MostRatedGenres

April 1, 2021

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0.1 IMDB Movie Dataset

The data are contained in six files `links.csv`, `movies.csv`, `ratings.csv` and `tags.csv` etc.In this project we will use only two given data set which is **movies.csv** and **ratings.csv**. Our research question is **What types of movies genres user viewed and rated most than other movies genres ?**

DataSet can be get from this site : http://grouplens.org/datasets/

Importing Pandas

[1]:
```
import pandas as pd
%matplotlib inline
```

Importing or Acquiring movies.csv and ratings.csv data sets

[2]:
```
movies = pd.read_csv('/home/roshan/Documents/datascience/edX/
 ↪PythonForDataScience/ml-25m/movies.csv',sep = ',')
ratings = pd.read_csv('/home/roshan/Documents/datascience/edX/
 ↪PythonForDataScience/ml-25m/ratings.csv',sep = ',')
```

[3]:
```
movies.head() # showing the first 15 items in csv file
```

[3]:

	movieId	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

[4]:
```
ratings.head(5) # showing the first 15 items in csv file
```

[4]:

	userId	movieId	rating	timestamp
0	1	296	5.0	1147880044
1	1	306	3.5	1147868817
2	1	307	5.0	1147868828
3	1	665	5.0	1147878820
4	1	899	3.5	1147868510

[5]:
```
# deleting the timestamp and userId coloumns
del ratings['timestamp']
del ratings['userId']
```

[6]:
```
ratings.head() # after
```

[6]:

	movieId	rating
0	296	5.0
1	306	3.5
2	307	5.0
3	665	5.0
4	899	3.5

Merge Dataframes

[7]:
```
# take the average ratings value and group them by concern MovieId ....
avg_ratings = ratings.groupby('movieId', as_index=False).mean()
avg_ratings.head()
```

[7]:

	movieId	rating
0	1	3.893708
1	2	3.251527
2	3	3.142028
3	4	2.853547
4	5	3.058434

[8]:
```
# we can visualize ratings values by box ploting
avg_ratings.boxplot(column='rating',figsize=(10,5))
```

[8]:
```
<matplotlib.axes._subplots.AxesSubplot at 0x7f3cec33bdd0>
```

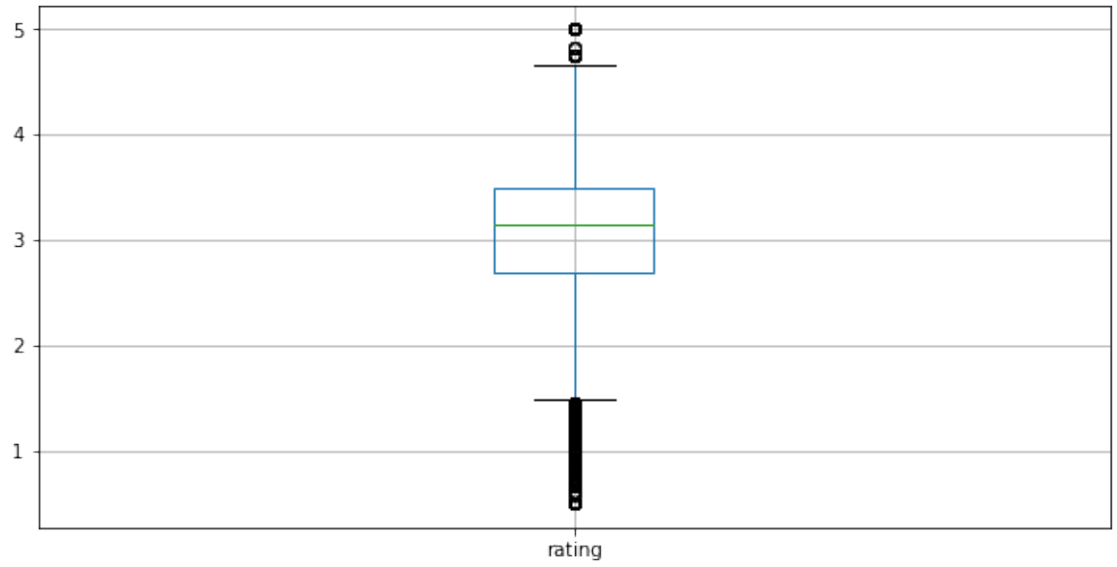


Fig : Visualize Rating in Box Plot

[9]:
```
# extract the launching year of each movies and make a new columns named year
movies['year'] = movies['title'].str.extract('.*\((.*)\).*', expand=True)
movies.tail()
```

[9]:

	movieId	title	genres	year
62418	209157	We (2018)	Drama	2018
62419	209159	Window of the Soul (2001)	Documentary	2001
62420	209163	Bad Poems (2018)	Comedy|Drama	2018
62421	209169	A Girl Thing (2001)	(no genres listed)	2001
62422	209171	Women of Devil's Island (1962)	Action|Adventure|Drama	1962

Merge the previous avg_ratings and movies data set

[10]:
```
#  merging....
join_datasets = movies.merge(avg_ratings, on='movieId', how='inner')
join_datasets.tail()
```

[10]:

	movieId	title	genres	year	\
59042	209157	We (2018)	Drama	2018	
59043	209159	Window of the Soul (2001)	Documentary	2001	
59044	209163	Bad Poems (2018)	Comedy|Drama	2018	
59045	209169	A Girl Thing (2001)	(no genres listed)	2001	
59046	209171	Women of Devil's Island (1962)	Action|Adventure|Drama	1962	

	rating
59042	1.5

```
59043     3.0
59044     4.5
59045     3.0
59046     3.0
```

[11]:
```
join_datasets.columns # coloumn in new data set
```

[11]:
```
Index(['movieId', 'title', 'genres', 'year', 'rating'], dtype='object')
```

[12]:
```
# get rid of the title column
del join_datasets['title']
```

[13]:
```
join_datasets.head()
```

[13]:

	movieId	genres	year	rating
0	1	Adventure|Animation|Children|Comedy|Fantasy	1995	3.893708
1	2	Adventure|Children|Fantasy	1995	3.251527
2	3	Comedy|Romance	1995	3.142028
3	4	Comedy|Drama|Romance	1995	2.853547
4	5	Comedy	1995	3.058434

Data Cleaning: Handling Missing Data

[14]:
```
# Find the shape of the join data set
join_datasets.shape
```

[14]:
```
(59047, 4)
```

[15]:
```
# check is there any null value , if so , true boolean value will be return
join_datasets.isnull().any()
```

[15]:
```
movieId    False
genres     False
year        True
rating     False
dtype: bool
```

Hmm , there're some null value , we have to drop them out.

[16]:
```
# dropna () is used to drop out the null values
join_datasets = join_datasets.dropna()
```

[17]:
```
# again check the shape of the data sets , these time row number decrease
# indcating some rows are erased as they hold null values
join_datasets.shape
```

[17]:
```
(58678, 4)
```

[18]:
```
# again check is their any null values
join_datasets.isnull().any()
```

[18]:
```
movieId    False
genres     False
year       False
rating     False
dtype: bool
```

hmm , all null values are gone.It's Ok now.

Data Visualization

0.1.1 Comparing Genres VS Ratings value , to see the correlation plot in following. We will use general ploting diagram to visualize it ,where genres is alone X axes and ratings is along Y axes.

[19]:
```
join_datasets[-30:].plot(x='genres', y='rating', figsize=(15,5), grid=True ,␣
↪color ='g')
```

[19]:
```
<matplotlib.axes._subplots.AxesSubplot at 0x7f3cebb37a10>
```

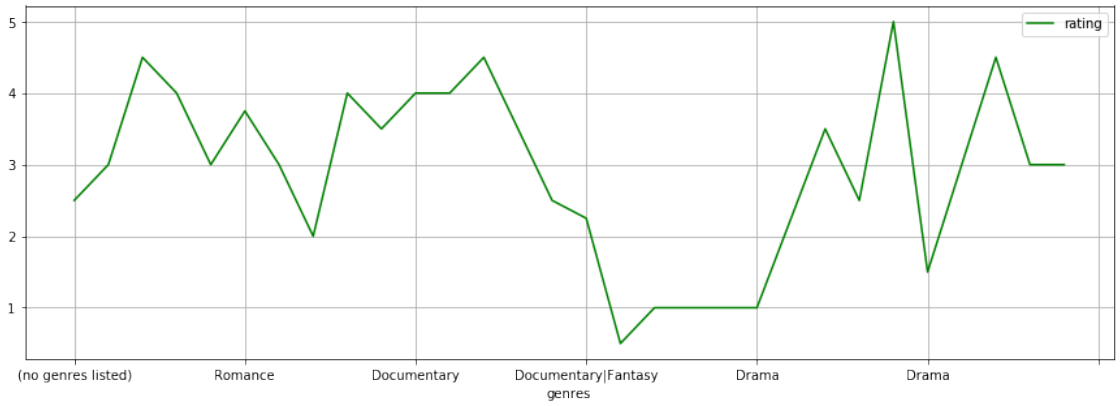


Fig : Ploting ratings VS genres . Drama genres tend to high than other movie genres

Comments On Plot

Here we can see , ploting **genres** and **ratings** values shows us that Drama type movies tends to rate more high than other movies genres.Other genres has average ratinsgs scale though comedy genres is following Drama genres.

0.1.2 For making to visualize more convineint , let's use pie plot.

[20]:
```
# using value_counts() on our join_datasets , we can also see Drama movies are␣
↪majority in numbers>
gen_count = join_datasets['genres'].value_counts()
gen_count[:10]
```

[20]:
```
Drama                     8621
Comedy                    5283
Documentary               4571
(no genres listed)        4325
Comedy|Drama              2308
Drama|Romance             2004
Horror                    1549
Comedy|Romance            1460
Comedy|Drama|Romance      1014
Drama|Thriller             893
Name: genres, dtype: int64
```

[21]:
```
# plot the most frequent genres
gen_count[:10].plot(
                    kind = 'pie', figsize=(10,8) , shadow = True,
                    explode =(0.1,0,0,0,0,0,0,0,0,0),
                    autopct = '%1.1f%%' , startangle = 45
                   )
```

[21]:
```
<matplotlib.axes._subplots.AxesSubplot at 0x7f3ceba0c2d0>
```

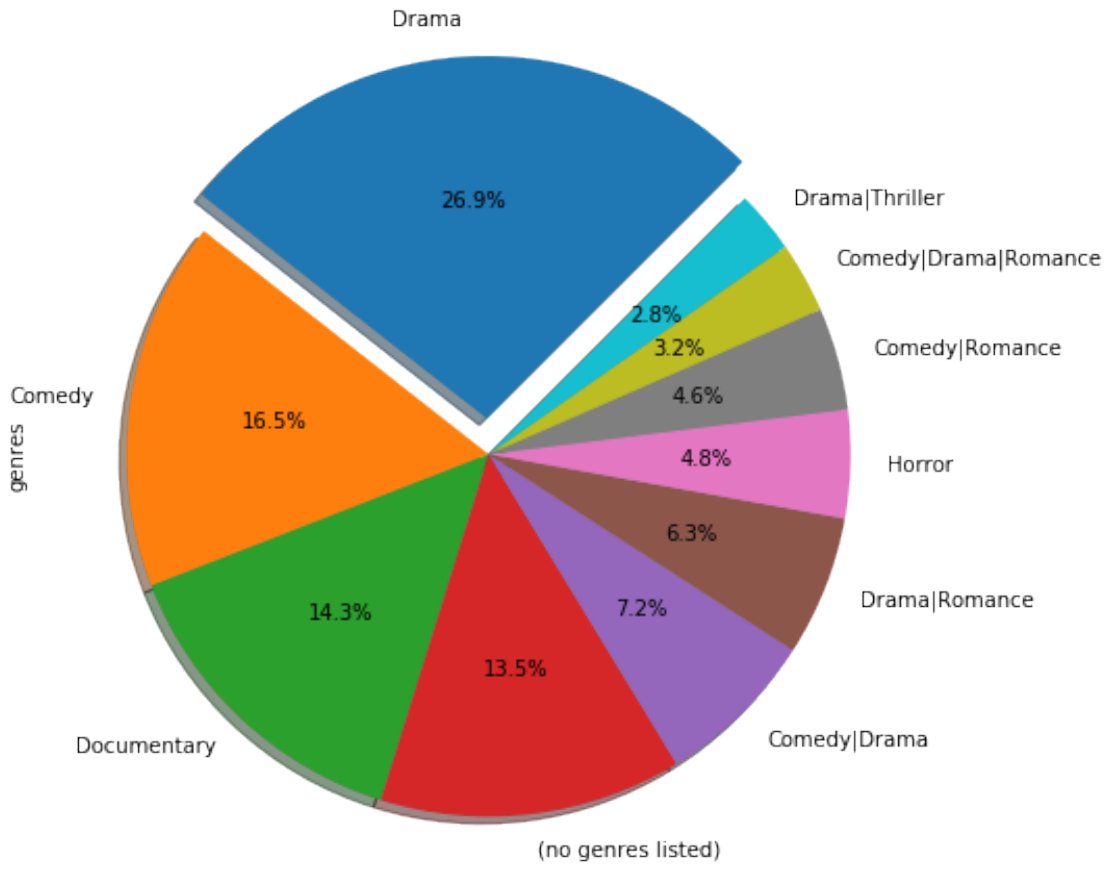
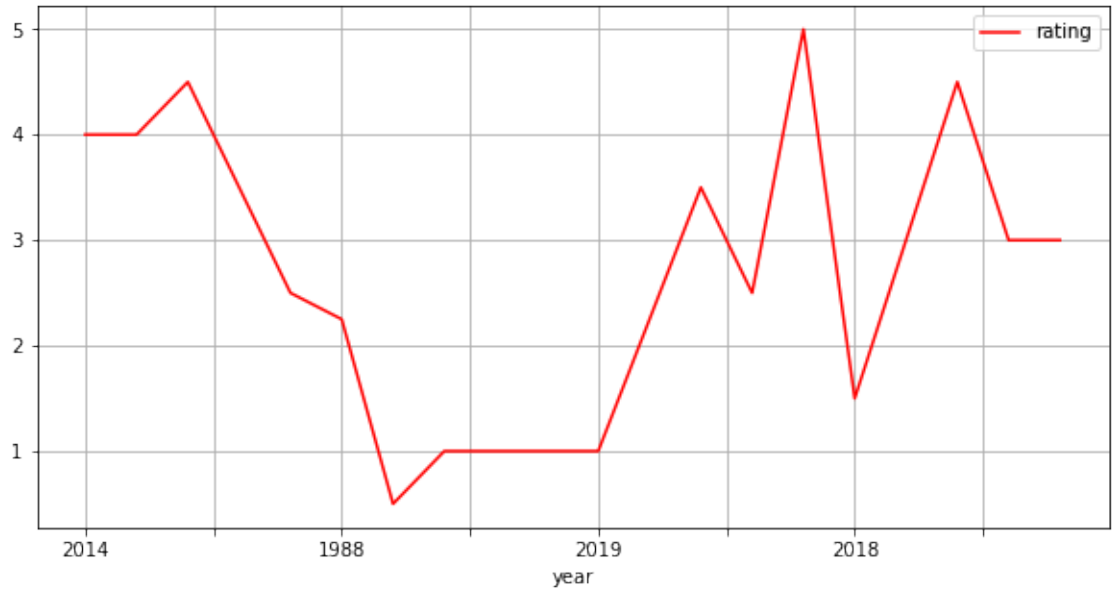


Fig : Ploting the most frequent genres , here which is Drama

0.1.3 We can also find is movie ratings are related of its concern launch year

[22]:
```
# takings our whole data set
join_datasets[-20:].plot(x='year', y='rating', figsize=(10,5), grid=True ,␣
↪color = 'r')
```

[22]: <matplotlib.axes._subplots.AxesSubplot at 0x7f3ceb487e90>



[23]:
```
# taking average of the year
average_year = join_datasets[['year','rating']].groupby('year', as_index=False).
↪mean()
```

[24]:
```
average_year[-20:].plot(x='year', y='rating', figsize=(10,5), grid=True ,␣
↪color='DarkBlue')
```

[24]: <matplotlib.axes._subplots.AxesSubplot at 0x7f3ceb05c850>

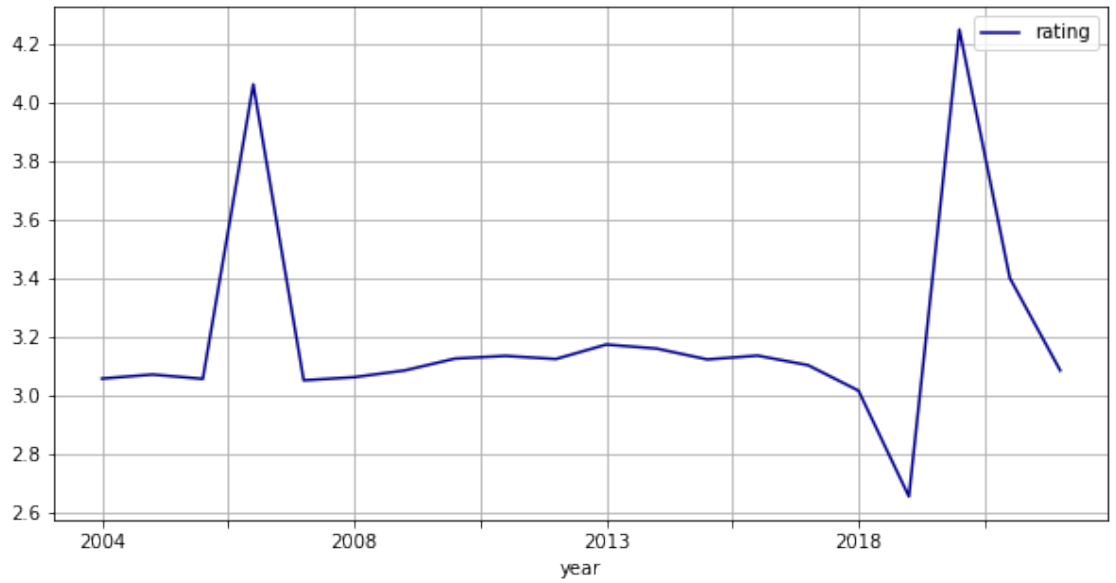


Fig : Average Movie Ratings over Time