

kaggle task

```
In [1]: import pandas as pd  
import matplotlib.pyplot as plt
```

```
In [2]: ratings=pd.read_csv(r'E:\rating.csv')
```

```
In [3]: ratings
```

```
Out[3]:
```

	userId	movieId	rating	timestamp
0	1	2	3.5	2005-04-02 23:53:47
1	1	29	3.5	2005-04-02 23:31:16
2	1	32	3.5	2005-04-02 23:33:39
3	1	47	3.5	2005-04-02 23:32:07
4	1	50	3.5	2005-04-02 23:29:40
...
20000258	138493	68954	4.5	2009-11-13 15:42:00
20000259	138493	69526	4.5	2009-12-03 18:31:48
20000260	138493	69644	3.0	2009-12-07 18:10:57
20000261	138493	70286	5.0	2009-11-13 15:42:24
20000262	138493	71619	2.5	2009-10-17 20:25:36

20000263 rows × 4 columns

```
In [4]: movies=pd.read_csv(r'E:\movie.csv')
```

```
In [5]: movies
```

Out[5]:

	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
...			
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 [6]: `tags=pd.read_csv(r'E:\tag.csv')`In [7]: `tags`

Out[7]:

	userId	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
3	65	521	noir thriller	2013-05-10 01:39:43
4	65	592	dark hero	2013-05-10 01:41:18
...				
465559	138446	55999	dragged	2013-01-23 23:29:32
465560	138446	55999	Jason Bateman	2013-01-23 23:29:38
465561	138446	55999	quirky	2013-01-23 23:29:38
465562	138446	55999	sad	2013-01-23 23:29:32
465563	138472	923	rise to power	2007-11-02 21:12:47

465564 rows × 4 columns

```
In [8]: print(tags.columns)
print(ratings.columns)
print(movies.columns)
```

```
Index(['userId', 'movieId', 'tag', 'timestamp'], dtype='object')
Index(['userId', 'movieId', 'rating', 'timestamp'], dtype='object')
Index(['movieId', 'title', 'genres'], dtype='object')
```

```
In [9]: del ratings['timestamp']
del tags['timestamp']
```

```
In [10]: print(tags.columns)
print(ratings.columns)
print(movies.columns)
```

```
Index(['userId', 'movieId', 'tag'], dtype='object')
Index(['userId', 'movieId', 'rating'], dtype='object')
Index(['movieId', 'title', 'genres'], dtype='object')
```

```
In [11]: tags.head(3)
```

```
Out[11]:   userId  movieId      tag
0         18      4141  Mark Waters
1         65       208  dark hero
2         65      353  dark hero
```

```
In [12]: row_0 = tags.iloc[0]
row_0
```

```
Out[12]: userId           18
movieId        4141
tag      Mark Waters
Name: 0, dtype: object
```

```
In [13]: row_0.index
```

```
Out[13]: Index(['userId', 'movieId', 'tag'], dtype='object')
```

```
In [14]: row_0['userId']
```

```
Out[14]: np.int64(18)
```

```
In [15]: 'rating' in row_0
```

```
Out[15]: False
```

```
In [16]: row_0.name
```

```
Out[16]: 0
```

```
In [17]: row_0=row_0.rename('firstRow')
row_0.name
```

```
Out[17]: 'firstRow'
```

```
In [18]: tags.head()
```

```
Out[18]:
```

	userId	movieId	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 [19]: tags.index
```

```
Out[19]: RangeIndex(start=0, stop=465564, step=1)
```

```
In [20]: tags.columns
```

```
Out[20]: Index(['userId', 'movieId', 'tag'], dtype='object')
```

```
In [21]: tags.iloc[[0,11,500]]
```

```
Out[21]:
```

	userId	movieId	tag
0	18	4141	Mark Waters
11	65	1783	noir thriller
500	342	55908	entirely dialogue

```
In [22]: #descriptive statistics
ratings ['rating'].describe()
```

```
Out[22]:
```

count	2.000026e+07
mean	3.525529e+00
std	1.051989e+00
min	5.000000e-01
25%	3.000000e+00
50%	3.500000e+00
75%	4.000000e+00
max	5.000000e+00
Name:	rating, dtype: float64

```
In [23]: ratings.describe()
```

Out[23]:

	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 [24]: ratings['rating'].min()

Out[24]: 0.5

In [25]: ratings['rating'].max()

Out[25]: 5.0

In [26]: ratings['rating'].std()

Out[26]: 1.051988919275684

In [27]: ratings['rating'].mode()

Out[27]: 0 4.0
Name: rating, dtype: float64

In [28]: ratings.corr()

	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 [29]: filter1=ratings['rating']>10
print(filter1)
filter1.any()

```
0      False
1      False
2      False
3      False
4      False
...
20000258  False
20000259  False
20000260  False
20000261  False
20000262  False
Name: rating, Length: 20000263, dtype: bool
```

```
Out[29]: np.False_
```

```
In [30]: filter2=ratings['rating']>0
filter2.all()
```

```
Out[30]: np.True_
```

```
In [31]: #Data cleaning:handling missing data
movies.shape
```

```
Out[31]: (27278, 3)
```

```
In [32]: movies.isnull().any().any()
```

```
Out[32]: np.False_
```

```
In [33]: ratings.shape
```

```
Out[33]: (20000263, 3)
```

```
In [34]: ratings.isnull().any().any()
```

```
Out[34]: np.False_
```

```
In [35]: tags.shape
```

```
Out[35]: (465564, 3)
```

```
In [36]: tags.isnull().any().any()
```

```
Out[36]: np.True_
```

```
In [37]: tags=tags.dropna()
```

```
In [38]: tags.isnull().any().any()
```

```
Out[38]: np.False_
```

```
In [39]: tags.shape
```

Out[39]: (465548, 3)

Data Visualization

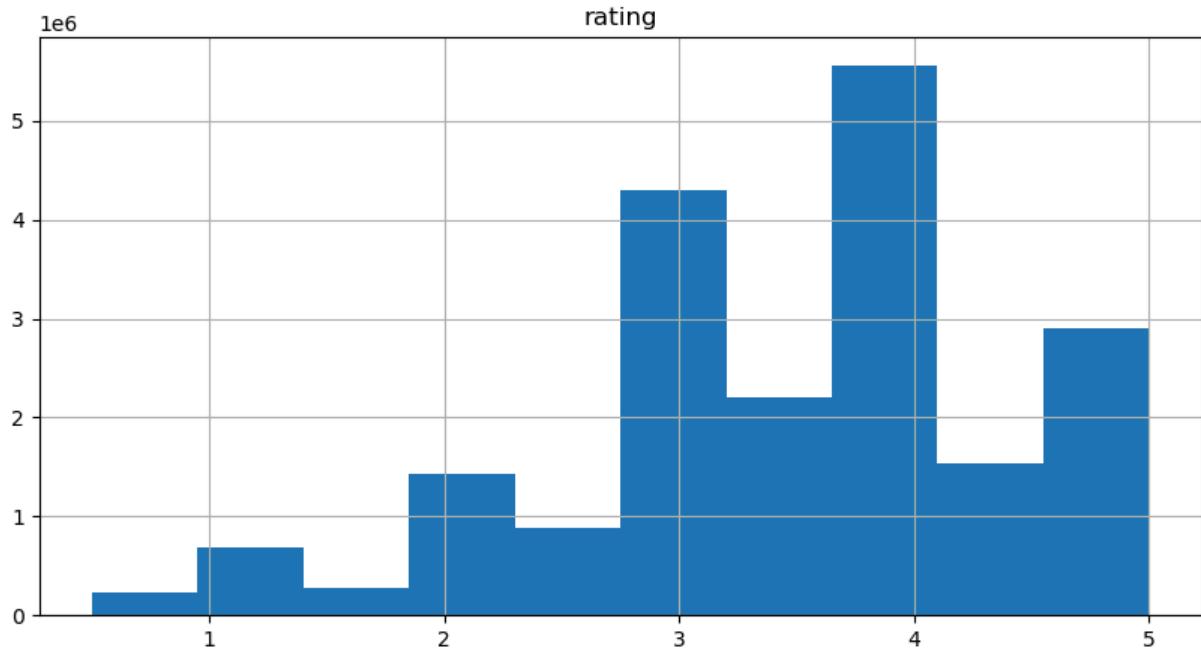
In [40]:

```
%matplotlib inline  
ratings.hist(column='rating', figsize=(10,5))
```

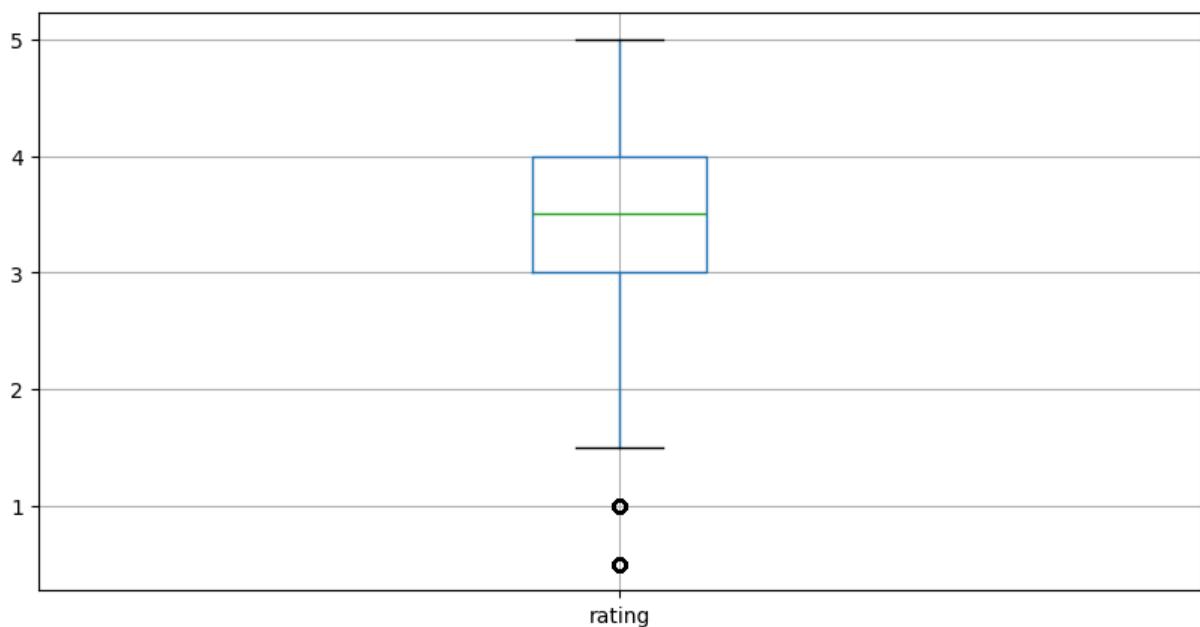
Out[40]: array([[[<Axes: title={'center': 'rating'}>]]], dtype=object)

In [41]:

```
plt.show()
```

In [42]:

```
ratings.boxplot(column='rating', figsize=(10,5))  
plt.show()
```



Slicing out Columns

```
In [43]: tags['tag'].head()
```

```
Out[43]: 0      Mark Waters
          1      dark hero
          2      dark hero
          3    noir thriller
          4      dark hero
Name: tag, dtype: object
```

```
In [44]: movies[['title', 'genres']].head()
```

	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 [45]: ratings[-10:]
```

Out[45]:

	userId	movieId	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 [46]:

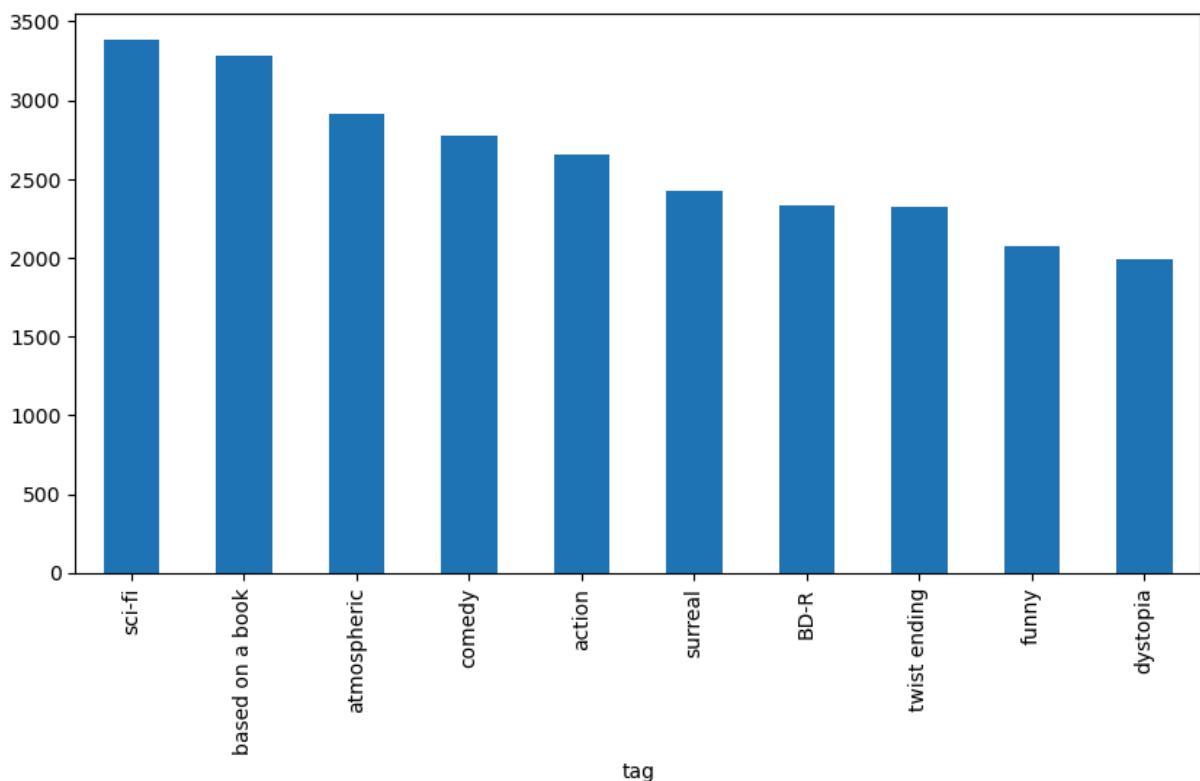
```
tags_counts=tags['tag'].value_counts()  
tags_counts[-10:]
```

Out[46]:

```
tag  
Hell naw          1  
This is my happy face    1  
I heel toe on Uday's house 1  
Why?             1  
Bobo              1  
Diamond Dallas Page    1  
I'm Devon Butler!      1  
No arguement        1  
Really Bad          1  
Botox              1  
Name: count, dtype: int64
```

In [47]:

```
tags_counts[:10].plot(kind='bar', figsize=(10,5))  
plt.show()
```



filters for selecting rows

```
In [52]: is_highly_rated=ratings['rating']>=5.0  
ratings[is_highly_rated][30:50]
```

Out[52]:

	userId	movieId	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 [53]:

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

Out[53]:

	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 [54]:

movies[is_action].head(15)

Out[54]:

	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 [56]: ratings_count=ratings[['movieId','rating']].groupby('rating').count()  
ratings_count
```

Out[56]:

rating	movield
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 [58]: average_rating=ratings[['movieId','rating']].groupby('movieId').mean()  
average_rating.head()
```

Out[58]:

rating	movield
1	3.921240
2	3.211977
3	3.151040
4	2.861393
5	3.064592

```
In [59]: movie_count=ratings[['movieId','rating']].groupby('movieId').count()  
movie_count.head()
```

