Investigate_a_Dataset

September 16, 2020

1 Project: TMDb Movie Data Analysis

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Introduction

I choose TMDb Movie Data for this project.

There's a description of this dataset from Kaggle:

"What can we say about the success of a movie before it is released? Are there certain companies (Pixar?) that have found a consistent formula? Given that major films costing over \\$100 million to produce can still flop, this question is more important than ever to the industry. Film aficionados might have different interests. Can we predict which films will be highly rated, whether or not they are a commercial success? This is a great place to start digging in to those questions, with data on the plot, cast, crew, budget, and revenues of several thousand films.

The following questions could be answered using these data: 1. Which genres are most popular from year to year?

- 2. How sharp is the divide between major film studios and the independents?
- 3. What kinds of properties are associated with movies that have high revenues?
- 4. Other questions which will arise during Data Cleaning.

Link to data:

 $https://www.google.com/url?q=https://d17h27t6h515a5.cloudfront.net/topher/2017/October/59dd1c4c_tmconvies/tmdb-movies.csv\&sa=D\&ust=1532469042115000$

```
In [1]: # Import statements for all of the packages that will be used.

# + 'magic word' so that visualizations are plotted
# inline with the notebook. See this page for more:
# http://ipython.readthedocs.io/en/stable/interactive/magics.html

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
% matplotlib inline
pd.set_option('display.max_columns', 100)
## Data Wrangling
```

1.1.1 General Properties

This data set contains information about 10,000 movies collected from The Movie Database (TMDb), including user ratings and revenue.

```
In [2]: # Load your data and print out a few lines. Perform operations to inspect data
            types and look for instances of missing or possibly errant data.
        df = pd.read_csv('tmdb-movies.csv', parse_dates=['release_date'])
        # release_date is date type column, so, when load data parse_dates was used to nterpret
        df.head(3)
Out[2]:
               id
                     imdb_id popularity
                                             budget
                                                                      original_title \
                                                        revenue
          135397 tt0369610
                               32.985763
                                          150000000
                                                                      Jurassic World
                                                     1513528810
        0
           76341 tt1392190
                               28.419936
                                          150000000
                                                      378436354 Mad Max: Fury Road
        2 262500 tt2908446
                               13.112507
                                          110000000
                                                       295238201
                                                                           Insurgent
                                                         cast \
           Chris Pratt|Bryce Dallas Howard|Irrfan Khan|Vi...
           Tom Hardy | Charlize Theron | Hugh Keays-Byrne | Nic...
           Shailene Woodley | Theo James | Kate Winslet | Ansel...
                                                 homepage
                                                                    director \
        0
                            http://www.jurassicworld.com/
                                                             Colin Trevorrow
                              http://www.madmaxmovie.com/
                                                              George Miller
        1
          http://www.thedivergentseries.movie/#insurgent Robert Schwentke
                              tagline \
        0
                    The park is open.
        1
                   What a Lovely Day.
        2 One Choice Can Destroy You
                                                    keywords \
          monster|dna|tyrannosaurus rex|velociraptor|island
            future|chase|post-apocalyptic|dystopia|australia
           based on novel|revolution|dystopia|sequel|dyst...
                                                     overview runtime \
          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
```

genres \

```
Action | Adventure | Science Fiction | Thriller
  Action | Adventure | Science Fiction | Thriller
          Adventure | Science Fiction | Thriller
                                 production_companies release_date vote_count \
O Universal Studios | Amblin Entertainment | Legenda...
                                                         2015-06-09
                                                                            5562
  Village Roadshow Pictures | Kennedy Miller Produ...
                                                         2015-05-13
                                                                            6185
   Summit Entertainment | Mandeville Films | Red Wago...
                                                         2015-03-18
                                                                            2480
   vote_average release_year
                                  budget_adj
                                                revenue_adj
0
            6.5
                          2015 1.379999e+08 1.392446e+09
            7.1
                          2015
                                1.379999e+08 3.481613e+08
1
2
            6.3
                          2015 1.012000e+08 2.716190e+08
```

