In [106... import pandas as pd

In [107... movies = pd.read\_csv(r'C:\Users\HAI\Downloads\archive\movie.csv')
 movies

Out[107...

genres	title	movield	
Adventure Animation Children Comedy Fantasy	Toy Story (1995)	1	0
Adventure   Children   Fantasy	Jumanji (1995)	2	1
Comedy Romance	Grumpier Old Men (1995)	3	2
Comedy Drama Romance	Waiting to Exhale (1995)	4	3
Comedy	Father of the Bride Part II (1995)	5	4
			•••
Comedy	Kein Bund für's Leben (2007)	131254	27273
Comedy	Feuer, Eis & Dosenbier (2002)	131256	27274
Adventure	The Pirates (2014)	131258	27275
(no genres listed)	Rentun Ruusu (2001)	131260	27276
Adventure Fantasy Horror	Innocence (2014)	131262	27277

27278 rows × 3 columns

In [108...

tags= pd.read\_csv(r'C:\Users\HAI\Downloads\archive\tag.csv')
tags

$\cap$		+	Γ	1	a	0	
U	u	L	н	ь.	U	O	

	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 [109...

ratings = pd.read\_csv(r'C:\Users\HAI\Downloads\archive\rating.csv')
ratings

Out[109...

	userId	movield	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 [110...

movies.head(20)

Tandas With Data Goldhoo_Ai						
genres	title	movield	[110 n			
re Animation Children Comedy Fantasy	Toy Story (1995)	<b>0</b> 1				
Adventure Children Fantasy	Jumanji (1995)	2	1			
Comedy Romance	Grumpier Old Men (1995)	3	2			
Comedy Drama Romance	Waiting to Exhale (1995)	4	3			
Comedy	Father of the Bride Part II (1995)	5	4			
Action Crime Thriller	Heat (1995)	6	5			
Comedy Romance	Sabrina (1995)	7	6			
Adventure Children	Tom and Huck (1995)	8	7			
Action	Sudden Death (1995)	9	8			
Action Adventure Thriller	GoldenEye (1995)	10	9			
Comedy Drama Romance	American President, The (1995)	11	10			
Comedy Horror	Dracula: Dead and Loving It (1995)	12	11			
Adventure   Animation   Children	Balto (1995)	13	12			
Drama	Nixon (1995)	14	13			
Action Adventure Romance	Cutthroat Island (1995)	15	14			

Casino (1995)

Calls (1995)

Four Rooms (1995)

Money Train (1995)

Sense and Sensibility (1995)

Ace Ventura: When Nature

In [111...

tags.head()

15

16

17

18

19

16

17

18

19

20

Out[111...

	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

In [112...

ratings.head()

Crime|Drama

Comedy

Comedy

Drama|Romance

Action | Comedy | Crime | Drama | Thriller

```
Out[112...
              userld movield rating
                                              timestamp
           0
                   1
                            2
                                  3.5 2005-04-02 23:53:47
           1
                           29
                                  3.5 2005-04-02 23:31:16
           2
                   1
                           32
                               3.5 2005-04-02 23:33:39
           3
                           47
                                  3.5 2005-04-02 23:32:07
                   1
           4
                           50
                                  3.5 2005-04-02 23:29:40
In [113...
          del ratings['timestamp']
           del tags['timestamp']
           row_0 = tags.iloc[0]
In [114...
           type(row_0)
Out[114... pandas.core.series.Series
In [115...
          print(row_0)
                              18
         userId
         movieId
                            4141
                   Mark Waters
         Name: 0, dtype: object
In [116...
          row_0.index
Out[116... Index(['userId', 'movieId', 'tag'], dtype='object')
In [117...
           row_0['userId']
Out[117...
           18
In [118...
          'rating' in row_0
Out[118...
           False
In [119...
           row_0.name
Out[119...
In [120...
           row_0 = row_0.rename('firstRow')
           row_0.name
Out[120... 'firstRow'
In [121... tags.head()
```

```
Out[121...
              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 [122...
           tags.index
Out[122...
           RangeIndex(start=0, stop=465564, step=1)
In [123...
           tags.columns
           Index(['userId', 'movieId', 'tag'], dtype='object')
Out[123...
In [124...
           tags.iloc[ [0,11,500] ]
Out[124...
                 userld movield
                                              tag
              0
                     18
                            4141
                                      Mark Waters
             11
                    65
                            1783
                                       noir thriller
           500
                    342
                           55908 entirely dialogue
In [125...
           ratings['rating'].describe()
Out[125...
           count
                     2.000026e+07
                     3.525529e+00
           mean
           std
                     1.051989e+00
                     5.000000e-01
           min
           25%
                     3.000000e+00
                     3.500000e+00
           50%
           75%
                     4.000000e+00
                     5.000000e+00
           Name: rating, dtype: float64
In [126...
           ratings.describe()
```

rating

Out[126...

userId

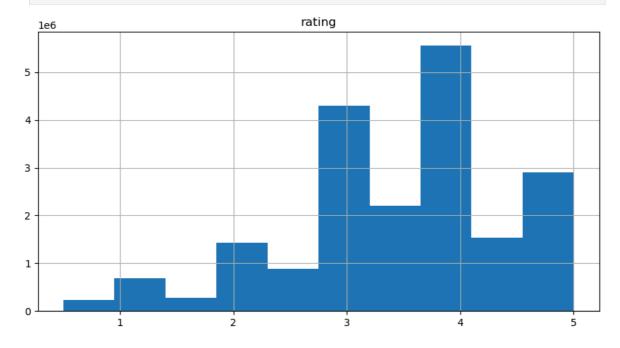
movield

```
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 [127...
           ratings['rating'].mean()
Out[127...
           3.5255285642993797
In [128...
           ratings.mean()
Out[128...
                      69045.872583
           userId
                       9041.567330
           movieId
                           3.525529
           rating
           dtype: float64
In [129...
           ratings['rating'].min()
Out[129...
           0.5
In [130...
           ratings['rating'].max()
Out[130...
           5.0
In [131...
           ratings['rating'].std()
Out[131...
           1.051988919275684
In [132...
           ratings['rating'].mode()
Out[132...
           Name: rating, dtype: float64
           ratings.corr()
In [133...
Out[133...
                       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 [134...
           filter1 = ratings['rating'] > 10
           print(filter1)
```

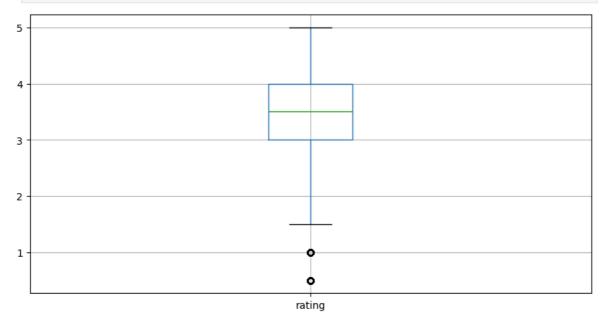
```
filter1.any()
         0
                      False
         1
                      False
          2
                      False
         3
                      False
                      False
                      . . .
         20000258
                      False
         20000259 False
         20000260 False
                      False
         20000261
         20000262
                      False
         Name: rating, Length: 20000263, dtype: bool
Out[134...
           False
In [135...
           filter2 = ratings['rating'] > 0
           filter2.all()
Out[135...
           True
In [136...
           movies.shape
Out[136...
           (27278, 3)
In [137...
           movies.isnull().any().any()
Out[137...
           False
In [138...
           ratings.shape
Out[138...
          (20000263, 3)
           ratings.isnull().any().any()
In [139...
Out[139...
           False
In [140...
           tags.shape
Out[140...
           (465564, 3)
In [141...
           tags.isnull().any().any()
Out[141...
           True
In [142...
           tags=tags.dropna()
In [143...
           tags.isnull().any().any()
Out[143...
           False
In [144...
           tags.shape
Out[144...
           (465548, 3)
In [145...
           %matplotlib inline
```

# import matplotlib.pyplot as plt

```
In [146... ratings.hist(column='rating', figsize=(10,5))
plt.show()
```



```
In [147... ratings.boxplot(column='rating', figsize=(10,5))
    plt.show()
```

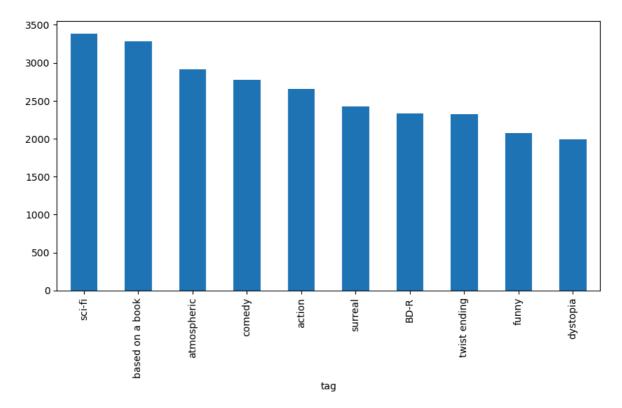


2 dark hero
3 noir thriller
4 dark hero

Name: tag, dtype: object

In [149... movies[['title', 'genres']].head()

