TMDB Analysis

January 18, 2018

```
In [1]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns

        %matplotlib inline

        data = 'data/tmdb-movies.csv'

        imported_data = pd.read_csv(data)
```

```
In [2]: imported_data.shape
```

```
Out[2]: (10866, 21)
```

1 Introduction

1.1 Description of Dataset

I selected The Movie Database dataset from Kaggle for my project. The dataset was created after IMDB requested a takedown of their data from Kaggle. As a consequence, an open-source alternative was selected with the caveat that there were some open questions about the data: * The currency of the budget amounts is not known. * The revenue might not consistently show global revenues. * The dataset, as a whole, has not undergone quality auditing. * There are 0s for numerous budget & revenue records. It is recommended that 0s are treated as missing values for these records. * Though not mentioned, because The Movie Database is a publicly curated dataset, it is subject to similar hesitations that other public, user-written databases are. The database is not necessarily peer-reviewed or sourced for accuracy.

With these caveats in mind, I decided to review the dataset to look at the general character of the data, as well as perform some experimental cleaning functions. The first question of interest to me is the scope of the movies in the dataset; what is the timespan of movies included, and how many are there?

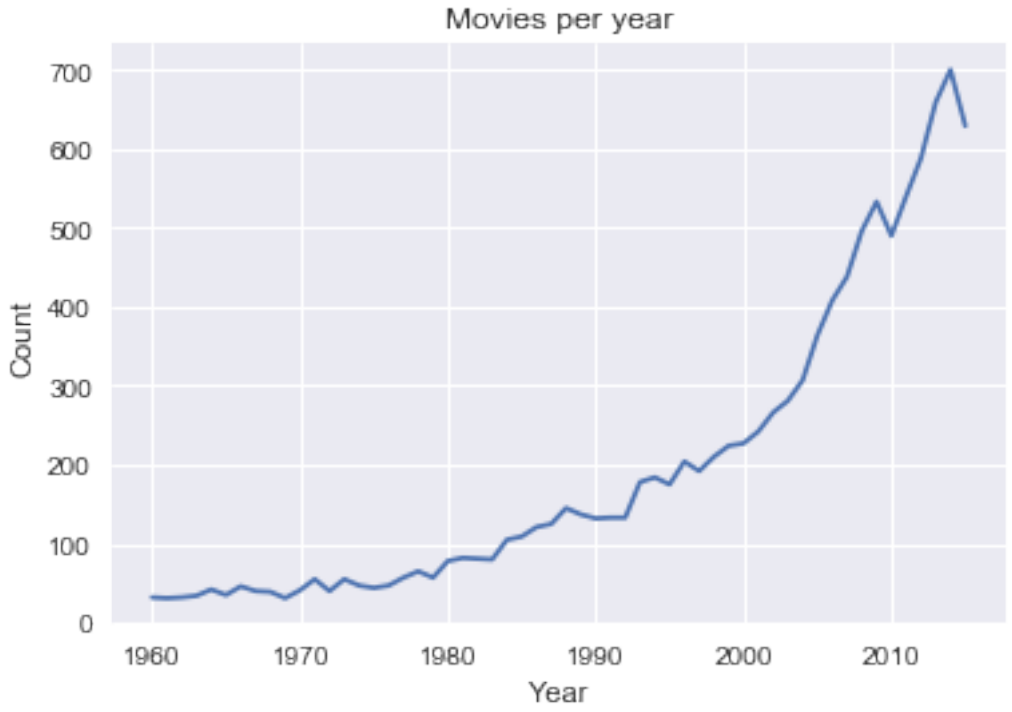
```
In [3]: imported_data['release_year'].agg(['min', 'max'])
```

```
Out[3]: min    1960
        max    2015
        Name: release_year, dtype: int64
```

```
In [4]: counts_by_year = imported_data.groupby('release_year')['id'].count()

        plt.plot(counts_by_year)
        plt.title('Movies per year')
        plt.xlabel('Year')
        plt.ylabel('Count')

Out[4]: <matplotlib.text.Text at 0x2242e5ed080>
```



We can see that the first movie included in the database is from 1960, and the latest is 2015; movies included seems to grow year-over-year on a seemingly logarithmic scale. One might suppose this is because the more recent a movie is, the more likely a user is contribute meaningfully to information about it. Similarly, it could be the case that more movies are actually produced as time goes on. Meaningfully answering these questions would require juxtaposing the dataset against other data – performing the quality assessment precluded in the Kaggle description – as well as more rigorous statistical testing than a line chart.

Interestingly, there are a few noticeable dips in movies per year. An immediate one of note is in 2010, possibly a consequence of a notable writer's strike. The strike occurred from late 2007 and into early 2008, and might account for the dip in movies.

```
In [5]: imported_data.query('release_year >= 2006 and release_year <= 2012').groupby('release_ye

Out[5]: release_year
        2006    408
        2007    438
```

```
2008    496
2009    533
2010    490
2011    540
2012    588
Name: id, dtype: int64
```

More information would be helpful to contextualize this fact of the data. Notably, it might be helpful to know the standard production time of a movie – to know when a movie being written, then optioned, is released theatrically! I'm sure there are many confounding variables beyond the scope of this dataset and investigation, but nonetheless it is an interesting point to speculate on.

1.2 Wrangling and Cleaning

One of my goals for this dataset is understanding what exactly is included, and experimenting with teasing out interesting questions that it might contribute to answering. To that end, I'll be looking at some of the standard techniques for assessing data: * Looking at the first and last 5 rows * getting a sense of how many values are missing * dropping some attributes that might not help with an analysis, either because they are poorly formatted, missing data, or require advanced techniques such as sentiment analysis.

```
In [6]: imported_data.head()
```

Out[6]:

	id	imdb_id	popularity	budget	revenue	original_title	cast	homepage	director
0	135397	tt0369610	32.985763	150000000	1513528810	Jurassic World	Chris Pratt|Bryce Dallas Howard|Irrfan Khan|Vi...	http://www.jurassicworld.com/	Colin Trevorrow
1	76341	tt1392190	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy|Charlize Theron|Hugh Keays-Byrne|Nic...	http://www.madmaxmovie.com/	George Miller
2	262500	tt2908446	13.112507	110000000	295238201	Insurgent	Shailene Woodley|Theo James|Kate Winslet|Ansel...	http://www.thedivergentseries.movie/#insurgent	Robert Schwentke
3	140607	tt2488496	11.173104	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford|Mark Hamill|Carrie Fisher|Adam D...	http://www.starwars.com/films/star-wars-episod...	J.J. Abrams
4	168259	tt2820852	9.335014	190000000	1506249360	Furious 7	Vin Diesel|Paul Walker|Jason Statham|Michelle ...		

```
4                          http://www.furious7.com/         James Wan

                         tagline      ...       \
0             The park is open.       ...
1            What a Lovely Day.       ...
2    One Choice Can Destroy You       ...
3  Every generation has a story.      ...
4            Vengeance Hits Home      ...

                                              overview runtime  \
0  Twenty-two years after the events of Jurassic ...     124
1  An apocalyptic story set in the furthest reach...     120
2  Beatrice Prior must confront her inner demons ...     119
3  Thirty years after defeating the Galactic Empi...     136
4  Deckard Shaw seeks revenge against Dominic Tor...     137

                                       genres  \
0  Action|Adventure|Science Fiction|Thriller
1  Action|Adventure|Science Fiction|Thriller
2         Adventure|Science Fiction|Thriller
3   Action|Adventure|Science Fiction|Fantasy
4                     Action|Crime|Thriller

                                production_companies release_date vote_count  \
0  Universal Studios|Amblin Entertainment|Legenda...       6/9/15       5562
1  Village Roadshow Pictures|Kennedy Miller Produ...      5/13/15       6185
2  Summit Entertainment|Mandeville Films|Red Wago...      3/18/15       2480
3          Lucasfilm|Truenorth Productions|Bad Robot     12/15/15       5292
4  Universal Pictures|Original Film|Media Rights ...       4/1/15       2947

   vote_average  release_year    budget_adj   revenue_adj
0           6.5          2015  1.379999e+08  1.392446e+09
1           7.1          2015  1.379999e+08  3.481613e+08
2           6.3          2015  1.012000e+08  2.716190e+08
3           7.5          2015  1.839999e+08  1.902723e+09
4           7.3          2015  1.747999e+08  1.385749e+09

[5 rows x 21 columns]
```

