import required libraries

```
In [3]: import pandas as pd
  import numpy as np
  import matplotlib.pyplot as plt
  import seaborn as sns
```

Read the dataset

In [197... movies = pd.read_csv(r'C:\Users\91939\Desktop\AI&DS\12thAug\archive\movie.csv')
movies

out[197		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
			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

```
Out[8]:
             userld movield
                                     tag
                                                  timestamp
          0
                18
                       4141 Mark Waters 2009-04-24 18:19:40
          1
                65
                         208
                                dark hero 2013-05-10 01:41:18
          2
                65
                         353
                                dark hero 2013-05-10 01:41:19
                               noir thriller 2013-05-10 01:39:43
          3
                65
                         521
          4
                65
                         592
                                dark hero 2013-05-10 01:41:18
 In [9]: tags.shape
 Out[9]: (465564, 4)
In [10]: tags.columns
Out[10]: Index(['userId', 'movieId', 'tag', 'timestamp'], dtype='object')
In [11]:
         ratings = pd.read_csv(r'C:\Users\91939\Desktop\AI&DS\12thAug\archive\rating.csv'
          ratings.head()
Out[11]:
             userld movield rating
                                            timestamp
                                3.5 2005-04-02 23:53:47
          0
                 1
                          2
                         29
                                3.5 2005-04-02 23:31:16
          1
          2
                 1
                          32
                                3.5 2005-04-02 23:33:39
                                3.5 2005-04-02 23:32:07
          3
                 1
                          47
                 1
          4
                          50
                                3.5 2005-04-02 23:29:40
In [12]: ratings.shape
Out[12]: (20000263, 4)
In [13]: ratings.columns
Out[13]: Index(['userId', 'movieId', 'rating', 'timestamp'], dtype='object')
In [17]: del tags['timestamp']
In [18]:
         tags.columns
Out[18]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [24]: #del ratings['timestamp']
In [20]:
         ratings.columns
Out[20]: Index(['userId', 'movieId', 'rating'], dtype='object')
In [25]: ratings.head(1)
```

```
Out[25]: userld movield rating

0 1 2 3.5
```

```
In [26]: tags.head(1)
```

Out[26]:		userId	movield	tag
	0	18	4141	Mark Waters

```
In [27]: tags.head()
```

Out[27]:		userId	movield	tag
	0	18	4141	Mark Waters
	1	65	208	dark hero
	2	65	353	dark hero
	3	65	521	noir thriller
	4	65	592	dark hero

iloc - returns the individual data by indexing

```
In [28]: tags.iloc[0]
Out[28]: userId
                              18
         movieId
                           4141
         tag
                    Mark Waters
         Name: 0, dtype: object
In [29]: tags.iloc[0,2]
Out[29]:
         'Mark Waters'
In [33]: tags.iloc[0][2]
Out[33]: 'Mark Waters'
In [34]:
         import warnings
         warnings.filterwarnings('ignore')
In [35]: tags.iloc[0:5]
```

Out[35]:		userId	movield	tag
	0	18	4141	Mark Waters
	1	65	208	dark hero
	2	65	353	dark hero
	3	65	521	noir thriller
	4	65	592	dark hero

Data Structures

Series

```
In [36]: row_0 = tags.iloc[0]
         type(row_0)
Out[36]: pandas.core.series.Series
In [37]: print(row_0)
                            18
        userId
        movieId
                          4141
                   Mark Waters
        tag
        Name: 0, dtype: object
In [38]: row_0.index
Out[38]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [39]: row_0['userId']
Out[39]: 18
        'rating' in row_0
In [40]:
Out[40]: False
         'movieId' in row_0
In [41]:
Out[41]: True
In [42]: row 0.name
Out[42]: 0
In [43]: row_0 = row_0.rename('firstRow')
         row_0.name
Out[43]: 'firstRow'
         Dataframes
In [44]:
         tags.head()
```

```
Out[44]:
               userld movield
                                          tag
           0
                   18
                           4141
                                 Mark Waters
           1
                   65
                            208
                                    dark hero
           2
                  65
                            353
                                    dark hero
           3
                   65
                            521
                                   noir thriller
           4
                   65
                            592
                                    dark hero
```

```
In [45]: tags.index
```

