2020-09-30-221223

September 30, 2020

1 Exploratory Data Analysis

- 2 The Miner League
- 2.1 Manav Agarwal:PES2201800025
- 2.2 Rishab Kashyap:PES2201800065
- 2.3 Shreya Venugopal:PES2201800688

```
[23]: import pandas as pd
import numpy as np
import seaborn as sn
import matplotlib.pyplot as plt
```

2.3.1 Reading data from all 4 tables

```
[3]: # Read data from file 'filename.csv'
# (in the same directory that your python process is based)
# Control delimiters, rows, column names with read_csv (see later)
Movies = pd.read_csv("IMDb movies.csv")
# Preview the first 5 lines of the loaded data
Movies.head()
```

```
[3]:
       imdb_title_id
                                                                   title \
           tt0000574
                                            The Story of the Kelly Gang
     1
           tt0001892
                                                          Den sorte drøm
     2
           tt0002101
                                                               Cleopatra
     3
           tt0002130
                                                               L'Inferno
     4
           tt0002199 From the Manger to the Cross; or, Jesus of Naz...
                                           original_title year date_published \
     0
                              The Story of the Kelly Gang 1906
                                                                     1906-12-26
     1
                                           Den sorte drøm 1911
                                                                     1911-08-19
     2
                                                Cleopatra 1912
                                                                     1912-11-13
     3
                                                 L'Inferno 1911
                                                                     1911-03-06
       From the Manger to the Cross; or, Jesus of Naz... 1912
                                                                         1913
```

```
duration
                                                         country language
                             genre
                                           70
     0
          Biography, Crime, Drama
                                                      Australia
                                                                      NaN
     1
                             Drama
                                           53
                                               Germany, Denmark
                                                                      NaN
     2
                    Drama, History
                                          100
                                                             USA
                                                                  English
     3
        Adventure, Drama, Fantasy
                                           68
                                                                  Italian
                                                           Italy
     4
                 Biography, Drama
                                           60
                                                             USA
                                                                  English
                                     director
                                Charles Tait
     0
     1
                                   Urban Gad ...
     2
                          Charles L. Gaskill ...
     3
       Francesco Bertolini, Adolfo Padovan ...
                               Sidney Olcott ...
                                                     actors \
     O Elizabeth Tait, John Tait, Norman Campbell, Be...
     1 Asta Nielsen, Valdemar Psilander, Gunnar Helse...
     2 Helen Gardner, Pearl Sindelar, Miss Fielding, ...
     3 Salvatore Papa, Arturo Pirovano, Giuseppe de L...
     4 R. Henderson Bland, Percy Dyer, Gene Gauntier,...
                                                description avg_vote votes
                                                                               budget \
        True story of notorious Australian outlaw Ned ...
                                                                6.1
                                                                             $ 2250
                                                                      537
       Two men of high rank are both wooing the beaut...
                                                                5.9
                                                                      171
                                                                                NaN
     2 The fabled queen of Egypt's affair with Roman ...
                                                                5.2
                                                                            $ 45000
                                                                      420
     3 Loosely adapted from Dante's Divine Comedy and...
                                                                7.0
                                                                     2019
                                                                                NaN
       An account of the life of Jesus Christ, based ...
                                                                5.7
                                                                      438
                                                                                NaN
        usa_gross_income worlwide_gross_income metascore reviews_from_users
     0
                      NaN
                                             NaN
                                                        NaN
                                                                            7.0
                      NaN
                                                                            4.0
     1
                                             NaN
                                                        NaN
     2
                                                                           24.0
                      NaN
                                             NaN
                                                        NaN
     3
                                                                           28.0
                      NaN
                                             NaN
                                                        NaN
     4
                                                                           12.0
                      NaN
                                             NaN
                                                        NaN
        reviews_from_critics
     0
                          7.0
                          2.0
     1
     2
                          3.0
     3
                         14.0
                          5.0
     [5 rows x 22 columns]
[4]: Names = pd.read_csv("IMDb_names.csv")
     # Preview the first 5 lines of the loaded data
     Names.head()
```

```
[4]:
       imdb_name_id
                                                                 height \
                                 name
                                                     birth_name
                                      Frederic Austerlitz Jr.
     0
          nm0000001
                         Fred Astaire
                                                                   177.0
     1
          nm0000002
                                             Betty Joan Perske
                                                                   174.0
                       Lauren Bacall
     2
          nm0000003
                     Brigitte Bardot
                                                Brigitte Bardot
                                                                   166.0
          nm0000004
     3
                         John Belushi
                                              John Adam Belushi
                                                                   173.0
          nm0000005
                       Ingmar Bergman
                                          Ernst Ingmar Bergman
                                                                   179.0
                                                        bio \
      Fred Astaire was born in Omaha, Nebraska, to J...
     1 Lauren Bacall was born Betty Joan Perske on Se...
     2 Brigitte Bardot was born on September 28, 1934...
     3 John Belushi was born in Chicago, Illinois, US...
     4 Ernst Ingmar Bergman was born July 14, 1918, t...
                                              birth_details
                                                             birth_year \
     0
                     May 10, 1899 in Omaha, Nebraska, USA
                                                                  1899.0
     1
        September 16, 1924 in The Bronx, New York City...
                                                               1924.0
     2
                      September 28, 1934 in Paris, France
                                                                  1934.0
     3
               January 24, 1949 in Chicago, Illinois, USA
                                                                  1949.0
            July 14, 1918 in Uppsala, Uppsala län, Sweden
                                                                  1918.0
       date of birth
                                                 place_of_birth
     0
          1899-05-10
                                           Omaha, Nebraska, USA
     1
          1924-09-16
                      The Bronx, New York City, New York, USA
     2
                                                  Paris, France
          1934-09-28
                                        Chicago, Illinois, USA
     3
          1949-01-24
          1918-07-14
                                  Uppsala, Uppsala län, Sweden
                                              death_details
                                                             death_year \
        June 22, 1987 in Los Angeles, California, USA ...
                                                                1987.0
        August 12, 2014 in New York City, New York, US...
                                                                2014.0
     1
     2
                                                        NaN
                                                                     NaN
        March 5, 1982 in Hollywood, Los Angeles, Calif...
     3
                                                                1982.0
              July 30, 2007 in Fårö, Gotlands län, Sweden
                                                                  2007.0
       date_of_death
                                                   place_of_death \
     0
          1987-06-22
                                  Los Angeles, California, USA
     1
          2014-08-12
                                  New York City, New York, USA
     2
                                                              NaN
                 NaN
          1982-03-05
     3
                      Hollywood, Los Angeles, California, USA
          2007-07-30
                                      Fårö, Gotlands län, Sweden
                               reason_of_death
                                                 spouses
                                                          divorces
     0
                                                       2
                                     pneumonia
                                                                  0
     1
                                        stroke
                                                                  1
     2
                                                       4
                                                                  3
        acute cocaine and heroin intoxication
                                                                  0
```

```
4
                                             NaN
                                                        5
                                                                   4
        spouses_with_children
                                 children
                                                        primary_profession \
     0
                                            soundtrack, actor, miscellaneous
                              1
                              2
                                        3
     1
                                                        actress, soundtrack
     2
                              1
                                        1
                                               actress, soundtrack, producer
     3
                              0
                                        0
                                                   actor, writer, soundtrack
     4
                              5
                                        8
                                                     writer, director, actor
                                 known_for_titles
       tt0050419,tt0053137,tt0072308,tt0043044
     1 tt0037382,tt0038355,tt0117057,tt0071877
     2 tt0054452,tt0059956,tt0057345,tt0049189
     3 tt0078723,tt0072562,tt0080455,tt0077975
     4 tt0050976,tt0083922,tt0069467,tt0050986
[5]: Ratings = pd.read_csv("IMDb_ratings.csv")
     # Preview the first 5 lines of the loaded data
     Ratings.head()
[5]:
       imdb_title_id weighted_average_vote total_votes
                                                             mean_vote
                                                                         median_vote \
     0
           tt0000574
                                           6.1
                                                         537
                                                                    6.3
                                                                                  6.0
     1
           tt0001892
                                           5.9
                                                         171
                                                                    6.1
                                                                                  6.0
     2
           tt0002101
                                           5.2
                                                        420
                                                                    5.2
                                                                                  5.0
                                           7.0
                                                                    6.9
                                                                                  7.0
     3
           tt0002130
                                                       2019
     4
           tt0002199
                                           5.7
                                                         438
                                                                    5.8
                                                                                  6.0
        votes_10
                   votes_9
                            votes_8
                                      votes_7
                                                votes_6
                                                         •••
                                                            females_30age_avg_vote
     0
              54
                        17
                                  55
                                           121
                                                    122
                                                                                 6.0
     1
                5
                         6
                                  17
                                            41
                                                     52
                                                                                 5.8
     2
              12
                         8
                                  16
                                           60
                                                     89
                                                                                 5.5
     3
             194
                       208
                                 386
                                                                                 7.3
                                           571
                                                    308
     4
              28
                        15
                                  42
                                            75
                                                    114
                                                                                 4.8
                               females_45age_avg_vote
        females_30age_votes
                                                        females 45age votes
     0
                        19.0
                                                   6.6
                                                                         14.0
     1
                         4.0
                                                   6.5
                                                                          8.0
                                                   6.2
                                                                         20.0
     2
                        14.0
     3
                        74.0
                                                   7.4
                                                                         75.0
     4
                                                   6.5
                        10.0
                                                                         15.0
        top1000_voters_rating
                                top1000_voters_votes
                                                        us_voters_rating \
     0
                           6.3
                                                                      6.0
                                                  64.0
                           5.9
                                                                       6.2
     1
                                                  29.0
     2
                           4.9
                                                  57.0
                                                                      5.5
     3
                           7.0
                                                 126.0
                                                                      7.1
     4
                                                                      5.9
                           5.7
                                                  56.0
```

```
us_voters_votes non_us_voters_rating non_us_voters_votes
     0
                   89.0
                                           6.2
                                                               309.0
                   27.0
                                           6.0
     1
                                                               114.0
     2
                  197.0
                                           4.7
                                                               103.0
     3
                  452.0
                                           7.0
                                                              1076.0
                  161.0
                                           5.7
                                                               164.0
     [5 rows x 49 columns]
[8]: TitleP = pd.read_csv("IMDb_title_principals.csv")
     # Preview the first 5 lines of the loaded data
     TitleP.head()
[8]:
       imdb title id ordering imdb name id category
                                                                     characters
                                                        job
           tt0000574
                              1
                                   nm0846887
                                                                 ["Kate Kelly"]
     0
                                               actress
                                                        NaN
                              2
                                                              ["School Master"]
     1
           tt0000574
                                   nm0846894
                                                        NaN
                                                 actor
     2
           tt0000574
                              3
                                   nm3002376
                                                        NaN
                                                                 ["Steve Hart"]
                                                 actor
     3
           tt0000574
                              4
                                   nm0170118
                                                                            NaN
                                               actress
                                                        NaN
           tt0000574
                                   nm0846879 director NaN
                              5
                                                                            NaN
    2.3.2 Filtering to movies only after 2005 for analysis
[9]: greater than 2005 = Movies['year']>2005
     new_movies=Movies[greater_than_2005]
     print(new_movies.shape)
     new movies.head()
    (34785, 22)
[9]:
           imdb_title_id
                                                title
                                                                    original_title \
     15090
               tt0069049
                          The Other Side of the Wind The Other Side of the Wind
     20324
               tt0085953
                                               Mo tai
                                                                            Mo tai
     24705
                                La Telenovela Errante
                                                             La Telenovela Errante
               tt0100275
     29800
               tt0119231
                                 The Junior Defenders
                                                              The Junior Defenders
               tt0119970
                                         Quality Time
     30117
                                                                      Quality Time
            year date_published
                                                   genre duration \
```

Comedy, Drama, Fantasy

Comedy, Drama, Sci-Fi

language

Cantonese

122

84

80

83

82

director ...

Orson Welles ...

Hung-Chuen Lau ...

Drama

Horror

Comedy

2018-11-02

2019-10-22

2018-09-06

2007-10-23

country 15090 France, Iran, USA English, German

Hong Kong

2008

15090

20324

24705

29800

30117

20324

2018

2019

2017

2007

2008

```
24705
                   Chile
                                   Spanish Raoul Ruiz, Valeria Sarmiento
                      USA
29800
                                   English
                                                             Keith Spiegel
30117
              India, USA
                                   English
                                                              Chris LaMont ...
                                                    actors \
       John Huston, Oja Kodar, Peter Bogdanovich, Sus...
15090
20324
       Eddie Chan, Yung-Chang Chin, Pak-Kwong Ho, Men...
       Luis Alarcón, Patricia Rivadeneira, Francisco ...
24705
29800
       Ally Sheedy, Brian O'Halloran, Justin Henry, J...
30117
       Corin Nemec, Bruce Weitz, Nancy Allen, John de...
                                               description avg_vote votes \
15090
       A Hollywood director emerges from semi-exile w...
                                                              6.9 5014
20324
       A woman becomes possessed and dies after a mon...
                                                              6.2
                                                                     262
       The film revolves around the concept of soap o...
24705
                                                              6.6
                                                                     121
29800
       A deranged fan goes on a mission to kidnap the...
                                                              4.3
                                                                     136
       In an apocalyptic future, a serial killer hold...
                                                                     100
30117
                                                              3.8
                 usa_gross_income worlwide_gross_income metascore
         budget
15090
            NaN
                                                                78.0
                               NaN
                                                      NaN
20324
            NaN
                               NaN
                                                      NaN
                                                                 NaN
24705
                            $ 3624
                                                   $ 3624
            NaN
                                                                 NaN
29800
       $ 125000
                                                                 NaN
                               NaN
                                                      NaN
30117
            NaN
                               NaN
                                                      NaN
                                                                 NaN
      reviews_from_users reviews_from_critics
15090
                     70.0
20324
                     16.0
                                            14.0
24705
                      NaN
                                            26.0
                      2.0
                                             1.0
29800
30117
                      2.0
                                             2.0
```

[5 rows x 22 columns]

37623

2.3.3 Removing Rows with NaN values

tt0259324

```
[10]: new_movies=new_movies.dropna()
      print(new_movies.shape)
      print(new_movies.head)
     (3262, 22)
     <bound method NDFrame.head of</pre>
                                           imdb_title_id
                                                                        title
     original_title year
     34961
                tt0200465
                                 The Bank Job
                                                     The Bank Job
                                                                    2008
     35307
                tt0206634
                              Children of Men
                                                  Children of Men
                                                                    2006
     35546
                tt0211933
                                        Awake
                                                             Awake
                                                                    2007
```

Ghost Rider

Ghost Rider 2007

```
38931
          tt0283503
                         Brooklyn Rules
                                             Brooklyn Rules
                                                               2007
80665
          tt8772262
                              Midsommar
                                                   Midsommar
                                                               2019
80858
          tt9024106
                              Unplanned
                                                   Unplanned
                                                               2019
                      Cold Blood Legacy
80908
          tt9082020
                                          Cold Blood Legacy
                                                               2019
80946
                      Playing with Fire
                                          Playing with Fire
                                                               2019
          tt9134216
81059
          tt9354944
                                    Jexi
                                                        Jexi
                                                               2019
      date_published
                                                    duration
                                            genre
          2008-02-29
34961
                                  Crime, Thriller
                                                         111
                                  Drama, Thriller
          2006-09-22
35307
                                                         109
35546
          2008-04-04
                           Crime, Drama, Mystery
                                                          84
37623
          2007-03-02
                       Action, Fantasy, Thriller
                                                         114
38931
          2009-05-28
                                     Crime, Drama
                                                          99
80665
          2019-07-03
                          Drama, Horror, Mystery
                                                         147
80858
          2019-10-04
                                 Biography, Drama
                                                         109
80908
          2019-05-15
                                 Action, Thriller
                                                          91
80946
          2019-11-08
                                   Comedy, Family
                                                          96
81059
          2019-10-11
                                           Comedy
                                                          84
                         country \
34961
             UK, USA, Australia
35307
                  USA, UK, Japan
35546
                             USA
37623
                  USA, Australia
                             USA
38931
           USA, Sweden, Hungary
80665
80858
                             USA
80908
       France, Ukraine, Belgium
80946
                             USA
81059
                             USA
                                                   language \
34961
                                            English, Latin
       English, German, Italian, Romanian, Spanish, A...
35307
35546
                                         English, Japanese
37623
                                                    English
38931
                                                    English
80665
                                          English, Swedish
80858
                                                    English
80908
                                           French, English
80946
                                                    English
81059
                                                    English
```

director ... \

```
34961
                      Roger Donaldson
35307
                       Alfonso Cuarón
35546
                          Joby Harold
37623
                 Mark Steven Johnson
38931
                     Michael Corrente
80665
                            Ari Aster
80858
       Chuck Konzelman, Cary Solomon
80908
                  Frédéric Petitjean
                         Andy Fickman
80946
81059
              Jon Lucas, Scott Moore
                                                    actors
34961
       Jason Statham, Saffron Burrows, Stephen Campbe...
35307
       Clive Owen, Juan Gabriel Yacuzzi, Michael Cain...
       Hayden Christensen, Jessica Alba, Terrence How...
35546
37623
       Matt Long, Raquel Alessi, Brett Cullen, Peter ...
38931
       Paulo Araujo, Daniel Tay, Ty Thomas Reed, Bria...
80665
       Florence Pugh, Jack Reynor, Vilhelm Blomgren, ...
80858
       Ashley Bratcher, Brooks Ryan, Robia Scott, Jar...
80908
       Jean Reno, Sarah Lind, Joe Anderson, David Gya...
80946
       John Cena, Keegan-Michael Key, John Leguizamo,...
       Adam Devine, Alexandra Shipp, Rose Byrne, Ron ...
81059
                                               description avg_vote
                                                                       votes
34961
       Martine offers Terry a lead on a foolproof ban...
                                                               7.2
                                                                    167682
35307
       In 2027, in a chaotic world in which women hav...
                                                               7.9
                                                                    437150
35546
       A wealthy young man undergoing heart transplan...
                                                               6.5
                                                                     70300
37623
       When motorcycle rider Johnny Blaze sells his s...
                                                               5.2
                                                                    212991
38931
       Brooklyn, 1985. With the mob world as a backdr...
                                                               6.3
                                                                      5347
                                                               7.2
80665
       A couple travels to Sweden to visit a rural ho...
                                                                     78830
80858
       All Abby Johnson ever wanted to do was help wo...
                                                               5.8
                                                                     10783
       A hit-man lives isolated in a cabin at the edg...
80908
                                                               4.5
                                                                      2193
       A crew of rugged firefighters meet their match...
                                                               4.5
                                                                      1067
       A comedy about what can happen when you love y...
                                                               6.6
                                                                      3674
                     usa_gross_income worlwide_gross_income metascore
            budget
34961
        $ 20000000
                           $ 30060660
                                                  $ 64828421
                                                                   69.0
35307
        $ 76000000
                           $ 35552383
                                                  $ 70099445
                                                                   84.0
35546
         $ 8600000
                           $ 14377198
                                                  $ 32685679
                                                                   33.0
37623
       $ 110000000
                          $ 115802596
                                                 $ 228738393
                                                                   35.0
38931
         $ 8000000
                             $ 458232
                                                    $ 458232
                                                                   53.0
80665
        $ 10000000
                           $ 27426361
                                                  $ 41123770
                                                                   72.0
80858
         $ 6000000
                           $ 19005109
                                                  $ 19320481
                                                                   10.0
80908
         $ 2700000
                               $ 5083
                                                   $ 1009888
                                                                   25.0
```

80946 81059	\$ 29900000 \$ 5000000	\$ 18847824 \$ 6546159	\$ 21788746 \$ 7257918	24.0 39.0
34961 35307 35546 37623	reviews_from_users 202.0 1310.0 207.0 703.0	reviews_from_critics 231.0 315.0 133.0 283.0		
38931	48.0	40.0		
 80665 80858 80908 80946 81059	 1454.0 531.0 72.0 49.0 46.0	365.0 32.0 26.0 16.0 24.0		
[3262	rows x 22 columns]>			

2.3.4 Analysing the correlation of various quantitative data

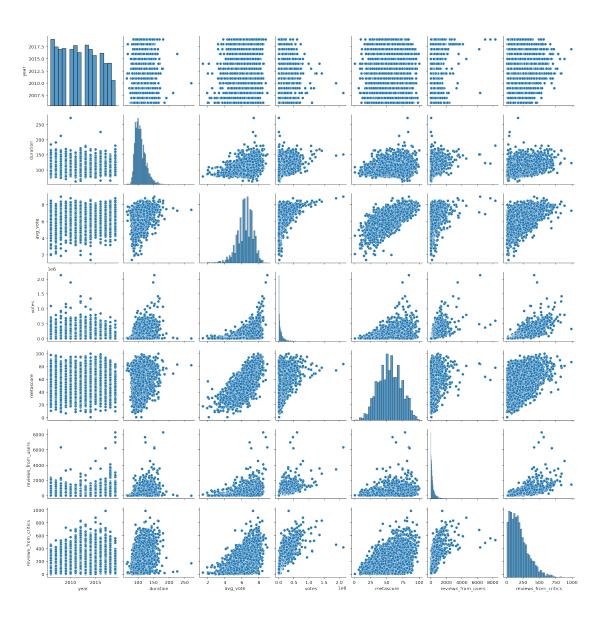
new_movies.corr()						
	year	duration	avg_vote	votes	metascore	\
year	1.000000	0.097212	0.033968	-0.023544	0.037554	
duration	0.097212	1.000000	0.399483	0.356525	0.267495	
avg_vote	0.033968	0.399483	1.000000	0.421946	0.715855	
votes	-0.023544	0.356525	0.421946	1.000000	0.287502	
metascore	0.037554	0.267495	0.715855	0.287502	1.000000	
reviews_from_users	0.143548	0.346309	0.259371	0.687401	0.188581	
reviews_from_critics	0.161950	0.364427	0.431228	0.742867	0.411304	
	reviews_f	rom_users	reviews_f	rom_critic	s	
year		0.143548		0.16195	0	
duration		0.346309		0.36442	:7	
avg_vote		0.259371		0.43122	8	
votes		0.687401		0.74286	7	
metascore		0.188581		0.41130	4	
reviews_from_users		1.000000		0.64745	0	
reviews_from_critics		0.647450		1.00000	0	

It is seen that the movies with greater critics reviews and greater user reviews have more votes. However, these two categories of reviews are highly correlated to each other which causes multi collinearity and also brings a need for multiple linear regression.

[32]: sn.pairplot(new_movies)

[32]: <seaborn.axisgrid.PairGrid at 0x7f7fff427c730>





[14]: new_names = Names.dropna() print(new_names.shape) print(new_names.head)

(3	654	,	20)
-1		•	

(0001, 20	,				
 bound me	thod NDFram	e.head of	imdb_name_id	name	
birth_nam	e height	\			
0	nm0000001	Fred Astaire	Frederic Austerli	tz Jr. 177.0)
1	nm0000002	Lauren Bacall	Betty Joan	Perske 174.0)
3	nm0000004	John Belushi	John Adam B	elushi 173.0)
5	nm0000006	Ingrid Bergman	Ingrid B	ergman 175.0)
6	nm0000007	Humphrey Bogart	Humphrey DeForest	Bogart 173.0)
	•••	•••	•••	•••	

```
nm5937849
                         Tatsuo Inoue
                                                    Tatsuo Inoue
                                                                    200.0
163420
                                                                    200.0
166483
          nm6649402
                          Salman Shah
                                        Shahriar Chowdhury Emon
167829
          nm7014999
                            Jay Bowdy
                                                       Jay Bowdy
                                                                    200.0
          nm7390393
                       Aachi Manorama
                                                     Gopishantha
                                                                    200.0
169090
                       Noelle Balfour
                                           Noëlle Marie Balfour
175063
          nm9626717
                                                                    200.0
                                                        bio \
0
        Fred Astaire was born in Omaha, Nebraska, to J...
        Lauren Bacall was born Betty Joan Perske on Se...
1
3
        John Belushi was born in Chicago, Illinois, US...
5
        Ingrid Bergman was one of the greatest actress...
6
        Humphrey DeForest Bogart was born in New York ...
        Tatsuo Inoue was born on November 8, 1941 in A...
163420
        Salman Shah was born on September 19, 1971 in ...
166483
        Frederick Jay Bowdy was born in San Antonio, T...
167829
169090
        Manorama is a veteran Tamil actress who has ac...
175063
       Noelle Balfour was born on December 11, 1984 i...
                                              birth details
                                                             birth year
                                                                  1899.0
0
                     May 10, 1899 in Omaha, Nebraska, USA
        September 16, 1924 in The Bronx, New York City...
1
                                                                1924.0
3
               January 24, 1949 in Chicago, Illinois, USA
                                                                  1949.0
5
                     August 29, 1915 in Stockholm, Sweden
                                                                  1915.0
6
        December 25, 1899 in New York City, New York, USA
                                                                  1899.0
              November 8, 1941 in Amagasaki, Hyogo, Japan
163420
                                                                  1941.0
166483
        September 19, 1971 in Daria para, Sylhet, Bang...
                                                                1971.0
               August 26, 1983 in San Antonio, Texas, USA
167829
                                                                  1983.0
169090
        May 26, 1937 in Mannargudi, Madras Presidency,...
                                                                1937.0
175063
                   December 11, 1984 in Reno, Nevada, USA
                                                                  1984.0
       date_of_birth
                                                      place_of_birth \
0
          1899-05-10
                                                Omaha, Nebraska, USA
                            The Bronx, New York City, New York, USA
1
          1924-09-16
                                              Chicago, Illinois, USA
3
          1949-01-24
5
          1915-08-29
                                                   Stockholm, Sweden
6
          1899-12-25
                                       New York City, New York, USA
163420
          1941-11-08
                                             Amagasaki, Hyogo, Japan
          1971-09-19
                                     Daria para, Sylhet, Bangladesh
166483
                                             San Antonio, Texas, USA
167829
          1983-08-26
169090
          1937-05-26
                      Mannargudi, Madras Presidency, British India
175063
          1984-12-11
                                                   Reno, Nevada, USA
                                              death_details
                                                            death_year \
0
        June 22, 1987 in Los Angeles, California, USA ...
                                                                1987.0
1
        August 12, 2014 in New York City, New York, US...
                                                                2014.0
```

```
3
        March 5, 1982 in Hollywood, Los Angeles, Calif...
                                                                 1982.0
5
        August 29, 1982 in Chelsea, London, England, U...
                                                                 1982.0
6
        January 14, 1957 in Los Angeles, California, U...
                                                                 1957.0
        October 5, 2016 in Nishinomiya, Hyôgo, Japan ...
163420
                                                                 2016.0
        September 6, 1996 in Dhaka, Bangladesh (suicide)
166483
                                                                   1996.0
167829
        January 23, 2017 in North Hollywood, Los Angel...
                                                                 2017.0
        October 10, 2015 in Chennai, Tamil Nadu, India...
169090
                                                                 2015.0
175063 October 11, 2017 in Escondido, California, USA...
                                                                 2017.0
                                                           place_of_death \
       date_of_death
          1987-06-22
                                         Los Angeles, California, USA
0
                                         New York City, New York, USA
1
          2014-08-12
3
                             Hollywood, Los Angeles, California, USA
          1982-03-05
                                         Chelsea, London, England, UK
5
          1982-08-29
6
          1957-01-14
                                         Los Angeles, California, USA
163420
          2016-10-05
                                             Nishinomiya, Hyôgo, Japan
          1996-09-06
                                                     Dhaka, Bangladesh
166483
167829
          2017-01-23
                       North Hollywood, Los Angeles, California, USA
169090
          2015-10-10
                                            Chennai, Tamil Nadu, India
          2017-10-11
                                            Escondido, California, USA
175063
                                             reason_of_death spouses
                                                                         divorces
0
                                                   pneumonia
                                                                     2
                                                                                0
                                                                     2
1
                                                                                1
                                                      stroke
3
                     acute cocaine and heroin intoxication
                                                                     1
                                                                                0
5
        lymphoma complications following a breast canc...
                                                                   3
                                                                              3
6
                                          esophageal cancer
                                                                     4
                                                                                3
163420
                                                                                0
                                                   emphysema
                                                                     0
166483
                                                     suicide
                                                                     1
                                                                                0
167829
                                               suicide - gun
                                                                     0
                                                                                0
                                                                                1
169090
                                                heart attack
                                                                     1
                                                                              0
175063
        complications from injuries sustained in a car...
                                                                   0
        spouses_with_children
                                 children
                                                            primary_profession
0
                                               soundtrack, actor, miscellaneous
                             2
                                        3
                                                            actress, soundtrack
1
3
                              0
                                        0
                                                      actor, writer, soundtrack
5
                              2
                                        4
                                                  actress, soundtrack, producer
6
                                        2
                                                    actor, soundtrack, producer
                              1
                                        0
163420
                              0
                                                                          actor
                                        0
166483
                              0
                                                       actor, music_department
167829
                              0
                                        0
169090
                              1
                                        1
                                                     actress, music_department
175063
                              0
                                           actress, casting_director, producer
```

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known_for_titles
     0
              tt0050419,tt0053137,tt0072308,tt0043044
     1
              tt0037382,tt0038355,tt0117057,tt0071877
     3
              tt0078723,tt0072562,tt0080455,tt0077975
     5
              tt0038109,tt0036855,tt0077711,tt0038787
     6
              tt0037382,tt0043265,tt0034583,tt0033870
     163420
              tt1105523,tt3825360,tt1438515,tt3178076
              tt5125172,tt5107670,tt3902736,tt5111408
     166483
     167829
                                   tt3903826,tt4943562
     169090
              tt0097416,tt0320736,tt0482389,tt1361809
     175063
              tt0942886,tt1723123,tt1332000,tt3467332
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[16]: new_ratings = Ratings.dropna()
      print(new_ratings.shape)
      print(new_ratings.head)
     (9948, 49)
     <bound method NDFrame.head of</pre>
                                            imdb_title_id weighted_average_vote
     total votes mean vote \
                tt0002130
                                               7.0
     3
                                                            2019
                                                                         6.9
                                               6.4
     45
                tt0004972
                                                           20885
                                                                         6.5
     73
                tt0006864
                                               7.8
                                                           13116
                                                                         7.9
                                               5.9
                                                             941
     132
                tt0009892
                                                                         7.1
                                               8.1
                                                                         7.9
     155
                tt0010323
                                                           51644
                                                                         2.2
     81227
                tt9783778
                                               1.7
                                                             998
     81234
                tt9799992
                                               6.6
                                                             867
                                                                         6.9
                                               7.9
     81242
                tt9817070
                                                            3210
                                                                         8.1
     81246
                                               5.0
                                                                         5.1
                tt9831136
                                                            1691
     81253
                tt9860728
                                               5.6
                                                            9816
                                                                         5.7
             median_vote
                           votes 10
                                     votes_9
                                               votes_8
                                                         votes 7
                                                                  votes 6
     3
                     7.0
                                194
                                          208
                                                   386
                                                             571
                                                                       308
     45
                     7.0
                               3548
                                         1668
                                                  3176
                                                            3789
                                                                      2680
     73
                     8.0
                               3372
                                         2133
                                                  3011
                                                            2080
                                                                      1085
     132
                    10.0
                                562
                                                     23
                                                              34
                                           13
                                                                        15
                     8.0
                                        10493
     155
                              10756
                                                 14688
                                                            8134
                                                                      3239
                                                               9
     81227
                     1.0
                                 51
                                            9
                                                     14
                                                                        16
     81234
                     7.0
                                 79
                                           97
                                                    174
                                                             216
                                                                       132
     81242
                     8.0
                                920
                                          580
                                                   831
                                                             429
                                                                       195
                     5.0
                                                             196
                                                                       279
     81246
                                115
                                           30
                                                    112
     81253
                     6.0
                                565
                                          277
                                                   757
                                                            1632
                                                                      2449
```

```
females_30age_avg_vote females_30age_votes females_45age_avg_vote \
3
                           7.3
                                                74.0
                                                                           7.4
45
                           5.9
                                              1000.0
                                                                           6.0
73
                           7.5
                                               599.0
                                                                           8.1
132
                           7.5
                                                60.0
                                                                           8.0
155
                           8.2
                                              3361.0
                                                                           8.1
81227
                                                32.0
                                                                           2.6
                           1.6
81234
                           6.6
                                                43.0
                                                                           7.1
81242
                           7.9
                                                81.0
                                                                           7.1
                                                49.0
81246
                           5.1
                                                                           5.3
81253
                           5.7
                                              1552.0
                                                                           5.9
       females_45age_votes top1000_voters_rating top1000_voters_votes \
                                                7.0
3
                       75.0
                                                                     126.0
45
                      687.0
                                                6.6
                                                                     357.0
73
                      637.0
                                                7.5
                                                                     324.0
132
                       36.0
                                                                      27.0
                                                6.1
155
                     1224.0
                                                7.5
                                                                     553.0
81227
                                                                      29.0
                       17.0
                                                1.5
81234
                       10.0
                                                4.7
                                                                      22.0
81242
                       13.0
                                                7.0
                                                                       9.0
81246
                       42.0
                                                4.9
                                                                      58.0
81253
                      339.0
                                                4.4
                                                                      59.0
       us_voters_rating us_voters_votes non_us_voters_rating \
                                                              7.0
3
                     7.1
                                    452.0
                                                              6.5
45
                     6.3
                                   7233.0
73
                     7.7
                                   4150.0
                                                              7.8
132
                     7.3
                                                              5.0
                                     112.0
155
                     8.0
                                   12670.0
                                                              8.1
81227
                     1.9
                                      86.0
                                                              1.7
81234
                     6.4
                                      63.0
                                                              6.4
                                                              7.7
81242
                     7.9
                                     125.0
                                                              4.9
81246
                     5.2
                                    534.0
                                                              5.4
81253
                     5.8
                                    1073.0
       non_us_voters_votes
3
                     1076.0
45
                     7837.0
73
                     5657.0
132
                      347.0
155
                    23911.0
81227
                      385.0
81234
                      354.0
```

	81242	1135.0
	81246	459.0
	81253	3843.0
	[9948 rows x	x 49 columns]>
[0]:		