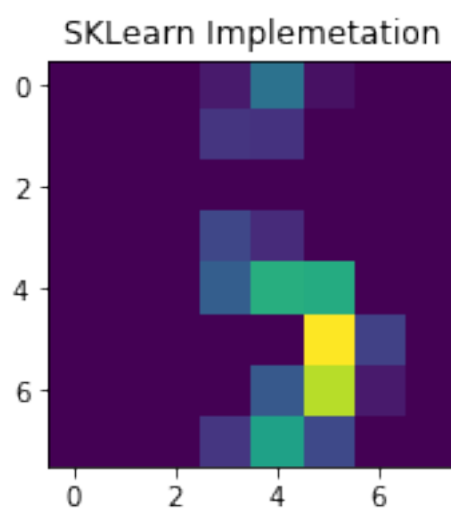
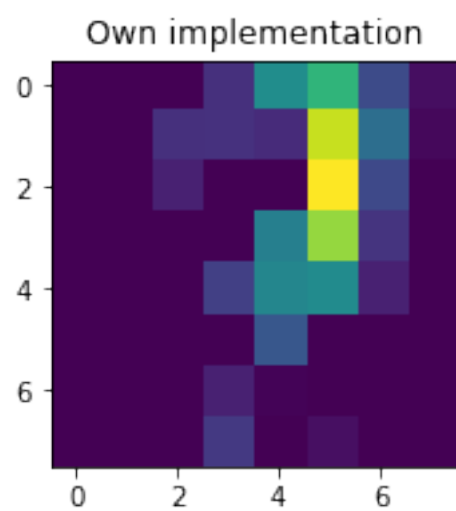
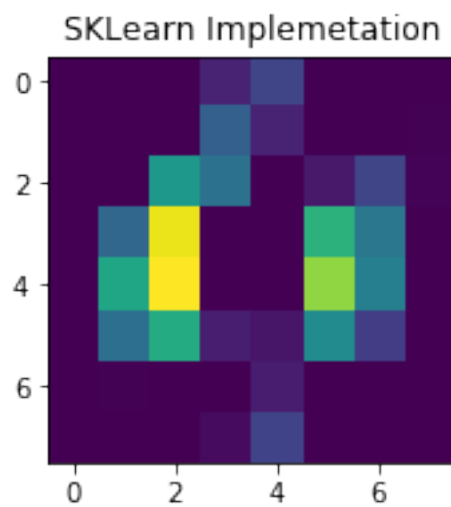
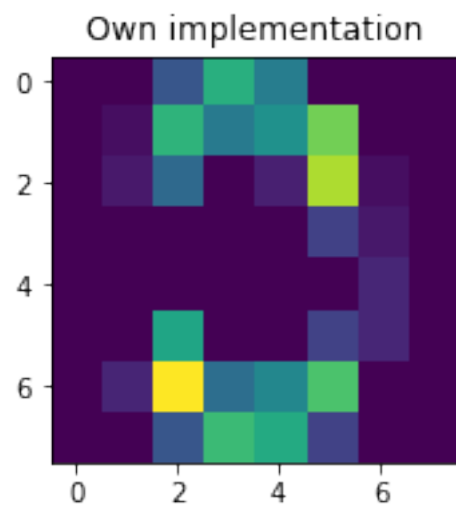
```
In [4]: model = decomposition.NMF(n_components=10, init='random', random_state=0)
        W_skl = model.fit_transform(X)
        H_skl = model.components_
```

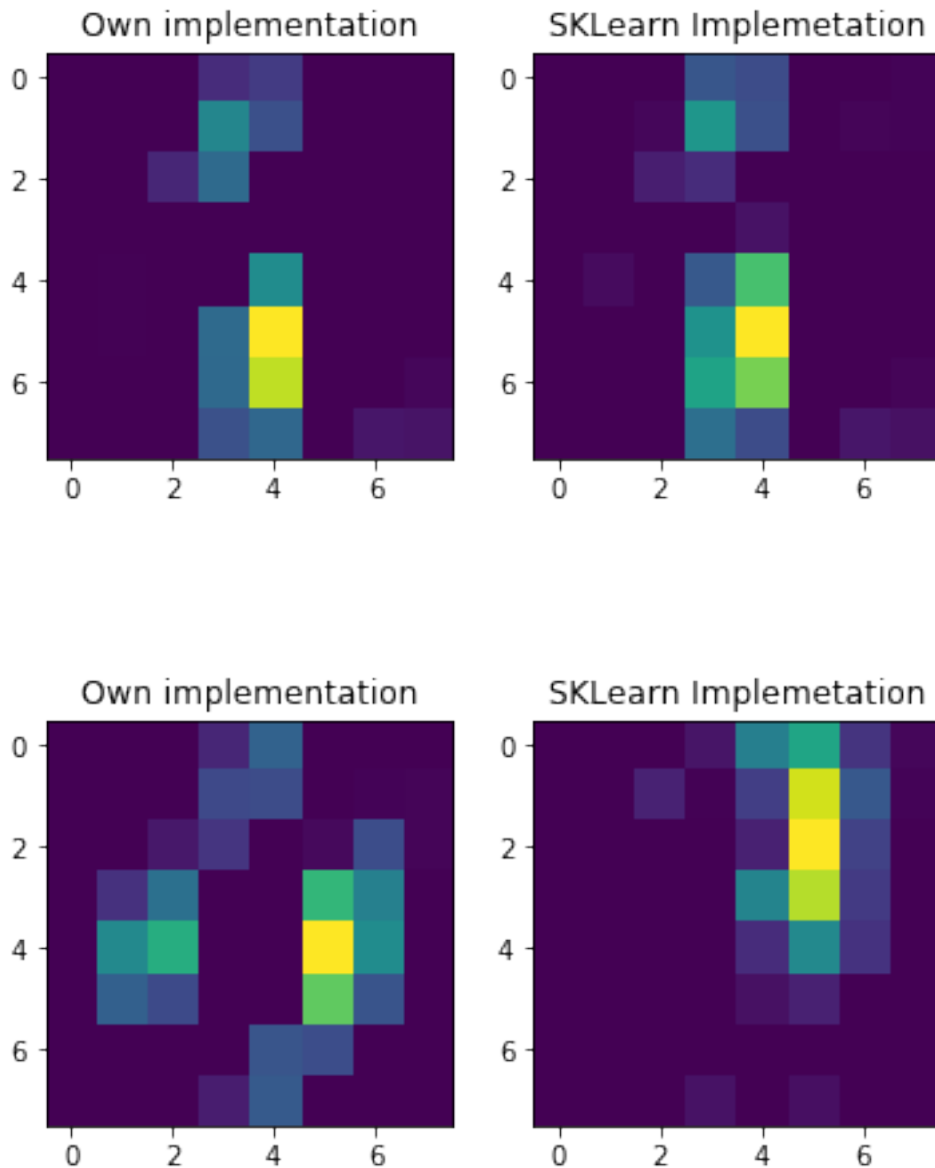
```
In [5]: for i in range(10):
        fig = plt.figure()
        a=fig.add_subplot(1,2,1)
        plt.imshow(H[i,:].reshape(8,8))
        a.set_title('Own implementation')
        a=fig.add_subplot(1,2,2)
        plt.imshow(H_skl[i,:].reshape(8,8))
        a.set_title('SKLearn Implemetation')
        plt.show()
```











The results are comparable, although there are some images that don't look similar. This might be the case, because of the random initialization of W and H : The NMF is not unique but you can see prototypical parts of the numbers.

1.2 5 Recommender System

In [6]: `import pandas as pd`

```
# column headers for the dataset
ratings_cols = ['user id', 'movie id', 'rating', 'timestamp']
movies_cols = ['movie id', 'movie title', 'release date',
               'video release date', 'IMDb URL', 'unknown', 'Action',
```

```

'Adventure','Animation','Childrens','Comedy','Crime',
' Documentary','Drama','Fantasy','Film-Noir','Horror',
'Musical','Mystery','Romance','Sci-Fi','Thriller',
'War' ,'Western']
users_cols = ['user id','age','gender','occupation',
'zip code']

users = pd.read_csv('ml-100k/u.user', sep='|',
                    names = users_cols, encoding='latin-1')
movies = pd.read_csv('ml-100k/u.item', sep='|',
                     names = movies_cols , encoding='latin-1')
ratings = pd.read_csv('ml-100k/u.data', sep='\t',
                      names = ratings_cols , encoding='latin-1 ')

# peek at the dataframes , if you like :)
users.head()
movies.head()
ratings.head()

# create a joint ratings dataframe for the matrix
fill_value = 0
rat_df = ratings.pivot(index = 'user id',
                        columns ='movie id', values = 'rating').fillna(fill_value)
rat_df.head()

```

```

Out[6]: movie id  1      2      3      4      5      6      7      8      9     10     ...  \
user id
1           5.0    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    ...

```

```

movie id  1673  1674  1675  1676  1677  1678  1679  1680  1681  1682
user id
1           0.0    0.0    0.0    0.0    0.0    0.0    0.0    0.0    0.0    0.0
2           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
4           0.0    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 [7]: W, H = non_negative(rat_df, 30)
        print(rat_df.shape, W.shape,H.shape)

```

(943, 1682) (943, 30) (30, 1682)

```

In [8]: def recommend_movies(reconstruction, user_id, movies, ratings):
    rec = reconstruction.iloc[user_id].as_matrix()
    ratings = ratings.as_matrix()
    rec_sort_idx = np.argsort(rec)[: -1]

    rec_idx = np.where(ratings[user_id-1, rec_sort_idx] == 0)[0][0]

    print("Recommended film based on your ratings: " + movies.iloc[rec_idx].at['movie ti
          " " + movies.iloc[rec_idx].at['IMDb URL'])

    reconstruction = pd.DataFrame(W @ H, columns = rat_df.columns)
    for i in range(200):
        predictions = recommend_movies(reconstruction, i+1, movies, rat_df)

Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Four Rooms (1995) http://us.imdb.com/M/title-exact?Four%
Recommended film based on your ratings: Shanghai Triad (Yao a yao yao dao waipo qiao) (1995) htt
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: Muppet Treasure Island (1996) http://us.imdb.com/M/title
Recommended film based on your ratings: Four Rooms (1995) http://us.imdb.com/M/title-exact?Four%
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Shanghai Triad (Yao a yao yao dao waipo qiao) (1995) htt
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Desperado (1995) http://us.imdb.com/M/title-exact?Desper
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: Dead Man Walking (1995) http://us.imdb.com/M/title-exact
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Four Rooms (1995) http://us.imdb.com/M/title-exact?Four%
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: GoldenEye (1995) http://us.imdb.com/M/title-exact?Golden
Recommended film based on your ratings: Usual Suspects, The (1995) http://us.imdb.com/M/title-ex
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20
Recommended film based on your ratings: Toy Story (1995) http://us.imdb.com/M/title-exact?Toy%20

```


[illegible]

[illegible]

[illegible]

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Shanghai Triad (Yao a yao yao dao waipo qiao) (1995) [http://us.imdb.com/M/title-exact?Shanghai%20Triad%20\(Yao%20a%20yao%20yao%20dao%20waipo%20qiao\)%20\(1995\)%20](http://us.imdb.com/M/title-exact?Shanghai%20Triad%20(Yao%20a%20yao%20yao%20dao%20waipo%20qiao)%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) [http://us.imdb.com/M/title-exact?GoldenEye%20\(1995\)%20](http://us.imdb.com/M/title-exact?GoldenEye%20(1995)%20)

Recommended film based on your ratings: Four Rooms (1995) [http://us.imdb.com/M/title-exact?Four%20Rooms%20\(1995\)%20](http://us.imdb.com/M/title-exact?Four%20Rooms%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Four Rooms (1995) [http://us.imdb.com/M/title-exact?Four%20Rooms%20\(1995\)%20](http://us.imdb.com/M/title-exact?Four%20Rooms%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Four Rooms (1995) [http://us.imdb.com/M/title-exact?Four%20Rooms%20\(1995\)%20](http://us.imdb.com/M/title-exact?Four%20Rooms%20(1995)%20)

Recommended film based on your ratings: GoldenEye (1995) [http://us.imdb.com/M/title-exact?GoldenEye%20\(1995\)%20](http://us.imdb.com/M/title-exact?GoldenEye%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

```
In [9]: W, H = non_negative(rat_df, 30)
        reconstruction = pd.DataFrame(W @ H, columns = rat_df.columns)
        for i in range(200):
            predictions = recommend_movies(reconstruction, i+1, movies, rat_df)
```

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Get Shorty (1995) [http://us.imdb.com/M/title-exact?Get%20Shorty%20\(1995\)%20](http://us.imdb.com/M/title-exact?Get%20Shorty%20(1995)%20)

Recommended film based on your ratings: Shanghai Triad (Yao a yao yao dao waipo qiao) (1995) [http://us.imdb.com/M/title-exact?Shanghai%20Triad%20\(Yao%20a%20yao%20yao%20dao%20waipo%20qiao\)%20\(1995\)%20](http://us.imdb.com/M/title-exact?Shanghai%20Triad%20(Yao%20a%20yao%20yao%20dao%20waipo%20qiao)%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) [http://us.imdb.com/M/title-exact?GoldenEye%20\(1995\)%20](http://us.imdb.com/M/title-exact?GoldenEye%20(1995)%20)

Recommended film based on your ratings: Muppet Treasure Island (1996) [http://us.imdb.com/M/title-exact?Muppet%20Treasure%20Island%20\(1996\)%20](http://us.imdb.com/M/title-exact?Muppet%20Treasure%20Island%20(1996)%20)

Recommended film based on your ratings: Get Shorty (1995) [http://us.imdb.com/M/title-exact?Get%20Shorty%20\(1995\)%20](http://us.imdb.com/M/title-exact?Get%20Shorty%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) [http://us.imdb.com/M/title-exact?GoldenEye%20\(1995\)%20](http://us.imdb.com/M/title-exact?GoldenEye%20(1995)%20)

Recommended film based on your ratings: Dead Man Walking (1995) [http://us.imdb.com/M/title-exact?Dead%20Man%20Walking%20\(1995\)%20](http://us.imdb.com/M/title-exact?Dead%20Man%20Walking%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Dolores Claiborne (1994) [http://us.imdb.com/M/title-exact?Dolores%20Claiborne%20\(1994\)%20](http://us.imdb.com/M/title-exact?Dolores%20Claiborne%20(1994)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) [http://us.imdb.com/M/title-exact?GoldenEye%20\(1995\)%20](http://us.imdb.com/M/title-exact?GoldenEye%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Copycat (1995) [http://us.imdb.com/M/title-exact?Copycat%20\(1995\)%20](http://us.imdb.com/M/title-exact?Copycat%20(1995)%20)

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

[illegible]

[illegible]

[illegible]

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Copycat (1995) <http://us.imdb.com/M/title-exact?Copycat%>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Richard III (1995) <http://us.imdb.com/M/title-exact?Rich>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>

Recommended film based on your ratings: Get Shorty (1995) <http://us.imdb.com/M/title-exact?Get%2>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

The random initialization has an influence on the recommendation. If I run the NMF several times with the same number of components, the recommendation changes. One reason could be that the NMF is not a unique factorization and the results can be different.

```
In [10]: # Try the recommendation with 100 components
         W, H = non_negative(rat_df, 100)
         print(rat_df.shape, W.shape, H.shape)
         reconstruction = pd.DataFrame(W @ H, columns = rat_df.columns)
         for i in range(200):
             predictions = recommend_movies(reconstruction, i+1, movies, rat_df)
```


(943, 1682) (943, 100) (100, 1682)

[illegible]

[illegible]

[illegible]

Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>
 Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>
 Recommended film based on your ratings: Four Rooms (1995) <http://us.imdb.com/M/title-exact?Four%20>
 Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>
 Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>
 Recommended film based on your ratings: GoldenEye (1995) <http://us.imdb.com/M/title-exact?Golden>
 Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>
 Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>
 Recommended film based on your ratings: Toy Story (1995) <http://us.imdb.com/M/title-exact?Toy%20>

Here one can see that the choose of the number of parameters has an influence on the recommendation.

