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In [5]: import pandas as pd
movies=pd.read_csv("C:/Users/admin/Desktop/dataset/movies.csv",usecols=["movieId","title"])

In [8]: movies.columns

Out[8]: Index(['movieId', 'title'], dtype='object')

In [6]: movies.head()

Out[6]:
   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 [7]: movies.isna().sum()

Out[7]:
movieId    0
title      0
dtype: int64

In [11]: rating=pd.read_csv("C:/Users/admin/Desktop/dataset/ratings.csv",usecols=["userId","movieId","rating"])

In [12]: rating.head()

Out[12]:
   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 [15]: movies_user=rating.pivot(index="movieId",columns="userId",values="rating").fillna(0)

In [16]: movies_user

Out[16]:
   userId  1  2  3  4  5  6  7  8  9  10  11  12  13  14  15  16  17  18  19  20  21  22  23  24  25  26  27  28  29  30  31  32  33  34  35  36  37  38  39  40  41  42  43  44  45  46
movieId
1         1  4.0  0.0  0.0  0.0  4.0  0.0  4.5  0.0  0.0  0.0  0.0  0.0  0.0  0.0  2.5  0.0  4.5  3.5  4.0  0.0  3.5  0.0  0.0  0.0  0.0  3.0  0.0  0.0  0.0  5.0  3.0  3.0  0.0  0.0  0.0  0.0  0.0  0.0  5.0  0.0  0.0  5.0  3.0  4.0  5.0
2         2  0.0  0.0  0.0  0.0  0.0  4.0  0.0  4.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  3.0  3.0  3.0  3.5  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  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
3         3  4.0  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  0.0  0.0  3.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  3.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  4.0  5.0  3.0  0.0  0.0
4         4  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  3.0  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  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
5         5  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  0.0  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  0.0  5.0  0.0  3.0  0.0
...      ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...  ...
193581    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  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  0.0  0.0  0.0  0.0  0.0
193583    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  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  0.0  0.0  0.0  0.0  0.0
193585    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  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  0.0  0.0  0.0  0.0  0.0
193587    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  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  0.0  0.0  0.0  0.0  0.0
193609    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  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  0.0  0.0  0.0  0.0  0.0

9724 rows x 610 columns

In [20]: from scipy.sparse import csr_matrix

In [22]: mat_Movies=csr_matrix(movies_user.values)

In [23]: mat_Movies

Out[23]:
<9724x610 sparse matrix of type '<class 'numpy.float64'>'
with 100836 stored elements in Compressed Sparse Row format>

In [25]: from sklearn.neighbors import NearestNeighbors
model=NearestNeighbors(metric="cosine",algorithm="brute",n_neighbors=20)
model.fit(movies_user)

Out[25]:
NearestNeighbors
NearestNeighbors(algorithm='brute', metric='cosine', n_neighbors=20)

In [28]: pip install fuzzywuzzy

Requirement already satisfied: fuzzywuzzy in c:\users\admin\appdata\local\anaconda3\lib\site-packages (0.18.0)
Note: you may need to restart the kernel to use updated packages.

In [30]: from warnings import filterwarnings
filterwarnings("ignore")

In [31]: from fuzzywuzzy import process

In [34]: def recommender(movie_name,data,n):
index=process.extractOne(movie_name,movies["title"])[2]
print("movie selected:",movies['title'][index],'index:',index)
print("searching for recommendation.....")
distance,indices=model.kneighbors(data[index],n_neighbors=n)
for i in indices:
print(movies['title'][i].where(i!=index))

In [35]: recommender('toy story',mat_Movies,10)

movie selected: Toy Story (1995) index: 0
searching for recommendation.....
0
2353      'night Mother (1986)
418      Jurassic Park (1993)
615      Independence Day (a.k.a. ID4) (1996)
224      Star Wars: Episode IV - A New Hope (1977)
314      Forrest Gump (1994)
322      Lion King, The (1994)
910      Once Upon a Time in the West (C'era una volta ...
546      Mission: Impossible (1996)
963      Diva (1981)
Name: title, dtype: object

In [37]: recommender('Forrest Gump',mat_Movies,10)

movie selected: Forrest Gump (1994) index: 314
searching for recommendation.....
314      NaN
277      Shawshank Redemption, The (1994)
418      Jurassic Park (1993)
257      Pulp Fiction (1994)
97      Braveheart (1995)
510      Silence of the Lambs, The (1991)
123      Apollo 13 (1995)
1938      Walk on the Moon, A (1999)
436      Mrs. Doubtfire (1993)
461      Schindler's List (1993)
Name: title, dtype: object

In [38]: recommender('Jumanji',mat_Movies,10)

movie selected: Jumanji (1995) index: 1
searching for recommendation.....
1      NaN
322      Lion King, The (1994)
436      Mrs. Doubtfire (1993)
325      Mask, The (1994)
418      Jurassic Park (1993)
504      Home Alone (1990)
483      Nightmare Before Christmas, The (1993)
506      Aladdin (1992)
512      Beauty and the Beast (1991)
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18         Ace Ventura: When Nature Calls (1995)
Name: title, dtype: object
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In [ ]:
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