PROJECT-4: KAGGLE INTRODUCTION & IMDB Movie Rating Analysis

```
In [1]: import pandas as pd
 In [2]: ratings = pd.read csv(r'E:\archive\rating.csv')
 In [3]: ratings.shape
Out[3]: (20000263, 4)
 In [4]: ratings.head(1)
 Out[4]:
            userld movield rating
                                            timestamp
         0
                 1
                          2
                                3.5 2005-04-02 23:53:47
 In [5]: tags = pd.read_csv(r'E:\archive\tag.csv')
 In [6]: tags.shape
 Out[6]: (465564, 4)
 In [7]: tags.head(1)
 Out[7]:
            userld movield
                                     tag
                                                 timestamp
                18
                       4141 Mark Waters 2009-04-24 18:19:40
 In [8]: movies = pd.read_csv(r'E:\archive\movie.csv')
         movies.head(1)
 Out[8]:
            movield
                                title
                                                                      genres
         0
                   1 Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy
 In [9]: # For current analysis, we will remove timestamp
         del ratings['timestamp']
         del tags['timestamp']
In [10]: ratings.columns
Out[10]: Index(['userId', 'movieId', 'rating'], dtype='object')
In [11]: tags.columns
```

```
Out[11]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [12]: tags.head()
Out[12]:
            userld movield
                                    tag
         0
                18
                       4141 Mark Waters
         1
                65
                        208
                               dark hero
         2
                65
                        353
                              dark hero
         3
                              noir thriller
                65
                        521
                        592
         4
                65
                               dark hero
         Data Structures
         Series
```

```
In [13]: row_0 = tags.iloc[0]
         type(row_0)
Out[13]: pandas.core.series.Series
In [14]: print(row_0)
                            18
        userId
        movieId
                          4141
        tag
                   Mark Waters
        Name: 0, dtype: object
In [15]: row_0.index
Out[15]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [16]: row_0['userId']
Out[16]: 18
In [17]: 'rating' in row_0
Out[17]: False
In [18]: row_0.name
Out[18]: 0
In [19]: row_0 = row_0.rename('firstRow')
         row_0.name
```

DataFrames

```
In [20]:
         tags.head()
Out[20]:
             userld movield
                                      tag
          0
                 18
                        4141
                              Mark Waters
                 65
                         208
                                 dark hero
          2
                 65
                         353
                                 dark hero
                                noir thriller
                 65
                         521
                 65
                         592
                                 dark hero
In [21]:
         tags.index
Out[21]: RangeIndex(start=0, stop=465564, step=1)
In [22]: tags.columns
Out[22]: Index(['userId', 'movieId', 'tag'], dtype='object')
In [23]: tags.iloc[ [0,11,500] ]
Out[23]:
               userld movield
                                            tag
            0
                   18
                          4141
                                    Mark Waters
                          1783
                                     noir thriller
           11
                   65
          500
                  342
                         55908 entirely dialogue
```

Descriptive Statistics

how the ratings are distributed!

```
In [24]: ratings['rating'].describe()
```

```
Out[24]: 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
                  5.000000e+00
         max
         Name: rating, dtype: float64
In [25]: ratings.describe()
Out[25]:
                      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
                3.439500e+04 9.020000e+02 3.000000e+00
           25%
           50% 6.914100e+04 2.167000e+03 3.500000e+00
               1.036370e+05 4.770000e+03 4.000000e+00
           max 1.384930e+05 1.312620e+05 5.000000e+00
In [26]: ratings['rating'].mean()
Out[26]: 3.5255285642993797
In [27]:
         ratings.mean()
                    69045.872583
Out[27]: userId
         movieId
                     9041.567330
                        3.525529
         rating
         dtype: float64
In [28]: ratings['rating'].min()
Out[28]: 0.5
In [29]: ratings['rating'].max()
Out[29]: 5.0
        ratings['rating'].std()
Out[30]: 1.051988919275684
In [31]: ratings['rating'].mode()
```

```
Out[31]: 0 4.0
         Name: rating, dtype: float64
In [32]: ratings.corr()
Out[32]:
                   userld
                           movield
                                      rating
          userId
                 1.000000
                          -0.000850 0.001175
         movield -0.000850
                          1.000000 0.002606
          rating
                  In [33]: filter1 = ratings['rating'] > 10
         print(filter1)
        filter1.any()
                  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[33]: False
In [34]: filter2 = ratings['rating'] > 0
         filter2.all()
Out[34]: True
```

Data Cleaning: Handling Missing Data

```
In [35]: movies.shape
Out[35]: (27278, 3)
In [36]: movies.isnull().any().any()
Out[36]: False
          No NULL values!
In [37]: ratings.shape
Out[37]: (20000263, 3)
```

```
In [38]: ratings.isnull().any().any()
Out[38]: False
     No NULL values in Tags!

In [39]: tags.shape
Out[39]: (465564, 3)

In [40]: tags.isnull().any().any()
Out[40]: True
     We have some tags which are NULL.

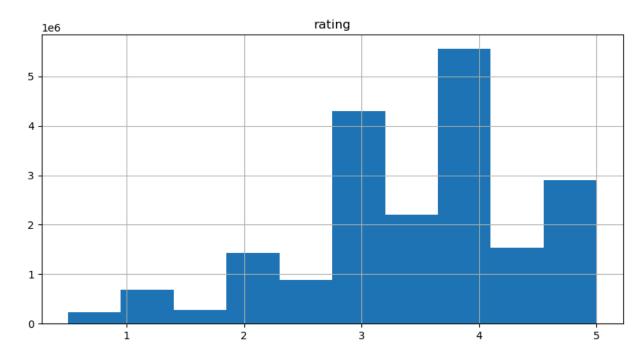
In [41]: tags=tags.dropna()

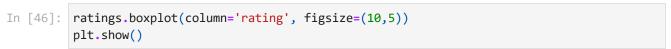
In [42]: tags.isnull().any().any()
Out[42]: False
In [43]: tags.shape
Out[43]: (465548, 3)
```

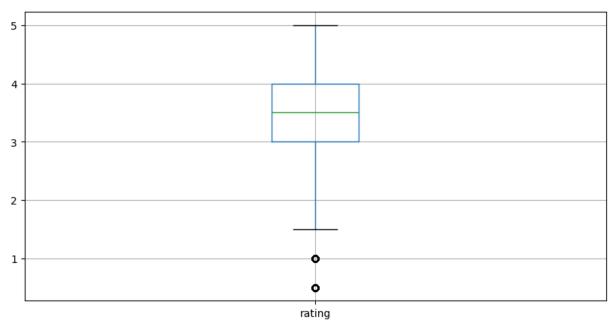
Thats nice! No NULL values! Notice the number of lines have reduced.

Data Visualization

```
In [45]: %matplotlib inline
    import matplotlib.pyplot as plt
    ratings.hist(column='rating', figsize=(10,5))
    plt.show()
```







Slicing Out Columns

```
In [48]: movies[['title','genres']].head()
Out[48]:
                                     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 [49]:
         ratings[-10:]
Out[49]:
                     userld 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
                               69526
                                         4.5
          20000259 138493
          20000260 138493
                               69644
                                         3.0
                               70286
          20000261 138493
                                         5.0
          20000262 138493
                               71619
                                         2.5
In [50]: tag_counts = tags['tag'].value_counts()
          tag_counts[-10:]
Out[50]: tag
                                            1
          missing child
          Ron Moore
                                            1
          Citizen Kane
          mullet
                                            1
                                            1
          biker gang
                                            1
          Paul Adelstein
          the wig
                                            1
          killer fish
                                            1
          genetically modified monsters
                                            1
          topless scene
          Name: count, dtype: int64
In [51]: colors = plt.cm.Paired.colors
          tag_counts[:10].plot(kind='bar', figsize=(10,5),color=colors)
```

