```
In [21]: row_0.index
Out[21]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [22]: row_0['userId']
Out[22]: 18
In [23]: 'rating' in row_0 # IN row_0 THERE IS NO RATING AS THERE IS NO RATING COLUMN IN TAGS
Out[23]: False
In [24]: row_0.name # NAME OF ZEROTH ROW
Out[24]: 0
In [25]: row_0 = row_0.rename('firstRow')
          row_0.name
Out[25]: 'firstRow'
In [26]: # DATA FRAMES
In [27]: tags.head()
Out[27]:
            userld movield
                                tag
                     4141 Mark Waters
                            dark hero
                            dark hero
                      521
                           noir thriller
                      592
                            dark hero
In [28]: tags.index
Out[28]: RangeIndex(start=0, stop=465564, step=1)
In [29]: tags.columns
Out[29]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [32]: tags.iloc[[0,11,500]]
Out[32]:
               userld movield
                                    tag
                18
                      4141
                              Mark Waters
                 65
                      1783
                               noir thriller
                342
                    55908 entirely dialogue
In [33]: # DESCRIPTIVE STATISTICS
In [34]: ratings['rating']
Out[34]: 0
                      3.5
                      3.5
                      3.5
                      3.5
                      3.5
                      3.5
                      4.0
                      4.0
                      4.0
                      4.0
         10
                      4.0
         11
                      4.0
          12
                      4.0
          13
                      3.5
         14
                      3.5
         15
                      4.0
         16
                      3.5
          17
                      3.5
         18
                      3.0
         19
                      3.5
          20
                      3.5
          21
                      3.5
          22
                      4.0
          23
                      4.0
          24
                      3.5
          25
                      3.5
          26
                      4.0
          27
                      4.0
          28
                      3.5
                      3.5
          20000233
                     3.5
          20000234
          20000235
                      4.5
          20000236
                     4.5
          20000237
                      4.0
          20000238
                     4.0
          20000239
                     4.0
          20000240
                     3.0
          20000241
                     4.0
          20000242
          20000243
                     4.5
          20000244
                     5.0
          20000245
                      5.0
          20000246
                      3.0
          20000247
                      3.0
          20000248
                     4.5
          20000249
                     4.0
          20000250
          20000251
                     5.0
          20000252
                     4.0
          20000253
                     4.5
          20000254
                      4.0
          20000255
                     4.5
          20000256
                     4.5
          20000257
                     4.5
          20000258
          20000259
                     4.5
          20000260
                     3.0
          20000261
                     5.0
                     2.5
          20000262
          Name: rating, Length: 20000263, dtype: float64
In [35]: ratings['rating'].describe() # DESCRIPTIVE DATA OF RATING COLUMN
Out[35]: count
                  2.000026e+07
                  3.525529e+00
          mean
                  1.051989e+00
          std
          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 [36]: ratings.describe() # DESCRIPTIVE DATA OF RATINGS DATASET
Out[36]:
                                           rating
                     userld
                               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 [37]: ratings['rating'].mean() # MEAN OF RATING COLUMN
Out[37]: 3.5255285642993797
In [38]: ratings.mean() # MEAN OF RATINGS DATASET
Out[38]: userId
                     69045.872583
                     9041.567330
          movieId
          rating
                         3.525529
         dtype: float64
In [39]: ratings['rating'].min() # MINIMUM VALUE OF RATING COLUMN
Out[39]: 0.5
In [40]: ratings['rating'].max() # MAXIMUM VALUE OF RATING COLUMN
Out[40]: 5.0
In [41]: ratings['rating'].std() # STANDARD DEVIATION : HOW DISPERSED THE DATA IS IN RELATION TO MEAN
Out[41]: 1.051988919275684
In [42]: ratings['rating'].mode() # MODE : REPEATED NO.
Out[42]: 0 4.0
         dtype: float64
In [43]: ratings.corr() # CALCULATES THE RELATIONSHIP B/W EACH COLUMN IN DATASET
          # CORRELATION : EXPRESS THE EXTENT TO WHICH 2 VARIABLES ARE LINEARLY RELATED (-1 TO +1)
Out[43]:
                    userld
                          movield
                                    rating
           userId 1.000000 -0.000850 0.001175
           movield -0.000850 1.000000 0.002606
            rating 0.001175 0.002606 1.000000
In [44]: filter1 = ratings['rating'] > 10
          print(filter1)
         0
                      False
                      False
         1
                      False
                      False
                      False
                      False
                      False
                      False
                      False
                      False
         9
         10
                      False
          11
                      False
         12
                      False
         13
                      False
          14
                      False
          15
                      False
         16
                      False
         17
                      False
          18
                      False
         19
                      False
          20
                      False
          21
                      False
          22
                      False
          23
                      False
         24
                      False
          25
                      False
          26
                      False
          27
                      False
          28
                      False
          29
                      False
                      . . .
          20000233
                      False
          20000234
                      False
          20000235
                      False
          20000236
                      False
          20000237
                      False
          20000238
                      False
          20000239
                      False
          20000240
                      False
          20000241
                      False
          20000242
                      False
          20000243
                      False
          20000244
                      False
          20000245
                      False
          20000246
                      False
          20000247
                      False
          20000248
                     False
          20000249
                      False
          20000250
                     False
          20000251
                     False
          20000252
                     False
          20000253
                      False
          20000254
                      False
          20000255
                      False
          20000256
                      False
          20000257
                      False
          20000258
                     False
          20000259
                      False
          20000260
                      False
          20000261
                      False
         20000262
                     False
          Name: rating, Length: 20000263, dtype: bool
In [45]: filter1.any() # PRINTS ONE VALUE FOR EACH COLUMN, True IF ANY VALUE IN THAT COLUMN IS TRUE OTHERWISE False
Out[45]: False
In [46]: filter2 = ratings['rating'] > 0 # PRINTS ONE VALUE FOR EACH COLUMN, True IF ALL VALUE IN THAT COLUMN ARE TRUE, OTHERWI
          SE False
          filter2.all()
Out[46]: True
In [47]: # DATA CLEANING : HANDLING MISSING DATA
In [48]: movies.shape # TOTAL NO. OF ROWS AND COLUMNS
Out[48]: (27278, 3)
In [49]: movies.isnull().any().any() # NO NULL VALUES # True-NULL VALUE, False-NO NULL VALUE
Out[49]: False
In [50]: ratings.shape
Out[50]: (20000263, 3)
In [51]: ratings.isnull().any().any()
Out[51]: False
In [52]: tags.shape
Out[52]: (465564, 3)
In [53]: tags.isnull().any().any() # HAVE NULL VALUE
Out[53]: True
In [54]: tags = tags.dropna() # REMOVE THE NULL VALUES
In [55]: tags.isnull().any().any()
Out[55]: False
In [56]: tags.shape
Out[56]: (465548, 3)
In [57]: # DATA VISUALIZATION
In [58]: %matplotlib inline
          ratings.hist(column='rating',figsize=(10,5)) # PLOTTING HISTOGRAM
Out[58]: array([[<matplotlib.axes._subplots.AxesSubplot object at 0x0000000043A87940>]],
                dtype=object)
                                                 rating
           5000000
           4000000
           3000000
           2000000
           1000000
In [59]: ratings.boxplot(column='rating', figsize=(10,5)) # PLOTTING BOXPLOT
Out[59]: <matplotlib.axes._subplots.AxesSubplot at 0x59fc3c88>
In [60]: # SLICING OUT COLUMNS
In [61]: tags['tag'].head() # TAG COLUMN FIRST 5 VALUES
Out[61]: 0
                Mark Waters
                  dark hero
         1
                   dark hero
              noir thriller
                  dark hero
         Name: tag, dtype: object
In [62]: movies[['title', 'genres']].head() # TITLE & GENRES COLUMN FIRST 5 VALUES
Out[62]:
          0
                       Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy
                                                Adventure|Children|Fantasy
          1
                        Jumanji (1995)
          2
                 Grumpier Old Men (1995)
                                                      Comedy|Romance
                  Waiting to Exhale (1995)
          3
                                                 Comedy|Drama|Romance
          4 Father of the Bride Part II (1995)
                                                             Comedy
In [63]: ratings[-10:]
Out[63]:
                   userld movield rating
                          60816
          20000253 138493
                                  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 [64]: tag_counts = tags['tag'].value_counts() # SERIES CONTAINING COUNTS OF UNIQUE VALUES
Out[64]: sci-fi
                                                              3384
                                                              3281
          based on a book
                                                              2917
          atmospheric
                                                              2779
          comedy
          action
                                                              2657
          surreal
                                                              2427
         BD-R
                                                              2334
          twist ending
                                                              2323
                                                              2072
          funny
          dystopia
                                                              1991
          stylized
                                                              1941
          quirky
                                                              1906
          dark comedy
                                                              1899
                                                              1769
         classic
                                                              1754
          psychology
          fantasy
                                                              1703
                                                              1549
          time travel
                                                              1534
          romance
          visually appealing
                                                              1509
          disturbing
                                                              1487
                                                              1428
          aliens
                                                              1422
          thought-provoking
                                                              1417
          social commentary
          Nudity (Topless)
                                                              1400
          violence
                                                              1336
                                                              1312
         drugs
                                                              1286
         Criterion
          true story
                                                              1276
         nudity (topless)
                                                              1245
         adventure
                                                              1243
                                                              . . .
          Sexual
                                                                 1
          silhouettes
                                                                 1
         I love the powerglove
                                                                 1
          Facing Mortality
                                                                 1
          color symbolism
                                                                 1
          in the true meaning of the word:awesome
                                                                 1
          too much sex
                                                                 1
          AFI #79
                                                                 1
1
          Farley Granger
          Morgan Neville
                                                                 1
          Henry de Sliva
          Imperialism
          mental illnesses vs. sanity
          Richard Rush
         consensual sex
          small budget
         Tim Burton Hits Another One Out Of The Cemetery
                                                                 1
          exploding helicopter
                                                                 1
          saliva song rocks
                                                                 1
          karrin's favorite
          Fletcher Markle
         Not Another Mingella Stretcher! We Hates Them!
          jonathan rhys meyers
         perfect entertainment
         brainless fun
         Restitution
          sailor and lula
         Detroit MI
         3-hours of sex and drugs
          rescue mission
         Name: tag, Length: 38643, dtype: int64
In [65]: tag_counts[-10:]
Out[65]: Fletcher Markle
                                                            1
          Not Another Mingella Stretcher! We Hates Them!
                                                            1
          jonathan rhys meyers
                                                             1
          perfect entertainment
                                                             1
          brainless fun
          Restitution
          sailor and lula
         Detroit MI
                                                             1
                                                            1
          3-hours of sex and drugs
                                                            1
          rescue mission
         Name: tag, dtype: int64
In [66]: tag_counts[:10].plot(kind='bar', figsize=(10,5))
Out[66]: <matplotlib.axes._subplots.AxesSubplot at 0x5bc7cf60>
           3500
           3000
           2500
           2000
           1500
           1000
           500
                                                          BD-R
```

In [1]: # DATASET :- MOVIELENS 20M

In [2]: # IMPORT LIBRARIES

In [4]: import warnings

In [9]: print(type(movies))

movield

1

2

3

4

5

6

8

9

10

11

12

13

14

15

16

17

18

20

IMPORT TAG DATASET

userld movield

In [12]: tags.head() # TOP 5 TAGS PRINTED

IMPORT RATING DATASET

userld movield rating

del tags['timestamp']

In [18]: # DATA STRUCTURES

userId movieId

tag

type(row_0)

Out[19]: pandas.core.series.Series

In [20]: print(row_0) # ZEROTH ROW ELEMENTS

Name: 0, dtype: object

4141

Mark Waters

0

1

3

5

9

10

11

12

13

14

15

16

17

18

19

In [16]: ratings.head()

Out[12]:

Out[16]:

Out[10]:

PREFERENCES AND MOVIE RATINGS

In [3]: import pandas as pd # IMPORTING PANDAS LIBRARY

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

In [10]: movies.head(20) # TOP 20 MOVIES DATA PRINTED

warnings.filterwarnings('ignore')

IMPORT MOVIE DATASET

PURPOSE :- ANALYZE THE RATINGS AND TAGS OF MOVIES

In [8]: | movies = pd.read_csv(r'C:\Users\Lenovo\Downloads\archive\movie.csv', sep=',')

Jumanji (1995)

Heat (1995)

Sabrina (1995)

Tom and Huck (1995)

Sudden Death (1995)

GoldenEye (1995)

Balto (1995)

Nixon (1995)

Casino (1995)

Cutthroat Island (1995)

Four Rooms (1995)

Money Train (1995)

In [11]: tags = pd.read_csv(r'C:\Users\Lenovo\Downloads\archive\tag.csv', sep=',')

timestamp

Sense and Sensibility (1995)

19 Ace Ventura: When Nature Calls (1995)

tag

4141 Mark Waters 2009-04-24 18:19:40

521 noir thriller 2013-05-10 01:39:43

2 3.5 2005-04-02 23:53:47

32 3.5 2005-04-02 23:33:39 47 3.5 2005-04-02 23:32:07 50 3.5 2005-04-02 23:29:40

3.5 2005-04-02 23:31:16

In [17]: del ratings['timestamp'] # REMOVE TIMESTAMP COLUMN FROM RATINGS & TAGS

In [19]: row_0 = tags.iloc[0] # iloc=> HELPS TO SELECT A SPECIFIC ROW OR COLUMN FROM DATASET

dark hero 2013-05-10 01:41:18 dark hero 2013-05-10 01:41:19

dark hero 2013-05-10 01:41:18

Grumpier Old Men (1995)

Waiting to Exhale (1995)

Father of the Bride Part II (1995)

American President, The (1995)

Dracula: Dead and Loving It (1995)

sep => SPECIFYA CUSTOM DELIMETER (CHARACTER LIKE COMMA etc THAT SEPARATES DATA)

Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy

MOVIELENS :- A WEB-BASED RECOMMENDATION SYSTEM & VIRTUAL COMMUNITY THAT MAKES MOVIE RECOMMENDATIONS BASED ON USER

genres

Comedy

Action

Drama

Adventure|Children|Fantasy

Comedy|Drama|Romance

Comedy|Romance

Action|Crime|Thriller

Comedy|Romance

Adventure|Children

Comedy|Horror

Crime|Drama

Comedy

Comedy

Drama|Romance

Action|Adventure|Thriller

Comedy|Drama|Romance

Adventure|Animation|Children

Action|Adventure|Romance

Action|Comedy|Crime|Drama|Thriller

In [13]: ratings = pd.read_csv(r'C:\Users\Lenovo\Downloads\archive\rating.csv', sep=',', parse_dates=['timestamp'])

parse_dates='timestamp' => IF TRUE, PANDAS WILL ATTEMPT TO GUESS THE FORMAT OF DATETIME STRINGS

MOVIELENS 20M :- 20 MILLION RATINGS AND 465564 TAG APPLICATIONS APPLIED TO 27278 MOVIESBY 138000 USERS