```
import pandas as pd
import numpy as np
df =
pd.read csv(r'https://raw.githubusercontent.com/Sarvesh-S-Patil/Datase
t/main/Movies%20Recommendation.csv')
df.head()
   Movie ID
                 Movie Title
                                                     Movie Genre
Movie Language
          1
                   Four Rooms
                                                    Crime Comedy
en
1
          2
                    Star Wars
                               Adventure Action Science Fiction
en
          3
                Finding Nemo
                                                Animation Family
2
en
          4
                Forrest Gump
                                            Comedy Drama Romance
3
en
          5
             American Beauty
4
                                                           Drama
en
   Movie Budget
                 Movie Popularity Movie Release Date
                                                       Movie Revenue
0
        4000000
                         22.876230
                                            09-12-1995
                                                               4300000
1
       11000000
                                                            775398007
                        126.393695
                                            25-05-1977
2
       94000000
                         85,688789
                                            30-05-2003
                                                            940335536
3
       55000000
                        138.133331
                                            06-07-1994
                                                            677945399
4
       15000000
                                            15-09-1999
                                                            356296601
                         80.878605
   Movie Runtime
                 Movie Vote
0
            98.0
                          6.5
1
           121.0
                          8.1
2
           100.0
                          7.6
3
           142.0
                          8.2
4
           122.0
                          7.9
                                       Movie Homepage ∖
0
1
   http://www.starwars.com/films/star-wars-episod...
2
               http://movies.disney.com/finding-nemo
3
                                                   NaN
4
                        http://www.dreamworks.com/ab/
                                       Movie Keywords
0
           hotel new year's eve witch bet hotel room
         android galaxy hermit death star lightsaber
1
   father son relationship harbor underwater fish...
3
   vietnam veteran hippie mentally disabled runni...
   male nudity female nudity adultery midlife cri...
```

```
Movie Overview \
  It's Ted the Bellhop's first night on the job....
  Princess Leia is captured and held hostage by ...
  Nemo, an adventurous young clownfish, is unexp...
  A man with a low IQ has accomplished great thi...
   Lester Burnham, a depressed suburban father in...
                                    Movie Production House
   [{"name": "Miramax Films", "id": 14}, {"name":...
   [{"name": "Lucasfilm", "id": 1}, {"name": "Twe...
1
       [{"name": "Pixar Animation Studios", "id": 3}]
2
             [{"name": "Paramount Pictures", "id": 4}]
3
   [{"name": "DreamWorks SKG", "id": 27}, {"name"...
                                 Movie Production Country \
   [{"iso_3166_1": "US", "name": "United States o...
1
                          Movie Spoken Language
   [{"iso 639 1": "en", "name": "English"}]
   [{"iso_639_1": "en",
                           "name": "English"}]
1
   [{"iso_639_1": "en", "name": "English"}]
[{"iso_639_1": "en", "name": "English"}]
[{"iso_639_1": "en", "name": "English"}]
3
                                              Movie Tagline \
   Twelve outrageous guests. Four scandalous requ...
         A long time ago in a galaxy far, far away...
   There are 3.7 trillion fish in the ocean, they...
3
   The world will never be the same, once you've ...
                                                Look closer.
                                                  Movie Cast
   Tim Roth Antonio Banderas Jennifer Beals Madon...
   Mark Hamill Harrison Ford Carrie Fisher Peter ...
   Albert Brooks Ellen DeGeneres Alexander Gould ...
   Tom Hanks Robin Wright Gary Sinise Mykelti Wil...
   Kevin Spacey Annette Bening Thora Birch Wes Be...
                                                  Movie Crew
                                                                  Movie Director
   [{'name': 'Allison Anders', 'gender': 1, 'depa... Allison Anders
   [{'name': 'George Lucas', 'gender': 2, 'depart...
1
                                                                    George Lucas
   [{'name': 'Andrew Stanton', 'gender': 2, 'depa... Andrew Stanton
```

```
[{'name': 'Alan Silvestri', 'gender': 2, 'depa... Robert Zemeckis
  [{'name': 'Thomas Newman', 'gender': 2, 'depar...
                                                            Sam Mendes
[5 rows x 21 columns]
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4760 entries, 0 to 4759
Data columns (total 21 columns):
#
     Column
                               Non-Null Count
                                               Dtype
- - -
     -----
                                                ----
 0
     Movie ID
                               4760 non-null
                                                int64
 1
     Movie Title
                               4760 non-null
                                               object
 2
     Movie_Genre
                               4760 non-null
                                               object
 3
     Movie Language
                               4760 non-null
                                               object
 4
     Movie Budget
                               4760 non-null
                                                int64
 5
     Movie Popularity
                               4760 non-null
                                                float64
 6
     Movie Release Date
                               4760 non-null
                                                object
 7
     Movie Revenue
                               4760 non-null
                                                int64
 8
     Movie Runtime
                               4758 non-null
                                                float64
 9
     Movie Vote
                               4760 non-null
                                                float64
 10
    Movie Vote Count
                               4760 non-null
                                                int64
 11
    Movie Homepage
                               1699 non-null
                                                object
 12
    Movie_Keywords
                               4373 non-null
                                                object
 13
    Movie Overview
                               4757 non-null
                                                object
    Movie Production House
                               4760 non-null
                                                object
 15 Movie Production Country
                               4760 non-null
                                                object
 16 Movie Spoken Language
                               4760 non-null
                                                object
 17
    Movie Tagline
                               3942 non-null
                                                object
 18 Movie Cast
                               4733 non-null
                                                object
 19
    Movie Crew
                               4760 non-null
                                                object
    Movie Director
                               4738 non-null
                                                object
dtypes: float64(3), int64(4), object(14)
memory usage: 781.1+ KB
df
      Movie ID
                                Movie Title
Movie Genre
             1
                                 Four Rooms
                                                                  Crime
Comedy
             2
                                  Star Wars
                                              Adventure Action Science
Fiction
             3
                               Finding Nemo
                                                              Animation
Family
             4
3
                               Forrest Gump
                                                          Comedy Drama
Romance
```

4 Drama	5		Americ	an Beauty		
4755	4784	Midnight Cabaret				
4757	4785	Growing Up Smith			Comedy	
	y Drama 4786			8 Days		
4758	ler Drama 4787	Running Forever				
Famil 4759 Docum	y 4788 entary	То В	e Frank, Sinat	ra at 100		
	Movie_Lang	uage	Movie_Budget	Movie_Popularity	Movie_Release_Date	
0		en	4000000	22.876230	09-12-1995	
1		en	11000000	126.393695	25-05-1977	
2		en	94000000	85.688789	30-05-2003	
3		en	55000000	138.133331	06-07-1994	
4		en	15000000	80.878605	15-09-1999	
4755		en	0	0.001389	01-01-1990	
4756		en	0	0.710870	03-02-2017	
4757		en	0	0.015295	15-06-2014	
4758		en	0	0.028756	27-10-2015	
4759		en	2	0.050625	12-12-2015	
0 1 2 3 4	Movie_Rev 430 77539 94033 67794 35629	0000 8007 5536 5399 6601	Movie_Runtime 98.0 121.0 100.0 142.0 122.0 	- 6.5 8.1 7.6 8.2 7.9 0.0		
4756		0	102.0	7.4		

```
4757
                               90.0
                  0
                                            0.0
4758
                  0
                               88.0
                                            0.0
                                            0.0
4759
                  0
                                NaN
                                          Movie Homepage
0
1
      http://www.starwars.com/films/star-wars-episod...
2
                  http://movies.disney.com/finding-nemo
3
4
                           http://www.dreamworks.com/ab/
4755
                                                      NaN
4756
                     http://www.growingupsmithmovie.com
4757
                                                      NaN
4758
                                                      NaN
4759
                                                      NaN
                                          Movie Keywords
0
              hotel new year's eve witch bet hotel room
1
            android galaxy hermit death star lightsaber
2
      father son relationship harbor underwater fish...
      vietnam veteran hippie mentally disabled runni...
3
4
      male nudity female nudity adultery midlife cri...
4755
                                                      NaN
4756
                                                      NaN
4757
                          christian film sex trafficking
4758
                                                      NaN
4759
      music actors legendary perfomer classic hollyw...
                                          Movie Overview
      It's Ted the Bellhop's first night on the job....
1
      Princess Leia is captured and held hostage by ...
2
      Nemo, an adventurous young clownfish, is unexp...
      A man with a low IQ has accomplished great thi...
3
4
      Lester Burnham, a depressed suburban father in...
4755
      A Broadway producer puts on a play with a Devi...
4756
      In 1979, an Indian family moves to America wit...
4757
      After sneaking to a party with her friends, 16...
4758
      After being estranged since her mother's death...
4759
                                  Movie Production House
      [{"name": "Miramax Films", "id": 14}, {"name":...
0
      [{"name": "Lucasfilm", "id": 1}, {"name": "Twe...
