IMDB movie data analysis using pandas

```
In [ ]: numpy - image to array (1d, 2d, 3d)
        pytorch -- python + torch
        pytorch - image to tensor
                1d tensor , 2d tensor , 3d tensor , nd tensor
        kaggle.com
        world - (kaggle.com )
        kaggle.com -- data anlysis, data scientist, business analyst
        kaggle.com -- multiple dataset
                banking, healthcare, insure, supplyIF
In [1]: #Import Libraries
        import pandas as pd
In [3]: # read the dataset
        movies = pd.read_csv(r'E:\Data Science & AI\Dataset files\archive\movie.csv')
In [5]: ratings = pd.read_csv(r'E:\Data Science & AI\Dataset files\archive\rating.csv')
In [6]: tags = pd.read_csv(r'E:\Data Science & AI\Dataset files\archive\tag.csv')
In [7]: print(type(movies))
        movies.head(20)
```

<class 'pandas.core.frame.DataFrame'>

Out[7]:		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
	5	6	Heat (1995)	Action Crime Thriller
	6	7	Sabrina (1995)	Comedy Romance
	7	8	Tom and Huck (1995)	Adventure Children
	8	9	Sudden Death (1995)	Action
	9	10	GoldenEye (1995)	Action Adventure Thriller
	10	11	American President, The (1995)	Comedy Drama Romance
	11	12	Dracula: Dead and Loving It (1995)	Comedy Horror
	12	13	Balto (1995)	Adventure Animation Children
	13	14	Nixon (1995)	Drama
	14	15	Cutthroat Island (1995)	Action Adventure Romance
	15	16	Casino (1995)	Crime Drama
	16	17	Sense and Sensibility (1995)	Drama Romance
	17	18	Four Rooms (1995)	Comedy
	18	19	Ace Ventura: When Nature Calls (1995)	Comedy
	19	20	Money Train (1995)	Action Comedy Crime Drama Thriller

In [8]: print(type(ratings))
 ratings.head(10)

<class 'pandas.core.frame.DataFrame'>

Out[8]:		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
	5	1	112	3.5	2004-09-10 03:09:00
	6	1	151	4.0	2004-09-10 03:08:54
	7	1	223	4.0	2005-04-02 23:46:13
	8	1	253	4.0	2005-04-02 23:35:40
	9	1	260	4.0	2005-04-02 23:33:46

In [9]: print(type(tags))
 tags.head(10)

<class 'pandas.core.frame.DataFrame'>

Out[9]:		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
	5	65	668	bollywood	2013-05-10 01:37:56
	6	65	898	screwball comedy	2013-05-10 01:42:40
	7	65	1248	noir thriller	2013-05-10 01:39:43
	8	65	1391	mars	2013-05-10 01:40:55
	9	65	1617	neo-noir	2013-05-10 01:43:37

In [10]: tags.head()

Out[10]:		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 [11]: ratings.head()
Out[11]:
            userId movieId rating
                                           timestamp
          0
                                3.5 2005-04-02 23:53:47
                         29
                               3.5 2005-04-02 23:31:16
          2
                 1
                         32
                               3.5 2005-04-02 23:33:39
          3
                               3.5 2005-04-02 23:32:07
                 1
                         50
                               3.5 2005-04-02 23:29:40
          4
In [12]: #For current analysis, we will remove timestamp
         del ratings['timestamp']
         del tags['timestamp']
         Data Structures:
         Series
In [14]: row_0 = tags.iloc[0]
         type(row_0)
Out[14]: pandas.core.series.Series
In [15]: print(row_0)
        userId
                            18
        movieId
                          4141
                 Mark Waters
        tag
        Name: 0, dtype: object
In [16]: row_0.index
Out[16]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [17]: row_0['userId']
Out[17]: 18
In [18]: 'rating' in row_0
Out[18]: False
In [19]: row_0.name
Out[19]: 0
In [20]: row_0 = row_0.rename('firstRow')
         row_0.name
Out[20]: 'firstRow'
```

```
In [21]: #DataFrames
         tags.head()
Out[21]:
            userld movield
                                    tag
          0
                18
                       4141 Mark Waters
          1
                65
                        208
                               dark hero
          2
                65
                        353
                               dark hero
          3
                              noir thriller
                65
                        521
          4
                65
                        592
                               dark hero
In [22]: tags.index
Out[22]: RangeIndex(start=0, stop=465564, step=1)
In [23]: tags.columns
Out[23]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [24]: tags.iloc[ [0,11,500] ]
Out[24]:
              userld movield
                                         tag
                  18
                         4141
                                  Mark Waters
           11
                  65
                         1783
                                   noir thriller
                        55908 entirely dialogue
          500
                 342
In [25]: ## Descriptive Statistics
         ratings['rating'].describe()
Out[25]: count
                   2.000026e+07
                 3.525529e+00
          mean
          std
                 1.051989e+00
          min
                 5.000000e-01
                 3.000000e+00
          25%
          50%
                 3.500000e+00
                   4.000000e+00
          75%
                   5.000000e+00
          max
          Name: rating, dtype: float64
In [26]: ratings.describe()
```

```
Out[26]:
                     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 [27]:
         ratings['rating'].mean()
Out[27]: 3.5255285642993797
In [28]: ratings.mean()
                  69045.872583
Out[28]: userId
         movieId
                    9041.567330
                   3.525529
         rating
         dtype: float64
In [29]: ratings['rating'].min()
Out[29]: 0.5
In [30]: ratings['rating'].max()
Out[30]: 5.0
In [31]: ratings['rating'].std()
Out[31]: 1.051988919275684
In [32]: ratings['rating'].mode()
Out[32]: 0
             4.0
         Name: rating, dtype: float64
In [33]: ratings.corr()
Out[33]:
                    userId
                            movield
                                      rating
                 1.000000
           userId
                          -0.000850 0.001175
         movield -0.000850
                            1.000000 0.002606
           rating
                  In [34]: filter1 = ratings['rating'] > 10
         print(filter1)
```

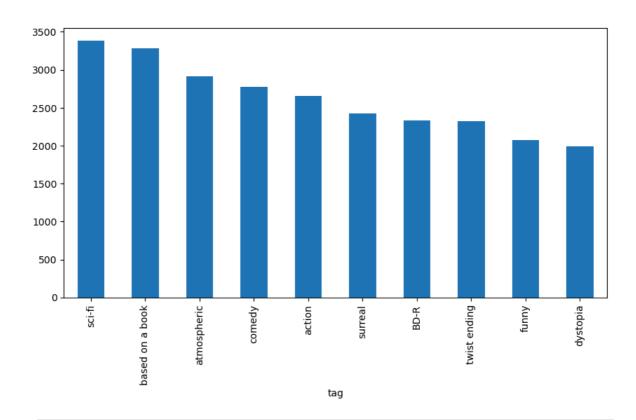
```
filter1.any()
        0
                   False
        1
                   False
        2
                   False
        3
                   False
                  False
                    . . .
        20000258 False
        20000259 False
        20000260 False
        20000261 False
        20000262
                  False
        Name: rating, Length: 20000263, dtype: bool
Out[34]: False
In [35]: filter2 = ratings['rating'] > 0
         filter2.all()
Out[35]: True
In [36]: # \ Data Cleaning: Handling Missing Data
         movies.shape
Out[36]: (27278, 3)
In [37]: movies.isnull().any().any()
Out[37]: False
In [38]: ratings.shape
Out[38]: (20000263, 3)
In [39]: ratings.isnull().any().any()
Out[39]: False
In [40]: tags.shape
Out[40]: (465564, 3)
In [41]: tags.isnull().any().any()
Out[41]: True
In [43]: tags=tags.dropna()
In [44]: tags.isnull().any().any()
Out[44]: False
In [53]: tags.shape
Out[53]: (465548, 3)
```

```
In [64]: # ** Data Visualization
         %matplotlib inline
         ratings.hist(column='rating', figsize=(10,5))
Out[64]: array([[<Axes: title={'center': 'rating'}>]], dtype=object)
                                               rating
        5
        4
        3
        2
        1
                                     2
In [66]: ratings.boxplot(column='rating', figsize=(10,5))
Out[66]: <Axes: >
        5
        3
        2
        1
                                                rating
In [67]: # ₫ Slicing Out Columns
         tags['tag'].head()
Out[67]: 0
                 Mark Waters
          1
                   dark hero
                   dark hero
          2
              noir thriller
                   dark hero
          Name: tag, dtype: object
In [74]: movies[['title','genres']].head()
```

```
Out[74]:
                                     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 [76]:
          ratings[-10:]
Out[76]:
                      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
                                          5.0
                               70286
          20000262 138493
                                71619
                                          2.5
In [89]:
          tag_counts = tags['tag'].value_counts()
          tag_counts[-10:]
Out[89]:
          tag
          missing child
                                              1
                                              1
          Ron Moore
          Citizen Kane
                                              1
          mullet
                                              1
          biker gang
                                              1
          Paul Adelstein
                                              1
                                             1
          the wig
          killer fish
          genetically modified monsters
                                             1
          topless scene
                                              1
          Name: count, dtype: int64
```

In [90]: tag_counts[:10].plot(kind='bar', figsize=(10,5))

Out[90]: <Axes: xlabel='tag'>



In []: