

```
In [6]: import pandas as pd
        from scipy.sparse import csr_matrix
        from sklearn.neighbors import NearestNeighbors
        from fuzzywuzzy import process
```

```
In [2]: !pip install python-Levenshtein

Collecting python-Levenshtein
  Downloading python_Levenshtein-0.20.5-py3-none-any.whl (9.4 kB)
Collecting Levenshtein==0.20.5
  Downloading Levenshtein-0.20.5-cp39-cp39-win_amd64.whl (98 kB)
----- 98.5/98.5 kB 156.8 kB/s eta 0:00:00
Collecting rapidfuzz<3.0.0,>=2.3.0
  Downloading rapidfuzz-2.11.1-cp39-cp39-win_amd64.whl (993 kB)
----- 993.6/993.6 kB 243.0 kB/s eta 0:00:00
Installing collected packages: rapidfuzz, Levenshtein, python-Levenshtein
Successfully installed Levenshtein-0.20.5 python-Levenshtein-0.20.5 rapidfuzz-2.11.1

[notice] A new release of pip available: 22.1.2 -> 22.2.2
[notice] To update, run: python.exe -m pip install --upgrade pip
```

```
In [24]: !pip install --upgrade pip

Requirement already satisfied: pip in c:\users\kmiyienda\appdata\local\programs\python\python39\lib\site-packages (22.2.2)
```

```
In [7]: #importing files in .csv

movies='movies.csv'
ratings='ratings.csv'

df_movies=pd.read_csv(movies, usecols=['movieId','title'], dtype={'movieId':'int32','title':'str'})
df_ratings=pd.read_csv(ratings, usecols=['userId','movieId','rating'],dtype={'userId':'int32','movieId':'int32','rating':'float32'})
```

```
In [8]: #checking to see if the files imported correctly

df_movies.head()
```

Out[8]:

	movieId	title
0	1	Toy Story (1995)
1	2	Jumanji (1995)
2	3	Grumpier Old Men (1995)
3	4	Waiting to Exhale (1995)
4	5	Father of the Bride Part II (1995)

```
In [9]: #checking to see if the files imported correctly

df_ratings.head()
```

Out[9]:

	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

```
In [17]: #sorting the data into a sparse matrix

movies_users=df_ratings.pivot(index='movieId', columns='userId',values='rating').fillna(0)

mat_movies_users=csr_matrix(movies_users.values)
```

```
In [11]: #invoking KNN and passing it various parameteres

model_knn= NearestNeighbors(metric='cosine', algorithm='brute', n_neighbors=20)
```

```
In [12]: model_knn.fit(mat_movies_users)
```

Out[12]:

▼

NearestNeighbors

NearestNeighbors(algorithm='brute', metric='cosine', n\_neighbors=20)

```
In [23]: #recommender function

def recommender(movie_name, data,model, n_recommendations ):
    model.fit(data)
    idx=process.extractOne(movie_name, df_movies['title'])[2]

    print('Movie Selected: ',df_movies['title'][idx], 'Index: ',idx)
    print('Here are your recommendations...')
    distances, indices=model.kneighbors(data[idx], n_neighbors=n_recommendations)

    for i in indices:
        print(df_movies['title'][i].where(i!=idx))

recommender('Independence Day', mat_movies_users, model_knn,20)

Movie Selected:  Independence Day (a.k.a. ID4) (1996) Index:  615
Here are your recommendations...
615                                     NaN
546                Mission: Impossible (1996)
418                Jurassic Park (1993)
594                Twister (1996)
910    Once Upon a Time in the West (C'era una volta ...
507                Terminator 2: Judgment Day (1991)
224                Star Wars: Episode IV - A New Hope (1977)
0                    Toy Story (1995)
592                Rock, The (1996)
31                Twelve Monkeys (a.k.a. 12 Monkeys) (1995)
938                Local Hero (1983)
314                Forrest Gump (1994)
897                Cheech and Chong's Up in Smoke (1978)
97                Braveheart (1995)
334                Speed (1994)
1182               Fall (1997)
1575               Outsiders, The (1983)
968                Arsenic and Old Lace (1944)
509                Batman (1989)
337                True Lies (1994)
Name: title, dtype: object
```