1
         [{"name": "Pixar Animation Studios", "id": 3}]
2
              [{"name": "Paramount Pictures", "id": 4}]
3
4
      [{"name": "DreamWorks SKG", "id": 27}, {"name"...
```

```
4755
                                                                     []
4756
4757
           [{"name": "After Eden Pictures", "id": 85248}]
         [{"name": "New Kingdom Pictures", "id": 41671}]
[{"name": "Eyeline Entertainment", "id": 60343}]
4758
4759
                                        Movie Production Country \
       [{"iso_3166_1": "US", "name": "United States o...
0
1
2
3
        [{"iso_3166_1": "US", "name": "United States o...
4
4755
                                                                     []
4756
                                                                     []
       [{"iso_3166_1": "US", "name": "United States o...
[{"iso_3166_1": "US", "name": "United States o...
4757
4758
4759
         [\{"iso_3166_1": "GB", "name": "United Kingdom"\}]
                                Movie Spoken Language
        [{"iso_639_1": "en", "name": "English"}]
[{"iso_639_1": "en", "name": "English"}]
0
        [{"iso_639_1": "en",
1
       [{"iso_639_1": "en", "name": "English"}]
[{"iso_639_1": "en", "name": "English"}]
2
3
4
        [{"iso_639_1": "en", "name": "English"}]
. . .
        [{"iso_639_1": "en", "name": "English"}]
4755
       [{"iso_639_1": "en", "name": "English"}]
4756
        [{"iso 639 1": "en", "name": "English"}]
4757
4758
                                                         []
4759
                                                         []
                                                      Movie Tagline \
0
       Twelve outrageous guests. Four scandalous requ...
1
              A long time ago in a galaxy far, far away...
2
       There are 3.7 trillion fish in the ocean, they...
3
       The world will never be the same, once you've ...
4
                                                        Look closer.
. . .
                          The hot spot where Satan's waitin'.
4755
4756
                   It's better to stand out than to fit in.
4757
                   She never knew it could happen to her...
4758
                                                                    NaN
4759
                                                                    NaN
                                                          Movie_Cast
0
       Tim Roth Antonio Banderas Jennifer Beals Madon...
1
       Mark Hamill Harrison Ford Carrie Fisher Peter ...
2
       Albert Brooks Ellen DeGeneres Alexander Gould ...
3
       Tom Hanks Robin Wright Gary Sinise Mykelti Wil...
```

```
Kevin Spacey Annette Bening Thora Birch Wes Be...
4
4755
      Lisa Hart Carroll Michael Des Barres Paul Drak...
4756
      Roni Akurati Brighton Sharbino Jason Lee Anjul...
      Nicole Smolen Kim Baldwin Ariana Stephens Brys...
4757
4758
4759
                                          Tony Oppedisano
                                                Movie Crew
Movie Director
      [{'name': 'Allison Anders', 'gender': 1, 'depa...
                                                               Allison
Anders
      [{'name': 'George Lucas', 'gender': 2, 'depart...
                                                                  George
Lucas
2
      [{'name': 'Andrew Stanton', 'gender': 2, 'depa...
                                                               Andrew
      [{'name': 'Alan Silvestri', 'gender': 2, 'depa...
                                                              Robert
Zemeckis
      [{'name': 'Thomas Newman', 'gender': 2, 'depar...
                                                                    Sam
Mendes
. . .
. . .
4755 [{'name': 'Pece Dingo', 'gender': 0, 'departme...
                                                                    Pece
Dingo
4756 [{'name': 'Frank Lotito', 'gender': 0, 'depart...
                                                                  Frank
Lotito
     [{'name': 'Jaco Booyens', 'gender': 0, 'depart...
4757
                                                                  Jaco
Boovens
4758
                                                        []
NaN
      [{'name': 'Simon Napier-Bell', 'gender': 0, 'd... Simon Napier-
4759
Bell
[4760 rows \times 21 columns]
df.shape
(4760, 21)
df.columns
Index(['Movie ID', 'Movie Title', 'Movie Genre', 'Movie Language',
       'Movie_Budget', 'Movie_Popularity', 'Movie_Release_Date',
       'Movie Revenue', 'Movie Runtime', 'Movie Vote',
'Movie Vote Count',
       'Movie Homepage', 'Movie Keywords', 'Movie Overview',
       'Movie_Production_House', 'Movie_Production_Country', 'Movie_Spoken_Language', 'Movie_Tagline', 'Movie_Cast',
'Movie Crew',
       'Movie Director'],
      dtype='object')
```

```
df_features = df[[ 'Movie_Genre', 'Movie_Keywords'
'Movie Tagline' ,'Movie Cast' , 'Movie Director' ]].fillna('')
df features.shape
(4760, 5)
df features
                           Movie Genre \
                           Crime Comedy
      Adventure Action Science Fiction
1
2
                      Animation Family
3
                  Comedy Drama Romance
4
                                  Drama
4755
                                 Horror
4756
                   Comedy Family Drama
4757
                        Thriller Drama
4758
                                 Family
4759
                           Documentary
                                          Movie Keywords \
              hotel new year's eve witch bet hotel room
0
1
            android galaxy hermit death star lightsaber
2
      father son relationship harbor underwater fish...
      vietnam veteran hippie mentally disabled runni...
3
      male nudity female nudity adultery midlife cri...
4
. . .
4755
4756
4757
                         christian film sex trafficking
4758
4759
      music actors legendary perfomer classic hollyw...
                                           Movie Tagline \
      Twelve outrageous guests. Four scandalous requ...
           A long time ago in a galaxy far, far away...
1
2
      There are 3.7 trillion fish in the ocean, they...
3
      The world will never be the same, once you've ...
4
                                            Look closer.
4755
                    The hot spot where Satan's waitin'.
4756
               It's better to stand out than to fit in.
               She never knew it could happen to her...
4757
4758
4759
                                              Movie Cast
Movie Director
      Tim Roth Antonio Banderas Jennifer Beals Madon...
                                                            Allison
```

```
Anders
      Mark Hamill Harrison Ford Carrie Fisher Peter ...
                                                                 George
1
Lucas
      Albert Brooks Ellen DeGeneres Alexander Gould ...
                                                               Andrew
Stanton
      Tom Hanks Robin Wright Gary Sinise Mykelti Wil...
                                                              Robert
Zemeckis
      Kevin Spacey Annette Bening Thora Birch Wes Be...
                                                                    Sam
Mendes
. . .
4755 Lisa Hart Carroll Michael Des Barres Paul Drak...
                                                                    Pece
Dingo
4756 Roni Akurati Brighton Sharbino Jason Lee Anjul...
                                                                 Frank
Lotito
4757 Nicole Smolen Kim Baldwin Ariana Stephens Brys...
                                                                 Jaco
Boovens
4758
4759
                                          Tony Oppedisano Simon Napier-
Bell
[4760 rows x 5 columns]
df features.loc[4759]
Movie Genre
                                                           Documentary
Movie Keywords music actors legendary perfomer classic hollyw...
Movie Tagline
Movie Cast
                                                       Tony Oppedisano
Movie Director
                                                     Simon Napier-Bell
Name: 4759, dtype: object
X = df_features['Movie_Genre'] + ' ' +df_features['Movie_Keywords'] +
' ' + df_features['Movie_Tagline'] + ' ' + df_features['Movie_Cast'] +
' ' + df features['Movie Director']
Χ
0
        Crime Comedy hotel new year's eve witch bet ho...
1
        Adventure Action Science Fiction android galax...
2
        Animation Family father son relationship harbo...
3
        Comedy Drama Romance vietnam veteran hippie me...
        Drama male nudity female nudity adultery midli...
4755
        Horror The hot spot where Satan's waitin'. Li...
4756
        Comedy Family Drama It's better to stand out ...
4757
        Thriller Drama christian film sex trafficking ...
4758
                                                  Family
        Documentary music actors legendary perfomer cl...
4759
Length: 4760, dtype: object
```

```
X.shape
(4760,)
from sklearn.feature extraction.text import TfidfVectorizer
tfidf = TfidfVectorizer()
X=tfidf.fit transform(X)
print(X)
  (0, 617) 0.1633382144407513
  (0, 492) 0.1432591540388685
  (0, 15413)
                 0.1465525095337543
  (0, 9675)
                 0.14226057295252661
  (0, 9465)
                 0.1659841367820977
  (0, 1390)
                 0.16898383612799558
  (0, 7825)
                 0.09799561597509843
  (0, 1214)
                 0.13865857545144072
  (0, 729) 0.13415063359531618
  (0, 13093)
                 0.1432591540388685
  (0, 15355)
                 0.10477815972666779
  (0, 9048)
                 0.0866842116160778
  (0, 11161)
                 0.06250380151644369
  (0, 16773)
                 0.17654247479915475
  (0, 5612)
                 0.08603537588547631
  (0, 16735)
                 0.10690083751525419
  (0, 7904)
                 0.13348000542112332
  (0, 15219)
                 0.09800472886453934
  (0, 11242)
                 0.07277788238484746
  (0, 3878)
                 0.11998399582562203
  (0, 5499)
                 0.11454057510303811
  (0, 7071)
                 0.19822417598406614
  (0, 7454)
                 0.14745635785412262
  (0, 1495)
                 0.19712637387361423
  (0, 9206)
                 0.15186283580984414
  (4757, 5455)
                 0.12491480594769522
  (4757, 2967)
                 0.16273475835631626
  (4757, 8464)
                 0.23522565554066333
  (4757, 6938)
                 0.17088173678136628
  (4757, 8379)
                 0.17480603856721913
  (4757, 15303)
                 0.07654356007668191
  (4757, 15384)
                 0.09754322497537371
  (4757, 7649)
                 0.11479421494340192
  (4757, 10896)
                0.14546473055066447
  (4757, 4494)
                 0.05675298448720501
  (4758, 5238)
                 1.0
  (4759, 11264) 0.33947721804318337
  (4759, 11708) 0.33947721804318337
```

```
(4759, 8902)
               0.3040290704566037
  (4759, 14062) 0.3237911628497312
  (4759, 3058) 0.2812896191863103
  (4759, 7130)
               0.26419662449963793
  (4759, 10761) 0.3126617295732147
  (4759, 4358) 0.18306542312175342
  (4759, 14051) 0.20084315377640435
  (4759, 5690)
               0.19534291014627303
  (4759, 15431) 0.19628653185946862
  (4759, 1490) 0.21197258705292082
  (4759, 10666) 0.15888268987343043
from sklearn.metrics.pairwise import cosine_similarity
Similarity Score = cosine similarity(X)
Similarity Score
                 , 0.01351235, 0.03570468, ..., 0. , 0.
array([[1.
      [0.01351235, 1. , 0.00806674, ..., 0.
                                                        . 0.
      [0.03570468, 0.00806674, 1. , ..., 0.
0.08014876,
       0.
                 ],
       . . . ,
                      , 0. , ..., 1.
       [0.
                 , 0.
                                                        , 0.
       0.
                         , 0.08014876, ..., 0.
                 , 0.
      [0.
                                                        , 1.
       0.
                 ],
                            , 0. , ..., 0.
                 , 0.
                                                        , 0.
      [0.
       1.
                 11)
Similarity Score.shape
(4760, 4760)
Favourite Movie Name = input( "Enter your favourite Movie Name : " )
Enter your favourite Movie Name : avtaar
All Movie Title List = df['Movie Title'].tolist()
import difflib
Movie Recommendation =
difflib.get close matches(Favourite Movie Name,All Movie Title List)
```

(4759, 205)

0.3237911628497312

```
print(Movie Recommendation)
['Avatar', 'Gattaca']
Close Match = Movie Recommendation[0]
print(Close Match)
Avatar
Index_of_Close_Match_Movie = df[df.Movie_Title == Close Match]
['Movie ID'].values[0]
print(Index of Close Match Movie)
2692
Recommendation Score =
list(enumerate(Similarity Score[Index of Close Match Movie]))
# print(Recommendation Score)
len(Recommendation Score)
4760
Sorted Similar Movies = sorted(Recommendation Score , key = lambda
x:x[1], reverse = True)
# print(Sorted Similar Movies)
print("Top 30 Movies suggested for you are : ")
for movie in Sorted Similar Movies:
  index = movie[0];
  title from index = df[df.index == index]['Movie Title'].values[0]
  if(i<31):
      print(i, '.',title_from_index)
Top 30 Movies suggested for you are :
1 . Niagara
2 . Caravans
3 . My Week with Marilyn
4 . Brokeback Mountain
5 . Harry Brown
6 . Night of the Living Dead
7 . The Curse of Downers Grove
8 . The Boy Next Door
9 . Back to the Future
10 . The Juror
11 . Some Like It Hot
12 . Enough
13 . The Kentucky Fried Movie
```

- 14 . Eye for an Eye
- 15 . Welcome to the Sticks
- 16 . Alice Through the Looking Glass
- 17 . Superman III
- 18 . The Misfits
- 19 . Premium Rush
- 20 . Duel in the Sun
- 21 . Sabotage
- 22 . Small Soldiers
- 23 . All That Jazz
- 24 . Camping Sauvage
- 25 . The Raid
- 26 . Beyond the Black Rainbow
- 27 . To Kill a Mockingbird
- 28 . World Trade Center
- 29 . The Dark Knight Rises
- 30 . Tora! Tora! Tora!