In [7]:
```
imported_data.tail()
```

Out[7]:
```
          id    imdb_id  popularity  budget  revenue  \
10861     21  tt0060371    0.080598       0        0
10862  20379  tt0060472    0.065543       0        0
10863  39768  tt0060161    0.065141       0        0
10864  21449  tt0061177    0.064317       0        0
10865  22293  tt0060666    0.035919   19000        0
```

```
               original_title  \
10861        The Endless Summer
10862                Grand Prix
10863       Beregis Avtomobilya
10864    What's Up, Tiger Lily?
10865  Manos: The Hands of Fate

                                                 cast homepage  \
10861  Michael Hynson|Robert August|Lord 'Tally Ho' B...      NaN
10862  James Garner|Eva Marie Saint|Yves Montand|Tosh...      NaN
10863  Innokentiy Smoktunovskiy|Oleg Efremov|Georgi Z...      NaN
10864  Tatsuya Mihashi|Akiko Wakabayashi|Mie Hama|Joh...      NaN
10865  Harold P. Warren|Tom Neyman|John Reynolds|Dian...      NaN

               director                                              tagline  \
10861        Bruce Brown                                                 NaN
10862  John Frankenheimer  Cinerama sweeps YOU into a drama of speed and ...
10863     Eldar Ryazanov                                                 NaN
10864        Woody Allen                          WOODY ALLEN STRIKES BACK!
10865    Harold P. Warren       It's Shocking! It's Beyond Your Imagination!

        ...                                                overview runtime  \
10861     ...       The Endless Summer, by Bruce Brown, is one of ...      95
10862     ...       Grand Prix driver Pete Aron is fired by his te...     176
10863     ...       An insurance agent who moonlights as a carthie...      94
10864     ...       In comic Woody Allen's film debut, he took the...      80
10865     ...       A family gets lost on the road and stumbles up...      74

                    genres  \
10861           Documentary
10862  Action|Adventure|Drama
10863          Mystery|Comedy
10864           Action|Comedy
10865                  Horror

                                  production_companies release_date  \
10861                                   Bruce Brown      6/15/66
10862  Cherokee Productions|Joel Productions|Douglas ...     12/21/66
10863                                        Mosfilm       1/1/66
10864                          Benedict Pictures Corp.      11/2/66
10865                                      Norm-Iris     11/15/66

      vote_count  vote_average  release_year     budget_adj  revenue_adj
10861         11           7.4          1966       0.000000          0.0
10862         20           5.7          1966       0.000000          0.0
10863         11           6.5          1966       0.000000          0.0
10864         22           5.4          1966       0.000000          0.0
10865         15           1.5          1966  127642.279154          0.0
```

```
[5 rows x 21 columns]
```

This dataset seems to be structured as a table; that is, it contains no strange artifacts after the end of the dataset that would complicate an analysis. From a high-level view, it looks like the beginning and end of the data have appropriate values for each field.

```
In [8]: imported_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10866 entries, 0 to 10865
Data columns (total 21 columns):
id                      10866 non-null int64
imdb_id                 10856 non-null object
popularity              10866 non-null float64
budget                  10866 non-null int64
revenue                 10866 non-null int64
original_title          10866 non-null object
cast                    10790 non-null object
homepage                2936 non-null object
director                10822 non-null object
tagline                 8042 non-null object
keywords                9373 non-null object
overview                10862 non-null object
runtime                 10866 non-null int64
genres                  10843 non-null object
production_companies    9836 non-null object
release_date            10866 non-null object
vote_count              10866 non-null int64
vote_average            10866 non-null float64
release_year            10866 non-null int64
budget_adj              10866 non-null float64
revenue_adj             10866 non-null float64
dtypes: float64(4), int64(6), object(11)
memory usage: 1.7+ MB
```

On the other hand, there are several fields that are missing values. Furthermore, I know from the description of the dataset that the revenue and budget attributes ('budget', 'revenue', 'budget_adj', 'revenue_adj') contain 0 values that should be treated as missing rather than actual 0 values.

For the fields missing many values, I want to spot check the kind of information contained to see if it seems important. I particularly will be looking at the following: * 'homepage' * 'tagline' * 'keywords' * 'production_companies'

```
In [9]: imported_data['homepage'].value_counts()
```

```
Out[9]: http://phantasm.com
        http://www.missionimpossible.com/
```

http://www.thehungergames.movie/
http://www.kungfupanda.com/
http://www.transformersmovie.com/
http://www.thehobbit.com/
http://www.americanreunionmovie.com/
http://www.georgecarlin.com
http://www.jeffdunham.com
http://www.khartonline.com/
http://www.theamazingspiderman.com
http://www.harrypotter.com
http://www.howtotrainyourdragon.com/
http://stepupmovie.com/
http://eleanorrigby-movie.com/
http://www.lordoftherings.net/
http://disney.go.com/disneypictures/pirates/
http://www.riomovies.com/
http://www.apocalypsenow.com
http://www.miramax.com/movie/kill-bill-volume-1/
http://www.magpictures.com/nymphomaniac/
http://www.munkyourself.com/
http://disney.go.com/tron/
http://www.ironmanmovie.com/
http://www.zeitgeistmovie.com
http://www.beowulfmovie.com/
http://www.sexandthecitymovie.com/
http://www.indianajones.com
http://www.billandted.org/
http://magnetreleasing.com/survivalofthedead/

http://www.welcometohotelt.com
http://www.warnerbros.com/movies/home-entertainment/scanner-darkly-a/d7c290af-c285-41c4-
http://www.vivarockvegas.com/
http://www.23blast.com
http://movies.disney.com/planes-fire-and-rescue
http://www.repomenarecoming.com/
http://www.jupiterascending.com
http://www.universalpictures.com/bestman/
http://www.sonypictures.com/classics/dogtown/
http://www.thelegomovie.com
http://www.spacestation76.com
http://www.timerthemovie.com
http://www.grownups-movie.com/
http://www.lifetime.com
http://www.ifcfilms.com/viewFilm.htm?filmId=245
http://www.funnypeoplemovie.com
http://www.devilsplaygroundmovie.co.uk/
http://www.foxsearchlight.com/cedarrapids/
http://www.mgm.com/view/movie/529/Diamonds-Are-Forever/

```
http://tomandjerrythemovie.warnerbros.com
http://www.vivamachete.com/
http://theidenticalmovie.com/
http://www.ifcfilms.com/uncategorized/salvation-boulevard
https://www.facebook.com/finderskeepersdocumentary
http://www.graceisgone-themovie.com/
http://www.growththemovie.com/
http://womaninblack.com/
http://www.ifcfilms.com/films/house-of-pleasures
http://www.minionsmovie.com/
http://theluckyonemovie.warnerbros.com/
Name: homepage, Length: 2896, dtype: int64
```

```
In [10]: imported_data['tagline'].value_counts()
```

```
Out[10]: Based on a true story.
         Two Films. One Love.
         Be careful what you wish for.
         There are no clean getaways.
         The chase is on!
         One ordinary couple. One little white lie.
         Who is John Galt?
         There are two sides to every love story.
         Worlds Collide
         The end of the world is just the beginning.
         Survival is no game
         Some things are worth fighting for.
         Forget About Love
         It's A Trap
         Love is a force of nature.
         Fight Fire With Fire
         No one is above the law.
         Some lines should never be crossed.
         Victor Crowley Lives Again
         Misery loves family.
         -
         Some houses are born bad.
         Some things are better left buried.
         There is no turning back
         Who's next?
         Adapt or die.
         How far would you go?
         The legend comes to life.
         The timeless tale of a special place where magic, hope and love grow.
         Free your mind.

         They're Down On Their Luck And Up To Their Necks In Senoritas, Margaritas, Banditos And
         How much can a man take...before he gives back?
```

```
Unlock The Universe
No one is ever really prepared.
Let the sun shine in!
Don't Get Mad. Get Evil.
Fight or die!
From Backpacks to Strollers
A comedy about two brothers, a girl with a broken heart, a sex tape, an angel and a pig
Where everything seems possible and nothing is what it seems.
Sometimes you don't need more than one person to not feel alone
Spend Thanksgiving With Good Ol' Charlie Brown!
He took the job that no one wanted... and got the girl that everyone did.
Unlock The Ultimate Secret This October!
Space will never be the same.
To Life!
There comes a time to cut loose.
Take the ride.
Too Cool For The Rules!
The way back begins with a single chord.
You Stop You Die
A romantic comedy with a whole lot of drama.
New house. New family. What could possibly go wrong?
I don't feel I have to wipe everybody out, Tom. Just my enemies.
How to marry a billionaire
FLINT'S BACK In Action... In Danger... In the Virgin Islands... Where the Bad Guys... A
You'll be sorry you were ever born human
Willy Wonka is semi-sweet and nuts.
Game On.
Real War, Real Heroes
Name: tagline, Length: 7997, dtype: int64
```

```
In [11]: imported_data['keywords'].value_counts()
```

```
Out[11]: woman director                                  134
         independent film                                82
         sport                                           25
         suspense                                        24
         musical                                         24
         duringcreditsstinger                            24
         holiday                                         16
         stand-up|stand up comedy                        16
         biography                                       15
         independent film|woman director                 13
         stand up comedy                                  9
         found footage                                    7
         holiday|christmas                                7
         dystopia                                         7
         christmas                                        7
         based on novel                                   7
```

```
sequel                                                                  6
aftercreditsstinger                                                     6
cop|new england|jesse stone                                             5
crime solving                                                           5
aftercreditsstinger|duringcreditsstinger                                5
stand-up                                                                4
werewolf                                                                4
dinosaur                                                                4
zombies                                                                 4
baseball|sport                                                          4
based on video game                                                     4
possession                                                              4
independent film|duringcreditsstinger                                   4
haunted house                                                           3
                                                                      ...
nudity|boot camp|reality spoof                                          1
film producer|party                                                     1
graduation|ex husband|woman director                                    1
wife husband relationship|space travel|space|mission|pregnancy          1
ice|space marine|paranoia|snow storm|norwegian                          1
sport|skiing                                                            1
journalist|journalism|distrust|audio tape|wound                         1
haunted house|inheritance|fireplace|creature|demon                      1
sex|seduction|gigolo|callboy|party                                      1
fire|recession|diary|time travel|murder                                 1
prostitute|photographer|brothel|virgin|new orleans                      1
dc comics|based on comic book|superhero team|super powers               1
new york|concentration camp|holocaust|writer                            1
based on novel|attack|massacre|private                                  1
based on novel|biography|based on true story                            1
seduction|college|love|friends|betrayal                                 1
holy grail|camelot|round table|wizardry|merlin                          1
robbery|inventor|penguin|telecontrol|surrealism                         1
individual|slum|suicide|tattoo|alcohol                                  1
ventriloquist|doll|ventriloquist dummy                                  1
soulmates|vampire|forbidden love|immortality|trust                      1
rock star|heavy metal|recording studio|psychologist|conflict            1
adventure|farscape|tv mini-series                                       1
comedian|stand-up|stand up comedy|clean comedy                          1
sport|figure skating|olympics                                           1
female nudity|robbery|mail order bride|bank clerk                      1
spy|airport|gas station|garage|pilot                                    1
nurse|patriotism|hawaii|world war ii|pilot                              1
corruption|terrorist|explosive|police|kidnapping                        1
sea|fireworks|prince|kingdom|daughter                                   1
Name: keywords, Length: 8804, dtype: int64
```

```
In [12]: imported_data['production_companies'].value_counts()
```

Out[12]: Paramount Pictures
Universal Pictures
Warner Bros.
Walt Disney Pictures
Metro-Goldwyn-Mayer (MGM)
Columbia Pictures
New Line Cinema
Touchstone Pictures
20th Century Fox
Twentieth Century Fox Film Corporation
TriStar Pictures
Orion Pictures
Miramax Films
Columbia Pictures Corporation
DreamWorks Animation
Pixar Animation Studios
Walt Disney Productions
Dimension Films
United Artists
Marvel Studios
Imagine Entertainment|Universal Pictures
The Asylum
Lions Gate Films
Walt Disney Pictures|Pixar Animation Studios
New World Pictures
American International Pictures (AIP)
Disney Channel
Hammer Film Productions
Hollywood Pictures
Walt Disney Pictures|Walt Disney Animation Studios

Haven Entertainment|Sandia Media|Minerva Productions
Kanzaman S.A.|Scion Films Limited|Millenium Films|Black Forest Films|Double Edge Entert
Yari Film Group|Furst Films
Twentieth Century Fox Film Corporation|SLM Production Group|Silver Pictures
Company Films|Voltage Pictures
New Line Cinema|Goldsmith-Thomas Productions|Red Om Films|HBO Films|Picturehouse
Dune Entertainment|DiNovi Pictures
Paramount Pictures|Twentieth Century Fox Film Corporation|Lightstorm Entertainment
Columbia Pictures Corporation|Stonebridge Entertainment
Das Films|Living Out Loud Films|Elephant Eye Films
Institution, The
Likely Story|ATO Pictures|Olympus Pictures
Legendary Pictures|Green Hat Films|Warner Bros.|IFP Westcoast Erste
Producers Circle|Incorporated Television Company
Fountainbridge Films|Warner Bros.|Lee Rich Productions
Myriad Pictures|CJ Entertainment|Toiion
BBC Films|Aramid Entertainment Fund

```
Lifetime Network|Sony Pictures Television|Woodridge Productions
Raven Banner Entertainment|De Angeles Films
Columbia Pictures Corporation|Tom Ward Enterprises|Rastar Productions
Columbia Pictures|Centropolis Film Productions
Universal Pictures|Saga Film|Focus Films|Gold Circle Films|Chambara Pictures
Columbia Pictures|Imagine Entertainment|Revolution Studios
Screen Gems|Olive Bridge Entertainment
Antzworks
Paramount Pictures|Mutual Film Company|Skydance Productions|TC Productions
Twentieth Century Fox Film Corporation|Lawrence Gordon Productions|Davis Entertainment|
Mad Circus Films|Lions Gate Entertainments|Mr. X
Walt Disney Pictures|Robert Simonds Productions
Universal Pictures|Chernin Entertainment|Relativity Media|Monolith Pictures (III)|Radic
Name: production_companies, Length: 7445, dtype: int64
```

OK, so those values make sense for these attributes, but they don't really seem helpful to my goals. Maybe the homepages would be interesting if I were interested in having seeds for a web crawler, or the taglines to run through a natural language analysis – maybe something to correlate sentiment to viewership. The keywords are too sporadic and unstructured to be particularly helpful, I would think.

I'm going to drop these attributes from my dataframe. It'd probably be OK to leave them there and ignore then, but why keep them in memory if I don't have to?

```
In [13]: imported_data.drop(['homepage', 'tagline', 'keywords', 'production_companies'], axis=1,
         imported_data.drop_duplicates(inplace=True)

In [14]: imported_data.shape

Out[14]: (10865, 17)
```

Awesome! Now, another thing I noticed about is that the release year isn't exactly helpful, so I want to change it to a datetime object in case I end up needing to do anything with the date.

```
In [15]: from datetime import datetime

         def change_to_date(date_string):
             return datetime.strptime(date_string, "%m/%d/%y")

         imported_data['release_date'] = imported_data['release_date'].apply(change_to_date)

In [16]: imported_data

Out[16]:            id    imdb_id  popularity     budget     revenue  \
         0      135397  tt0369610   32.985763  150000000  1513528810
         1       76341  tt1392190   28.419936  150000000   378436354
         2      262500  tt2908446   13.112507  110000000   295238201
         3      140607  tt2488496   11.173104  200000000  2068178225
         4      168259  tt2820852    9.335014  190000000  1506249360
         5      281957  tt1663202    9.110700  135000000   532950503
```

6	87101	tt1340138	8.654359	155000000	440603537
7	286217	tt3659388	7.667400	108000000	595380321
8	211672	tt2293640	7.404165	74000000	1156730962
9	150540	tt2096673	6.326804	175000000	853708609
10	206647	tt2379713	6.200282	245000000	880674609
11	76757	tt1617661	6.189369	176000003	183987723
12	264660	tt0470752	6.118847	15000000	36869414
13	257344	tt2120120	5.984995	88000000	243637091
14	99861	tt2395427	5.944927	280000000	1405035767
15	273248	tt3460252	5.898400	44000000	155760117
16	260346	tt2446042	5.749758	48000000	325771424
17	102899	tt0478970	5.573184	130000000	518602163
18	150689	tt1661199	5.556818	95000000	542351353
19	131634	tt1951266	5.476958	160000000	650523427
20	158852	tt1964418	5.462138	190000000	209035668
21	307081	tt1798684	5.337064	30000000	91709827
22	254128	tt2126355	4.907832	110000000	470490832
23	216015	tt2322441	4.710402	40000000	569651467
24	318846	tt1596363	4.648046	28000000	133346506
25	177677	tt2381249	4.566713	150000000	682330139
26	214756	tt2637276	4.564549	68000000	215863606
27	207703	tt2802144	4.503789	81000000	403802136
28	314365	tt1895587	4.062293	20000000	88346473
29	294254	tt4046784	3.968891	61000000	311256926
...	...	...	...	...	...
10836	38720	tt0061170	0.239435	0	0
10837	19728	tt0060177	0.291704	0	0
10838	22383	tt0060862	0.151845	0	0
10839	13353	tt0060550	0.276133	0	0
10840	34388	tt0060437	0.102530	0	0
10841	42701	tt0062262	0.264925	75000	0
10842	36540	tt0061199	0.253437	0	0
10843	29710	tt0060588	0.252399	0	0
10844	23728	tt0059557	0.236098	0	0
10845	5065	tt0059014	0.230873	0	0
10846	17102	tt0059127	0.212716	0	0
10847	28763	tt0060548	0.034555	0	0
10848	2161	tt0060397	0.207257	5115000	12000000
10849	28270	tt0060445	0.206537	0	0
10850	26268	tt0060490	0.202473	0	0
10851	15347	tt0060182	0.342791	0	0
10852	37301	tt0060165	0.227220	0	0
10853	15598	tt0060086	0.163592	0	0
10854	31602	tt0060232	0.146402	0	0
10855	13343	tt0059221	0.141026	700000	0
10856	20277	tt0061135	0.140934	0	0
10857	5921	tt0060748	0.131378	0	0
10858	31918	tt0060921	0.317824	0	0

```
10859   20620  tt0060955    0.089072          0           0
10860    5060  tt0060214    0.087034          0           0
10861      21  tt0060371    0.080598          0           0
10862   20379  tt0060472    0.065543          0           0
10863   39768  tt0060161    0.065141          0           0
10864   21449  tt0061177    0.064317          0           0
10865   22293  tt0060666    0.035919      19000           0

                                       original_title  \
0                                      Jurassic World
1                                  Mad Max: Fury Road
2                                           Insurgent
3                        Star Wars: The Force Awakens
4                                           Furious 7
5                                         The Revenant
6                                  Terminator Genisys
7                                        The Martian
8                                            Minions
9                                          Inside Out
10                                           Spectre
11                                   Jupiter Ascending
12                                          Ex Machina
13                                              Pixels
14                           Avengers: Age of Ultron
15                                 The Hateful Eight
16                                          Taken 3
17                                           Ant-Man
18                                         Cinderella
19                The Hunger Games: Mockingjay - Part 2
20                                       Tomorrowland
21                                          Southpaw
22                                        San Andreas
23                               Fifty Shades of Grey
24                                       The Big Short
25                   Mission: Impossible - Rogue Nation
26                                             Ted 2
27                         Kingsman: The Secret Service
28                                           Spotlight
29                        Maze Runner: The Scorch Trials
...                                                ...
10836                                 Walk Don't Run
10837                                   The Blue Max
10838                               The Professionals
10839            It's the Great Pumpkin, Charlie Brown
10840                               Funeral in Berlin
10841                                    The Shooting
10842               Winnie the Pooh and the Honey Tree
10843                                        Khartoum
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10844                                    Our Man Flint
10845                                  Carry On Cowboy
10846                        Dracula: Prince of Darkness
10847                                  Island of Terror
10848                                  Fantastic Voyage
10849                                           Gambit
10850                                           Harper
10851                                       Born Free
10852                    A Big Hand for the Little Lady
10853                                            Alfie
10854                                       The Chase
10855                             The Ghost & Mr. Chicken
10856                                  The Ugly Dachshund
10857                                       Nevada Smith
10858  The Russians Are Coming, The Russians Are Coming
10859                                         Seconds
10860                                Carry On Screaming!
10861                                 The Endless Summer
10862                                         Grand Prix
10863                                Beregis Avtomobilya
10864                              What's Up, Tiger Lily?
10865                            Manos: The Hands of Fate

                                                    cast  \
0      Chris Pratt|Bryce Dallas Howard|Irrfan Khan|Vi...
1      Tom Hardy|Charlize Theron|Hugh Keays-Byrne|Nic...
2      Shailene Woodley|Theo James|Kate Winslet|Ansel...
3      Harrison Ford|Mark Hamill|Carrie Fisher|Adam D...
4      Vin Diesel|Paul Walker|Jason Statham|Michelle ...
5      Leonardo DiCaprio|Tom Hardy|Will Poulter|Domhn...
6      Arnold Schwarzenegger|Jason Clarke|Emilia Clar...
7      Matt Damon|Jessica Chastain|Kristen Wiig|Jeff ...
8      Sandra Bullock|Jon Hamm|Michael Keaton|Allison...
9      Amy Poehler|Phyllis Smith|Richard Kind|Bill Ha...
10     Daniel Craig|Christoph Waltz|LÃľa Seydoux|Ralp...
11     Mila Kunis|Channing Tatum|Sean Bean|Eddie Redm...
12     Domhnall Gleeson|Alicia Vikander|Oscar Isaac|S...
13     Adam Sandler|Michelle Monaghan|Peter Dinklage|...
14     Robert Downey Jr.|Chris Hemsworth|Mark Ruffalo...
15     Samuel L. Jackson|Kurt Russell|Jennifer Jason ...
16     Liam Neeson|Forest Whitaker|Maggie Grace|Famke...
17     Paul Rudd|Michael Douglas|Evangeline Lilly|Cor...
18     Lily James|Cate Blanchett|Richard Madden|Helen...
19     Jennifer Lawrence|Josh Hutcherson|Liam Hemswor...
20     Britt Robertson|George Clooney|Raffey Cassidy|...
21     Jake Gyllenhaal|Rachel McAdams|Forest Whitaker...
22     Dwayne Johnson|Alexandra Daddario|Carla Gugino...
23     Dakota Johnson|Jamie Dornan|Jennifer Ehle|Eloi...
```

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24     Christian Bale|Steve Carell|Ryan Gosling|Brad ...
25     Tom Cruise|Jeremy Renner|Simon Pegg|Rebecca Fe...
26     Mark Wahlberg|Seth MacFarlane|Amanda Seyfried|...
27     Taron Egerton|Colin Firth|Samuel L. Jackson|Mi...
28     Mark Ruffalo|Michael Keaton|Rachel McAdams|Lie...
29     Dylan O'Brien|Kaya Scodelario|Thomas Brodie-Sa...
...                                                  ...
10836  Cary Grant|Samantha Eggar|Jim Hutton|John Stan...
10837  George Peppard|James Mason|Ursula Andress|Jere...
10838  Burt Lancaster|Lee Marvin|Robert Ryan|Woody St...
10839  Christopher Shea|Sally Dryer|Kathy Steinberg|A...
10840  Michael Caine|Paul Hubschmid|Oskar Homolka|Eva...
10841  Will Hutchins|Millie Perkins|Jack Nicholson|Wa...
10842  Sterling Holloway|Junius Matthews|Sebastian Ca...
10843  Charlton Heston|Laurence Olivier|Richard Johns...
10844  James Coburn|Lee J. Cobb|Gila Golan|Edward Mul...
10845  Sid James|Jim Dale|Angela Douglas|Kenneth Will...
10846  Christopher Lee|Barbara Shelley|Andrew Keir|Fr...
10847  Peter Cushing|Edward Judd|Carole Gray|Eddie By...
10848  Stephen Boyd|Raquel Welch|Edmond O'Brien|Donal...
10849  Michael Caine|Shirley MacLaine|Herbert Lom|Joh...
10850  Paul Newman|Lauren Bacall|Julie Harris|Arthur ...
10851  Virginia McKenna|Bill Travers|Geoffrey Keen|Pe...
10852  Henry Fonda|Joanne Woodward|Jason Robards|Paul...
10853  Michael Caine|Shelley Winters|Millicent Martin...
10854  Marlon Brando|Jane Fonda|Robert Redford|E.G. M...
10855  Don Knotts|Joan Staley|Liam Redmond|Dick Sarge...
10856  Dean Jones|Suzanne Pleshette|Charles Ruggles|K...
10857  Steve McQueen|Karl Malden|Brian Keith|Arthur K...
10858  Carl Reiner|Eva Marie Saint|Alan Arkin|Brian K...
10859  Rock Hudson|Salome Jens|John Randolph|Will Gee...
10860  Kenneth Williams|Jim Dale|Harry H. Corbett|Joa...
10861  Michael Hynson|Robert August|Lord 'Tally Ho' B...
10862  James Garner|Eva Marie Saint|Yves Montand|Tosh...
10863  Innokentiy Smoktunovskiy|Oleg Efremov|Georgi Z...
10864  Tatsuya Mihashi|Akiko Wakabayashi|Mie Hama|Joh...
10865  Harold P. Warren|Tom Neyman|John Reynolds|Dian...

                          director  \
0                    Colin Trevorrow
1                      George Miller
2                   Robert Schwentke
3                        J.J. Abrams
4                          James Wan
5      Alejandro GonzÃąlez IÃśÃąrritu
6                         Alan Taylor
7                        Ridley Scott
8              Kyle Balda|Pierre Coffin
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9	Pete Docter
10	Sam Mendes
11	Lana Wachowski|Lilly Wachowski
12	Alex Garland
13	Chris Columbus
14	Joss Whedon
15	Quentin Tarantino
16	Olivier Megaton
17	Peyton Reed
18	Kenneth Branagh
19	Francis Lawrence
20	Brad Bird
21	Antoine Fuqua
22	Brad Peyton
23	Sam Taylor-Johnson
24	Adam McKay
25	Christopher McQuarrie
26	Seth MacFarlane
27	Matthew Vaughn
28	Tom McCarthy
29	Wes Ball
...	...
10836	Charles Walters
10837	John Guillermin
10838	Richard Brooks
10839	Bill Melendez
10840	Guy Hamilton
10841	Monte Hellman
10842	Wolfgang Reitherman
10843	Basil Dearden|Eliot Elisofon
10844	Daniel Mann
10845	Gerald Thomas
10846	Terence Fisher
10847	Terence Fisher
10848	Richard Fleischer
10849	Ronald Neame
10850	Jack Smight
10851	James Hill
10852	Fielder Cook
10853	Lewis Gilbert
10854	Arthur Penn
10855	Alan Rafkin
10856	Norman Tokar
10857	Henry Hathaway
10858	Norman Jewison
10859	John Frankenheimer
10860	Gerald Thomas
10861	Bruce Brown

```
10862              John Frankenheimer
10863                  Eldar Ryazanov
10864                     Woody Allen
10865                Harold P. Warren

                                                 overview  runtime  \
0      Twenty-two years after the events of Jurassic ...      124
1      An apocalyptic story set in the furthest reach...      120
2      Beatrice Prior must confront her inner demons ...      119
3      Thirty years after defeating the Galactic Empi...      136
4      Deckard Shaw seeks revenge against Dominic Tor...      137
5      In the 1820s, a frontiersman, Hugh Glass, sets...      156
6      The year is 2029. John Connor, leader of the r...      125
7      During a manned mission to Mars, Astronaut Mar...      141
8      Minions Stuart, Kevin and Bob are recruited by...       91
9      Growing up can be a bumpy road, and it's no ex...       94
10     A cryptic message from Bondâs past sends him...      148
11     In a universe where human genetic material is ...      124
12     Caleb, a 26 year old coder at the world's larg...      108
13     Video game experts are recruited by the milita...      105
14     When Tony Stark tries to jumpstart a dormant p...      141
15     Bounty hunters seek shelter from a raging bliz...      167
16     Ex-government operative Bryan Mills finds his ...      109
17     Armed with the astonishing ability to shrink i...      115
18     When her father unexpectedly passes away, youn...      112
19     With the nation of Panem in a full scale war, ...      136
20     Bound by a shared destiny, a bright, optimisti...      130
21     Billy "The Great" Hope, the reigning junior mi...      123
22     In the aftermath of a massive earthquake in Ca...      114
23     When college senior Anastasia Steele steps in ...      125
24     The men who made millions from a global econom...      130
25     Ethan and team take on their most impossible m...      131
26     Newlywed couple Ted and Tami-Lynn want to have...      115
27     The story of a super-secret spy organization t...      130
28     The true story of how The Boston Globe uncover...      128
29     Thomas and his fellow Gladers face their great...      132
...                                                  ...      ...
10836  British industrialist Sir William Rutland - "B...      114
10837  A young pilot in the German air force of 1918,...      156
10838  The Professionals is a 1966 American Western f...      117
10839  This classic "Peanuts" tale focuses on the thu...       25
10840  Colonel Stok, a Soviet intelligence officer re...      102
10841  A hired gun seeks to enact revenge on a group ...       82
10842  Christopher Robin's bear attempts to raid a be...       25
10843  English General Charles George Gordon, a devou...      134
10844  When scientists use eco-terrorism to impose th...      108
10845  Stodge City is in the grip of the Rumpo Kid an...       93
10846  Whilst vacationing in the Carpathian Mountain,...       90
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10847  A small island community is overrun with creep...       89
10848  The science of miniaturization has been unlock...      100
10849  Harry Dean (Michael Caine) has a perfect plan ...      109
10850  Harper is a cynical private eye in the best tr...      121
10851  Born Free (1966) is an Open Road Films Ltd./Co...       95
10852  A naive traveler in Laredo gets involved in a ...       95
10853  The film tells the story of a young man who le...      114
10854  Most everyone in town thinks that Sheriff Cald...      135
10855  Luther Heggs aspires to being a reporter for h...       90
10856  The Garrisons (Dean Jones and Suzanne Pleshett...       93
10857  Nevada Smith is the young son of an Indian mot...      128
10858  Without hostile intent, a Soviet sub runs agro...      126
10859  A secret organisation offers wealthy people a ...      100
10860  The sinister Dr Watt has an evil scheme going...       87
10861  The Endless Summer, by Bruce Brown, is one of ...       95
10862  Grand Prix driver Pete Aron is fired by his te...      176
10863  An insurance agent who moonlights as a carthie...       94
10864  In comic Woody Allen's film debut, he took the...       80
10865  A family gets lost on the road and stumbles up...       74

                                              genres release_date  \
0            Action|Adventure|Science Fiction|Thriller   2015-06-09
1            Action|Adventure|Science Fiction|Thriller   2015-05-13
2                   Adventure|Science Fiction|Thriller   2015-03-18
3             Action|Adventure|Science Fiction|Fantasy   2015-12-15
4                               Action|Crime|Thriller   2015-04-01
5                    Western|Drama|Adventure|Thriller   2015-12-25
6            Science Fiction|Action|Thriller|Adventure   2015-06-23
7                      Drama|Adventure|Science Fiction   2015-09-30
8                    Family|Animation|Adventure|Comedy   2015-06-17
9                               Comedy|Animation|Family   2015-06-09
10                               Action|Adventure|Crime   2015-10-26
11            Science Fiction|Fantasy|Action|Adventure   2015-02-04
12                               Drama|Science Fiction   2015-01-21
13                          Action|Comedy|Science Fiction   2015-07-16
14                       Action|Adventure|Science Fiction   2015-04-22
15                            Crime|Drama|Mystery|Western   2015-12-25
16                                 Crime|Action|Thriller   2015-01-01
17                       Science Fiction|Action|Adventure   2015-07-14
18                           Romance|Fantasy|Family|Drama   2015-03-12
19                          War|Adventure|Science Fiction   2015-11-18
20     Action|Family|Science Fiction|Adventure|Mystery   2015-05-19
21                                        Action|Drama   2015-06-15
22                               Action|Drama|Thriller   2015-05-27
23                                       Drama|Romance   2015-02-11
24                                        Comedy|Drama   2015-12-11
25                                              Action   2015-07-23
26                                              Comedy   2015-06-25
```

27	Crime|Comedy|Action|Adventure	2015-01-24
28	Drama|Thriller|History	2015-11-06
29	Action|Science Fiction|Thriller	2015-09-09
...	...	...
10836	Comedy|Romance	2066-01-01
10837	War|Action|Adventure|Drama	2066-06-21
10838	Action|Adventure|Western	2066-11-01
10839	Family|Animation	2066-10-27
10840	Thriller	2066-12-22
10841	Western	2066-10-23
10842	Animation|Family	2066-01-01
10843	Adventure|Drama|War|History|Action	2066-06-09
10844	Adventure|Comedy|Fantasy|Science Fiction	2066-01-16
10845	Comedy|Western	2066-03-01
10846	Horror	2066-01-09
10847	Science Fiction|Horror	2066-06-20
10848	Adventure|Science Fiction	2066-08-24
10849	Action|Comedy|Crime	2066-12-16
10850	Action|Drama|Thriller|Crime|Mystery	2066-02-23
10851	Adventure|Drama|Action|Family|Foreign	2066-06-22
10852	Western	2066-05-31
10853	Comedy|Drama|Romance	2066-03-29
10854	Thriller|Drama|Crime	2066-02-17
10855	Comedy|Family|Mystery|Romance	2066-01-20
10856	Comedy|Drama|Family	2066-02-16
10857	Action|Western	2066-06-10
10858	Comedy|War	2066-05-25
10859	Mystery|Science Fiction|Thriller|Drama	2066-10-05
10860	Comedy	2066-05-20
10861	Documentary	2066-06-15
10862	Action|Adventure|Drama	2066-12-21
10863	Mystery|Comedy	2066-01-01
10864	Action|Comedy	2066-11-02
10865	Horror	2066-11-15

	vote_count	vote_average	release_year	budget_adj	revenue_adj
0	5562	6.5	2015	1.379999e+08	1.392446e+09
1	6185	7.1	2015	1.379999e+08	3.481613e+08
2	2480	6.3	2015	1.012000e+08	2.716190e+08
3	5292	7.5	2015	1.839999e+08	1.902723e+09
4	2947	7.3	2015	1.747999e+08	1.385749e+09
5	3929	7.2	2015	1.241999e+08	4.903142e+08
6	2598	5.8	2015	1.425999e+08	4.053551e+08
7	4572	7.6	2015	9.935996e+07	5.477497e+08
8	2893	6.5	2015	6.807997e+07	1.064192e+09
9	3935	8.0	2015	1.609999e+08	7.854116e+08
10	3254	6.2	2015	2.253999e+08	8.102203e+08
11	1937	5.2	2015	1.619199e+08	1.692686e+08

12	2854	7.6	2015	1.379999e+07	3.391985e+07
13	1575	5.8	2015	8.095996e+07	2.241460e+08
14	4304	7.4	2015	2.575999e+08	1.292632e+09
15	2389	7.4	2015	4.047998e+07	1.432992e+08
16	1578	6.1	2015	4.415998e+07	2.997096e+08
17	3779	7.0	2015	1.195999e+08	4.771138e+08
18	1495	6.8	2015	8.739996e+07	4.989630e+08
19	2380	6.5	2015	1.471999e+08	5.984813e+08
20	1899	6.2	2015	1.747999e+08	1.923127e+08
21	1386	7.3	2015	2.759999e+07	8.437300e+07
22	2060	6.1	2015	1.012000e+08	4.328514e+08
23	1865	5.3	2015	3.679998e+07	5.240791e+08
24	1545	7.3	2015	2.575999e+07	1.226787e+08
25	2349	7.1	2015	1.379999e+08	6.277435e+08
26	1666	6.3	2015	6.255997e+07	1.985944e+08
27	3833	7.6	2015	7.451997e+07	3.714978e+08
28	1559	7.8	2015	1.839999e+07	8.127872e+07
29	1849	6.4	2015	5.611998e+07	2.863562e+08
...	...	...	...	...	...
10836	11	5.8	1966	0.000000e+00	0.000000e+00
10837	12	5.5	1966	0.000000e+00	0.000000e+00
10838	21	6.0	1966	0.000000e+00	0.000000e+00
10839	49	7.2	1966	0.000000e+00	0.000000e+00
10840	13	5.7	1966	0.000000e+00	0.000000e+00
10841	12	5.5	1966	5.038511e+05	0.000000e+00
10842	12	7.9	1966	0.000000e+00	0.000000e+00
10843	12	5.8	1966	0.000000e+00	0.000000e+00
10844	13	5.6	1966	0.000000e+00	0.000000e+00
10845	15	5.9	1966	0.000000e+00	0.000000e+00
10846	16	5.7	1966	0.000000e+00	0.000000e+00
10847	13	5.3	1966	0.000000e+00	0.000000e+00
10848	42	6.7	1966	3.436265e+07	8.061618e+07
10849	14	6.1	1966	0.000000e+00	0.000000e+00
10850	14	6.0	1966	0.000000e+00	0.000000e+00
10851	15	6.6	1966	0.000000e+00	0.000000e+00
10852	11	6.0	1966	0.000000e+00	0.000000e+00
10853	26	6.2	1966	0.000000e+00	0.000000e+00
10854	17	6.0	1966	0.000000e+00	0.000000e+00
10855	14	6.1	1966	4.702610e+06	0.000000e+00
10856	14	5.7	1966	0.000000e+00	0.000000e+00
10857	10	5.9	1966	0.000000e+00	0.000000e+00
10858	11	5.5	1966	0.000000e+00	0.000000e+00
10859	22	6.6	1966	0.000000e+00	0.000000e+00
10860	13	7.0	1966	0.000000e+00	0.000000e+00
10861	11	7.4	1966	0.000000e+00	0.000000e+00
10862	20	5.7	1966	0.000000e+00	0.000000e+00
10863	11	6.5	1966	0.000000e+00	0.000000e+00
10864	22	5.4	1966	0.000000e+00	0.000000e+00

```
10865          15           1.5          1966  1.276423e+05  0.000000e+00

[10865 rows x 17 columns]
```

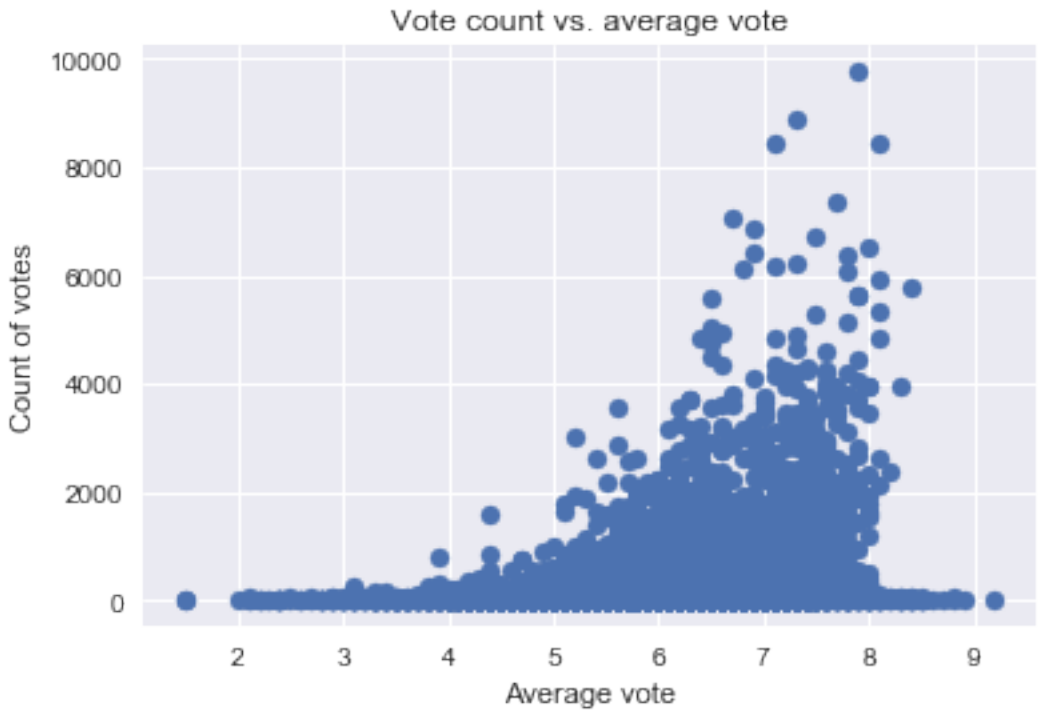
There we go; that seems to be a good start. There are some annoying facts about the data – like that 'cast' and 'genres' are stored as pipe-delimited values. That seems extremely unnecessary, but I'm not the architect, so I'll refrain from further judgment at this time.

Spoiler alert: this does end up causing me a lot of frustration, especially because of how Pandas interacts with strings.

2 Exploring the Data

An easy way to start exploring data is to look at some correlations. We can't make any statistical inferences from scatterplots alone, but we might see that data trends in a particular way. I'll start by looking at the relationship between the number of votes, and the average rating. Maybe we'll see a common adage, "people are more likely to express an opinion for something they dislike."

```
In [17]: vote_counts = imported_data['vote_count']
         vote_averages = imported_data['vote_average']

         plt.scatter(vote_averages, vote_counts)
         plt.title("Vote count vs. average vote")
         plt.xlabel("Average vote")
         plt.ylabel("Count of votes")

Out[17]: <matplotlib.text.Text at 0x2242ea66cf8>
```

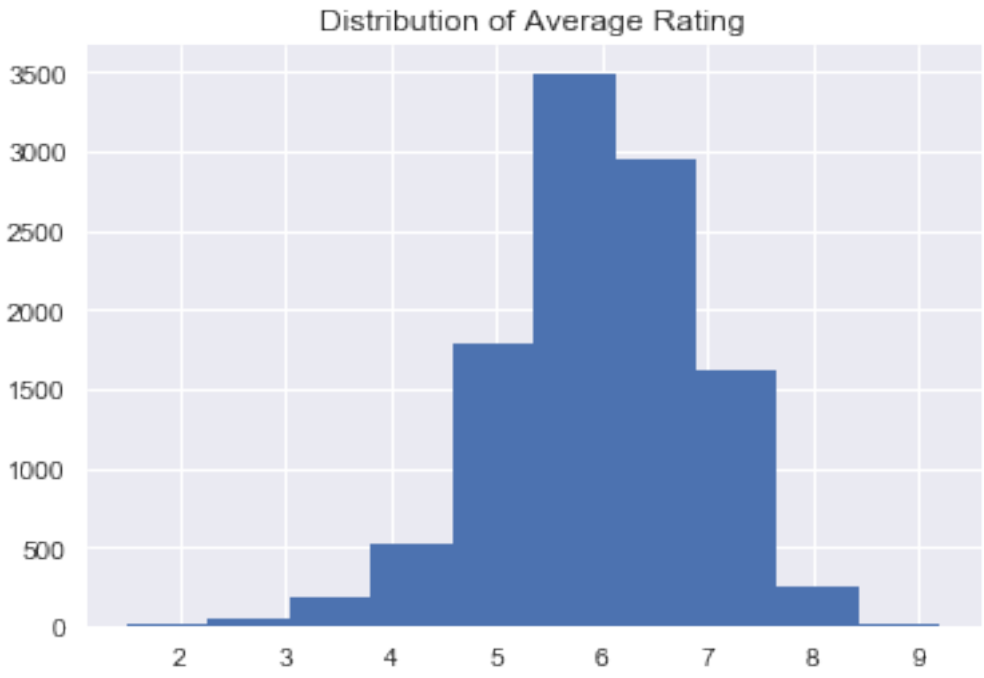


Whoa! There's a lot of records hovering at an extremely low number of votes. You can almost claim a positive correlation between the count and average score, but I'm extremely hesitant to do so since so many entries

```
In [18]: low_votes = len(imported_data.query('vote_count < 100'))
         high_votes = len(imported_data.query('vote_count >= 100'))
         low_votes, high_votes, low_votes/len(imported_data)

Out[18]: (7537, 3328, 0.6936953520478601)

In [19]: len(imported_data.query('vote_count <= 10'))/len(imported_data)

Out[19]: 0.0461113667740451
```

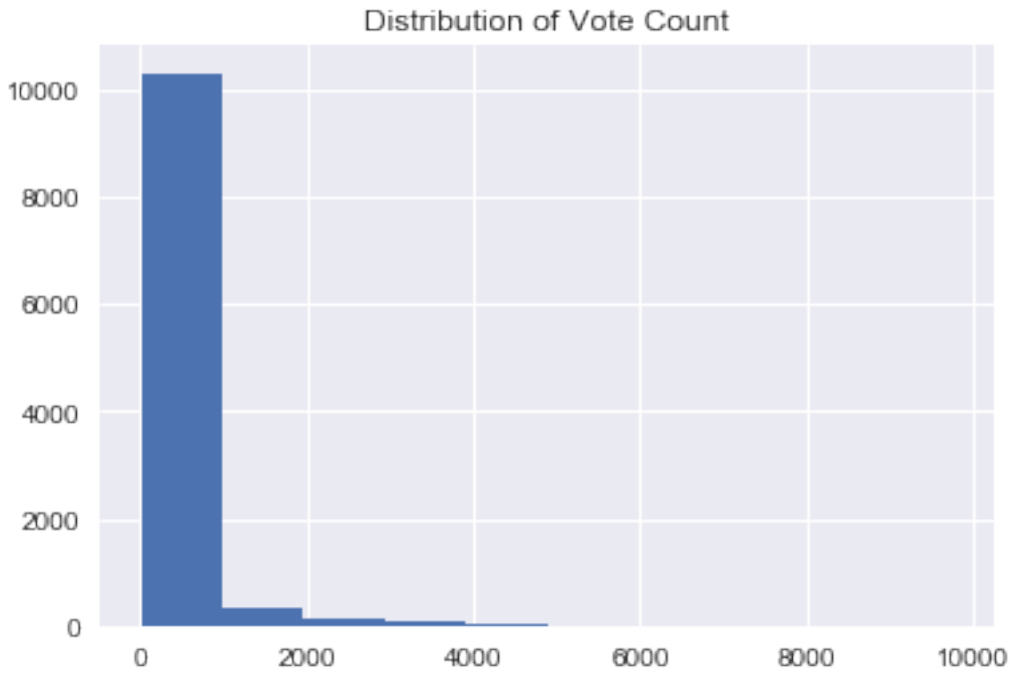
Interesting. Almost 70% of the movies have less than 100 votes. It's hard to define what might even be considered an outlier with such a density of low votes. 5% of the movies have 10 or less votes! I wonder what the distribution of ratings looks like...

```
In [20]: plt.hist(vote_averages)
         plt.title('Distribution of Average Rating')

Out[20]: <matplotlib.text.Text at 0x2242eb28e10>
```



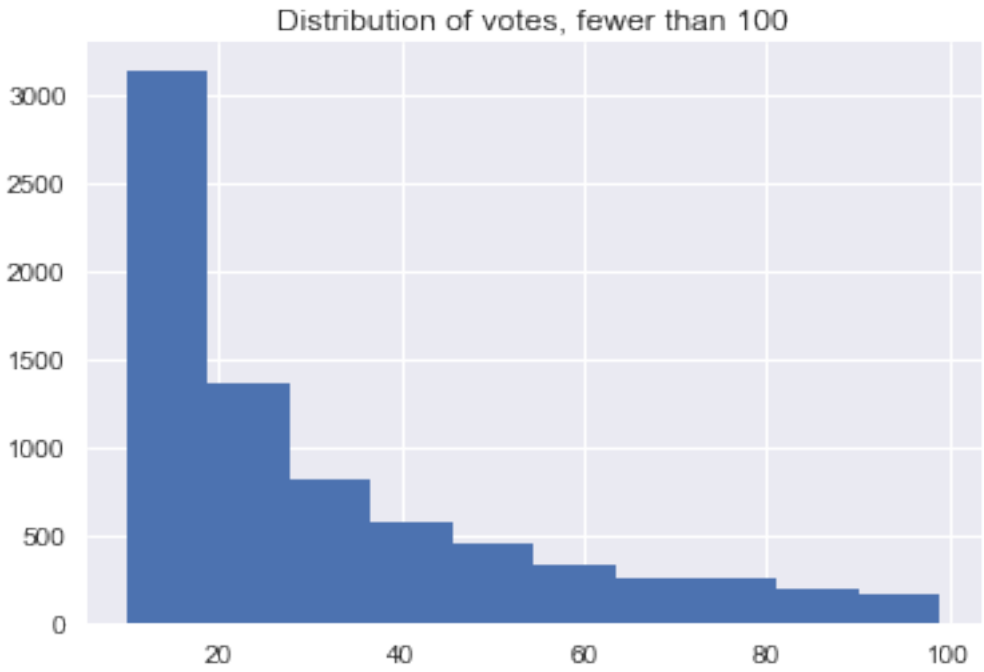
Maybe unsurprisingly, the distribution seems to match the general count of votes. At least the averages are normally distributed.

```
In [21]: plt.hist(vote_counts)
         plt.title('Distribution of Vote Count')

Out[21]: <matplotlib.text.Text at 0x2242ebffd30>
```



The distribution of vote counts, on the other hand, is so extremely right skewed I wonder if it's even usable as a datapoint.

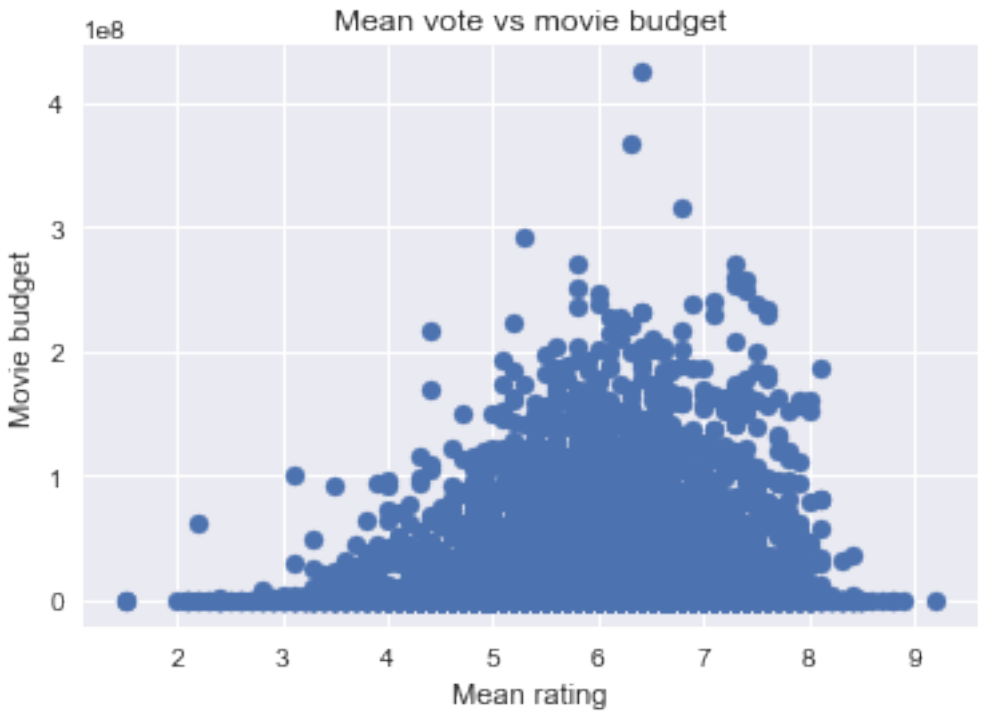
```
In [22]: vote_counts_less_than_100 = imported_data.query('vote_count < 100')['vote_count']

         plt.hist(vote_counts_less_than_100)
         plt.title('Distribution of votes, fewer than 100')

Out[22]: <matplotlib.text.Text at 0x2242ecb4160>
```



We can see that the distribution is the same when only considering records with fewer than 100 votes. We could probably keep drilling down, but I'm not sure that would be a productive exercise.

Instead, let's look at some of the other data points, like whether the budget has any impact on the rating.

```
In [23]: budgets = imported_data['budget_adj']

         plt.scatter(vote_averages, budgets)
         plt.title("Mean vote vs movie budget")
         plt.xlabel("Mean rating")
         plt.ylabel("Movie budget")

Out[23]: <matplotlib.text.Text at 0x2242ed52048>
```



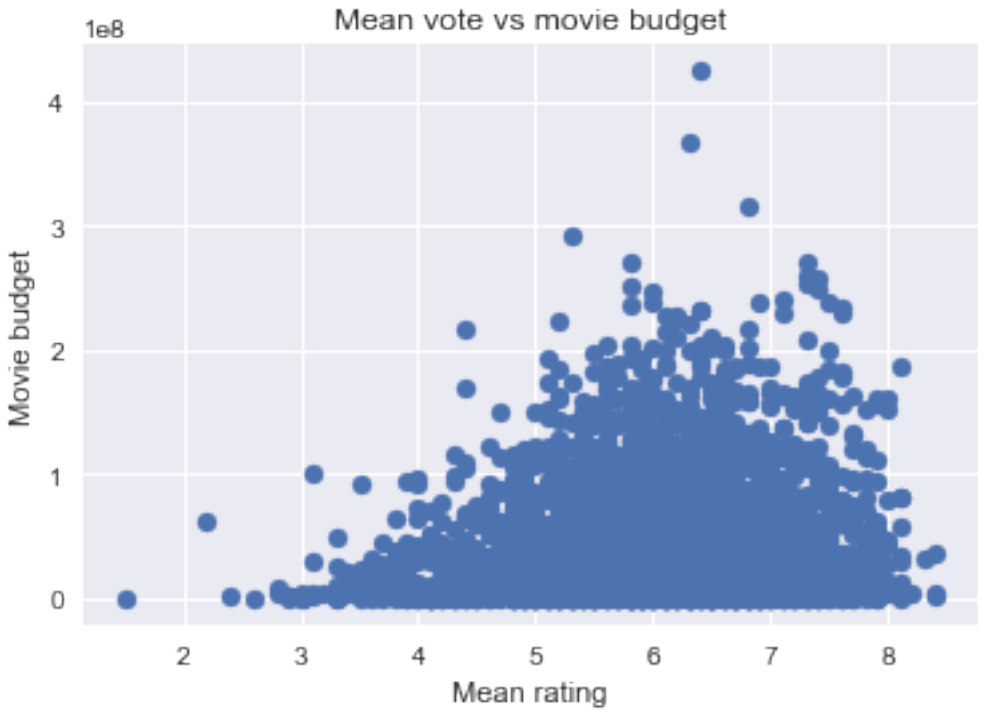
Yikes! We can see that this value is similarly clustered around a low budget.

```
In [24]: imported_data['budget_adj'].agg(['min', 'max'])

Out[24]: min            0.0
         max    425000000.0
         Name: budget_adj, dtype: float64
```

OK, let's remove the 0 values from the budget, which is something we know should be discounted from the description of the data.

```
In [25]: budget_df = imported_data[['budget_adj', 'vote_average']].query('budget_adj != 0')
         budget_df['budget_adj'].agg(['min', 'max'])

Out[25]: min    9.210911e-01
         max    4.250000e+08
         Name: budget_adj, dtype: float64

In [26]: budgets = budget_df['budget_adj']
         vote_averages = budget_df['vote_average']

         plt.scatter(vote_averages, budgets)
         plt.title("Mean vote vs movie budget")
         plt.xlabel("Mean rating")
         plt.ylabel("Movie budget")
```

```
Out[26]: <matplotlib.text.Text at 0x2242edb66a0>
```



Interestingly, the shape of this scatterplot doesn't seem to significantly change. I wonder if these numerical data are useful in the context of an analysis.

Perhaps it's best to switch tactics. Let's take a look at the genres included for each movie, to get a sense of how commonly a label is applied to a movie. I'm sure it's a simple matter to split a series in Pandas to aggregate a count of each value after splitting the value of each cell.

Ha, ha! Just kidding. This exercise was actually one of the more painful things I've ever tried to do with Pandas. I'm not sure the exact reason I was having trouble, but I believe it might be related to one of two implementation details with the library: 1. Pandas stores strings as pointers, and lazily evaluates the items in a series. The library doesn't intelligently apply the split function to the string but rather the pointer value.
2. Pandas lazily evaluates strings and iteritems(), and despite a standard Python pattern to retrieve only the value from the iterator, the library always returns the index value first. When attempting to split the string in the iterator, the split function is applied to the index value.

Of course, neither of these hypotheses could be correct and I'm uncertain how to go about discovering more or reporting a bug (feature?) to the author. For one, this specific technique might be rather atypical in data analysis and an incoherent strategy to extract meaning from a dataset. A differently-delimited field inside a delimited file is definitely an insane way to store data.

After much wrestling with the internal structure and a consideration of reloading the file using an entirely different library, I managed to use a standard Python pattern of counting values in a dictionary by forcing Pandas to access the string value by printing it. It might look sloppy, but at least I will have accessed the data.

In [27]:
```
# NOTE TO EVALUATOR
# Why does this loop work, but if you comment out the print statement it does not?
# The error is: AttributeError: 'float' object has no attribute 'split'

genres = imported_data['genres'].iteritems()
genre_counts = {}

for _, row in genres:
    for v in row.split('|'):
        print(v)
        if v not in genre_counts.keys():
            genre_counts[v] = 1
        else:
            genre_counts[v] += 1
```

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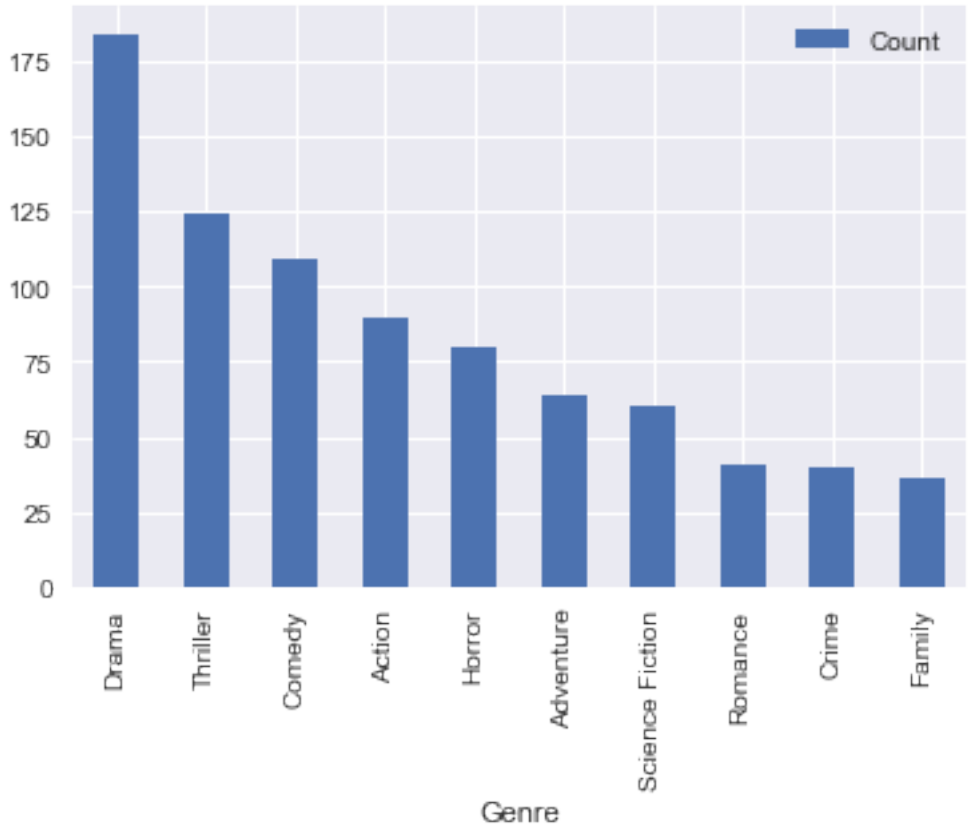
```
AttributeError                            Traceback (most recent call last)

<ipython-input-27-32cb2bc3360b> in <module>()
  7
  8 for _, row in genres:
----> 9     for v in row.split('|'):
 10         print(v)
 11         if v not in genre_counts.keys():


AttributeError: 'float' object has no attribute 'split'
```

Finally, let's make a bar chart to compare the relative frequencies of the genres, now that we've aggregated the totals from the 'genres' series.

```
In [28]: genre_counts_df = pd.DataFrame(list(genre_counts.items()), columns=['Genre', 'Count'])

         genre_counts_df.set_index('Genre').nlargest(10, 'Count').plot.bar()

Out[28]: <matplotlib.axes._subplots.AxesSubplot at 0x2242ebd76d8>
```



Interesting. It looks like the most movies are tagged as dramas. Not to overemphasize my own struggle, but that seems like an appropriate end to my own exploration. Certainly, I felt my struggle with making this simple chart a dramatic struggle worthy of a Greek playwright.