IMPORT THE LIBRARY

In [1]: import pandas as pd #import the libarry

READ THE DATASET

In [2]: movie = pd.read_csv(r'C:\Users\91996\Desktop\python\M movie.csv.csv')

In [3]: movie

title Out[3]: movield genres 0 1 Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy 1 2 Jumanji (1995) Adventure|Children|Fantasy 2 Grumpier Old Men (1995) 3 Comedy|Romance 3 Waiting to Exhale (1995) Comedy|Drama|Romance 4 Father of the Bride Part II (1995) Comedy 27273 131254 Kein Bund für's Leben (2007) Comedy 27274 131256 Feuer, Eis & Dosenbier (2002) Comedy 27275 131258 The Pirates (2014) Adventure 27276 131260 Rentun Ruusu (2001) (no genres listed) 27277 131262 Innocence (2014) Adventure|Fantasy|Horror

27278 rows × 3 columns

In [4]: movie.head()

ut[4]:		movield	title	genres
	0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy
	1	2	Jumanji (1995)	Adventure Children Fantasy
	2	3	Grumpier Old Men (1995)	Comedy Romance
	3	4	Waiting to Exhale (1995)	Comedy Drama Romance
	4	5	Father of the Bride Part II (1995)	Comedy

In [5]: tags = pd.read_csv(r'C:\Users\91996\Desktop\python\M tag.csv.csv')

In [6]: tags.head(1)

 Out[6]:
 userId
 movield
 tag
 timestamp

 0
 18
 4141
 Mark Waters
 2009-04-24 18:19:40

```
In [7]: | tags.info
        tag
                                                                                         timest
Out[7]:
                         4141
                                Mark Waters 2009-04-24 18:19:40
        0
                    18
                                   dark hero 2013-05-10 01:41:18
        1
                    65
                           208
        2
                   65
                            353
                                    dark hero 2013-05-10 01:41:19
        3
                  65
                           521 noir thriller 2013-05-10 01:39:43
                   65
                           592
                                   dark hero 2013-05-10 01:41:18
                   . . .
                           . . .
                                           . . .
        . . .
                       55999
                                    dragged 2013-01-23 23:29:32
        465559 138446
        465560 138446 55999 Jason Bateman 2013-01-23 23:29:38
        465561 138446
                                      quirky 2013-01-23 23:29:38
                        55999
        465562 138446
                         55999
                                       sad 2013-01-23 23:29:32
        465563 138472
                          923 rise to power 2007-11-02 21:12:47
        [465564 rows x 4 columns]>
In [8]: | rating=pd.read_csv(r'C:\Users\91996\Desktop\python\M rating.csv.csv')
In [9]: rating.head(1)
Out[9]:
         userId movieId rating
                                    timestamp
           1
                      2
                          3.5 2005-04-02 23:53:47
In [10]: rating.shape
        (20000263, 4)
Out[10]:
In [11]: rating.describe()
Out[11]:
                   userId
                             movield
                                         rating
        count 2.000026e+07 2.000026e+07 2.000026e+07
        mean 6.904587e+04 9.041567e+03 3.525529e+00
          std 4.003863e+04 1.978948e+04 1.051989e+00
          min 1.000000e+00 1.000000e+00 5.000000e-01
         25% 3.439500e+04 9.020000e+02 3.000000e+00
         50% 6.914100e+04 2.167000e+03 3.500000e+00
         75% 1.036370e+05 4.770000e+03 4.000000e+00
         max 1.384930e+05 1.312620e+05 5.000000e+00
In [12]: movie.columns
        Index(['movieId', 'title', 'genres'], dtype='object')
Out[12]:
In [13]:
        rating.columns
        Index(['userId', 'movieId', 'rating', 'timestamp'], dtype='object')
Out[13]:
        tags.columns
In [14]:
        Index(['userId', 'movieId', 'tag', 'timestamp'], dtype='object')
Out[14]:
In [15]: del rating['timestamp']
```

```
In [16]: rating.columns
         Index(['userId', 'movieId', 'rating'], dtype='object')
Out[16]:
         len(rating.columns)
In [17]:
Out[17]:
         del tags['timestamp']
In [18]:
         tags.columns
In [19]:
         Index(['userId', 'movieId', 'tag'], dtype='object')
Out[19]:
         DATA STRUCTURES
In [20]: tags.head()
            userld movield
Out[20]:
                                 tag
               18
                     4141 Mark Waters
                      208
                            dark hero
         2
               65
                      353
                            dark hero
                      521
                           noir thriller
               65
                      592
                            dark hero
         row 0 = tags.iloc[0]
In [21]:
         row 0
In [22]:
         userId
                              18
Out[22]:
         movieId
                            4141
                    Mark Waters
         tag
         Name: 0, dtype: object
         type(row 0)
In [23]:
         pandas.core.series.Series
Out[23]:
         row 0.index
In [24]:
         Index(['userId', 'movieId', 'tag'], dtype='object')
Out[24]:
```

row 0['userId']

movie['movieId']

1

2 3 4

In [25]:

Out[25]:

In [26]:

Out[26]:

```
27273
                    131254
          27274
                   131256
          27275
                   131258
                   131260
          27276
                   131262
          27277
         Name: movieId, Length: 27278, dtype: int64
          'tag' in row 0
In [27]:
          True
Out[27]:
          row 0.tag
In [28]:
          'Mark Waters'
Out[28]:
          tags.iloc[ [0,11,300] ]
In [29]:
Out[29]:
               userld movield
                                                               tag
            0
                        4141
                                                        Mark Waters
                  18
                        1783
                                                         noir thriller
           11
                  65
          300
                 316
                       45186 Ethan Hunt Should Stop Hogging The Screen!
```

DESCRIPTIVE STATS

. . .

```
In [30]: movie.describe
         <bound method NDFrame.describe of</pre>
                                                  movieId
                                                                                            title
Out[30]:
                                            Toy Story (1995)
                      1
                                              Jumanji (1995)
         1
                      2
                     3
         2
                                    Grumpier Old Men (1995)
         3
                      4
                                    Waiting to Exhale (1995)
                      5 Father of the Bride Part II (1995)
         27273 131254
                               Kein Bund für's Leben (2007)
         27274 131256
                             Feuer, Eis & Dosenbier (2002)
         27275
                131258
                                          The Pirates (2014)
               131260
                                         Rentun Ruusu (2001)
         27276
         27277
               131262
                                            Innocence (2014)
                                                       genres
                Adventure | Animation | Children | Comedy | Fantasy
         1
                                  Adventure | Children | Fantasy
         2
                                              Comedy | Romance
         3
                                        Comedy|Drama|Romance
         4
                                                       Comedy
         27273
                                                       Comedy
         27274
                                                       Comedy
         27275
                                                   Adventure
         27276
                                          (no genres listed)
         27277
                                    Adventure|Fantasy|Horror
         [27278 rows x 3 columns]>
In [31]: rating.describe
         <bound method NDFrame.describe of</pre>
                                                       userId movieId rating
Out[31]:
```

```
1
                         1
                                 29
                                         3.5
         2
                                         3.5
                         1
                                 32
                         1
                                 47
                                        3.5
                                        3.5
         4
                         1
                                 50
                       . . .
                                        . . .
         . . .
                                . . .
         20000258 138493
                             68954
                                        4.5
         20000259 138493
                              69526
                                        4.5
         20000260 138493
                              69644
                                         3.0
         20000261 138493
                              70286
                                        5.0
         20000262 138493
                              71619
                                        2.5
         [20000263 rows x 3 columns]>
In [48]: rating.mean()
                    69045.872583
         userId
Out[48]:
         movieId
                     9041.567330
                         3.525529
         rating
         dtype: float64
In [33]: rating['rating'].mean()
         3.5255285642993797
Out[33]:
         rating['rating'].min()
In [34]:
Out[34]:
In [36]: rating['rating'].max()
Out[36]:
         rating['rating'].std()
In [37]:
         1.051988919275684
Out[37]:
         rating['rating'].mode()
In [38]:
              4.0
Out[38]:
         Name: rating, dtype: float64
         rating.corr()
In [39]:
Out[39]:
                   userId
                          movield
                                    rating
                1.000000
                         -0.000850 0.001175
          userId
                         1.000000 0.002606
                -0.000850
         movield
                 rating
In [40]: filter1 = rating['rating'] > 10
         print(filter1)
         filter1.any()
                     False
         1
                     False
         2
                     False
         3
                     False
                     False
                     . . .
         20000258
                     False
         20000259
                     False
```

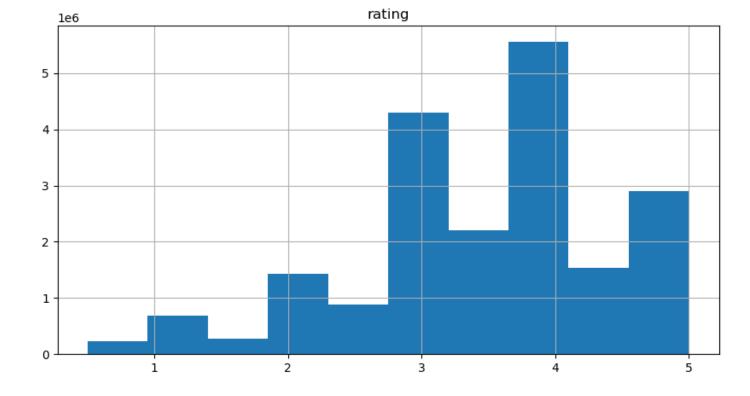
```
20000260
                     False
         20000261
                     False
         20000262
                    False
         Name: rating, Length: 20000263, dtype: bool
Out[40]:
In [41]:
         filter2 = rating['rating'] > 0
         filter2.all()
         True
Out[41]:
         movie.shape
In [43]:
         (27278, 3)
Out[43]:
```

Data Cleaning: Handling Missing Data

```
movie.isnull().any().any()
In [44]:
         False
Out[44]:
In [45]:
         rating.shape
         (20000263, 3)
Out[45]:
In [46]:
         rating.isnull().any().any()
         False
Out[46]:
         tags.shape
In [48]:
          (465564, 3)
Out[48]:
         tags.isnull().any().any()
In [49]:
Out[49]:
         tags=tags.dropna()
In [50]:
         tags.isnull().any().any()
In [51]:
         False
Out[51]:
In [52]:
         tags.shape
         (465548, 3)
Out[52]:
```

Data Visualization

```
In [53]: %matplotlib inline
    rating.hist(column='rating', figsize=(10,5))
Out[53]: array([[<AxesSubplot:title={'center':'rating'}>]], dtype=object)
```



```
In [54]: rating.boxplot(column='rating', figsize=(10,5))
Out[54]:

5
4
4
3
```

Slicing Out Columns

2

1

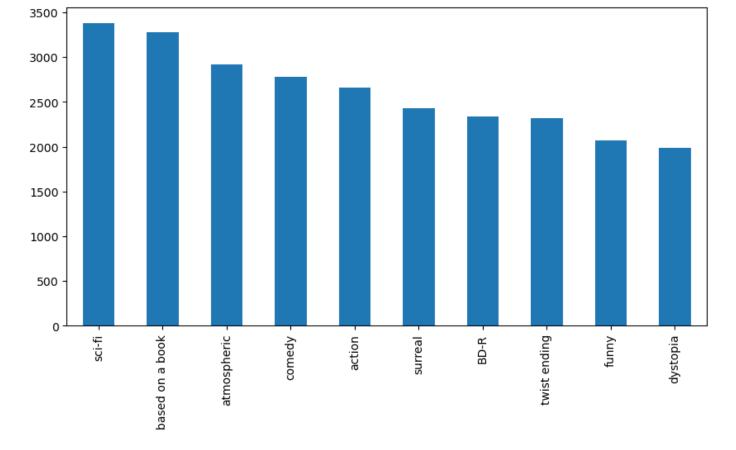
rating

```
Out[56]:
                                   title
                                                                        genres
          0
                          Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy
          1
                          Jumanji (1995)
                                                        Adventure|Children|Fantasy
          2
                  Grumpier Old Men (1995)
                                                               Comedy|Romance
          3
                   Waiting to Exhale (1995)
                                                         Comedy|Drama|Romance
          4 Father of the Bride Part II (1995)
                                                                       Comedy
In [57]:
          rating[-10:]
                     userId movieId rating
Out[57]:
          20000253 138493
                              60816
                                       4.5
          20000254 138493
                              61160
                                       4.0
          20000255 138493
                              65682
                                       4.5
          20000256 138493
                              66762
                                       4.5
          20000257 138493
                              68319
                                       4.5
          20000258 138493
                              68954
                                       4.5
          20000259 138493
                              69526
                                       4.5
          20000260 138493
                              69644
                                        3.0
          20000261 138493
                              70286
                                        5.0
          20000262 138493
                              71619
                                        2.5
          tag counts = tags['tag'].value counts()
In [58]:
          tag counts[-10:]
                                                 1
          missing child
Out[58]:
          Ron Moore
                                                 1
          Citizen Kane
                                                 1
          mullet
                                                 1
          biker gang
                                                 1
          Paul Adelstein
                                                 1
                                                 1
          the wig
          killer fish
                                                 1
          genetically modified monsters
                                                 1
          topless scene
                                                 1
          Name: tag, dtype: int64
          tag counts[:10].plot(kind='bar', figsize=(10,5))
In [59]:
```

In [56]: movie[['title','genres']].head()

<AxesSubplot:>

Out[59]:



In [60]: is_highly_rated = rating['rating'] >= 5.0
 rating[is_highly_rated][30:50]

Out[60]:		userId	movield	rating
	239	3	50	5.0

239	3	50	5.0
242	3	175	5.0
244	3	223	5.0
245	3	260	5.0
246	3	316	5.0
247	3	318	5.0
248	3	329	5.0
252	3	457	5.0
253	3	480	5.0
254	3	490	5.0
256	3	541	5.0
258	3	593	5.0
263	3	858	5.0
264	3	904	5.0
267	3	924	5.0
268	3	953	5.0
271	3	1060	5.0
272	3	1073	5.0
275	3	1084	5.0

In [61]: is_action= movie['genres'].str.contains('Action')
 movie[is_action][5:15]

Out[61]:		movield	title	genres
	22	23	Assassins (1995)	Action Crime Thriller
	41	42	Dead Presidents (1995)	Action Crime Drama
	43	44	Mortal Kombat (1995)	Action Adventure Fantasy
	50	51	Guardian Angel (1994)	Action Drama Thriller
	65	66	Lawnmower Man 2: Beyond Cyberspace (1996)	Action Sci-Fi Thriller
	69	70	From Dusk Till Dawn (1996)	Action Comedy Horror Thriller
	70	71	Fair Game (1995)	Action
	75	76	Screamers (1995)	Action Sci-Fi Thriller
	77	78	Crossing Guard, The (1995)	Action Crime Drama Thriller
	85	86	White Squall (1996)	Action Adventure Drama

In [62]: movie[is_action].head(15)

Out[62]:

	movield	title	genres
5	6	Heat (1995)	Action Crime Thriller
8	9	Sudden Death (1995)	Action
9	10	GoldenEye (1995)	Action Adventure Thriller
14	15	Cutthroat Island (1995)	Action Adventure Romance
19	20	Money Train (1995)	Action Comedy Crime Drama Thriller
22	23	Assassins (1995)	Action Crime Thriller
41	42	Dead Presidents (1995)	Action Crime Drama
43	44	Mortal Kombat (1995)	Action Adventure Fantasy
50	51	Guardian Angel (1994)	Action Drama Thriller
65	66	Lawnmower Man 2: Beyond Cyberspace (1996)	Action Sci-Fi Thriller
69	70	From Dusk Till Dawn (1996)	Action Comedy Horror Thriller
70	71	Fair Game (1995)	Action
75	76	Screamers (1995)	Action Sci-Fi Thriller
77	78	Crossing Guard, The (1995)	Action Crime Drama Thriller
85	86	White Squall (1996)	Action Adventure Drama

Group By and Aggregate

```
Out[63]:
                movield
          rating
            0.5
                 239125
                680732
            1.0
                279252
            1.5
            2.0 1430997
            2.5
                883398
            3.0 4291193
            3.5 2200156
            4.0 5561926
            4.5 1534824
            5.0 2898660
In [64]: average_rating = rating[['movieId', 'rating']].groupby('movieId').mean()
          average rating.head()
Out[64]:
                    rating
         movield
```

1 3.921240

2 3.211977

3 3.151040

4 2.861393

5 3.064592

```
In [65]: movie_count = rating[['movieId','rating']].groupby('movieId').count()
    movie_count.head()
    rating
```

Out[65]:

	userId	movield	rating
0	1	2	3.5
1	1	29	3.5
2	1	32	3.5
3	1	47	3.5
4	1	50	3.5
20000258	138493	68954	4.5
20000259	138493	69526	4.5
20000260	138493	69644	3.0
20000261	138493	70286	5.0
20000262	138493	71619	2.5

Toy Story

(1995)

2

```
In [67]:
            movie count = rating[['movieId','rating']].groupby('movieId').count()
           movie count.tail()
Out[67]:
                    rating
           movield
           131254
                         1
           131256
           131258
                         1
           131260
           131262
          Merge Dataframes
           tags.head()
In [69]:
Out[69]:
              userld movield
                                      tag
          0
                 18
                         4141
                               Mark Waters
                          208
                                 dark hero
                 65
           2
                 65
                          353
                                 dark hero
           3
                 65
                          521
                                noir thriller
                 65
                          592
                                 dark hero
           movie.head()
In [70]:
                                              title
Out[70]:
              movield
                                                                                      genres
          0
                    1
                                    Toy Story (1995)
                                                    Adventure|Animation|Children|Comedy|Fantasy
                    2
                                      Jumanji (1995)
                                                                     Adventure|Children|Fantasy
           2
                    3
                            Grumpier Old Men (1995)
                                                                             Comedy|Romance
           3
                             Waiting to Exhale (1995)
                                                                      Comedy|Drama|Romance
                    5 Father of the Bride Part II (1995)
                                                                                     Comedy
           t = movie.merge(tags, on='movieId', how='inner')
In [71]:
           t.head()
Out[71]:
             movield
                                title
                                                                        genres
                                                                               userId
                                                                                                                  tag
                            Toy Story
          0
                    1
                                      Adventure|Animation|Children|Comedy|Fantasy
                                                                                 1644
                                                                                                              Watched
                               (1995)
                            Toy Story
                    1
                                      Adventure|Animation|Children|Comedy|Fantasy
                                                                                 1741
                                                                                                    computer animation
                               (1995)
```

Adventure|Animation|Children|Comedy|Fantasy

1741

Disney animated feature

```
1 Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy 1741 Pixar animation (1995)

1 Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy 1741

Téa Leoni does not star in this movie
```

Combine aggreagation, merging, and filters to get useful analytics

```
In [74]: avg_rating= rating.groupby('movieId', as_index=False).mean()
    del avg_rating['userId']
    avg_rating.head()
```

```
      Out[74]:
      movield
      rating

      0
      1
      3.921240

      1
      2
      3.211977

      2
      3
      3.151040

      3
      4
      2.861393

      4
      5
      3.064592
```

```
In [75]: box_office = movie.merge(avg_ratings, on='movieId', how='inner')
box_office.tail()
```

Out[75]:		movield	title	genres	userId	rating
	26739	131254	Kein Bund für's Leben (2007)	Comedy	79570.0	4.0
	26740	131256	Feuer, Eis & Dosenbier (2002)	Comedy	79570.0	4.0
	26741	131258	The Pirates (2014)	Adventure	28906.0	2.5
	26742	131260	Rentun Ruusu (2001)	(no genres listed)	65409.0	3.0
	26743	131262	Innocence (2014)	Adventure Fantasy Horror	133047.0	4.0

```
In [76]: is_highly_rated = box_office['rating'] >= 4.0
box_office[is_highly_rated][-5:]
```

Out[76]:		movield	title	genres	userId	rating
	26737	131250	No More School (2000)	Comedy	79570.0	4.0
	26738	131252	Forklift Driver Klaus: The First Day on the Jo	Comedy Horror	79570.0	4.0
	26739	131254	Kein Bund für's Leben (2007)	Comedy	79570.0	4.0
	26740	131256	Feuer, Eis & Dosenbier (2002)	Comedy	79570.0	4.0
	26743	131262	Innocence (2014)	Adventure Fantasy Horror	133047.0	4.0

```
In [77]: is_Adventure = box_office['genres'].str.contains('Adventure')
box_office[is_Adventure][:5]
```

Out[77]:		movield	title	genres	userId	rating
	0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	69282.396821	3.921240
			, , ,	1 1 1 21 2		

1	2	Jumanji (1995)	Adventure Children Fantasy	69169.928202	3.211977
7	8	Tom and Huck (1995)	Adventure Children	68677.092580	3.142049
9	10	GoldenEye (1995)	Action Adventure Thriller	69161.741045	3.430029
12	13	Balto (1995)	Adventure Animation Children	70136.308693	3.272416

```
In [78]: box_office[is_Adventure & is_highly_rated][-5:]
```

	movield	title	genres	userId	rating
26611	130586	Itinerary of a Spoiled Child (1988)	Adventure Drama	130784.0	4.5
26655	130996	The Beautiful Story (1992)	Adventure Drama Fantasy	31411.0	5.0
26667	131050	Stargate SG-1 Children of the Gods - Final Cut	Adventure Sci-Fi Thriller	79570.0	5.0
26736	131248	Brother Bear 2 (2006)	Adventure Animation Children Comedy Fantasy	79570.0	4.0
26743	131262	Innocence (2014)	Adventure Fantasy Horror	133047.0	4.0

Vectorized String Operations

Out[78]:

[n [80]:	mo	movie.head()				
Out[80]:	movield		title	genres		
	0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy		
	1	2	Jumanji (1995)	Adventure Children Fantasy		
	2	3	Grumpier Old Men (1995)	Comedy Romance		
	3	4	Waiting to Exhale (1995)	Comedy Drama Romance		
	4	5	Father of the Bride Part II (1995)	Comedy		

Split 'genres' into multiple columns

```
In [81]:
          movie genres = movie['genres'].str.split('|', expand=True)
          movie_genres[:10]
In [82]:
                                        2
                                                3
                                                               5
                                                                                        9
Out[82]:
                               1
                                                        4
                                                                     6
                                                                           7
                                                                                  8
                                  Children Comedy
            Adventure
                       Animation
                                                   Fantasy
                                                                  None
                                                                        None
             Adventure
                         Children
                                   Fantasy
                                             None
                                                     None
                                                           None
                                                                 None
                                                                        None
                                                                              None
                                                                                     None
          2
               Comedy
                                                                                     None
                        Romance
                                    None
                                             None
                                                     None
                                                           None
                                                                 None
                                                                        None
                                                                              None
          3
               Comedy
                                 Romance
                                             None
                                                           None
                                                                  None
                                                                              None
                                                                                     None
                          Drama
                                                     None
                                                                        None
               Comedy
                           None
                                    None
                                             None
                                                     None
                                                           None
                                                                 None
                                                                        None
                                                                              None
                                                                                     None
          5
                Action
                           Crime
                                   Thriller
                                                                                     None
                                             None
                                                     None
                                                           None
                                                                 None
                                                                        None
                                                                              None
                                             None
                                                                              None
                                                                                     None
               Comedy
                        Romance
                                                           None
                                                                  None
                                                                        None
                                    None
                                                     None
             Adventure
                         Children
                                    None
                                             None
                                                           None
                                                                 None
                                                                        None
                                                                              None
                                                                                     None
```

Add a new column for comedy genre flag

[83]:	<pre>movie_genres['isComedy'] = movie['genres'].str.contains('Comedy')</pre>												
[84]:	mc	movie_genres[:10]											
[84]:		0	1	2	3	4	5	6	7	8	9	isComedy	
	0	Adventure	Animation	Children	Comedy	Fantasy	None	None	None	None	None	True	
	1	Adventure	Children	Fantasy	None	None	None	None	None	None	None	False	
	2	Comedy	Romance	None	None	None	None	None	None	None	None	True	
	3	Comedy	Drama	Romance	None	None	None	None	None	None	None	True	
	4	Comedy	None	None	None	None	None	None	None	None	None	True	
	5	Action	Crime	Thriller	None	None	None	None	None	None	None	False	
	6	Comedy	Romance	None	None	None	None	None	None	None	None	True	
	7	Adventure	Children	None	None	None	None	None	None	None	None	False	
	8	Action	None	None	None	None	None	None	None	None	None	False	
	9	Action	Adventure	Thriller	None	None	None	None	None	None	None	False	

Extract year from title e.g. (2007)

```
movie['year'] = movie['title'].str.extract('.*\((.*)\).*', expand=True)
In [86]:
          movie.tail()
In [88]:
Out[88]:
                  movield
                                                title
                                                                     genres year
           27273
                  131254
                            Kein Bund für's Leben (2007)
                                                                    Comedy 2007
                  131256
                           Feuer, Eis & Dosenbier (2002)
                                                                    Comedy 2002
           27274
           27275
                   131258
                                    The Pirates (2014)
                                                                  Adventure 2014
           27276
                   131260
                                  Rentun Ruusu (2001)
                                                            (no genres listed) 2001
          27277
                   131262
                                     Innocence (2014) Adventure|Fantasy|Horror 2014
```