```
import numpy as np
In [142]: | df = pd.read_csv('ratings.csv')
In [143]: df.head()
Out[143]:
               userld movield rating timestamp
                                4.0 964982703
                                4.0 964981247
                               4.0 964982224
                                5.0 964983815
                          50 5.0 964982931
In [144]: | movies_titles = pd.read_csv('movies.csv')
In [145]: movies_titles = movies_titles[['movieId', 'title']]
            # movies titles.head()
In [146]: movies_titles.head()
Out[146]:
               movield
                                   Toy Story (1995)
                    2
                                   Jumanji (1995)
                            Grumpier Old Men (1995)
                            Waiting to Exhale (1995)
                    5 Father of the Bride Part II (1995)
In [147]: df = pd.merge(df, movies_titles, on='movieId')
In [149]: df.head()
Out[149]:
                                                         title
               userld movield rating
                                     timestamp
                                     964982703 Toy Story (1995)
                                4.0 847434962 Toy Story (1995)
                                4.5 1106635946 Toy Story (1995)
                  15
                                2.5 1510577970 Toy Story (1995)
                  17
                               4.5 1305696483 Toy Story (1995)
In [151]: import matplotlib.pyplot as plt
In [152]: import seaborn as sns
In [153]: sns.set_style('white')
In [154]: %matplotlib inline
In [156]: df.groupby('title')['rating'].mean().sort_values(ascending=False).head()
Out[156]: title
           Karlson Returns (1970)
                                                                  5.0
           Winter in Prostokvashino (1984)
                                                                  5.0
                                                                  5.0
           My Love (2006)
           Sorority House Massacre II (1990)
           Winnie the Pooh and the Day of Concern (1972)
           Name: rating, dtype: float64
In [157]: | df.groupby('title')['rating'].count().sort_values(ascending=False).head()
Out[157]: title
           Forrest Gump (1994)
                                                    329
           Shawshank Redemption, The (1994)
                                                    317
           Pulp Fiction (1994)
                                                    307
           Silence of the Lambs, The (1991)
                                                    279
           Matrix, The (1999)
                                                    278
           Name: rating, dtype: int64
In [158]: ratings = pd.DataFrame(df.groupby('title')['rating'].mean())
In [159]: ratings.head()
Out[159]:
                                             rating
                                        title
                                   '71 (2014)
                                               4.0
            'Hellboy': The Seeds of Creation (2004)
                                               4.0
                         'Round Midnight (1986)
                                               3.5
                            'Salem's Lot (2004)
                                               5.0
                       'Til There Was You (1997)
                                               4.0
In [161]: ratings['no of ratings'] = pd.DataFrame(df.groupby('title')['rating'].count())
In [163]: ratings.head()
Out[163]:
                                             rating no of ratings
                                        title
                                   '71 (2014)
                                               4.0
            'Hellboy': The Seeds of Creation (2004)
                                               4.0
                         'Round Midnight (1986)
                                               3.5
                            'Salem's Lot (2004)
                                               5.0
                       'Til There Was You (1997)
In [164]: ratings['no of ratings'].hist(bins=70)
Out[164]: <matplotlib.axes._subplots.AxesSubplot at 0x215eedad160>
            6000
            5000
            4000
             3000
            2000
             1000
In [165]: ratings['rating'].hist(bins=70)
Out[165]: <matplotlib.axes._subplots.AxesSubplot at 0x215eeee67b8>
             1000
             800
             600
             400
             200
In [166]: sns.jointplot(x='rating', y='no of ratings', data=ratings, alpha=0.5)
Out[166]: <seaborn.axisgrid.JointGrid at 0x215f10ebf60>
               300
               250
               200
               150
               100
               50
In [167]: df.head()
Out[167]:
                                                         title
               userld movield rating
                                     timestamp
                                     964982703 Toy Story (1995)
                                     847434962 Toy Story (1995)
                                   1106635946 Toy Story (1995)
                  15
                                2.5 1510577970 Toy Story (1995)
                  17
                                4.5 1305696483 Toy Story (1995)
In [168]: moviemat = df.pivot_table(index='userId', columns='title', values='rating')
In [169]: moviemat.head()
Out[169]:
                          'Hellboy':
                                                       'Til 'Tis the
                                                                                    (500)
                                                                                         *batteries
                                                          Season 'burbs,
                                    'Round 'Salem's There
                                                                          'night
                              The
                                                                                 Days of
                                                                                                       Zulu [REC] [REC]<sup>2</sup>
                                                                                               not
               title
                          Seeds of
                                   Midnight
                                                     Was
                                                                     The Mother
                                                Lot
                                                              for
                   (2014)
                                                                                          included
                                                                                                      (2013) (2007) (2009)
                                                                                 Summer
                                             (2004)
                          Creation
                                    (1986)
                                                     You
                                                             Love (1989) (1986)
                                                                                   (2009)
                                                                                            (1987)
                                                    (1997)
                            (2004)
                                                           (2015)
            userld
                1
                    NaN
                              NaN
                                      NaN
                                               NaN
                                                     NaN
                                                             NaN
                                                                    NaN
                                                                           NaN
                                                                                    NaN
                                                                                              NaN ...
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                                                                                                              NaN
                                                                                                                     NaN
                2
                    NaN
                              NaN
                                      NaN
                                               NaN
                                                     NaN
                                                             NaN
                                                                    NaN
                                                                            NaN
                                                                                    NaN
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                                                                                                              NaN
                                                                                                                     NaN
                3
                    NaN
                                                             NaN
                                                                    NaN
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                                                                                                                     NaN
                              NaN
                                      NaN
                                               NaN
                                                     NaN
                4
                    NaN
                              NaN
                                      NaN
                                              NaN
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                                                             NaN
                                                                    NaN
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                    NaN
                              NaN
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                                                                    NaN
                                                                           NaN
                                                                                    NaN
                                                                                              NaN
                                                                                                       NaN
                                                                                                              NaN
                                                                                                                     NaN
                                               NaN
                                                     NaN
           5 rows × 9719 columns
In [170]: ratings.sort_values('no of ratings', ascending=False).head(10)
Out[170]:
                                                   rating no of ratings
                                           title
                              Forrest Gump (1994) 4.164134
                                                                329
                                                                317
                  Shawshank Redemption, The (1994) 4.429022
                                Pulp Fiction (1994) 4.197068
                                                                307
                    Silence of the Lambs, The (1991) 4.161290
                                                                279
                                Matrix, The (1999) 4.192446
                                                                278
            Star Wars: Episode IV - A New Hope (1977) 4.231076
                                                                251
                              Jurassic Park (1993) 3.750000
                                                                238
                                Braveheart (1995) 4.031646
                                                                237
                  Terminator 2: Judgment Day (1991) 3.970982
                                                                224
                            Schindler's List (1993) 4.225000
                                                                220
In [172]: forrest_gump_ratings = moviemat['Forrest Gump (1994)']
            matrix_ratings = moviemat['Matrix, The (1999)']
In [173]: | similar_to_forrest_gump = moviemat.corrwith(forrest_gump_ratings)
           C:\Users\Vikas\Anaconda3\lib\site-packages\numpy\lib\function_base.py:2522: RuntimeWarning: Degr
            ees of freedom <= 0 for slice
             c = cov(x, y, rowvar)
           C:\Users\Vikas\Anaconda3\lib\site-packages\numpy\lib\function_base.py:2451: RuntimeWarning: divi
           de by zero encountered in true_divide
             c *= np.true_divide(1, fact)
In [174]: similar_to_matrix = moviemat.corrwith(matrix_ratings)
In [176]: corr_forrest_gump = pd.DataFrame(similar_to_forrest_gump, columns=['Correlation'])
            corr_forrest_gump.dropna(inplace=True)
In [177]: corr_forrest_gump.head()
Out[177]:
                                      Correlation
                      'burbs, The (1989)
                                       0.197712
            (500) Days of Summer (2009)
                                       0.234095
            *batteries not included (1987)
                                       0.892710
               ...And Justice for All (1979)
                                       0.928571
                   10 Cent Pistol (2015) -1.000000
In [178]: | corr_forrest_gump.sort_values('Correlation', ascending=False).head()
Out[178]:
                                        Correlation
                                    title
                       Lost & Found (1999)
                                               1.0
               Century of the Self, The (2002)
                                               1.0
                       The 5th Wave (2016)
                                               1.0
            Play Time (a.k.a. Playtime) (1967)
                                               1.0
                Memories (Memorîzu) (1995)
                                               1.0
In [179]: corr_forrest_gump = corr_forrest_gump.join(ratings['no of ratings'])
In [180]: corr forrest gump.head()
Out[180]:
                                      Correlation no of ratings
                                title
                     'burbs, The (1989)
                                       0.197712
                                                        17
            (500) Days of Summer (2009)
                                       0.234095
                                                        42
            *batteries not included (1987)
                                       0.892710
               ...And Justice for All (1979)
                                       0.928571
                                                         3
                   10 Cent Pistol (2015)
                                      -1.000000
In [181]: corr forrest gump[corr forrest gump['no of ratings']>100].sort values('Correlation', ascending=False
            ).head()
Out[181]:
                                   Correlation no of ratings
                                                     329
                 Forrest Gump (1994)
                                     1.000000
              Good Will Hunting (1997)
                                     0.484042
                                                     141
                      Aladdin (1992)
                                     0.464268
                                                     183
            American History X (1998)
                                     0.457287
                                                     129
             Truman Show, The (1998)
                                                     125
                                     0.432556
In [182]: corr_matrix = pd.DataFrame(similar_to_matrix, columns=['Correlation'])
In [184]: corr_matrix.dropna(inplace=True)
            corr_matrix.head()
Out[184]:
                                      Correlation
                                 title
                                      -0.160843
                      'burbs, The (1989)
            (500) Days of Summer (2009)
                                       0.302316
            *batteries not included (1987)
                                       0.392232
                                       0.654654
              ...And Justice for All (1979)
                   10 Cent Pistol (2015) -1.000000
In [187]: corr_matrix = corr_matrix.join(ratings['no of ratings'])
In [188]: corr matrix.head()
Out[188]:
                                      Correlation
                                                  rating no of ratings
                                title
                     'burbs, The (1989)
                                      -0.160843 3.176471
                                                                 17
            (500) Days of Summer (2009)
                                                                 42
                                       0.302316 3.666667
            *batteries not included (1987)
                                       0.392232 3.285714
              ...And Justice for All (1979)
                                       0.654654 3.166667
                                                                  3
                   10 Cent Pistol (2015) -1.000000 1.250000
In [190]: corr_matrix[corr_matrix['no of ratings']>100].sort_values('Correlation', ascending=False).head()
Out[190]:
                                          rating no of ratings
                             Correlation
                        title
             Matrix, The (1999)
                              1.000000 4.192446
                                                       278
              Die Hard (1988)
                              0.544466 3.862069
                                                       145
```

Inception (2010)

Aliens (1986)

Braveheart (1995)

In []:

0.514767 4.066434

0.496045 4.031646

0.470865 3.964286

143

237

126

In [1]: import pandas as pd