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Out[45]: RangeIndex(start=0, stop=465564, step=1)

In [46]: tags.columns

Out[46]: Index(['userId', 'movieId', 'tag'], dtype='object')

In [47]: tags.iloc[[0,11,500]]

 Out[47]:
 userId
 movield
 tag

 0
 18
 4141
 Mark Waters

 11
 65
 1783
 noir thriller

 500
 342
 55908
 entirely dialogue

In [48]: tags.describe()

Out[48]: userId movield **count** 465564.000000 465564.000000 68712.354263 32627.762920 mean 41877.674053 36080.241157 std min 18.000000 1.000000 25% 28780.000000 2571.000000 **50%** 70201.000000 7373.000000 **75**% 107322.000000 62235.000000 138472.000000 131258.000000

Descriptive Statistics

In [49]: ratings

Out[49]:		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

20000263 rows × 3 columns

```
In [50]: ratings['rating']
Out[50]: 0
                     3.5
         1
                     3.5
         2
                     3.5
         3
                     3.5
                     3.5
                     . . .
         20000258
                    4.5
         20000259
                     4.5
         20000260 3.0
         20000261
                   5.0
         20000262
                     2.5
         Name: rating, Length: 20000263, dtype: float64
In [51]: ratings['rating'].describe()
Out[51]: count
                  2.000026e+07
         mean
                  3.525529e+00
                  1.051989e+00
         std
                  5.000000e-01
         min
         25%
                  3.000000e+00
         50%
                  3.500000e+00
         75%
                  4.000000e+00
                  5.000000e+00
         max
         Name: rating, dtype: float64
In [52]: ratings.columns
Out[52]: Index(['userId', 'movieId', 'rating'], dtype='object')
In [53]: ratings['userId'].describe()
```

```
Out[53]: count
                  2.000026e+07
         mean
                  6.904587e+04
         std
                  4.003863e+04
                  1.000000e+00
         min
         25%
                  3.439500e+04
         50%
                  6.914100e+04
         75%
                  1.036370e+05
                  1.384930e+05
         max
         Name: userId, dtype: float64
         ratings.describe()
In [54]:
Out[54]:
                                  movield
                      userId
                                                 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 [55]:
         ratings['rating'].mean()
Out[55]: 3.5255285642993797
In [56]: ratings['rating'].min()
Out[56]: 0.5
In [57]: ratings['rating'].max()
Out[57]: 5.0
In [58]:
         ratings['rating'].std()
Out[58]: 1.051988919275684
         ratings['rating'].mode()
In [59]:
Out[59]: 0
              4.0
         Name: rating, dtype: float64
In [60]: ratings.mode()
Out[60]:
             userld movield rating
         0 118205
                        296
                               4.0
In [61]:
         ratings.corr()
```

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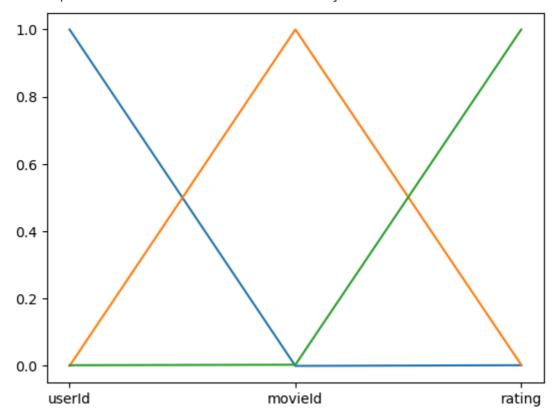
```
        userId
        movield
        rating

        userId
        1.000000
        -0.000850
        0.001175

        movield
        -0.000850
        1.000000
        0.002606

        rating
        0.001175
        0.002606
        1.000000
```

```
In [62]: plt.plot(ratings.corr())
```



```
In [63]: filter1 = ratings['rating'] > 10
         print(filter1)
        0
                     False
                     False
        1
        2
                     False
        3
                     False
                     False
        20000258
                     False
        20000259
                     False
        20000260
                     False
                     False
        20000261
        20000262
                     False
        Name: rating, Length: 20000263, dtype: bool
```

```
In [64]: filter1.any()
```

Out[64]: False

```
In [65]: filter2 = ratings['rating'] > 0
  filter2.all()
```

Out[65]: True

```
Data Cleaning: Handling Missing Data
In [66]: movies.shape
Out[66]: (27278, 3)
In [67]: movies.isnull()
Out[67]:
                  movield
                           title genres
               0
                     False False
                                   False
                     False False
                                   False
               2
                     False False
                                   False
                     False False
                                   False
               4
                     False False
                                   False
          27273
                     False False
                                   False
          27274
                     False False
                                   False
          27275
                     False False
                                   False
          27276
                     False False
                                   False
          27277
                     False False
                                   False
         27278 rows × 3 columns
In [68]: movies.isnull().any()
Out[68]:
          movieId
                      False
          title
                      False
```

```
Out[68]: movieId False
    title False
    genres False
    dtype: bool

In [69]: ratings.shape
Out[69]: (20000263, 3)

In [71]: ratings.isnull().any().any()
Out[71]: False

In [72]: ratings.isnull().any()
```

Out[72]: userId False movieId False rating False dtype: bool

In [73]: ratings.isnull()

Out[73]:

	userId	movield	rating
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False
20000258	False	False	False
20000259	False	False	False
20000260	False	False	False
20000261	False	False	False
20000262	False	False	False

20000263 rows × 3 columns

```
In [75]: tags.shape
Out[75]: (465564, 3)
In [76]: tags.isnull().any().any()
Out[76]: True
In [77]: tags=tags.dropna()
In [78]: tags
```

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Out[78]:		userId	movield	tag
	0	18	4141	Mark Waters
	1	65	208	dark hero
	2	65	353	dark hero
	3	65	521	noir thriller
	4	65	592	dark hero
	•••			
	465559	138446	55999	dragged
	465560	138446	55999	Jason Bateman
	465561	138446	55999	quirky
	465562	138446	55999	sad
	465563	138472	923	rise to power

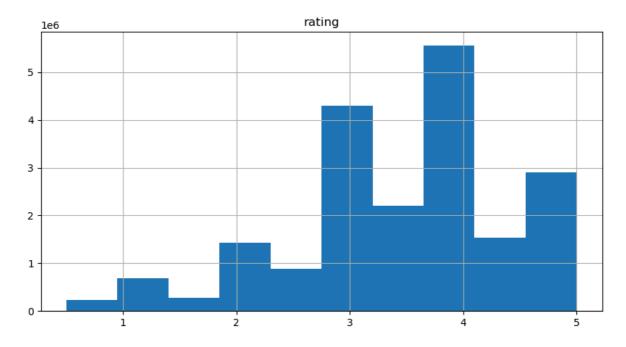
465548 rows × 3 columns

```
In [79]: tags.isnull().any().any()
Out[79]: False
In [80]: tags.shape
Out[80]: (465548, 3)
```

Data Visualization

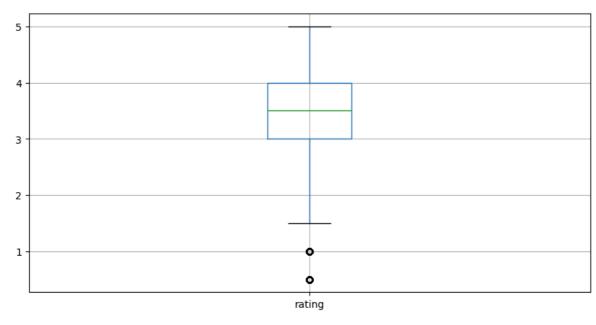
```
In [81]: %matplotlib inline
    ratings.hist(column='rating', figsize=(10,5))
```

Out[81]: array([[<Axes: title={'center': 'rating'}>]], dtype=object)



```
In [82]: ratings.boxplot(column='rating', figsize=(10,5))
```

Out[82]: <Axes: >



```
In [83]: ratings['rating'].min()
```

Out[83]: 0.5

```
In [84]: ratings['rating'].max()
```

Out[84]: 5.0

```
In [85]: ratings['rating'].median()
```

Out[85]: 3.5

Slicing Out Columns

```
In [86]: tags['tag'].head()
```

```
Out[86]: 0 Mark Waters
1 dark hero
2 dark hero
3 noir thriller
4 dark hero
Name: tag, dtype: object
```

In [89]: movies[['title', 'genres']].head()

Out[89]:	titla	genre

genres	titie	
Adventure Animation Children Comedy Fantasy	Toy Story (1995)	0
Adventure Children Fantasy	Jumanji (1995)	1
Comedy Romance	Grumpier Old Men (1995)	2
Comedy Drama Romance	Waiting to Exhale (1995)	3
Comedy	Father of the Bride Part II (1995)	4

In [90]: ratings.head(10)

Out[90]:

	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
5	1	112	3.5
6	1	151	4.0
7	1	223	4.0
8	1	253	4.0
9	1	260	4.0

In [91]: ratings[-10:]

Out[91]:

	userId	movield	rating
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

In [92]: ratings

Out[92]:

	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

20000263 rows × 3 columns

In [93]: tags

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Out[93]:		userId	movield	tag
	0	18	4141	Mark Waters
	1	65	208	dark hero
	2	65	353	dark hero
	3	65	521	noir thriller
	4	65	592	dark hero
	•••			
	465559	138446	55999	dragged
	465560	138446	55999	Jason Bateman
	465561	138446	55999	quirky
	465562	138446	55999	sad
	465563	138472	923	rise to power

465548 rows × 3 columns

```
In [94]: tags['tag'][-10:]
Out[94]: 465554
                                    visually appealing
                                       family friendly
           465555
           465556
                     Scary Movies To See on Halloween
           465557
                                             Peter Pan
                                    visually appealing
           465558
           465559
                                               dragged
           465560
                                         Jason Bateman
           465561
                                                quirky
           465562
                                                   sad
           465563
                                         rise to power
           Name: tag, dtype: object
In [102...
          tag_counts = tags['tag'].value_counts()
          tag_counts[:10]
Out[102...
           tag
           sci-fi
                              3384
           based on a book
                              3281
           atmospheric
                              2917
           comedy
                              2779
           action
                              2657
           surreal
                              2427
           BD-R
                              2334
           twist ending
                              2323
           funny
                              2072
                              1991
           dystopia
           Name: count, dtype: int64
In [103...
          tag_counts = tags['tag'].unique()
          tag_counts[:10]
```

```
Out[103...
           array(['Mark Waters', 'dark hero', 'noir thriller', 'bollywood',
                   'screwball comedy', 'mars', 'neo-noir', 'jesus', 'dragon',
                   'conspiracy theory'], dtype=object)
In [104...
           tag_counts = tags.tag.unique()
           tag_counts[:10]
           array(['Mark Waters', 'dark hero', 'noir thriller', 'bollywood',
Out[104...
                   'screwball comedy', 'mars', 'neo-noir', 'jesus', 'dragon',
                  'conspiracy theory'], dtype=object)
In [110...
           'dark hero' in tags
Out[110...
           False
          tags[tags['tag'] == 'dark hero']
In [133...
Out[133...
                    userld movield
                                         tag
                1
                       65
                               208 dark hero
                2
                               353 dark hero
                       65
                4
                       65
                               592 dark hero
               21
                              6874 dark hero
                       65
             1353
                      558
                              5146 dark hero
           460229 136015
                              6874 dark hero
           462399 137717
                              5146 dark hero
           462400 137717
                              7373 dark hero
           462466 137717
                             60684 dark hero
           462467 137717
                             91529 dark hero
          194 rows × 3 columns
In [123...
          print(len('Mark Waters'))
         11
In [117...
          tags[tags['tag'] == 'sci-fi']
```

\cap	14-	г	1	1	7	
U	ЛL	Ŀ	_	Τ	/	••

	userId	movield	tag
162	129	4878	sci-fi
259	190	2011	sci-fi
303	318	260	sci-fi
386	342	2571	sci-fi
438	342	8914	sci-fi
•••			
464719	138301	27904	sci-fi
464793	138301	52328	sci-fi
464865	138301	70286	sci-fi
464961	138301	109487	sci-fi
464970	138301	112623	sci-fi

3384 rows × 3 columns

In [107...

tags

Out[107...

tag	movield	userId	
Mark Waters	4141	18	0
dark hero	208	65	1
dark hero	353	65	2
noir thriller	521	65	3
dark hero	592	65	4
			•••
dragged	55999	138446	465559
Jason Bateman	55999	138446	465560
quirky	55999	138446	465561
sad	55999	138446	465562
rise to power	923	138472	465563

465548 rows × 3 columns

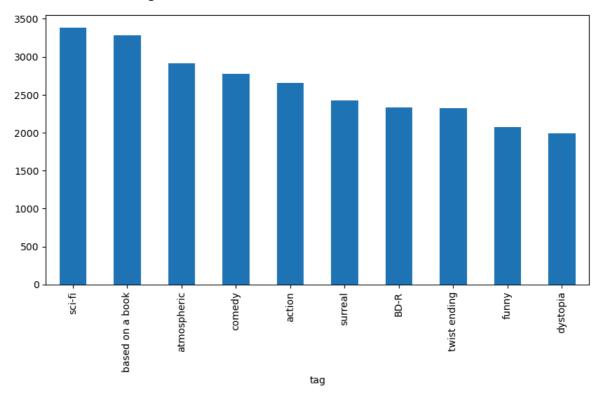
```
In [136...
```

```
tag_counts = tags['tag'].value_counts()
tag_counts
```

Out[136... tag sci-fi 3384 based on a book 3281 atmospheric 2917 comedy 2779 action 2657 . . . Paul Adelstein 1 the wig 1 killer fish 1 genetically modified monsters topless scene Name: count, Length: 38643, dtype: int64

In [144... tag_counts[:10].plot(kind='bar', figsize=(10,5))

Out[144... <Axes: xlabel='tag'>



In [172... tags.head(10)

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Out[172...

tag	movield	userId	
Mark Waters	4141	18	0
dark hero	208	65	1
dark hero	353	65	2
noir thriller	521	65	3
dark hero	592	65	4
bollywood	668	65	5
screwball comedy	898	65	6
noir thriller	1248	65	7
mars	1391	65	8
neo-noir	1617	65	9

Filters for Selecting Rows

```
In [145...
```

```
is_highly_rated = ratings['rating'] >= 5.0
ratings[is_highly_rated][30:50]
```

Out[145...

	userId	movield	rating
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
276	3	1089	5.0

```
In [146... is_action= movies['genres'].str.contains('Action')
    movies[is_action][5:15]
```

Out[146...

	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 [147...

movies[is_action].head(15)

Out[147...

r	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

```
In [148... ratings_count = ratings[['movieId','rating']].groupby('rating').count()
ratings_count
```

```
Out[148... movield
```

rating				
0.5	239125			
1.0	680732			
1.5	279252			
2.0	1430997			
2.5	883398			
3.0	4291193			
3.5	2200156			
4.0	5561926			
4.5	1534824			
5.0	2898660			

```
In [149... average_rating = ratings[['movieId','rating']].groupby('movieId').mean()
average_rating.head()
```

Out[149...

rating

movield

- **1** 3.921240
- **2** 3.211977
- **3** 3.151040
- **4** 2.861393
- **5** 3.064592

```
In [150... average_rating = ratings[['movieId','rating']].groupby('movieId').median()
average_rating.head()
```

Out[150...

rating

```
      4.0

      2
      3.0

      3
      3.0

      4
      3.0

      5
      3.0
```

```
In [151... movie_count = ratings[['movieId','rating']].groupby('movieId').count()
movie_count.head()
```

Out[151... rating

```
      movield

      1
      49695

      2
      22243

      3
      12735

      4
      2756

      5
      12161
```

```
In [152... movie_count = ratings[['movieId','rating']].groupby('movieId').count()
movie_count.tail()
```

Out[152... rating

movield	
131254	1
131256	1
131258	1
131260	1
131262	1

merge dataframes

In [153... tags.head()

Out[153...

	userId	movield	tag
0	18	4141	Mark Waters
1	65	208	dark hero
2	65	353	dark hero
3	65	521	noir thriller
4	65	592	dark hero

```
In [154... movies.head()
```

Out[154... genres movield title 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 4 Waiting to Exhale (1995) Comedy|Drama|Romance Father of the Bride Part II 5 4 Comedy (1995)In [161... movies.merge(tags, on='movieId', how='inner') t.head() Out[161... movield title genres userld tag Toy 0 1 Story Adventure|Animation|Children|Comedy|Fantasy 1644 Watched (1995)Toy computer 1 1 Adventure|Animation|Children|Comedy|Fantasy Story 1741 animation (1995)Toy Disney 2 Adventure|Animation|Children|Comedy|Fantasy 1741 animated Story (1995)feature Toy Pixar 3 1 Adventure|Animation|Children|Comedy|Fantasy Story 1741 animation (1995)Téa Leoni Toy does not Adventure|Animation|Children|Comedy|Fantasy 4 1 Story 1741 star in this (1995)

In [165... t[t['tag'] == 'Mark Waters']

movie

Out[165...

	movield	title	genres	userId	tag
84036	1648	House of Yes, The (1997)	Comedy Drama	70201	Mark Waters
162637	4141	Head Over Heels (2001)	Comedy Romance	18	Mark Waters
162639	4141	Head Over Heels (2001)	Comedy Romance	70201	Mark Waters
212960	6593	Freaky Friday (2003)	Children Comedy Fantasy	70201	Mark Waters
234375	7451	Mean Girls (2004)	Comedy	70201	Mark Waters
234395	7451	Mean Girls (2004)	Comedy	77463	Mark Waters
281094	36525	Just Like Heaven (2005)	Comedy Fantasy Romance	70201	Mark Waters
336445	58105	Spiderwick Chronicles, The (2008)	Adventure Children Drama Fantasy IMAX	70201	Mark Waters
336458	58105	Spiderwick Chronicles, The (2008)	Adventure Children Drama Fantasy IMAX	107711	Mark Waters
413659	87483	Mr. Popper's	Comedy	70201	Mark

Combine aggreagation, merging, and filters to get useful analytics

Comedy

70201

Waters

```
In [166...
          avg_ratings= ratings.groupby('movieId', as_index=False).mean()
          del avg_ratings['userId']
          avg_ratings.head()
```

Out[166...

n	novield	rating
0	1	3.921240
1	2	3.211977
2	3	3.151040
3	4	2.861393
4	5	3.064592

Penguins

(2011)

```
In [167...
            box office = movies.merge(avg ratings, on='movieId', how='inner')
            box_office.tail()
Out[167...
                    movield
                                                      title
                                                                             genres rating
            26739
                     131254
                               Kein Bund für's Leben (2007)
                                                                            Comedy
                                                                                         4.0
            26740
                     131256
                              Feuer, Eis & Dosenbier (2002)
                                                                            Comedy
                                                                                         4.0
            26741
                     131258
                                         The Pirates (2014)
                                                                          Adventure
                                                                                         2.5
            26742
                     131260
                                       Rentun Ruusu (2001)
                                                                   (no genres listed)
                                                                                         3.0
                                          Innocence (2014) Adventure|Fantasy|Horror
            26743
                     131262
                                                                                         4.0
            is_highly_rated = box_office['rating'] >= 4.0
In [168...
            box_office[is_highly_rated][-5:]
Out[168...
                    movield
                                                              title
                                                                                             rating
                                                                                     genres
                     131250
                                            No More School (2000)
            26737
                                                                                    Comedy
                                                                                                 4.0
                                Forklift Driver Klaus: The First Day on
                     131252
            26738
                                                                             Comedy|Horror
                                                                                                 4.0
                                                           the Jo...
            26739
                     131254
                                        Kein Bund für's Leben (2007)
                                                                                    Comedy
                                                                                                 4.0
                                       Feuer, Eis & Dosenbier (2002)
            26740
                      131256
                                                                                    Comedy
                                                                                                 4.0
            26743
                     131262
                                                  Innocence (2014) Adventure|Fantasy|Horror
                                                                                                 4.0
            is_highly_rated = box_office['rating'] >= 4.0
In [169...
            box_office[is_highly_rated].tail()
Out[169...
                    movield
                                                              title
                                                                                     genres
                                                                                              rating
            26737
                     131250
                                            No More School (2000)
                                                                                                 4.0
                                                                                    Comedy
                                Forklift Driver Klaus: The First Day on
                     131252
            26738
                                                                             Comedy|Horror
                                                                                                 4.0
                                                           the Jo...
            26739
                     131254
                                        Kein Bund für's Leben (2007)
                                                                                    Comedy
                                                                                                 4.0
            26740
                      131256
                                       Feuer, Eis & Dosenbier (2002)
                                                                                    Comedy
                                                                                                 4.0
            26743
                     131262
                                                  Innocence (2014) Adventure|Fantasy|Horror
                                                                                                 4.0
```

is_Adventure = box_office['genres'].str.contains('Adventure')

box office[is Adventure][:5]

In [170...

Out[170...

 mo	vield	title	genres	rating
0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	3.921240
1	2	Jumanji (1995)	Adventure Children Fantasy	3.211977
7	8	Tom and Huck (1995)	Adventure Children	3.142049
9	10	GoldenEye (1995)	Action Adventure Thriller	3.430029
12	13	Balto (1995)	Adventure Animation Children	3.272416

In [171...

box_office[is_Adventure & is_highly_rated][-5:]

Out[171...

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

Vectorized String Operations

In [172...

movies.head()

Out[172...

m	ovield	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 [175... movies['genres']

8/18/24, 10:49 PM IMDBData analysis

```
Adventure | Animation | Children | Comedy | Fantasy
Out[175...
           0
                                        Adventure | Children | Fantasy
           1
           2
                                                     Comedy | Romance
           3
                                              Comedy | Drama | Romance
           4
                                                             Comedy
           27273
                                                             Comedy
           27274
                                                             Comedy
           27275
                                                          Adventure
           27276
                                                 (no genres listed)
           27277
                                          Adventure | Fantasy | Horror
           Name: genres, Length: 27278, dtype: object
           movie_genres = movies['genres'].str.split('|', expand=True)
In [181...
In [179...
           movie_genres[:10]
Out[179...
                      0
                                 1
                                           2
                                                     3
                                                             4
                                                                    5
                                                                           6
                                                                                 7
                                                                                        8
                                                                                               9
                                              Comedy
              Adventure
                         Animation
                                     Children
                                                        Fantasy
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
              Adventure
                           Children
                                      Fantasy
                                                 None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
           2
                Comedy
                          Romance
                                        None
                                                 None
                                                                                           None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
           3
                Comedy
                                    Romance
                             Drama
                                                 None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
           4
                Comedy
                                                 None
                             None
                                        None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
           5
                             Crime
                                       Thriller
                  Action
                                                 None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
           6
                Comedy
                           Romance
                                        None
                                                 None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
           7
              Adventure
                           Children
                                        None
                                                 None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
           8
                  Action
                              None
                                        None
                                                 None
                                                         None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
           9
                  Action
                         Adventure
                                       Thriller
                                                 None
                                                          None
                                                                None
                                                                       None
                                                                              None
                                                                                     None
                                                                                           None
```

Add a new column for comedy genre flag

```
In [185... movie_genres['isComedy'] = movies['genres'].str.contains('Comedy')
In [187... movie_genres[:10]
```

Out[187...

	0	1	2	3	4	5	6	7	8	9
0	Adventure	Animation	Children	Comedy	Fantasy	None	None	None	None	None
1	Adventure	Children	Fantasy	None	None	None	None	None	None	None
2	Comedy	Romance	None	None	None	None	None	None	None	None
3	Comedy	Drama	Romance	None	None	None	None	None	None	None
4	Comedy	None	None	None	None	None	None	None	None	None
5	Action	Crime	Thriller	None	None	None	None	None	None	None
6	Comedy	Romance	None	None	None	None	None	None	None	None
7	Adventure	Children	None	None	None	None	None	None	None	None
8	Action	None	None	None	None	None	None	None	None	None
9	Action	Adventure	Thriller	None	None	None	None	None	None	None
4										>

Extract year from title e.g. (2007)

```
In [198...
           movies.columns
            Index(['movieId', 'title', 'genres'], dtype='object')
Out[198...
In [199...
            movies.head()
Out[199...
               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
                               Waiting to Exhale (1995)
                                                                            Comedy|Drama|Romance
                               Father of the Bride Part II
            4
                      5
                                                                                           Comedy
                                                (1995)
            movies['year'] = movies['title'].str.extract('.*\((.*)\).*', expand=True)
In [200...
In [193...
            movies.tail()
Out[193...
                    movield
                                                     title
                                                                            genres
                                                                                    year
            27273
                     131254
                               Kein Bund für's Leben (2007)
                                                                           Comedy
                                                                                    2007
            27274
                     131256
                              Feuer, Eis & Dosenbier (2002)
                                                                           Comedy
                                                                                    2002
                                                                                   2014
            27275
                     131258
                                        The Pirates (2014)
                                                                         Adventure
```

Rentun Ruusu (2001)

Innocence (2014)

(no genres listed)

Adventure|Fantasy|Horror 2014

2001

27276

27277

131260

131262

Parsing Timestamps

```
tags = pd.read csv(r'C:\Users\91939\Desktop\AI&DS\12thAug\archive\tag.csv')
In [201...
In [202...
           tags.dtypes
Out[202...
           userId
                           int64
           movieId
                           int64
           tag
                          object
           timestamp
                         object
           dtype: object
In [203...
           tags.head(5)
Out[203...
              userld movield
                                                     timestamp
                                        tag
           0
                  18
                          4141
                                Mark Waters
                                             2009-04-24 18:19:40
           1
                  65
                           208
                                  dark hero
                                             2013-05-10 01:41:18
           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
           4
           tags['parsed_time'] = pd.to_datetime(tags['timestamp'])
In [206...
In [207...
           tags['parsed_time'].dtype
Out[207...
           dtype('<M8[ns]')</pre>
In [208...
           tags.head(2)
Out[208...
              userId movieId
                                        tag
                                                     timestamp
                                                                        parsed_time
           0
                                Mark Waters 2009-04-24 18:19:40
                  18
                          4141
                                                                 2009-04-24 18:19:40
                  65
                           208
                                  dark hero 2013-05-10 01:41:18 2013-05-10 01:41:18
           greater_than_t = tags['parsed_time'] > '2015-02-01'
In [209...
           selected_rows = tags[greater_than_t]
           tags.shape, selected_rows.shape
Out[209...
           ((465564, 5), (12130, 5))
In [213...
           tags.sort_values(by='parsed_time', ascending=True)[:10]
```

Out[213...

	userId	movield	tag	timestamp	parsed_time
333932	100371	2788	monty python	2005-12-24 13:00:10	2005-12-24 13:00:10
333927	100371	1732	coen brothers	2005-12-24 13:00:36	2005-12-24 13:00:36
333924	100371	1206	stanley kubrick	2005-12-24 13:00:48	2005-12-24 13:00:48
333923	100371	1193	jack nicholson	2005-12-24 13:02:51	2005-12-24 13:02:51
333939	100371	5004	peter sellers	2005-12-24 13:03:19	2005-12-24 13:03:19
333922	100371	47	morgan freeman	2005-12-24 13:03:32	2005-12-24 13:03:32
333921	100371	47	brad pitt	2005-12-24 13:03:32	2005-12-24 13:03:32
333936	100371	4011	brad pitt	2005-12-24 13:03:51	2005-12-24 13:03:51
333937	100371	4011	guy ritchie	2005-12-24 13:03:51	2005-12-24 13:03:51
333920	100371	32	bruce willis	2005-12-24 13:04:02	2005-12-24 13:04:02

In [214...

average_rating = ratings[['movieId','rating']].groupby('movieId', as_index=False
average_rating.tail()

Out[214...

	movield	rating
26739	131254	4.0
26740	131256	4.0
26741	131258	2.5
26742	131260	3.0
26743	131262	4.0