[149			ti	itle	genres
	0	To	y Story (19	95) Adv	venture Animation Children Comedy Fantasy
	1		Jumanji (19	95)	Adventure Children Fantasy
	<b>2</b> Gi	rumpier O	ld Men (19	95)	Comedy Romance
	3	Waiting to	Exhale (19	95)	Comedy Drama Romance
	<b>4</b> Father o	of the Bride	e Part II (19	95)	Comedy
[150	ratings[-1	10:]			
150		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	
151	tag_counts		['tag'].v	alue_co	unts()
[151	missing control Ron Moore Citizen Kon Moore Citizen Kon Mullet biker gang Paul Adel the wig killer fingenetical topless son Name: countrol Rong Rong Rong Rong Rong Rong Rong Rong	ane g stein sh ly modif cene		ers	1 1 1 1 1 1 1 1 1
152				'bar',	figsize=(10,5))
152	<axes: th="" xl<=""><td>abel='ta</td><td>g'&gt;</td><td></td><td></td></axes:>	abel='ta	g'>		
153	plt.show()	)			



In [154... is\_highly\_rated = ratings['rating'] >= 5.0
 ratings[is\_highly\_rated][30:50]

$\cap$		+	Γ	1		/	
U	u	L		_	J	+	

	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 [155...
```

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

$\cap$		+	Γ	1	5	5	
U	и	L	L	+	J	J	•••

	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 [156...

movies[is\_action].head(15)

Out[156...

n	novield	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

In [157...

ratings\_count = ratings[['movieId','rating']].groupby('rating').count()
ratings\_count

Out[157... 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 [158... average_rating = ratings[['movieId','rating']].groupby('movieId').mean()
    average_rating.head()
```

Out[158...

## rating

#### movield

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

```
In [159... movie_count = ratings[['movieId','rating']].groupby('movieId').count()
movie_count.head()
```

Out[159...

## rating

## movield

- **1** 49695
- **2** 22243
- **3** 12735
- **4** 2756
- **5** 12161

```
In [160... movie_count = ratings[['movieId','rating']].groupby('movieId').count()
movie_count.tail()
```

Out[160... rating

movield	
131254	1
131256	1
131258	1
131260	1
131262	1

In [161...

tags.head()

Out[161...

	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 [162... movies.head()

Out[162...

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

In [163...

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

Out[163	mov	/ield	title		genres	userl	d	tag
	0	1	Toy Story (1995)	Adventure Animation Child	lren Comedy Fantasy	164	4 Wa	tched
	1	1	Toy Story (1995)	Adventure Animation Child	Iren Comedy Fantasy	174	I	nputer nation
	2	1	Toy Story (1995)	Adventure Animation Child	Iren Comedy Fantasy	174	1 anir	Disney mated eature
	3	1	Toy Story (1995)	Adventure Animation Child	Iren Comedy Fantasy	174	1 anim	Pixar nation
	4	1	Toy Story (1995)	Adventure Animation Chilo	Iren   Comedy   Fantasy	174	stari	Leoni es not in this movie
In [164	del avg	g_rat:	= ratingings['uangle].head()	gs.groupby('movieId', a serId']	as_index= <b>False</b> ).me	ean()		
Out[164	mov	/ield	ratin	9				
	0	1	3.92124	0				
	1	2	3.21197	7				
	2	3	3.15104	0				
	3	4	2.86139	3				
	4	5	3.06459	2				
In [165	box_off			s.merge(avg_ratings, or	n='movieId', how='	inner	')	
Out[165		movi	ield	title	ge	enres	rating	
		121	254 Ke	ein Bund für's Leben (2007)	Coi	medy	4.0	
	26739	131						
	26739 26740			uer, Eis & Dosenbier (2002)	Cor	medy	4.0	
		131		uer, Eis & Dosenbier (2002) The Pirates (2014)	Coı		4.0 2.5	
	26740	131 131	256 Fe			nture		
	26740 26741	131 131 131	256 Fei 258	The Pirates (2014) Rentun Ruusu (2001)	Adve	nture isted)	2.5	

$\cap$	1.1	+	Г	1	6	6	
U	и	L	L	+	U	U	•••

		movield	title	genres	rating
26	6737	131250	No More School (2000)	Comedy	4.0
26	6738	131252	Forklift Driver Klaus: The First Day on the Jo	Comedy Horror	4.0
26	6739	131254	Kein Bund für's Leben (2007)	Comedy	4.0
26	6740	131256	Feuer, Eis & Dosenbier (2002)	Comedy	4.0
26	6743	131262	Innocence (2014)	Adventure Fantasy Horror	4.0

In [167...

is\_Adventure = box\_office['genres'].str.contains('Adventure')
box\_office[is\_Adventure][:5]

Out[167...

	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 [168...

box\_office[is\_Adventure & is\_highly\_rated][-5:]

Out[168...

	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

In [169...

movies.head()

Out[169		movield			title					ge	nres
	0	1		Toy Story	(1995) Ad	dventure A	Animatic	n Child	ren Com	nedy Fan	itasy
	1	2		Jumanji	(1995)			Advent	ure Child	dren Fan	itasy
	2	3	Grumpie	er Old Men	(1995)				Come	dy Roma	ance
	3	4	Waitin	g to Exhale	(1995)			Come	edy Dran	na Roma	ance
	4	5	Father	of the Bride	Part II (1995)					Com	nedy
In [170	mo	vie_genres	= movies[	'genres'].	str.spli	t(' ', e	xpand=	True)			
In [171	mo	vie_genres	[:10]								
Out[171		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 [172		vie_genres vie_genres	_	] = movie	es['genre	s'].str.	contai	ns('Co	medy')		

Out[172		0	1	2	3	4	5	6	7	8	9
	<b>0</b> Ac	dventure	Animation	Children	Comedy	Fantasy	None	None	None	None	None
	<b>1</b> Ac	dventure	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
	<b>7</b> Ac	dventure	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										•
In [173	movie	s['year	'] = movie	s['title']	.str.ext	ract('.*	\((.*)	\).*',	expand	l=True)	
	<>:1: 5 C:\Use valid	SyntaxWa rs\HAI\A escape s	rning: inv rning: inv appData\Loc equence '\ '] = movie	valid esca al\Temp\i √('	pe sequer pykernel_	nce '\(' _10184\76					-
In [174	movie	s.tail(	)								
Out[174		movie	ld		title			genr	es yea	ır	
	2727	<b>3</b> 1312	54 Kein Bu	ınd für's Lek	oen (2007)			Come	dy 200	7	
	27274	<b>4</b> 1312	56 Feuer, E	is & Dosenb	oier (2002)	Comedy		dy 200	2		
	2727	<b>5</b> 1312	58	The Pira	ates (2014)		A	Adventu	re 201	4	
	27276	<b>5</b> 1312	60	Rentun Ru	usu (2001)		(no gen	res liste	d) 200	1	
	27277	<b>7</b> 1312	62	Innoce	nce (2014)	Adventu	ıre Fanta	asy Horr	or 201	4	
In [175	tags=	pd.rea	ad_csv(r'C	:\Users\HA	AI\Downlo	ads\arch	ive\ta	g.csv'	)		

Out[175...

	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 [176... tags.dtypes Out[176... userId int64 movieId int64 object tag timestamp object dtype: object In [177... tags.head(5) Out[177... userId movieId tag timestamp 0 18 4141 Mark Waters 2009-04-24 18:19:40 208 1 65 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

```
In [178... tags['parsed_time'] = pd.to_datetime(tags['timestamp'])
In [179... tags['parsed_time'].dtype
Out[179... dtype('<M8[ns]')
In [180... tags.head(2)</pre>
```

```
Out[180...
              userld movield
                                        tag
                                                     timestamp
                                                                       parsed_time
           0
                  18
                         4141 Mark Waters
                                             2009-04-24 18:19:40 2009-04-24 18:19:40
           1
                  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 [181...
           selected_rows = tags[greater_than_t]
           tags.shape, selected_rows.shape
           ((465564, 5), (12130, 5))
Out[181...
           tags.sort values(by='parsed time', ascending=True)[:10]
In [182...
Out[182...
                     userld 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 [183...
           average_rating = ratings[['movieId', 'rating']].groupby('movieId', as_index=False
           average rating.tail()
Out[183...
                   movield rating
           26739
                    131254
                               4.0
           26740
                    131256
                               4.0
           26741
                    131258
                               2.5
           26742
                    131260
                               3.0
           26743
                    131262
                               4.0
In [184...
           joined = movies.merge(average rating, on='movieId', how='inner')
           joined.head()
           joined.select_dtypes(include='number').corr()
```

Out[184		movield	rating
	movield	1.000000	-0.090369
	rating	-0.090369	1.000000

In [ ]: