KAGGLE PROJECT (MOVIES Rating Analysis)

```
import pandas as pd
In [1]:
                                    # here importing libraries
In [2]:
          # READ THE DATASET
          # in this notebeek , we will be using 3 CSV files
          # RATINGS.CSV : userId, movieId, rating, timestamp
          # TAGS.CSV : userId, movieId, tag, timestamp
          # MOVIES.CSV : movieId, title, genres
         movies = pd.read_csv(r'C:\Users\Hanshu\Desktop\EXCEL 1 kaggale\movie.csv')
In [4]:
         movies
Out[4]:
                 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
                            Waiting to Exhale (1995)
                                                                      Comedy|Drama|Romance
                           Father of the Bride Part II
              4
                                                                                     Comedy
                                           (1995)
                              Kein Bund für's Leben
                  131254
         27273
                                                                                     Comedy
                                           (2007)
                             Feuer, Eis & Dosenbier
         27274
                  131256
                                                                                     Comedy
                                           (2002)
         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 [5]: print(type(movies))
        <class 'pandas.core.frame.DataFrame'>
In [7]: movies.head(20) # (0-19 records displyed)
```

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]: tags = pd.read_csv(r'C:\Users\Hanshu\Desktop\EXCEL 1 kaggale\tag.csv')
 tags

Out[8]:		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 [9]:	tags.head()	#	bу	default	5	records	(0-4)
---------	-------------	---	----	---------	---	---------	-------

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

In [11]: ratings = pd.read_csv(r'C:\Users\Hanshu\Desktop\EXCEL 1 kaggale\rating.csv')
ratings

	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 [12]: del ratings['timestamp']
    del tags['timestamp']
```

In [16]: ratings # here timestamp remove bcz above step we are using DEL

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Out[11]:

	userId	movield	rating
0	1	2	3.5
1	1	29	3.5
2	1	32	3.5
3	1	47	3.5
4	1	50	3.5
•••			
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

20000263 rows × 3 columns

```
In [15]: tags.head()
```

Out[15]:		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

DATA STRUCTURES

Series

```
In [21]:
                                  # in TAGS , first record information
         row_0 = tags.iloc[0]
         row_0
                                   # iloc[0] -- first record infor (oth index)
                                  # get output in the form of pandas SERIES
Out[21]:
                             18
         userId
                           4141
         movieId
                    Mark Waters
         Name: 0, dtype: object
In [18]: type(row_0)
Out[18]: pandas.core.series.Series
In [19]: print(row_0)
        userId
                            18
        movieId
                          4141
                   Mark Waters
        tag
        Name: 0, dtype: object
In [23]: row_0.index # we get columns names of pands SERIES
Out[23]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [24]: row_0['userId']
Out[24]: 18
         'rating' in row_0
In [26]:
                              # here used membership operators
Out[26]: False
In [27]:
        row_0.name
Out[27]: 0
In [29]:
         row 0 = row 0.rename('firstROW')
                                             # here rename the na,e as 'firstROW'
         row_0
```

Out[29]: userId 18

movieId 4141 tag Mark Waters

Name: firstROW, dtype: object

DataFrames

```
tags.head()
In [30]:
Out[30]:
             userld movield
                                       tag
          0
                 18
                        4141
                              Mark Waters
                         208
                                 dark hero
          1
                 65
          2
                 65
                         353
                                 dark hero
          3
                                noir thriller
                 65
                         521
          4
                 65
                         592
                                 dark hero
In [32]: tags.index
Out[32]: RangeIndex(start=0, stop=465564, step=1)
In [33]: tags.columns
Out[33]: Index(['userId', 'movieId', 'tag'], dtype='object')
          tags.iloc[ [0,11,500] ]
In [34]:
Out[34]:
                userld movield
                                             tag
            0
                   18
                           4141
                                    Mark Waters
                           1783
                                      noir thriller
          500
                  342
                         55908 entirely dialogue
```

Descriptive Statistics

```
In [35]: ratings['rating'].describe()
Out[35]: count
                   2.000026e+07
                   3.525529e+00
          mean
          std
                   1.051989e+00
                   5.000000e-01
          min
          25%
                   3.000000e+00
          50%
                   3.500000e+00
          75%
                   4.000000e+00
                   5.000000e+00
          Name: rating, dtype: float64
In [36]: ratings.describe()
```

```
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 [39]: ratings['rating'].mean()
Out[39]: 3.5255285642993797
In [40]: ratings.mean()
                  69045.872583
Out[40]: userId
         movieId
                   9041.567330
                 3.525529
         rating
         dtype: float64
In [41]: ratings['rating'].min()
Out[41]: 0.5
In [42]: ratings['rating'].max()
Out[42]: 5.0
In [43]: ratings['rating'].std()
Out[43]: 1.051988919275684
In [44]: ratings['rating'].mode()
Out[44]: 0
             4.0
         Name: rating, dtype: float64
In [45]: ratings.corr()
Out[45]:
                    userId
                            movield
                                      rating
           userId
                 1.000000 -0.000850 0.001175
         movield -0.000850
                          1.000000 0.002606
           rating
                  In [47]: filter1 = ratings['rating'] > 10
         filter1
```

Out[36]:

userld

movield

```
Out[47]: 0
                    False
                    False
                    False
         3
                    False
                  False
         20000258 False
         20000259 False
         20000260 False
         20000261 False
         20000262 False
         Name: rating, Length: 20000263, dtype: bool
In [49]: print(filter1)
        filter1.any()
                  False
       1
                  False
       2
                  False
                 False
                 False
                 . . .
       20000258 False
       20000259 False
       20000260 False
       20000261 False
       20000262 False
       Name: rating, Length: 20000263, dtype: bool
Out[49]: False
In [50]: filter2 = ratings['rating'] > 0
        filter2.all()
Out[50]: True
```

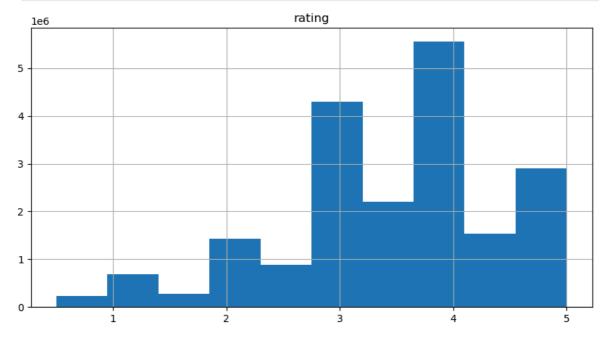
Data Cleaning: Handling Missing Data

```
In [51]: movies.shape
Out[51]: (27278, 3)
In [53]: movies.isnull().any().any() # NO NULL Values
Out[53]: False
In [54]: ratings.shape
Out[54]: (20000263, 3)
In [55]: tags.shape
Out[55]: (465564, 3)
In [57]: tags.isnull().any().any() # here NULL Values will have
```

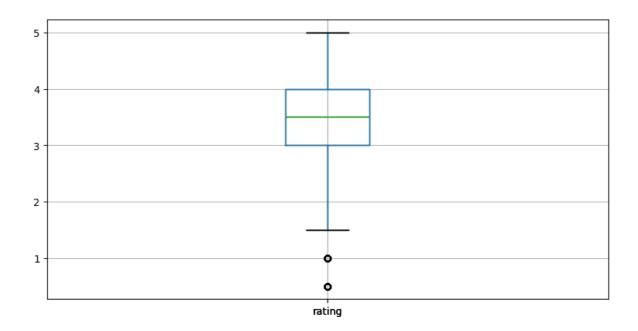
```
Out[57]: True
In [58]: tags = tags.dropna()
In [59]: tags.isnull().any().any() # now no NULL values BCZ in above step we are droppi
Out[59]: False
In [60]: tags.shape # NO NULL VALUES , check before step(tags.shape) & now this step numb
Out[60]: (465548, 3)
```

Data Visualization

```
In [69]: import matplotlib.pyplot as plt
         %matplotlib inline
         ratings.hist(column = 'rating' , figsize = (10,5))
         plt.show()
```



```
In [72]: ratings.boxplot(column = 'rating' , figsize=(10,5))
         plt.show()
```



Slicing Out Columns

```
In [77]: tags['tag'].head()
Out[77]:
                  Mark Waters
                    dark hero
                    dark hero
               noir thriller
                    dark hero
          Name: tag, dtype: object
In [81]: movies[['title', 'genres']] . head()
Out[81]:
                                      title
                                                                               genres
           0
                           Toy Story (1995)
                                            Adventure|Animation|Children|Comedy|Fantasy
                             Jumanji (1995)
                                                             Adventure|Children|Fantasy
                   Grumpier Old Men (1995)
           2
                                                                     Comedy|Romance
                    Waiting to Exhale (1995)
                                                               Comedy|Drama|Romance
             Father of the Bride Part II (1995)
                                                                              Comedy
In [82]:
         ratings[-10:]
```

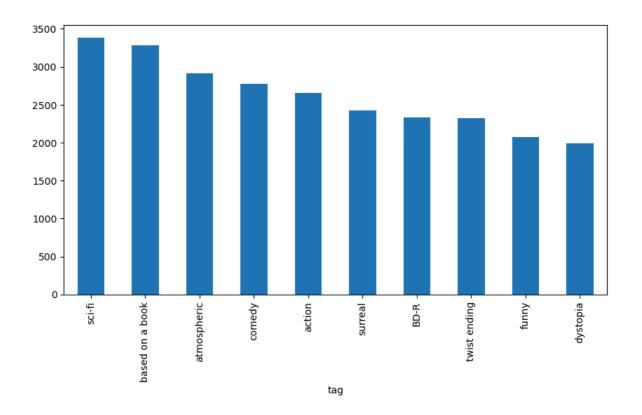
Out[82]:		userId	movield	rating
	20000253	138493	60816	4.5
	20000254	138493	61160	4.0
	20000255	138493	65682	4.5
	20000256	138493	66762	4.5
	20000257	138493	68319	4.5
	20000258	138493	68954	4.5
	20000259	138493	69526	4.5
	20000260	138493	69644	3.0
	20000261	138493	70286	5.0

20000262 138493

71619

2.5

```
In [83]: tag_counts = tags['tag'].value_counts()
          tag_counts[-10:]
Out[83]: tag
          missing child
                                             1
          Ron Moore
                                             1
          Citizen Kane
                                             1
          mullet
                                             1
          biker gang
                                             1
          Paul Adelstein
                                             1
          the wig
          \quad \hbox{killer fish} \quad
          genetically modified monsters
          topless scene
                                             1
          Name: count, dtype: int64
In [89]: tag_counts[:10].plot(kind = 'bar' , figsize= (10,5))
          plt.show()
```



Filter for Selecting Rows

```
In [90]: is_high_rated = ratings['rating'] >= 5.0
    ratings[is_high_rated][30:50]
```

Out[90]	•	userlo

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

	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

Out[92]:

In [93]: movies[is_action].head(15)

Out[93]:

	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

```
Out[94]:
                 movield
         rating
            0.5
                  239125
            1.0
                  680732
                  279252
            1.5
                1430997
            2.0
            2.5
                  883398
            3.0 4291193
            3.5 2200156
            4.0 5561926
            4.5 1534824
            5.0 2898660
         average_rating = ratings[['movieId', 'rating']].groupby('movieId').count()
In [97]:
         average_rating.head()
Out[97]:
                  rating
          movield
                1 49695
                   22243
                  12735
                    2756
                  12161
In [98]: average_rating = ratings[['movieId','rating']].groupby('movieId').mean()
         average_rating.head()
Out[98]:
                     rating
          movield
                1 3.921240
                2 3.211977
                3 3.151040
                4 2.861393
                5 3.064592
         movie_count = ratings[['movieId','rating']].groupby('movieId').count()
In [99]:
         movie_count.head()
```

```
rating
          movield
                 1 49695
                    22243
                 3 12735
                     2756
                 5 12161
In [100...
          movie_count = ratings[['movieId','rating']].groupby('movieId').count()
          movie_count.tail()
Out[100...
                    rating
           movield
           131254
                        1
           131256
           131258
                        1
           131260
           131262
                        1
```

Merge Dataframes

Out[99]:

```
In [101...
           tags.head()
Out[101...
               userld movield
                                         tag
            0
                   18
                          4141 Mark Waters
            1
                   65
                           208
                                    dark hero
            2
                   65
                            353
                                    dark hero
            3
                   65
                            521
                                  noir thriller
                   65
                            592
                                    dark hero
In [102...
          movies.head()
```

```
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
                               Father of the Bride Part II
                      5
                                                                                            Comedy
            4
                                                (1995)
In [103...
                movies.merge(tags, on = 'movieId' , how = 'inner')
            t.head()
Out[103...
                                                                                                tag
                movield
                            title
                                                                       genres
                                                                               userld
                             Toy
            0
                      1
                           Story
                                  Adventure|Animation|Children|Comedy|Fantasy
                                                                                 1644
                                                                                            Watched
                          (1995)
                             Toy
                                                                                           computer
            1
                      1
                           Story
                                  Adventure|Animation|Children|Comedy|Fantasy
                                                                                 1741
                                                                                           animation
                          (1995)
                             Toy
                                                                                              Disney
            2
                                  Adventure|Animation|Children|Comedy|Fantasy
                           Story
                                                                                 1741
                                                                                           animated
                          (1995)
                                                                                             feature
                             Toy
                                                                                                Pixar
            3
                      1
                                  Adventure|Animation|Children|Comedy|Fantasy
                                                                                 1741
                           Story
                                                                                          animation
                          (1995)
                                                                                         Téa Leoni
                             Toy
                                                                                            does not
            4
                      1
                                  Adventure|Animation|Children|Comedy|Fantasy
                                                                                 1741
                           Story
                                                                                          star in this
                          (1995)
                                                                                              movie
In [106...
            avg_ratings = ratings.groupby('movieId',as_index = False).mean()
            del avg_ratings['userId']
            avg_ratings.head()
Out[106...
                movield
                            rating
            0
                         3.921240
            1
                         3.211977
            2
                         3.151040
                         2.861393
            3
                         3.064592
            4
  In [ ]:
```

title

genres

Out[102...

movield