```
In [1]: import os
    import numpy as np
    import pandas as pd
    import matplotlib.pyplot as plt
    import seaborn as sns

In [2]: os.getcwd()

Out[2]: 'C:\\Users\\user\\Documents\\ML Projects\\Recommender Systems Projects\\Movie Recommendation System'

In [3]: movies = pd.read_csv("Datasets/tmdb_5000_movies.csv")
    credits = pd.read_csv("Datasets/tmdb_5000_credits.csv")

In [4]: print(movies.shape)
    print(credits.shape)
    (4803, 20)
    (4803, 4)
```

In [5]: movies.head()

Out[5]:

•	budget	genres	homepage	id	keywords	original_language	original_title	overview	popularity	production_companies	production_cou
	<b>0</b> 237000000	[{"id": 28, "name": "Action"}, {"id": 12, "nam	http://www.avatarmovie.com/	19995	[{"id": 1463, "name": "culture clash"}, {"id":	en	Avatar	In the 22nd century, a paraplegic Marine is di	150.437577	[{"name": "Ingenious Film Partners", "id": 289	[{"iso_3166_1": "name": "United \$
	1 300000000	[{"id": 12, "name": "Adventure"}, {"id": 14, "	http://disney.go.com/disneypictures/pirates/	285	[{"id": 270, "name": "ocean"}, {"id": 726, "na	en	Pirates of the Caribbean: At World's End	Captain Barbossa, long believed to be dead, ha	139.082615	[{"name": "Walt Disney Pictures", "id": 2}, {"	[{"iso_3166_1": "name": "United \$
	<b>2</b> 245000000	[{"id": 28, "name": "Action"}, {"id": 12, "nam	http://www.sonypictures.com/movies/spectre/	206647	[{"id": 470, "name": "spy"}, {"id": 818, "name	en	Spectre	A cryptic message from Bond's past sends him o	107.376788	[{"name": "Columbia Pictures", "id": 5}, {"nam	[{"iso_3166_1": "name": "\ Kingd
	<b>3</b> 250000000	[{"id": 28, "name": "Action"}, {"id": 80, "nam	http://www.thedarkknightrises.com/	49026	[{"id": 849, "name": "dc comics"}, {"id": 853,	en	The Dark Knight Rises	Following the death of District Attorney Harve	112.312950	[{"name": "Legendary Pictures", "id": 923}, {"	[{"iso_3166_1": "name": "United (
	<b>4</b> 260000000	[{"id": 28, "name": "Action"}, {"id": 12, "nam	http://movies.disney.com/john-carter	49529	[{"id": 818, "name": "based on novel"}, {"id":	en	John Carter	John Carter is a war- weary, former military ca	43.926995	[{"name": "Walt Disney Pictures", "id": 2}]	[{"iso_3166_1": "name": "United {

```
In [6]: credits.head()
Out[6]:
               movie_id
                                                           title
                                                                                                       cast
                                                                                                                                                      crew
                                                                  [{"cast_id": 242, "character": "Jake Sully", "... [{"credit_id": "52fe48009251416c750aca23", "de...
            0
                  19995
                                                         Avatar
                          Pirates of the Caribbean: At World's End
            1
                     285
                                                                 [{"cast_id": 4, "character": "Captain Jack Spa... [{"credit_id": "52fe4232c3a36847f800b579", "de...
            2
                  206647
                                                       Spectre
                                                                 [{"crast_id": 1, "character": "James Bond", "cr... [{"credit_id": "54805967c3a36829b5002c41", "de...
            3
                  49026
                                          The Dark Knight Rises [{"cast_id": 2, "character": "Bruce Wayne / Ba... [{"credit_id": "52fe4781c3a36847f81398c3", "de...
                  49529
                                                    John Carter
                                                                  [{"cast id": 5, "character": "John Carter", "c... [{"credit id": "52fe479ac3a36847f813eaa3", "de...
In [7]: movies.duplicated().sum()
Out[7]: 0
In [8]: credits.duplicated().sum()
Out[8]: 0
```

No Duplicate rows in **movies** and **credits**.

## In [9]: movies.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4803 entries, 0 to 4802
Data columns (total 20 columns):

memory usage: 750.6+ KB

#	Column	Non-Null Count	Dtype						
	hdeat	400211							
0	budget	4803 non-null	int64						
1	genres	4803 non-null	object						
2	homepage	1712 non-null	object						
3	id	4803 non-null	int64						
4	keywords	4803 non-null	object						
5	original_language	4803 non-null	object						
6	original_title	4803 non-null	object						
7	overview	4800 non-null	object						
8	popularity	4803 non-null	float64						
9	<pre>production_companies</pre>	4803 non-null	object						
10	production_countries	4803 non-null	object						
11	release_date	4802 non-null	object						
12	revenue	4803 non-null	int64						
13	runtime	4801 non-null	float64						
14	spoken_languages	4803 non-null	object						
15	status	4803 non-null	object						
16	tagline	3959 non-null	object						
17	title	4803 non-null	object						
18	vote_average	4803 non-null	float64						
19	vote_count	4803 non-null	int64						
dtyp	<pre>dtypes: float64(3), int64(4), object(13)</pre>								

```
In [10]: isnull ser movies = movies.isna().sum()
         print(isnull ser movies/len(movies)*100)
         print(isnull ser movies[isnull ser movies != 0].index)
         budget
                                   0.000000
         genres
                                   0.000000
         homepage
                                 64.355611
         id
                                   0.000000
         keywords
                                   0.000000
         original language
                                   0.000000
         original title
                                   0.000000
         overview
                                   0.062461
         popularity
                                   0.000000
         production companies
                                   0.000000
         production countries
                                   0.000000
         release date
                                   0.020820
         revenue
                                   0.000000
         runtime
                                   0.041641
         spoken languages
                                   0.000000
         status
                                   0.000000
         tagline
                                 17.572351
         title
                                   0.000000
                                   0.000000
         vote average
         vote count
                                   0.000000
         dtype: float64
         Index(['homepage', 'overview', 'release date', 'runtime', 'tagline'], dtype='object')
         In movies dataframe: 'homepage', 'overview', 'release date', 'runtime', 'tagline' columns contain null values.
In [11]: credits.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 4803 entries, 0 to 4802
         Data columns (total 4 columns):
              Column
                        Non-Null Count Dtype
              movie id 4803 non-null int64
                        4803 non-null object
              title
          1
          2
              cast
                        4803 non-null object
              crew
                        4803 non-null
                                        object
         dtypes: int64(1), object(3)
         memory usage: 150.2+ KB
```

```
In [12]: # movie_id is the common column b/w movies and credits that has no null values.
         movies.rename(columns={'id': 'movie id'}, inplace=True)
In [13]: movies['movie_id'].nunique()
Out[13]: 4803
In [14]: credits['movie_id'].nunique()
Out[14]: 4803
In [15]: movies['title'].nunique()
Out[15]: 4800
In [16]: credits['title'].nunique()
Out[16]: 4800
In [17]: movies['status'].value_counts()
Out[17]: status
         Released
                            4795
         Rumored
                               5
         Post Production
                               3
         Name: count, dtype: int64
```

```
In [18]: movies['original_language'].value_counts()
Out[18]: original_language
              4505
         en
         fr
                70
         es
                32
         zh
                27
                27
         de
                19
         hi
                16
         ja
         it
                14
         cn
                12
                11
         ru
                11
         ko
                  9
         pt
                  7
         da
         sv
         nl
                  4
         fa
                  4
         th
                  3
                  3
         he
                  2
         ta
         cs
                  2
                  2
         ro
         id
                  2
                  2
         ar
         νi
                  1
         sl
                  1
         ps
                  1
         no
                  1
         ky
                  1
                  1
         hu
         pl
                  1
         af
                  1
         nb
                  1
                  1
         tr
         is
                  1
         хx
                  1
         te
                  1
                  1
         el
         Name: count, dtype: int64
```

title	original_title	
Shin Godzilla	シン・ゴジラ	97
Fantastic 4: Rise of the Silver Surfer	4: Rise of the Silver Surfer	215
Asterix at the Olympic Games	Astérix aux Jeux Olympiques	235
The Flowers of War	金陵十三釵	317
Evolution	Évolution	474
The World Is Mine	Lumea e a mea	4699
The Married Woman	Une femme mariée: Suite de fragments d'un film	4719
Gabriela	Gabriela, Cravo e Canela	4751
The Circle	دايره	4790
Cure	キュア	4792

261 rows × 2 columns

```
In [21]: credits['cast'][0]
```

Out[21]: '[{"cast id": 242, "character": "Jake Sully", "credit id": "5602a8a7c3a3685532001c9a", "gender": 2, "id": 65731, "name": "Sam Worthington", "order": 0}, {"cast id": 3, "character": "Neytiri", "credit id": "52fe48009251416c750ac9cb", "gender": 1, "id": 8691, "name": "Zoe Saldana", "order": 1}, {"cast id": 25, "character": "Dr. Grace Augustine", "credit id": "52fe48009251416c750aca39", "gender": 1, "id": 10205, "name": "Sigourney Weaver", "order": 2}, {"cast id": 4, "character": "Col. Quaritch", "credit id": "52fe48009251416c750ac9cf", "gender": 2, "id": 327 47, "name": "Stephen Lang", "order": 3}, {"cast id": 5, "character": "Trudy Chacon", "credit id": "52fe48009251416c750ac9d3", "gender": 1, "i d": 17647, "name": "Michelle Rodriguez", "order": 4}, {"cast id": 8, "character": "Selfridge", "credit id": "52fe48009251416c750ac9e1", "gend er": 2, "id": 1771, "name": "Giovanni Ribisi", "order": 5}, {"cast id": 7, "character": "Norm Spellman", "credit id": "52fe48009251416c750ac9 dd", "gender": 2, "id": 59231, "name": "Joel David Moore", "order": 6}, {"cast id": 9, "character": "Moat", "credit id": "52fe48009251416c750 ac9e5", "gender": 1, "id": 30485, "name": "CCH Pounder", "order": 7}, {"cast id": 11, "character": "Eytukan", "credit id": "52fe48009251416c7 50ac9ed", "gender": 2, "id": 15853, "name": "Wes Studi", "order": 8}, {"cast id": 10, "character": "Tsu\'Tev", "credit id": "52fe48009251416c 750ac9e9", "gender": 2, "id": 10964, "name": "Laz Alonso", "order": 9}, {"cast id": 12, "character": "Dr. Max Patel", "credit id": "52fe48009 251416c750ac9f1", "gender": 2, "id": 95697, "name": "Dileep Rao", "order": 10}, {"cast id": 13, "character": "Lyle Wainfleet", "credit id": "52fe48009251416c750ac9f5", "gender": 2, "id": 98215, "name": "Matt Gerald", "order": 11}, {"cast id": 32, "character": "Private Fike", "cred it id": "52fe48009251416c750aca5b", "gender": 2, "id": 154153, "name": "Sean Anthony Moran", "order": 12}, {"cast id": 33, "character": "Cryo Vault Med Tech", "credit id": "52fe48009251416c750aca5f", "gender": 2, "id": 397312, "name": "Jason Whyte", "order": 13}, {"cast id": 34, "ch aracter": "Venture Star Crew Chief", "credit id": "52fe48009251416c750aca63", "gender": 2, "id": 42317, "name": "Scott Lawrence", "order": 1 4}, {"cast id": 35, "character": "Lock Up Trooper", "credit id": "52fe48009251416c750aca67", "gender": 2, "id": 986734, "name": "Kelly Kilgou r", "order": 15}, {"cast id": 36, "character": "Shuttle Pilot", "credit id": "52fe48009251416c750aca6b", "gender": 0, "id": 1207227, "name": "James Patrick Pitt", "order": 16}, {"cast id": 37, "character": "Shuttle Co-Pilot", "credit id": "52fe48009251416c750aca6f", "gender": 0, "i

## In [22]: credits['crew'][0]

Out[22]: '[{"credit id": "52fe48009251416c750aca23", "department": "Editing", "gender": 0, "id": 1721, "job": "Editor", "name": "Stephen E. Rivkin"}, {"credit id": "539c47ecc3a36810e3001f87", "department": "Art", "gender": 2, "id": 496, "job": "Production Design", "name": "Rick Carter"}, {"credit id": "54491c89c3a3680fb4001cf7", "department": "Sound", "gender": 0, "id": 900, "job": "Sound Designer", "name": "Christopher Boye s"}, {"credit id": "54491cb70e0a267480001bd0", "department": "Sound", "gender": 0, "id": 900, "job": "Supervising Sound Editor", "name": "Chr istopher Boyes"}, {"credit id": "539c4a4cc3a36810c9002101", "department": "Production", "gender": 1, "id": 1262, "job": "Casting", "name": "M ali Finn"}, {"credit id": "5544ee3b925141499f0008fc", "department": "Sound", "gender": 2, "id": 1729, "job": "Original Music Composer", "nam e": "James Horner"}, {"credit id": "52fe48009251416c750ac9c3", "department": "Directing", "gender": 2, "id": 2710, "job": "Director", "name": "James Cameron"}, {"credit id": "52fe48009251416c750ac9d9", "department": "Writing", "gender": 2, "id": 2710, "job": "Writer", "name": "James Cameron"}, {"credit id": "52fe48009251416c750aca17", "department": "Editing", "gender": 2, "id": 2710, "job": "Editor", "name": "James Camero n"}, {"credit id": "52fe48009251416c750aca29", "department": "Production", "gender": 2, "id": 2710, "job": "Producer", "name": "James Camero n"}, {"credit id": "52fe48009251416c750aca3f", "department": "Writing", "gender": 2, "id": 2710, "job": "Screenplay", "name": "James Camero n"}, {"credit id": "539c4987c3a36810ba0021a4", "department": "Art", "gender": 2, "id": 7236, "job": "Art Direction", "name": "Andrew Menzie s"}, {"credit id": "549598c3c3a3686ae9004383", "department": "Visual Effects", "gender": 0, "id": 6690, "job": "Visual Effects Producer", "na me": "Jill Brooks"}, {"credit id": "52fe48009251416c750aca4b", "department": "Production", "gender": 1, "id": 6347, "job": "Casting", "name": "Margery Simkin"}, {"credit id": "570b6f419251417da70032fe", "department": "Art", "gender": 2, "id": 6878, "job": "Supervising Art Director", "name": "Kevin Ishioka"}, {"credit id": "5495a0fac3a3686ae9004468", "department": "Sound", "gender": 0, "id": 6883, "job": "Music Editor", "n ame": "Dick Bernstein"}, {"credit id": "54959706c3a3686af3003e81", "department": "Sound", "gender": 0, "id": 8159, "job": "Sound Effects Edit or", "name": "Shannon Mills"}, {"credit id": "54491d58c3a3680fb1001ccb", "department": "Sound", "gender": 0, "id": 8160, "job": "Foley", "nam e": "Dennie Thorpe"}, {"credit\_id": "54491d6cc3a3680fa5001b2c", "department": "Sound", "gender": 0, "id": 8163, "job": "Foley", "name": "Jana

```
In [23]: movies[movies['original language'] == 'fr']['genres'][235]
Out[23]: '[{"id": 14, "name": "Fantasy"}, {"id": 12, "name": "Adventure"}, {"id": 35, "name": "Comedy"}, {"id": 10751, "name": "Family"}]'
In [24]: movies['genres'][0]
Out[24]: '[{"id": 28, "name": "Action"}, {"id": 12, "name": "Adventure"}, {"id": 14, "name": "Fantasy"}, {"id": 878, "name": "Science Fiction"}]'
In [25]: movies['keywords'][0]
Out[25]: '[{"id": 1463, "name": "culture clash"}, {"id": 2964, "name": "future"}, {"id": 3386, "name": "space war"}, {"id": 3388, "name": "space colony"},
         {"id": 3679, "name": "society"}, {"id": 3801, "name": "space travel"}, {"id": 9685, "name": "futuristic"}, {"id": 9840, "name": "romance"}, {"i
         d": 9882, "name": "space"}, {"id": 9951, "name": "alien"}, {"id": 10148, "name": "tribe"}, {"id": 10158, "name": "alien planet"}, {"id": 10987,
         "name": "cgi"}, {"id": 11399, "name": "marine"}, {"id": 13065, "name": "soldier"}, {"id": 14643, "name": "battle"}, {"id": 14720, "name": "love a
         ffair"}, {"id": 165431, "name": "anti war"}, {"id": 193554, "name": "power relations"}, {"id": 206690, "name": "mind and soul"}, {"id": 209714,
         "name": "3d"}]'
In [26]: movies['original language'][0]
Out[26]: 'en'
In [27]: movies['spoken languages'][0]
Out[27]: '[{"iso 639 1": "en", "name": "English"}, {"iso 639 1": "es", "name": "Espa\\u00f1ol"}]'
In [28]: movies['movie id'][0]
Out[28]: 19995
In [29]: credits['movie id'][0]
Out[29]: 19995
In [30]: movies['overview'][0]
Out[30]: 'In the 22nd century, a paraplegic Marine is dispatched to the moon Pandora on a unique mission, but becomes torn between following orders and pr
         otecting an alien civilization.'
```

```
In [31]: movies['tagline'][0]
Out[31]: 'Enter the World of Pandora.'
In [32]: movies['title'][0]
Out[32]: 'Avatar'
```

Building a recommender system that uses Content Based filtering:

## Significant variable for Content based filtering:

'genres', 'id', 'keywords', 'original language', 'overview', 'release date', 'runtime', 'spoken languages', 'tagline', 'title', 'cast', 'crew'

But since we do not have user profile information in this project therefore we would not use 'original language', 'spoken languages', 'runtime'.

We are dropping 'tagline' because over 17% missing values are there. And imputing them based on 'overview' means increased complexity of the recommender system.

Out[35]:

crew	cast	title	release_date	overview	keywords	movie_id	genres	
[{"credit_id": "52fe48009251416c750aca23", "de	[{"cast_id": 242, "character": "Jake Sully", "	Avatar	2009-12-10	In the 22nd century, a paraplegic Marine is di	[{"id": 1463, "name": "culture clash"}, {"id":	19995	[{"id": 28, "name": "Action"}, {"id": 12, "nam	0
[{"credit_id": "52fe4232c3a36847f800b579", "de	[{"cast_id": 4, "character": "Captain Jack Spa	Pirates of the Caribbean: At World's End	2007-05-19	Captain Barbossa, long believed to be dead, ha	[{"id": 270, "name": "ocean"}, {"id": 726, "na	285	[{"id": 12, "name": "Adventure"}, {"id": 14, "	1
[{"credit_id": "54805967c3a36829b5002c41", "de	[{"cast_id": 1, "character": "James Bond", "cr	Spectre	2015-10-26	A cryptic message from Bond's past sends him o	[{"id": 470, "name": "spy"}, {"id": 818, "name	206647	[{"id": 28, "name": "Action"}, {"id": 12, "nam	2
[{"credit_id": "52fe4781c3a36847f81398c3", "de	[{"cast_id": 2, "character": "Bruce Wayne / Ba	The Dark Knight Rises	2012-07-16	Following the death of District Attorney Harve	[{"id": 849, "name": "dc comics"}, {"id": 853,	49026	[{"id": 28, "name": "Action"}, {"id": 80, "nam	3
[{"credit_id": "52fe479ac3a36847f813eaa3", "de	[{"cast_id": 5, "character": "John Carter", "c	John Carter	2012-03-07	John Carter is a war- weary, former military ca	[{"id": 818, "name": "based on novel"}, {"id":	49529	[{"id": 28, "name": "Action"}, {"id": 12, "nam	4

Dropping the rows having null value for 'release\_date' and 'overview'.

In [36]: movies1[movies1['release\_date'].isnull()]

Out[36]:

	genres	movie_id	keywords	overview	release_date	title	cast	crew
4553		380097		1971 post civil rights San Francisco seemed li	NaN	America Is Still the Place		

In [37]: movies1[movies1['overview'].isnull()]

Out[37]:

		genres	movie_id	keywords	overview	release_date	title	cast	crew
	2656	[{"id": 18, "name": "Drama"}]	370980	[{"id": 717, "name": "pope"}, {"id": 5565, "na	NaN	2015-12-03	Chiamatemi Francesco - II Papa della gente	[{"cast_id": 5, "character": "Jorge Mario Berg	[{"credit_id": "5660019ac3a36875f100252b", "de
,	4140	[{"id": 99, "name": "Documentary"}]	459488	[{"id": 6027, "name": "music"}, {"id": 225822,	NaN	2015-12-12	To Be Frank, Sinatra at 100	[{"cast_id": 0, "character": "Narrator", "cred	[{"credit_id": "592b25e4c3a368783e065a2f", "de
	4431	[{"id": 99, "name": "Documentary"}]	292539	0	NaN	2014-04-26	Food Chains	0	[{"credit_id": "5470c3b1c3a368085e000abd", "de

```
In [38]: movies1.dropna(inplace=True)
In [39]: movies1.isnull().sum()
Out[39]: genres
                         0
         movie id
                         0
         keywords
                         0
         overview
         release date
                         0
         title
                         0
         cast
                         0
                         0
         crew
         dtype: int64
In [40]: # Coverting the string into the right data type, extracting the value
         # corresponding to 'name' key and adding it into a list.
         def extract_name(s):
             list1 = []
             for el in eval(s):
                 list1.append(el["name"])
             return list1
In [41]: def extract director name(s):
             list1 = []
             for el in eval(s):
                 if el["job"] == "Director":
                     list1.append(el["name"])
             return list1
```

```
In [42]: # Extracting the top 4 main casts.
         # Note: According to the meta data casts are listed in the order
         # they appear in the credits.
         def extract main cast info(s):
             list1 = []
             count = 1
             for el in eval(s):
                 list1.append(el["character"])
                 list1.append(el["name"])
                 if count == 4:
                     break
                 count += 1
             return list1
In [43]: from datetime import datetime
In [44]: def extract release year(s):
             return [str(datetime.strptime(s, '%Y-%m-%d').year)]
In [45]: import nltk
         nltk.download('punkt')
         # Stopwords.
         from nltk.corpus import stopwords
         nltk.download('stopwords')
         from nltk.stem import PorterStemmer
         import string
         [nltk data] Downloading package punkt to
                         C:\Users\user\AppData\Roaming\nltk data...
         [nltk data]
                       Package punkt is already up-to-date!
         [nltk data]
         [nltk data] Downloading package stopwords to
                         C:\Users\user\AppData\Roaming\nltk data...
         [nltk data]
         [nltk data] Package stopwords is already up-to-date!
```

```
In [46]: translator = str.maketrans('', '', string.punctuation)
         porter stemmer = PorterStemmer()
In [47]: sentence = "In the 22nd century, a paraplegic Marine is dispatched to the moon Pandora on a unique mission, but becomes torn between following order
         words = nltk.word tokenize(sentence.lower().translate(translator))
         words = [porter stemmer.stem(word) for word in words if word not in stopwords.words('english')]
         print(words)
         ['22nd', 'centuri', 'parapleg', 'marin', 'dispatch', 'moon', 'pandora', 'uniqu', 'mission', 'becom', 'torn', 'follow', 'order', 'protect', 'alie
         n', 'civil']
In [48]: def extract normalized words(s):
             words = nltk.word tokenize(s.lower().translate(translator))
             words = [porter stemmer.stem(word) for word in words if word not in stopwords.words('english')]
             return words
In [49]: movies1['genres'] = movies1['genres'].apply(extract name)
In [50]: movies1['keywords'] = movies1['keywords'].apply(extract name)
In [51]: movies1['cast'] = movies1['cast'].apply(extract main cast info)
In [52]: movies1['crew'] = movies1['crew'].apply(extract_director_name)
In [53]: movies1['release_date'] = movies1['release_date'].apply(extract_release_year)
In [54]: movies1['overview'] = movies1['overview'].apply(extract_normalized_words)
```

```
In [55]: movies1.isna().sum()
Out[55]: genres
                          0
         movie id
                          0
         keywords
                          0
         overview
                          0
         release_date
                          0
         title
                          0
         cast
                          0
                          0
         crew
         dtype: int64
In [56]: type(movies1['genres'][0])
Out[56]: list
In [57]: movies1['genres'][0]
Out[57]: ['Action', 'Adventure', 'Fantasy', 'Science Fiction']
In [58]: movies1['keywords'][0]
Out[58]: ['culture clash',
          'future',
           'space war',
          'space colony',
          'society',
          'space travel',
           'futuristic',
           'romance',
          'space',
          'alien',
          'tribe',
          'alien planet',
          'cgi',
          'marine',
          'soldier',
          'battle',
          'love affair',
          'anti war',
          'power relations',
          'mind and soul',
          '3d']
```

```
In [59]: movies1['overview'][0]
Out[59]: ['22nd',
           'centuri',
           'parapleg',
           'marin',
           'dispatch',
           'moon',
           'pandora',
           'uniqu',
           'mission',
           'becom',
           'torn',
           'follow',
           'order',
           'protect',
           'alien',
           'civil']
In [60]: movies1['cast'][0]
Out[60]: ['Jake Sully',
           'Sam Worthington',
           'Neytiri',
           'Zoe Saldana',
           'Dr. Grace Augustine',
           'Sigourney Weaver',
           'Col. Quaritch',
           'Stephen Lang']
In [61]: movies1.columns
Out[61]: Index(['genres', 'movie id', 'keywords', 'overview', 'release date', 'title',
                  'cast', 'crew'],
                dtype='object')
          Creating a column named 'Context' which contains the attributes of each corresponding movie. These attributes are: 'genres', 'keywords', 'overview', 'release date', 'cast',
          'crew'
In [62]: movies1['context'] = (movies1['genres'] + movies1['keywords'] + movies1['overview']
                                 + movies1['release date'] + movies1['cast'] + movies1['crew'])
```

In [63]: movies1['context'][0]

```
Out[63]: ['Action',
          'Adventure',
          'Fantasy',
          'Science Fiction',
          'culture clash',
          'future',
          'space war',
          'space colony',
          'society',
          'space travel',
          'futuristic',
          'romance',
          'space',
          'alien',
          'tribe',
          'alien planet',
           'cgi',
          'marine',
          'soldier',
          'battle',
          'love affair',
          'anti war',
          'power relations',
          'mind and soul',
          '3d',
          '22nd',
          'centuri',
          'parapleg',
          'marin',
          'dispatch',
          'moon',
           'pandora',
          'uniqu',
          'mission',
          'becom',
           'torn',
           'follow',
          'order',
          'protect',
          'alien',
          'civil',
          '2009',
          'Jake Sully',
          'Sam Worthington',
          'Neytiri',
          'Zoe Saldana',
          'Dr. Grace Augustine',
          'Sigourney Weaver',
          'Col. Quaritch',
```

```
'Stephen Lang',
           'James Cameron']
In [64]: movies2 = movies1.copy()
          movies2.drop(columns=['genres', 'keywords', 'overview', 'release date',
                  'cast', 'crew'], inplace=True)
In [65]: movies2['context'] = movies2['context'].apply(lambda x: " ".join(x))
In [66]: movies2.head()
Out[66]:
             movie_id
                                                 title
                                                                                       context
          0
                19995
                                                        Action Adventure Fantasy Science Fiction cultu...
                                                Avatar
          1
                  285 Pirates of the Caribbean: At World's End Adventure Fantasy Action ocean drug abuse exot...
               206647
                                               Spectre Action Adventure Crime spy based on novel secr...
           2
                                   The Dark Knight Rises
                                                        Action Crime Drama Thriller dc comics crime fi...
           3
                49026
                                            John Carter Action Adventure Science Fiction based on nove...
                49529
In [67]: movies2['context'][0]
Out[67]: 'Action Adventure Fantasy Science Fiction culture clash future space war space colony society space travel futuristic romance space alien tribe a
          lien planet cgi marine soldier battle love affair anti war power relations mind and soul 3d 22nd centuri parapleg marin dispatch moon pandora uni
          qu mission becom torn follow order protect alien civil 2009 Jake Sully Sam Worthington Neytiri Zoe Saldana Dr. Grace Augustine Sigourney Weaver C
          ol. Quaritch Stephen Lang James Cameron'
In [68]: from sklearn.feature extraction.text import CountVectorizer
In [69]: cv = CountVectorizer(stop words = 'english')
Out[69]:
                       CountVectorizer
          CountVectorizer(stop words='english')
```

```
In [70]: # Learning the vocabulary and creating a token count matrix in one go.
         # By default each token would be a word n-gram.
         words = cv.fit transform(movies2['context'])
         words
Out[70]: <4799x33621 sparse matrix of type '<class 'numpy.int64'>'
                with 255042 stored elements in Compressed Sparse Row format>
In [71]: # Shape of the token count matrix.
         words.shape
Out[71]: (4799, 33621)
In [72]: from sklearn.metrics.pairwise import cosine similarity
In [73]: similarity matrix = cosine similarity(words)
In [74]: similarity matrix
Out[74]: array([[1.
                          , 0.04868645, 0.05229764, ..., 0.04546629, 0.03892495,
                                  , 0.02864459, ..., 0.03735437, 0.
               [0.04868645, 1.
               [0.05229764, 0.02864459, 1. , ..., 0.02675002, 0.0763381 ,
                0. 1,
               [0.04546629, 0.03735437, 0.02675002, ..., 1. , 0.0132733 ,
                0.02381628],
                                  , 0.0763381 , ..., 0.0132733 , 1.
               [0.03892495, 0.
                0.01359318],
                                  , 0. , ..., 0.02381628, 0.01359318,
                         , 0.
                1.
                         ]])
In [75]: similarity matrix.shape
Out[75]: (4799, 4799)
In [76]: similarity_matrix[0]
Out[76]: array([1.
                         , 0.04868645, 0.05229764, ..., 0.04546629, 0.03892495,
               0.
                        1)
```

```
In [77]: movies2[movies2['title'] == 'Avatar'].index[0]
Out[77]: 0
In [78]: a = np.random.randint(1,31, (3,6))
         e = enumerate(a[0])
         e
Out[78]: <enumerate at 0x18a390e1800>
In [79]: list1 = list(e)
         list1
         sorted(list1, reverse=True, key=lambda x: x[1])
Out[79]: [(3, 29), (4, 29), (0, 28), (1, 23), (5, 16), (2, 11)]
In [80]: len(list1)
Out[80]: 6
In [81]: # movie name: a valid movie name that should be there in the dataset
         # count: specifies the number of similar movies to be recommended as
         # per the movie name
         # Print movies similar to the movie name
         def search related movies by name(movie name, count):
             try:
                 movie_index = movies2[movies2['title'].apply(lambda title: title.lower()) ==
                         movie name.lower()].index[0]
                 distances = sorted(list(enumerate(similarity matrix[movie index])),
                                    reverse=True, key = lambda x :x[1])
                 for t in distances[1:count+1]:
                   # print(f"movie: {movies2['title'][t[0]]}")
                   print(f"movie: {movies2['title'][t[0]]}, similarity score: {t[1]}")
             except IndexError:
                 print(f"'{movie name}', was not found! Try some other movie name")
```

```
In [83]: # query: specifies the tpe of movies to be recommended
         # count: specifies the number of similar movies to be recommended as per the query
         def search related movies by query(query, count):
             normalized_query = " ".join(extract_normalized_words(query))
             print(f'normalized query: {normalized query}\n')
             matrix = cv.transform([query])
             result = cosine similarity(matrix, words)[0]
             sorted result = sorted(list(enumerate(result)), reverse=True, key=lambda x: x[1])
             # print(sorted result)
             for t in sorted result[0:count]:
                 if t[1] == 0:
                     print(f'\nCould not find {count} related movies because the similarity scores of the query with the remaining movies are 0.')
                 # print(f"movie: {movies2['title'][t[0]]}")
                 print(f"movie: {movies2['title'][t[0]]}, similarity score: {t[1]}")
In [84]: def movie recommender(query, is movie name=True, count=5):
             if is movie name:
                 search related movies by name(query, count)
             else:
                 search related movies by query(query, count)
In [85]: movie recommender('Avatar', is movie name=True)
         movie: Aliens, similarity score: 0.3531887144873645
         movie: Lifeforce, similarity score: 0.3409971697352367
         movie: Moonraker, similarity score: 0.3167762968124701
         movie: Lockout, similarity score: 0.3105295017040594
         movie: Mission to Mars, similarity score: 0.30988989340045614
In [86]: movie recommender('Avatar', is movie name=False)
         normalized query: avatar
         movie: The Last Airbender, similarity score: 0.25
         movie: Bronson, similarity score: 0.1270001270001905
         movie: 16 to Life, similarity score: 0.125
         Could not find 5 related movies because the similarity scores of the query with the remaining movies are 0.
```

```
In [87]: movie recommender('batman', is movie name=True)
         movie: The Dark Knight, similarity score: 0.41640438621500747
         movie: Batman & Robin, similarity score: 0.3960590171906697
         movie: Batman Returns, similarity score: 0.39088155249705214
         movie: Batman Begins, similarity score: 0.38118124993124386
         movie: The Dark Knight Rises, similarity score: 0.3765330442186538
In [88]: movie recommender('batman', is movie name=False)
         normalized query: batman
         movie: Batman Returns, similarity score: 0.4029114820126901
         movie: 2:13, similarity score: 0.36650833306891567
         movie: The Dark Knight Rises, similarity score: 0.33567254331867563
         movie: The Dark Knight, similarity score: 0.26413527189768715
         movie: Batman Forever, similarity score: 0.2508726030021272
In [89]: movie recommender('the godfather', is movie name=True)
         movie: Center Stage, similarity score: 0.24123649045751025
         movie: Act of Valor, similarity score: 0.21774229673875098
         movie: Step Up 2: The Streets, similarity score: 0.2098217272655633
         movie: Take the Lead, similarity score: 0.203738641308575
         movie: Gone with the Wind, similarity score: 0.20191135200894694
In [90]: movie recommender('the godfather', is movie name=False)
         normalized query: godfath
         Could not find 5 related movies because the similarity scores of the query with the remaining movies are 0.
In [91]: movie recommender('Sherlock Holmes', is movie name=True)
         movie: Sherlock Holmes: A Game of Shadows, similarity score: 0.44150314702736076
         movie: Young Sherlock Holmes, similarity score: 0.40422604172722154
         movie: Shaft, similarity score: 0.18845713897261818
         movie: Midnight in the Garden of Good and Evil, similarity score: 0.16643851232744458
         movie: Die Hard: With a Vengeance, similarity score: 0.16141167423097186
```

```
In [92]: movie recommender('Sherlock Holmes', is movie name=False)
         normalized query: sherlock holm
         movie: Sherlock Holmes, similarity score: 0.42874646285627205
         movie: Young Sherlock Holmes, similarity score: 0.39283710065919303
         movie: Sherlock Holmes: A Game of Shadows, similarity score: 0.19802950859533483
         movie: The Claim, similarity score: 0.18107149208503706
         movie: Eulogy, similarity score: 0.12126781251816646
In [93]: movie recommender('Detective genre', is movie name=False)
         normalized query: detect genr
         movie: Se7en, similarity score: 0.29851115706299675
         movie: The Eclipse, similarity score: 0.2857142857142857
         movie: Bad Boys, similarity score: 0.2822162605150792
         movie: Girl 6, similarity score: 0.26037782196164777
         movie: In the Valley of Elah, similarity score: 0.24433888871261045
In [94]: movie recommender('action and suspense', is movie name=False)
         normalized query: action suspens
         movie: The Transporter Refueled, similarity score: 0.32232918561015206
         movie: Special, similarity score: 0.3162277660168379
         movie: Better Luck Tomorrow, similarity score: 0.26726124191242434
         movie: The Blair Witch Project, similarity score: 0.254000254000381
         movie: The Man with the Golden Gun, similarity score: 0.2357022603955158
In [95]: movie recommender('comedy movies', is movie name=False)
         normalized query: comedi movi
         movie: There's Something About Mary, similarity score: 0.32232918561015206
         movie: Special, similarity score: 0.3162277660168379
         movie: Manito, similarity score: 0.3
         movie: Twin Falls Idaho, similarity score: 0.2721655269759087
         movie: American Hero, similarity score: 0.25537695922762454
```

```
In [96]: movie recommender('movies involving magic', is movie name=False)
         normalized query: movi involv magic
         movie: Last Action Hero, similarity score: 0.21650635094610965
         movie: Practical Magic. similarity score: 0.21213203435596423
         movie: The Oogieloves in the Big Balloon Adventure, similarity score: 0.18749999999999997
         movie: Harry Potter and the Chamber of Secrets, similarity score: 0.18731716231633877
         movie: Oz: The Great and Powerful, similarity score: 0.18569533817705183
In [97]: movie recommender('war movies', is movie name=False)
         normalized query: war movi
         movie: The Hunting Party, similarity score: 0.429197537639476
         movie: Unbroken, similarity score: 0.4029114820126901
         movie: Saving Private Ryan, similarity score: 0.37947331922020544
         movie: Awake, similarity score: 0.36901248321155405
         movie: The Great Raid, similarity score: 0.3646624787447363
In [98]: movie recommender('adventure movies', is movie name=False)
         normalized query: adventur movi
         movie: Inkheart, similarity score: 0.20628424925175867
         movie: Year One, similarity score: 0.19611613513818402
         movie: 30 Minutes or Less, similarity score: 0.19611613513818402
         movie: The Adventures of Rocky & Bullwinkle, similarity score: 0.16222142113076252
         movie: Puss in Boots, similarity score: 0.159111456835146
In [99]: movie recommender('science fiction movies', is movie name=False)
         normalized query: scienc fiction movi
         movie: Special, similarity score: 0.5163977794943223
         movie: Timecop, similarity score: 0.30323921743156135
         movie: Ender's Game, similarity score: 0.25819888974716115
         movie: Red Planet, similarity score: 0.24743582965269678
         movie: An Ideal Husband, similarity score: 0.24618298195866548
```

```
In [100]: movie recommender('kung fu', is movie name=False)
          normalized query: kung fu
          movie: City of Life and Death, similarity score: 0.6135719910778963
          movie: Bulletproof Monk, similarity score: 0.3563483225498991
          movie: Kung Fu Panda 2, similarity score: 0.3475706678180953
          movie: Kung Fu Panda, similarity score: 0.24164883733207076
          movie: The Grandmaster, similarity score: 0.2332847374079217
In [101]: movie recommender('mummy', is movie name=False)
          normalized query: mummi
          movie: The Forest, similarity score: 0.13245323570650439
          movie: Hotel Transylvania 2, similarity score: 0.11043152607484653
          movie: Hotel Transylvania, similarity score: 0.10721125348377948
          movie: Home Run, similarity score: 0.09166984970282113
          Could not find 5 related movies because the similarity scores of the query with the remaining movies are 0.
In [102]: movie recommender('spiderman', is movie name=False)
          normalized query: spiderman
          movie: Spider-Man 3, similarity score: 0.211999576001272
          movie: Spider-Man 2, similarity score: 0.10314212462587934
          movie: The Amazing Spider-Man 2, similarity score: 0.09053574604251853
          movie: The Amazing Spider-Man, similarity score: 0.08137884587711594
          Could not find 5 related movies because the similarity scores of the query with the remaining movies are 0.
```

```
In [103]: movies2['title'][:40]
Out[103]: 0
                                                     Avatar
          1
                   Pirates of the Caribbean: At World's End
          2
                                                    Spectre
                                      The Dark Knight Rises
          3
                                                John Carter
          5
                                               Spider-Man 3
                                                    Tangled
                                    Avengers: Age of Ultron
          8
                     Harry Potter and the Half-Blood Prince
                         Batman v Superman: Dawn of Justice
          10
                                           Superman Returns
          11
                                          Quantum of Solace
          12
                 Pirates of the Caribbean: Dead Man's Chest
          13
                                            The Lone Ranger
                                               Man of Steel
          14
          15
                   The Chronicles of Narnia: Prince Caspian
          16
                                               The Avengers
          17
                Pirates of the Caribbean: On Stranger Tides
          18
                                             Men in Black 3
                  The Hobbit: The Battle of the Five Armies
          19
          20
                                     The Amazing Spider-Man
          21
                                                 Robin Hood
          22
                        The Hobbit: The Desolation of Smaug
                                         The Golden Compass
          23
          24
                                                  King Kong
          25
                                                    Titanic
          26
                                 Captain America: Civil War
          27
                                                 Battleship
          28
                                             Jurassic World
          29
                                                    Skyfall
          30
                                               Spider-Man 2
          31
                                                 Iron Man 3
          32
                                        Alice in Wonderland
          33
                                      X-Men: The Last Stand
          34
                                        Monsters University
          35
                        Transformers: Revenge of the Fallen
                            Transformers: Age of Extinction
          36
          37
                                 Oz: The Great and Powerful
          38
                                   The Amazing Spider-Man 2
                                               TRON: Legacy
```

Name: title, dtype: object