1.2.1 Try to find prototypical users

```
In [11]: comp = 10
         W, H = non_negative(rat_df, comp)

In [12]: movies_cols = np.array(movies_cols)
         H_sorted = np.argsort(H, axis = 1)[:,:-1]

         for c in range(comp):
             print("Top ten genres of prototypical user {}".format(c+1))
             a = H_sorted[c,0:9]
             for idx in a:
                 print(movies_cols[movies.iloc[idx].as_matrix() == 1])
```

Top ten genres of prototypical user 1:

```
['Action' 'Crime' 'Drama']
['Drama' 'War']
['Drama']
['Drama']
['Drama' 'Thriller']
['Action' 'Adventure']
['Action' 'Crime' 'Drama']
['Comedy' 'Romance' 'War']
['Crime' 'Drama']
```

Top ten genres of prototypical user 2:

```
['Action' 'Adventure' 'Romance' 'Sci-Fi' 'War']
['Action' 'Adventure' 'Romance' 'Sci-Fi' 'War']
['Action' 'Adventure' 'Comedy' 'Romance']
['Action' 'Adventure' 'Sci-Fi']
['Action' 'Adventure']
['Action' 'Adventure' 'Sci-Fi']
['Action' 'Adventure' 'Drama' 'Romance' 'Sci-Fi' 'War']
['Comedy']
['movie id' 'Animation' 'Childrens' 'Comedy']
```

Top ten genres of prototypical user 3:

```

['Comedy' 'Romance']
['Drama']
['Comedy' 'Romance']
['Drama']
['Comedy' 'Romance']
['Drama']
['Drama' 'War']
['Comedy' 'Romance']
['Drama']
Top ten genres of prototypical user 4:
['Action' 'Drama' 'Romance']
['Action' 'Thriller']
['Drama' 'Sci-Fi']
['Crime' 'Film-Noir' 'Mystery' 'Thriller']
['Action' 'Mystery' 'Romance' 'Thriller']
['Drama' 'Romance' 'War']
['Drama']
['Horror' 'Thriller']
['Mystery' 'Thriller']
Top ten genres of prototypical user 5:
['Drama']
['Crime' 'Thriller']
['Crime' 'Drama']
['Comedy']
['Crime' 'Thriller']
['Sci-Fi']
['Drama' 'Sci-Fi']
['Comedy']
['Comedy' 'Drama' 'Musical']
Top ten genres of prototypical user 6:
['Animation' 'Childrens' 'Musical']
['Animation' 'Childrens' 'Musical']
['Animation' 'Childrens' 'Musical']
['Animation' 'Childrens' 'Comedy' 'Musical']
['Childrens' 'Comedy' 'Musical']
['Animation' 'Childrens' 'Musical']
['Animation' 'Childrens' 'Musical']
['Comedy']
['Animation' 'Childrens']
Top ten genres of prototypical user 7:
['Drama' 'Romance']
['Drama']
['Drama']
['Drama' 'Romance']
['Drama' 'Romance' 'War']
['Drama' 'Mystery']
['Crime' 'Drama' 'Thriller']
['Comedy']

```

```

['Comedy']
Top ten genres of prototypical user 8:
['Action' 'Thriller']
['Action' 'Thriller']
['Action' 'Sci-Fi' 'Thriller']
['Action' 'Adventure' 'Comedy' 'Romance']
['Action' 'Adventure' 'Thriller']
['Action' 'Sci-Fi' 'Thriller']
['Action']
['Action' 'Thriller']
['Action' 'Drama' 'War']
Top ten genres of prototypical user 9:
['Mystery' 'Thriller']
['Film-Noir' 'Mystery']
['Comedy' 'Thriller']
['Action' 'Adventure' 'Romance' 'War']
['Drama' 'War']
['Drama' 'Romance' 'War']
['Mystery' 'Thriller']
['Comedy' 'Crime']
['Drama']
Top ten genres of prototypical user 10:
['Action' 'Adventure' 'Thriller']
['Drama' 'Romance']
['Action' 'Sci-Fi' 'War']
['movie id' 'Animation' 'Childrens' 'Comedy']
['Drama' 'Thriller']
['Action' 'Adventure' 'Mystery']
['Action' 'Adventure' 'Romance' 'Sci-Fi' 'War']
['Crime' 'Drama' 'Thriller']
['Action' 'Adventure' 'Romance' 'Sci-Fi' 'War']

```

Here one can see that there are several prototypical users e.g. : - User 1 likes action movies - User 3 likes crime and thriller - User 5 likes comedy and romance - User 7 likes animation and children - etc.