Out[59]:

rating

movield	
1	49695
2	22243
3	12735
4	2756
5	12161

```
In [60]: movie_count=ratings[['movieId','rating']].groupby('movieId').count()
movie_count.tail()
```

Out[60]:

rating

movield	
131254	1
131256	1
131258	1
131260	1
131262	1

Merge Dataframes

```
In [61]: tags.head()
```

Out[61]:

	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 [62]: movies.head()
```

Out[62]:

	moviedb	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 [68]:

```
t=movies.merge(tags,on='movieId',how='inner')
t.head()
```

Out[68]:

	moviedb	title	genres	userId	tag
0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1644	Watched
1	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	computer animation
2	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	Disney animated feature
3	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	Pixar animation
4	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	TÃ©la Leoni does not star in this movie

Combine aggregation,merging and filters to get useful analytics

In [71]:

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

```
Out[71]:
```

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

```
In [74]: box_office=movies.merge(avg_ratings, on='movieId', how='inner')
box_office.tail()
```

```
Out[74]:
```

	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
26743	131262	Innocence (2014)	Adventure Fantasy Horror	4.0

```
In [75]: is_highly Rated = box_office['rating'] >= 4.0
box_office[is_highly Rated][-5:]
```

```
Out[75]:
```

	movield	title	genres	rating
26737	131250	No More School (2000)	Comedy	4.0
26738	131252	Forklift Driver Klaus: The First Day on the Jo...	Comedy Horror	4.0
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

```
In [76]: is_Adventure = box_office['genres'].str.contains('Adventure')
box_office[is_Adventure][:5]
```

Out[76]:

	movield	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 [78]: `box_office[is_Adventure & is_highly Rated][-5:]`

Out[78]:

	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 Operation

In [79]: `movies.head()`

Out[79]:

	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 [84]: `#split genres into multiple columns`
`movies_genres=movies['genres'].str.split('|', expand=True)`
`movies_genres[:10]`

Out[84]:

	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

In [86]:

```
#add a new column for comedy genre flag
movies_genres['isComedy']=movies['genres'].str.contains('Comedy')
movies_genres[:10]
```

Out[86]:

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



In [89]:

```
#extract year from title
movies['year']=movies['title'].str.extract('.*\((.*))\.*',expand=True)
movies.tail()
```

Out[89]:

	moviedb	title	genres	year
27273	131254	Kein Bund für's Leben (2007)	Comedy	2007
27274	131256	Feuer, Eis & Dosenbier (2002)	Comedy	2002
27275	131258	The Pirates (2014)	Adventure	2014
27276	131260	Rentun Ruusu (2001)	(no genres listed)	2001
27277	131262	Innocence (2014)	Adventure Fantasy Horror	2014

In [90]: tags.dtypes

```
Out[90]: userId      int64
movieId     int64
tag         object
dtype: object
```

In [92]: tags.head(10)

	userId	moviedb	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
5	65	668	bollywood
6	65	898	screwball comedy
7	65	1248	noir thriller
8	65	1391	mars
9	65	1617	neo-noir

```
In [95]: average_rating=ratings[['movieId','rating']].groupby('movieId',as_index=False).mean
average_rating.tail()
```

Out[95]:

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

```
In [103...]: joined=movies.merge(average_rating,on='movieId',how='inner')
joined.head()
```

Out[103...]:

	movield	title	genres	year	rating
0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1995	3.921240
1	2	Jumanji (1995)	Adventure Children Fantasy	1995	3.211977
2	3	Grumpier Old Men (1995)	Comedy Romance	1995	3.151040
3	4	Waiting to Exhale (1995)	Comedy Drama Romance	1995	2.861393
4	5	Father of the Bride Part II (1995)	Comedy	1995	3.064592

```
In [101...]: average_rating.corr()
```

Out[101...]:

	movield	rating
movield	1.000000	-0.090369
rating	-0.090369	1.000000

In []: