Movielens

November 11, 2022

```
[28]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      %matplotlib inline
[29]: #Import the three datasets
      read_movies=pd.read_csv('movies.dat',sep='\s+\|\s+')
      read_movies.to_csv('movies.csv', index=None)
      read_movies
     /usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:2: ParserWarning:
     Falling back to the 'python' engine because the 'c' engine does not support
     regex separators (separators > 1 char and different from '\s+' are interpreted
     as regex); you can avoid this warning by specifying engine='python'.
[29]:
           1::Toy Story (1995)::Animation|Children's|Comedy
      0
            2::Jumanji (1995)::Adventure|Children's|Fantasy
      1
                 3::Grumpier Old Men (1995)::Comedy|Romance
      2
                  4::Waiting to Exhale (1995)::Comedy|Drama
      3
              5::Father of the Bride Part II (1995)::Comedy
      4
                      6::Heat (1995)::Action|Crime|Thriller
      3877
                      3948:: Meet the Parents (2000):: Comedy
      3878
                    3949::Requiem for a Dream (2000)::Drama
                              3950::Tigerland (2000)::Drama
      3879
```

[3882 rows x 1 columns]

3880

3881

```
[30]: read_ratings = pd.read_csv('ratings.dat')
read_ratings.to_csv ('ratings.csv', index=None)
read_ratings
```

3951::Two Family House (2000)::Drama

3952::Contender, The (2000)::Drama|Thriller

```
[30]:
                  1::1193::5::978300760
                   1::661::3::978302109
      0
      1
                   1::914::3::978301968
      2
                  1::3408::4::978300275
                  1::2355::5::978824291
      3
      4
                  1::1197::3::978302268
      1000203 6040::1091::1::956716541
      1000204 6040::1094::5::956704887
      1000205
                6040::562::5::956704746
      1000206 6040::1096::4::956715648
      1000207 6040::1097::4::956715569
      [1000208 rows x 1 columns]
[31]: read_users = pd.read_csv('users.dat')
      read_users.to_csv ('users.csv', index=None)
      read_users
[31]:
                1::F::1::10::48067
      0
               2::M::56::16::70072
      1
               3::M::25::15::55117
                4::M::45::7::02460
      3
               5::M::25::20::55455
                6::F::50::9::55117
      4
      6034 6036::F::25::15::32603
             6037::F::45::1::76006
      6035
      6036
             6038::F::56::1::14706
      6037
             6039::F::45::0::01060
      6038
             6040::M::25::6::11106
      [6039 rows x 1 columns]
[32]: read_ratings.head()
[32]:
         1::1193::5::978300760
        1::661::3::978302109
        1::914::3::978301968
      2 1::3408::4::978300275
      3 1::2355::5::978824291
      4 1::1197::3::978302268
[33]: read_ratings.columns
[33]: Index(['1::1193::5::978300760'], dtype='object')
```

```
[]:
[34]: read movies.columns
[34]: Index(['1::Toy Story (1995)::Animation|Children's|Comedy'], dtype='object')
[35]: read_users.columns
[35]: Index(['1::F::1::10::48067'], dtype='object')
[36]: #Convert single column dataset into multiple column dataset
      read_ratings.dropna(inplace = True)
      new = read_ratings['1::1193::5::978300760'].str.split("::", n = 3, expand =__
      →True)
      read ratings["UserID"] = new[0]
      read_ratings["MovieID"] = new[1]
      read_ratings["Rating"] = new[2]
      read_ratings["Timestamp"] = new[3]
      read_ratings.drop(columns =['1::1193::5::978300760'], inplace = True)
      # df display
      read_ratings
[36]:
              UserID MovieID Rating Timestamp
                   1
                         661
                                  3 978302109
                   1
                                  3 978301968
      1
                         914
      2
                   1
                        3408
                                  4 978300275
      3
                        2355
                                  5 978824291
                   1
      4
                   1
                        1197
                                  3 978302268
      1000203
                6040
                                  1 956716541
                        1091
      1000204
                6040
                        1094
                                  5 956704887
      1000205
                6040
                         562
                                  5 956704746
      1000206
                6040
                        1096
                                  4 956715648
      1000207
                6040
                        1097
                                  4 956715569
      [1000208 rows x 4 columns]
[37]: #Convert single column dataset into multiple column dataset
      read movies.dropna(inplace = True)
      new = read_movies["1::Toy Story (1995)::Animation|Children's|Comedy"].str.
       ⇒split("::", n =2, expand = True)
      read movies["MovieID"] = new[0]
      read_movies["Title"] = new[1]
      read_movies["Genres"] = new[2]
      read movies.drop(columns = ["1::Toy Story (1995)::Animation|Children's|Comedy"], __
       →inplace = True)
      # df display
```

```
read_movies
[37]:
           MovieID
                                                    Title
                                                                                   Genres
                                          Jumanji (1995)
                                                           Adventure | Children's | Fantasy
      1
                  3
                                 Grumpier Old Men (1995)
                                                                          Comedy | Romance
      2
                  4
                               Waiting to Exhale (1995)
                                                                            Comedy | Drama
                  5
      3
                     Father of the Bride Part II (1995)
                                                                                   Comedy
      4
                  6
                                             Heat (1995)
                                                                   Action|Crime|Thriller
      3877
              3948
                                 Meet the Parents (2000)
                                                                                   Comedy
                             Requiem for a Dream (2000)
                                                                                    Drama
      3878
              3949
      3879
              3950
                                        Tigerland (2000)
                                                                                    Drama
      3880
              3951
                                 Two Family House (2000)
                                                                                    Drama
                                                                          Drama|Thriller
      3881
              3952
                                   Contender, The (2000)
      [3882 rows x 3 columns]
[38]: #Convert single column dataset into multiple column dataset
      read_users.dropna(inplace = True)
      new = read_users["1::F::1::10::48067"].str.split("::", n =4, expand = True)
      read_users["UserID"] = new[0]
      read_users["Gender"] = new[1]
      read_users["Age"] = new[2]
      read_users["Occupation"] = new[3]
      read_users["Zip-code"] = new[4]
      read_users.drop(columns =["1::F::1::10::48067"], inplace = True)
      # df display
      read_users
[38]:
           UserID Gender Age Occupation Zip-code
      0
                 2
                        М
                          56
                                       16
                                             70072
      1
                 3
                           25
                                       15
                                             55117
                        M
      2
                 4
                          45
                                        7
                                             02460
      3
                 5
                           25
                        M
                                       20
                                             55455
                 6
                        F
                                        9
                           50
                                             55117
             6036
                        F
                           25
                                             32603
      6034
                                       15
                                             76006
      6035
             6037
                        F
                           45
                                        1
      6036
             6038
                        F
                           56
                                        1
                                             14706
      6037
                        F
                                        0
                                             01060
             6039
                           45
      6038
             6040
                        Μ
                           25
                                        6
                                             11106
      [6039 rows x 5 columns]
[56]:
      # using merge function to combine ratings and movie dataset by setting_{f U}
```

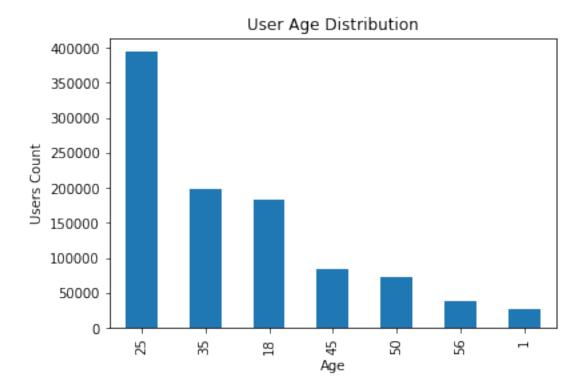
→how='inner'

```
→read_movies[['Title','MovieID','Genres']],
                          on='MovieID',
                          how='inner')
      # displaying result
      print(rating_movie_join)
            MovieID UserID Rating
                                                                             Title \
     0
                 661
                          1
                                                 James and the Giant Peach (1996)
                 661
                         23
                                  2
                                                 James and the Giant Peach (1996)
     1
                 661
                                                 James and the Giant Peach (1996)
     2
                         49
                                  3
                                                 James and the Giant Peach (1996)
     3
                 661
                         53
                                  5
                 661
                                                 James and the Giant Peach (1996)
     4
                         57
                                  4
     998126
                2198
                       5949
                                  5
                                                               Modulations (1998)
     998127
                2703
                       5675
                                  3
                                                            Broken Vessels (1998)
                                                                White Boys (1999)
     998128
                2845
                       5780
                                  1
                                  5
     998129
                3607
                       5851
                                                         One Little Indian (1973)
     998130
                2909
                       5938
                                     Five Wives, Three Secretaries and Me (1998)
                                     Genres
              Animation | Children's | Musical
     0
     1
              Animation | Children's | Musical
     2
              Animation | Children's | Musical
     3
              Animation | Children's | Musical
     4
              Animation|Children's|Musical
     998126
                                Documentary
     998127
                                      Drama
     998128
                                      Drama
                      Comedy | Drama | Western
     998129
     998130
                                Documentary
     [998131 rows x 5 columns]
[74]: # using merge function to combine rating movie join and user dataset by setting
       →how='inner'
      Master_Data= pd.
       →merge(rating_movie_join[['MovieID','Title','UserID','Rating','Genres']], __
       →read_users[['UserID','Gender','Age','Occupation']],
                          on='UserID',
                          how='inner')
      # displaying result
      print(Master_Data)
```

rating_movie_join= pd.merge(read_ratings[['MovieID', 'UserID', 'Rating']],_

```
MovieID
                                                                    Title UserID \
     0
                 661
                                        James and the Giant Peach (1996)
                                                                               23
                2355
                                                    Bug's Life, A (1998)
     1
                                                                               23
     2
                 595
                                             Beauty and the Beast (1991)
                                                                               23
     3
                2797
                                                                               23
                                                               Big (1988)
     4
                 720
                      Wallace & Gromit: The Best of Aardman Animatio...
                                                                             23
                                                         Footloose (1984)
     998075
                3791
                                                                             4211
     998076
                3806
                                                  MacKenna's Gold (1969)
                                                                             4211
     998077
                3840
                                                      Pumpkinhead (1988)
                                                                             4211
     998078
                3766
                                                Missing in Action (1984)
                                                                             4211
                                                      Bronco Billy (1980)
     998079
                3834
                                                                             4211
            Rating
                                            Genres Gender Age Occupation
                     Animation|Children's|Musical
                                                            35
     0
                                                                         0
     1
                      Animation | Children's | Comedy
                                                            35
     2
                  3 Animation|Children's|Musical
                                                         M 35
                                                                         0
     3
                                    Comedy | Fantasy
                                                        M 35
                                                                         0
     4
                  4
                                         Animation
                                                         M 35
                                                                         0
                  2
     998075
                                             Drama
                                                         М
                                                           45
                                                                        5
     998076
                  3
                                                         M 45
                                                                        5
                                           Western
     998077
                                            Horror
                                                        M 45
                                                                         5
                                                                         5
     998078
                                        Action|War
                                                        M 45
     998079
                  2
                          Adventure | Drama | Romance
                                                        M 45
                                                                         5
     [998080 rows x 8 columns]
[75]: #Explore the datasets using visual representations (graphs or tables), also
       → include your comments on the following:
[76]: #User Age Distribution
      Master_Data['Age'].value_counts()
[76]: 25
            394766
      35
            198580
      18
            183088
      45
             83490
      50
             72382
      56
             38727
      1
             27047
      Name: Age, dtype: int64
[77]: Master_Data['Age'].value_counts().plot(kind='bar')
      plt.xlabel("Age")
      plt.title("User Age Distribution")
      plt.ylabel('Users Count')
```

plt.show()



[78]: #User rating of the movie "Toy Story"

toystoryRating=Master_Data[Master_Data['Title'].str.contains('Toy Story')==True]
toystoryRating

[78]:		MovieID				Title	UserID	Rating	Genres \
[,0].	315	3114	Tov	Story	2	(1999)	49	4	Animation Children's Comedy
	438	3114	•	•		(1999)	53	5	Animation Children's Comedy
	1179	3114	•	•		(1999)	58	5	Animation Children's Comedy
	1855	3114	•	•		(1999)	75	5	Animation Children's Comedy
	2022	3114	•	·		(1999)	78	4	Animation Children's Comedy
			J					_	
	990324		Tov	Storv		(1999)	1898	5	Animation Children's Comedy
	990350		•	•		(1999)	1970	4	Animation Children's Comedy
	990402	3114	•	•		(1999)	4741	4	Animation Children's Comedy
	990424	3114	•	•		(1999)	5168	4	Animation Children's Comedy
	990443	3114	•	•		(1999)	5713	4	Animation Children's Comedy
			·	·					·
		Gender A	Age O	ccupat	ior	ı			
	315	М	18	-	12				
	438	M	25		()			

```
1179
           M 25
                         2
1855
           F
              1
                         10
2022
           F 45
                         1
990324
           M 25
                         12
990350
           M 50
                         13
990402
           M 35
                         7
990424
           M 35
                         7
990443
           F 50
                         7
```

[1584 rows x 8 columns]

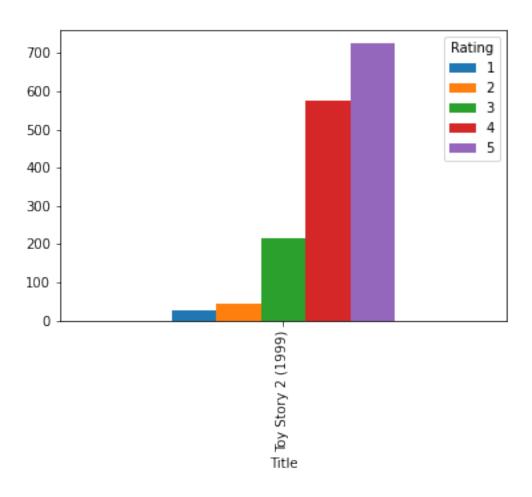
```
[79]: toystoryRating.groupby(['Title','Rating']).size()
```

```
[79]: Title Rating
Toy Story 2 (1999) 1 25
2 44
3 214
4 577
5 724
```

dtype: int64

```
[80]: toystoryRating.groupby(['Title','Rating']).size().unstack().plot(kind='bar')
```

[80]: <AxesSubplot:xlabel='Title'>

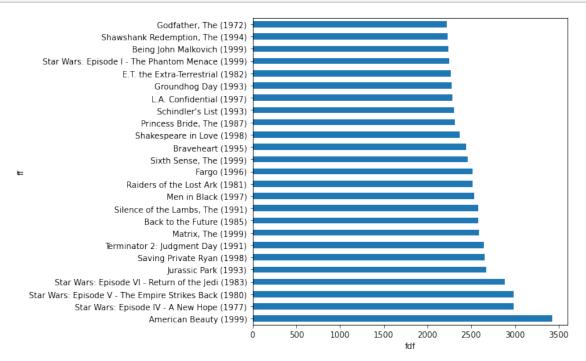


[81]: #Top 25 movies by viewership rating topmovie=Master_Data.groupby(['Title']).size().sort_values(ascending=False)[:25] topmovie

[81]: Title American Beauty (1999) 3428 Star Wars: Episode IV - A New Hope (1977) 2990 Star Wars: Episode V - The Empire Strikes Back (1980) 2990 Star Wars: Episode VI - Return of the Jedi (1983) 2883 Jurassic Park (1993) 2672 Saving Private Ryan (1998) 2652 Terminator 2: Judgment Day (1991) 2649 Matrix, The (1999) 2590 Back to the Future (1985) 2582 Silence of the Lambs, The (1991) 2578 Men in Black (1997) 2538 Raiders of the Lost Ark (1981) 2514 Fargo (1996) 2512 Sixth Sense, The (1999) 2458

```
2443
Braveheart (1995)
Shakespeare in Love (1998)
                                                           2369
Princess Bride, The (1987)
                                                           2317
Schindler's List (1993)
                                                           2303
L.A. Confidential (1997)
                                                           2288
Groundhog Day (1993)
                                                           2278
E.T. the Extra-Terrestrial (1982)
                                                           2268
Star Wars: Episode I - The Phantom Menace (1999)
                                                           2250
Being John Malkovich (1999)
                                                           2241
Shawshank Redemption, The (1994)
                                                           2227
Godfather, The (1972)
                                                           2223
dtype: int64
```

[82]: topmovie.plot(kind='barh',figsize=(7,7))
 plt.xlabel('fdf')
 plt.ylabel('ff')
 plt.show()



[83]: #Find the ratings for all the movies reviewed by for a particular user of user $\rightarrow id = 2696$

[84]: user_2696 = Master_Data[Master_Data['UserID'] == 2696]
user_2696

[84]: Empty DataFrame

Columns: [MovieID, Title, UserID, Rating, Genres, Gender, Age, Occupation]

Index: []

```
[85]: Master_Data[Master_Data['UserID']=='2696']
```

[85]:		MovieID		Title	UserID	Rating	\
	940348	1270	Back to the Future	(1985)	2696	2	
	940349	1097	E.T. the Extra-Terrestrial	(1982)	2696	3	
	940350	1617	L.A. Confidential	(1997)	2696	4	
	940351	800	Lone Star	(1996)	2696	5	
	940352	3386	JFK	(1991)	2696	1	
	940353	3176	Talented Mr. Ripley, The	(1999)	2696	4	
	940354	1711	Midnight in the Garden of Good and Evil	(1997)	2696	4	
	940355	1589	Cop Land	(1997)	2696	3	
	940356	1783	Palmetto	(1998)	2696	4	
	940357	1892	Perfect Murder, A	(1998)	2696	4	
	940358	1625	Game, The	(1997)	2696	4	
	940359	1644	I Know What You Did Last Summer	(1997)	2696	2	
	940360	1645	Devil's Advocate, The	(1997)	2696	4	
	940361	2389	Psycho	(1998)	2696	4	
	940362	1805	Wild Things	(1998)	2696	4	
	940363	1092	Basic Instinct	(1992)	2696	4	
	940364	2713	Lake Placid	(1999)	2696	1	
	940365	1258	Shining, The	(1980)	2696	4	
	940366	2338	I Still Know What You Did Last Summer	(1998)	2696	2	
	940367	350	Client, The	(1994)	2696	3	
			Genres Gender Age Occi	upation			
	940348		Comedy Sci-Fi M 25	7			
	940349	Childr	en's Drama Fantasy Sci-Fi M 25	7			
	940350	Crime F	ilm-Noir Mystery Thriller M 25	7			
	940351		Drama Mystery M 25	7			
	940352		Drama Mystery M 25	7			
	940353		Drama Mystery Thriller M 25	7			
	940354	C	omedy Crime Drama Mystery M 25	7			
	940355		Crime Drama Mystery M 25	7			
	940356	F	ilm-Noir Mystery Thriller M 25	7			
	940357		Mystery Thriller M 25	7			
	940358		Mystery Thriller M 25	7			
	940359		Horror Mystery Thriller M 25	7			
	940360	Crim	e Horror Mystery Thriller M 25	7			
	940361		Crime Horror Thriller M 25	7			
	940362	Cri	me Drama Mystery Thriller M 25	7			
	940363		Mystery Thriller M 25	7			
	940364		Horror Thriller M 25	7			
	940365		Horror M 25	7			

```
Drama|Mystery|Thriller
                                                     M 25
                                                                     7
      940367
[86]: #Feature Engineering
[87]: #Find out all the unique genres
      #(Hint: split the data in column genre making a list and
      #then process the data to find out only the unique categories of genres)
[88]: Master_Data['Genres'].unique()
[88]: array(["Animation|Children's|Musical", "Animation|Children's|Comedy",
             'Comedy|Fantasy', 'Animation', "Children's|Drama|Fantasy|Sci-Fi",
             'Animation|Comedy|Thriller', 'Thriller', 'Action|Crime|Romance',
             'Action|Adventure|Fantasy|Sci-Fi', 'Crime|Drama|Thriller',
             'Action|Adventure|Mystery', 'Crime|Drama',
             'Action|Adventure|Sci-Fi|Thriller', 'Action|Drama',
             'Comedy|Drama|Romance', 'Action|Thriller', 'Drama|Thriller',
             'Comedy|Romance', 'Action|Crime|Thriller', 'Drama|Sci-Fi',
             'Action|Sci-Fi|Thriller', 'Action|Horror|Sci-Fi', 'Comedy|Drama',
             'Action|Sci-Fi', 'Adventure|Drama|Romance|Sci-Fi',
             'Action|Adventure|Sci-Fi', 'Action|Adventure|Drama|Sci-Fi|War',
             'Action|Crime|Drama|Thriller', 'Comedy', 'Action|Comedy|Drama',
             'Drama|War', 'Action|Sci-Fi|War', 'Action|Adventure|Comedy|Sci-Fi',
             'Action|Adventure|Horror|Thriller', 'Adventure|Thriller',
             'Action|Horror|Sci-Fi|Thriller', 'Crime|Drama|Romance|Thriller',
             'Crime|Thriller', 'Crime|Film-Noir|Mystery|Thriller',
             'Mystery|Sci-Fi|Thriller', 'Action|Romance|Thriller',
             'Documentary', 'Horror|Sci-Fi', 'Comedy|Crime',
             'Drama|Sci-Fi|Thriller', "Children's|Comedy",
             "Children's | Comedy | Musical", 'Action | Comedy | Crime | Drama', 'Action',
             'Adventure | Sci-Fi', "Children's | Comedy | Drama",
             'Comedy|Fantasy|Romance', 'Drama|Romance|Thriller', 'Horror',
             'Action|Crime|Drama', 'Drama', 'Drama|Thriller|War',
             'Adventure | Animation | Sci-Fi | Thriller', 'Action | Crime',
             'Action|Comedy|Sci-Fi|Thriller', 'Comedy|Crime|Fantasy',
             'Animation|Comedy', "Adventure|Children's|Comedy|Fantasy|Sci-Fi",
             'Film-Noir | Mystery | Thriller', "Adventure | Children's | Fantasy",
             'Comedy|Mystery|Thriller', 'Action|Adventure|Crime|Drama',
             "Children's|Fantasy|Sci-Fi", 'Comedy|Sci-Fi',
             'Comedy|Horror|Musical|Sci-Fi', "Children's|Comedy|Fantasy",
             'Sci-Fi|War', "Children's|Sci-Fi", 'Mystery|Thriller',
             "Adventure | Children's | Fantasy | Sci-Fi", 'Adventure | Fantasy',
             'Drama|Mystery|Sci-Fi|Thriller', 'Crime|Drama|Sci-Fi',
             'Action|Adventure|Fantasy', 'Action|Sci-Fi|Thriller|War',
             'Adventure|War', 'Action|Drama|Fantasy|Romance', 'Sci-Fi|Thriller',
             'Adventure | Comedy | Sci-Fi', 'Comedy | Sci-Fi | Western',
```

Horror | Mystery | Thriller

M 25

7

940366

```
'Action|Adventure|Comedy|Horror|Sci-Fi',
"Adventure | Children's | Comedy | Fantasy", 'Film-Noir | Sci-Fi',
'Fantasy|Sci-Fi', 'Adventure|Fantasy|Sci-Fi',
'Action|Adventure|Sci-Fi|War', 'Comedy|Fantasy|Romance|Sci-Fi',
'Adventure|Fantasy|Romance', 'Action|Comedy|Crime|Horror|Thriller',
'Crime|Horror|Mystery|Thriller', 'Crime|Drama|Mystery|Thriller',
"Animation|Children's", 'Crime|Drama|Film-Noir',
'Horror|Sci-Fi|Thriller', 'Sci-Fi', 'Mystery|Sci-Fi',
'Action|Crime|Sci-Fi', 'Action|Romance|Sci-Fi',
'Action | Adventure | Animation | Horror | Sci-Fi',
'Action|Sci-Fi|Thriller|Western', 'Comedy|Crime|Thriller',
'Film-Noir|Sci-Fi|Thriller', 'Action|Comedy|Musical|Sci-Fi',
'Action|Drama|Sci-Fi|Thriller', 'Action|Comedy|Fantasy',
'Horror|Thriller', 'Comedy|Documentary',
'Action|Adventure|Animation', 'Action|Drama|War',
'Crime|Film-Noir|Thriller', 'Crime|Film-Noir',
'Film-Noir|Thriller', 'Comedy|Thriller', 'Comedy|Horror',
'Comedy|Crime|Drama', 'Action|Adventure|Comedy|Crime',
'Mystery|Romance|Thriller', 'Comedy|Mystery|Romance|Thriller',
'Action|Adventure|Sci-Fi|Thriller|War', 'Action|Horror',
'Action|Comedy', 'Crime', 'Adventure|Crime|Sci-Fi|Thriller',
'Drama|Fantasy', "Adventure|Animation|Children's|Comedy|Musical",
'Action|Adventure|Romance|Sci-Fi|War', 'Action|Romance|War',
'Action | Adventure | Comedy | Romance', 'Action | Adventure | Thriller',
'Action | Adventure | Romance',
"Animation|Children's|Comedy|Musical|Romance", 'Drama|Romance',
'Action|Romance', 'Animation|Sci-Fi', "Action|Children's|Fantasy",
'Animation|Musical', 'Action|Adventure|Crime',
'Action|Adventure|Drama|Thriller', 'Musical|Romance',
"Adventure | Children's | Drama | Musical", 'Musical',
"Animation | Children's | Musical | Romance",
"Animation | Children's | Comedy | Musical", 'Comedy | Drama | Western',
'Action|Comedy|Western', 'Adventure|Comedy|Drama',
'Action|Drama|Romance|Thriller', 'Action|Drama|Romance',
'Adventure | Drama | Western', 'Action | Adventure | Romance | Thriller',
'Action|Mystery|Romance|Thriller', 'Drama|Romance|War',
'Comedy | Romance | War', 'Action | Adventure | Horror',
'Action|Adventure|Comedy|Horror', 'Western', 'Comedy|Western',
'Action|Western', 'Adventure|Animation|Film-Noir',
'Action|Adventure|Comedy', 'Horror|Mystery|Thriller',
'Film-Noir | Mystery', 'Comedy | Crime | Mystery | Thriller',
'Drama | Musical', "Adventure | Animation | Children's | Musical",
'Comedy|Musical', "Children's|Fantasy|Musical",
'Comedy|Musical|Romance', 'Comedy|Crime|Drama|Mystery',
'Drama|Horror', 'Crime|Drama|Mystery', 'Comedy|Romance|Thriller',
"Animation|Children's|Drama|Fantasy", 'Horror|Romance',
"Animation|Children's|Fantasy|Musical", 'Comedy|War',
```

```
'Comedy|Horror|Musical', "Adventure|Children's|Musical",
'Action | Adventure | Romance | War',
"Action | Adventure | Children's | Comedy",
"Adventure|Children's|Sci-Fi", 'Comedy|Drama|War',
'Adventure | Animation | Sci-Fi', 'Action | Adventure | Mystery | Sci-Fi',
'Crime|Horror|Thriller', 'Adventure|Sci-Fi|Thriller',
'Drama|Fantasy|Romance|Thriller', 'Comedy|Drama|Sci-Fi',
"Adventure | Animation | Children's | Sci-Fi", "Children's | Drama",
'Action|War', 'Action|Comedy|Sci-Fi|War', 'Action|Adventure|Drama',
'Comedy | Crime | Horror', 'Action | Comedy | Musical',
'Action|Mystery|Thriller', 'Comedy|Mystery',
"Action | Animation | Children's | Sci-Fi | Thriller | War",
'Crime | Drama | Romance', "Animation | Children's | Fantasy | War",
'Drama | Mystery | Romance', 'Adventure | Drama | Romance', 'Film-Noir',
'Mystery', 'Adventure | Comedy | Musical', 'Action | Drama | Thriller',
'Romance|Thriller', 'Action|Mystery|Sci-Fi|Thriller',
'Action|Comedy|Romance|Thriller', 'Comedy|Mystery|Romance',
'Action|Crime|Mystery', 'Comedy|Drama|Thriller', 'Adventure|Drama',
'Drama|Mystery', "Children's|Comedy|Sci-Fi", 'Action|Adventure',
'Adventure | Musical', 'Adventure | Comedy | Romance', "Children's",
'Drama | Mystery | Thriller', 'Action | Horror | Thriller',
'Comedy|Drama|Musical', 'Action|Drama|Mystery',
'Adventure | Romance | Sci-Fi', 'Action | Sci-Fi | Western',
'Drama | Musical | War', 'Action | Drama | Mystery | Romance | Thriller',
"Adventure | Children's", 'Crime | Drama | Film-Noir | Thriller',
'Crime | Horror', 'Adventure', 'Film-Noir | Romance | Thriller',
'Romance', 'Musical|Romance|War', 'Drama|Western', 'Romance|War',
'Musical|War', 'War', "Adventure|Children's|Comedy",
"Adventure | Children's | Drama", "Children's | Drama | Fantasy",
"Children's | Musical", "Action | Adventure | Children's | Sci-Fi",
"Adventure | Animation | Children's | Comedy | Fantasy",
"Adventure | Children's | Comedy | Fantasy | Romance",
"Adventure | Children's | Comedy | Musical", "Children's | Comedy | Mystery",
"Animation|Children's|Comedy|Romance",
"Action | Adventure | Children's", "Action | Children's",
"Adventure | Children's | Romance", "Adventure | Animation | Children's",
"Action | Adventure | Animation | Children's | Fantasy",
'Action|Crime|Mystery|Thriller', 'Adventure|Comedy',
'Adventure | Romance', 'Adventure | Drama | Thriller',
'Comedy|Horror|Thriller', 'Action|Adventure|Drama|Romance',
'Action|Adventure|Western', 'Action|Adventure|Comedy|War',
"Adventure | Animation | Children's | Fantasy", 'Documentary | Musical',
"Children's|Horror", 'Crime|Film-Noir|Mystery', 'Drama|Film-Noir',
'Crime | Mystery', 'Drama | Romance | Sci-Fi', 'Drama | Horror | Thriller',
"Children's | Comedy | Western", 'Drama | Romance | War | Western',
'Action|Comedy|War', 'Action|Drama|Western', 'Documentary|Drama',
'Action|Thriller|War', "Action|Adventure|Children's|Fantasy",
```

```
'Adventure|Western', 'Sci-Fi|Thriller|War', 'Comedy|Horror|Sci-Fi',
'Horror|Mystery', 'Action|Drama|Thriller|War',
'Action|Adventure|Crime|Thriller', 'Drama|Film-Noir|Thriller',
'Documentary|War', 'Comedy|Romance|Sci-Fi', "Children's|Fantasy",
"Adventure|Children's|Drama|Romance", 'Action|Adventure|War',
'Action|Comedy|Crime', 'Adventure|Musical|Romance',
'Animation|Mystery', 'Drama|Romance|Western', 'Romance|Western',
'Comedy|Film-Noir|Thriller', 'Fantasy', 'Film-Noir|Horror'],
dtype=object)

]:
```

[]:	
[]:	