There are 21 columns and 10866 movies in dataset.

```
In [3]: df.shape
Out[3]: (10866, 21)
```

Some column types are correct (like popularity or director, and probably "id" column type isn't correct (should be string).

```
In [4]: df.dtypes
Out[4]: id
                                           int64
        imdb_id
                                          object
        popularity
                                         float64
                                           int64
        budget
        revenue
                                           int64
                                          object
        original_title
                                          object
        cast
        homepage
                                          object
        director
                                          object
        tagline
                                          object
        keywords
                                          object
        overview
                                          object
        runtime
                                           int64
        genres
                                          object
        production_companies
                                          object
        release_date
                                 datetime64[ns]
                                           int64
        vote_count
        vote_average
                                         float64
                                           int64
        release_year
        budget_adj
                                         float64
        revenue_adj
                                         float64
        dtype: object
```

Descriptive statistics of all numeric columns is shown below.

In [5]: df.describe()

Out[5]:		id	popularity	budget	revenue	runtime	\
	count	10866.000000	10866.000000	1.086600e+04	1.086600e+04	10866.000000	
	mean	66064.177434	0.646441	1.462570e+07	3.982332e+07	102.070863	
	std	92130.136561	1.000185	3.091321e+07	1.170035e+08	31.381405	
	min	5.000000	0.000065	0.000000e+00	0.000000e+00	0.000000	
	25%	10596.250000	0.207583	0.000000e+00	0.000000e+00	90.000000	
	50%	20669.000000	0.383856	0.000000e+00	0.000000e+00	99.000000	
	75%	75610.000000	0.713817	1.500000e+07	2.400000e+07	111.000000	
	max	417859.000000	32.985763	4.250000e+08	2.781506e+09	900.000000	
		vote_count	vote_average	release_year	budget_adj	revenue_adj	
	count	10866.000000	10866.000000	10866.000000	1.086600e+04	1.086600e+04	
	mean	217.389748	5.974922	2001.322658	1.755104e+07	5.136436e+07	
	std	575.619058	0.935142	12.812941	3.430616e+07	1.446325e+08	
	min	10.000000	1.500000	1960.000000	0.000000e+00	0.000000e+00	
	25%	17.000000	5.400000	1995.000000	0.000000e+00	0.000000e+00	
	50%	38.000000	6.000000	2006.000000	0.000000e+00	0.000000e+00	
	75%	145.750000	6.600000	2011.000000	2.085325e+07	3.369710e+07	
	max	9767.000000	9.200000	2015.000000	4.250000e+08	2.827124e+09	

Most of columns don't have missing values. And only cast, homepage, director, and other string columns which describe movies have some missings. However, numeric columns "budget", "revenue", "runtime", "budget_adj", and "revenue_adj" have no missing data. So, zeros will be interpreted as missing data.

```
In [6]: df.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
                        10866 non-null float64
popularity
                        10866 non-null int64
budget
revenue
                         10866 non-null int64
                        10866 non-null object
original_title
                        10790 non-null object
cast
homepage
                        2936 non-null object
                        10822 non-null object
director
tagline
                        8042 non-null object
keywords
                        9373 non-null object
overview
                        10862 non-null object
                        10866 non-null int64
runtime
                        10843 non-null object
genres
production_companies
                        9836 non-null object
release_date
                        10866 non-null datetime64[ns]
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: datetime64[ns](1), float64(4), int64(6), object(10)
memory usage: 1.7+ MB
```

1.1.2 Data Cleaning

From Kaggle:

Certain columns, like "cast" and "genres", contain multiple values separated by pipe (|) characters. There are some odd characters in the "cast" column. Don't worry about cleaning them. You can leave them as is.

The final two columns ending with "_adj" _show the budget and revenue of the associated movie in terms of 2010 dollars, accounting for inflation over time.

In the IMDb version it was necessary to treat values of zero in the budget field as missing. ... (It's probably a good idea to keep treating zeros as missing, with the caveat that missing budgets much more likely to have been from small budget films in the first place).

```
In [7]: # Check are column release_year and year in relaese_date differ
        df['different_years'] = df.release_date.dt.year - df.release_year
        print('Columns with different years in relaese_date and release_year columns')
        print(df['different_years'].sum())
        df.sort_values('different_years', ascending=False).head(3)
Columns with different years in relaese_date and release_year columns
36200
Out[7]:
                  id
                        imdb_id popularity
                                              budget
                                                      revenue
        10865 22293 tt0060666
                                   0.035919
                                               19000
                                                            0
        10155 43040 tt0054292
                                   0.269428
                                                   0
                                                            0
        10153 23439 tt0053925
                                   0.323180
                                             270000
                                                            0
                         original_title \
        10865 Manos: The Hands of Fate
                      Sergeant Rutledge
        10155
        10153
                         House of Usher
                                                             cast homepage
        10865 Harold P. Warren | Tom Neyman | John Reynolds | Dian...
                                                                       NaN
               Jeffrey Hunter | Woody Strode | Constance Towers | B...
        10155
                                                                       NaN
               Vincent Price | Mark Damon | Myrna Fahey | Harry Ell...
        10153
                                                                       NaN
                       director
                                                                            tagline \
        10865 Harold P. Warren
                                      It's Shocking! It's Beyond Your Imagination!
        10155
                      John Ford Forget all the suspense you have ever seen! Fo...
```

```
10153
                   Roger Corman Edgar Allan Poe's demonic tale of The Ungodly...
                                           keywords \
               fire|gun|drive|sacrifice|flashlight
        10865
        10155
                                rape | court martial
        10153
                 curse|new england|edgar allan poe
                                                         overview runtime
              A family gets lost on the road and stumbles up...
        10865
                                                                         74
               Respected black cavalry Sergeant Brax Rutledge...
        10155
                                                                        111
               After a long journey, Philip arrives at the Us...
        10153
                                                                         79
                        genres
                                   production_companies release_date
                                              Norm-Iris
        10865
                        Horror
                                                          2066-11-15
                                                                               15
                 Crime | Western
                                  John Ford Productions
                                                          2060-05-18
        10155
                                                                               12
        10153
               Horror|Thriller Alta Vista Productions
                                                          2060-06-21
                                                                               28
                                                                        different_years
               vote_average release_year
                                              budget_adj
                                                          revenue_adj
        10865
                        1.5
                                      1966
                                           1.276423e+05
                                                                   0.0
                                                                                    100
        10155
                        5.2
                                      1960
                                           0.000000e+00
                                                                   0.0
                                                                                    100
                                      1960
                                           1.990701e+06
        10153
                        5.9
                                                                   0.0
                                                                                    100
In [8]: # Create new release_date where year is replaced with release_year
        df['release_date_corrected'] = pd.to_datetime(df.release_year.astype(str) + '-' + df.rel
                                                       + '-' + df.release_date.dt.day.astype(str)
        # Drop useless columns
        df.drop(['release_year', 'release_date', 'different_years'], axis=1, inplace=True)
        # Rename release_date_corrected
        df.rename(columns={'release_date_corrected': 'release_date'}, inplace=True)
        df.head(3)
Out [8]:
               id
                     imdb_id popularity
                                              budget
                                                         revenue
                                                                       original_title \
                                                                       Jurassic World
          135397
                   tt0369610
                               32.985763
                                           150000000
                                                      1513528810
                   tt1392190
                                28.419936
                                           150000000
                                                       378436354 Mad Max: Fury Road
           76341
           262500
                               13.112507
                   tt2908446
                                           110000000
                                                       295238201
                                                                            Insurgent
                                                         cast \
          Chris Pratt Bryce Dallas Howard Irrfan Khan Vi...
           Tom Hardy | Charlize Theron | Hugh Keays-Byrne | Nic...
        2 Shailene Woodley|Theo James|Kate Winslet|Ansel...
                                                  homepage
                                                                     director \
        0
                            http://www.jurassicworld.com/
                                                             Colin Trevorrow
                               http://www.madmaxmovie.com/
                                                               George Miller
        1
         http://www.thedivergentseries.movie/#insurgent
                                                            Robert Schwentke
```

```
0
                    The park is open.
        1
                   What a Lovely Day.
        2 One Choice Can Destroy You
                                                     keywords \
           monster|dna|tyrannosaurus rex|velociraptor|island
            future|chase|post-apocalyptic|dystopia|australia
        2 based on novel|revolution|dystopia|sequel|dyst...
                                                     overview runtime \
           Twenty-two years after the events of Jurassic ...
                                                                   124
          An apocalyptic story set in the furthest reach...
                                                                   120
        2 Beatrice Prior must confront her inner demons ...
                                                                   119
                                              genres
          Action | Adventure | Science Fiction | Thriller
          Action | Adventure | Science Fiction | Thriller
                  Adventure|Science Fiction|Thriller
                                        production_companies vote_count \
        O Universal Studios | Amblin Entertainment | Legenda...
                                                                     5562
        1 Village Roadshow Pictures | Kennedy Miller Produ...
                                                                     6185
        2 Summit Entertainment | Mandeville Films | Red Wago...
                                                                     2480
           vote_average
                           budget_adj
                                        revenue_adj release_date
        0
                    6.5 1.379999e+08
                                      1.392446e+09
                                                       2015-06-09
                    7.1 1.379999e+08 3.481613e+08
                                                       2015-05-13
        1
                    6.3 1.012000e+08 2.716190e+08
                                                       2015-03-18
In [9]: # Check duplicated rows
        df[df.duplicated()]
Out [9]:
                 id
                       imdb_id popularity
                                              budget
                                                      revenue original_title \
        2090 42194 tt0411951
                                   0.59643 30000000
                                                        967000
                                                                       TEKKEN
                                                            cast homepage
              Jon Foo|Kelly Overton|Cary-Hiroyuki Tagawa|Ian...
                      director
                                             tagline
        2090
             Dwight H. Little Survival is no game
                                                        keywords \
        2090 martial arts|dystopia|based on video game|mart...
                                                        overview runtime \
```

tagline \

```
2090 In the year of 2039, after World Wars destroy ...
                                                                      92
                                                   genres
                                                             production_companies \
             Crime|Drama|Action|Thriller|Science Fiction Namco|Light Song Films
        2090
              vote_count vote_average budget_adj revenue_adj release_date
        2090
                                        30000000.0
                                                       967000.0
In [10]: df [df.id == 42194]
Out[10]:
                                                       revenue original_title
                  id
                        imdb_id popularity
                                               budget
                                                        967000
               42194
                     tt0411951
                                    0.59643
                                             3000000
                                                                       TEKKEN
                     tt0411951
                                             30000000
         2090 42194
                                    0.59643
                                                        967000
                                                                       TEKKEN
                                                            cast homepage \
         2089
               Jon Foo|Kelly Overton|Cary-Hiroyuki Tagawa|Ian...
                                                                      NaN
         2090
               Jon Foo|Kelly Overton|Cary-Hiroyuki Tagawa|Ian...
                                                                      NaN
                       director
                                             tagline
              Dwight H. Little Survival is no game
         2089
              Dwight H. Little Survival is no game
         2090
                                                        keywords \
               martial arts|dystopia|based on video game|mart...
         2089
         2090
              martial arts|dystopia|based on video game|mart...
                                                        overview runtime \
               In the year of 2039, after World Wars destroy ...
                                                                       92
         2090
              In the year of 2039, after World Wars destroy ...
                                                                       92
                                                              production_companies \
                                                    genres
         2089 Crime|Drama|Action|Thriller|Science Fiction Namco|Light Song Films
         2090 Crime|Drama|Action|Thriller|Science Fiction Namco|Light Song Films
               vote_count vote_average budget_adj
                                                     revenue_adj release_date
         2089
                      110
                                    5.0 30000000.0
                                                        967000.0
                                                                   2010-03-20
         2090
                                    5.0 30000000.0
                                                        967000.0
                      110
                                                                   2010-03-20
  From 10866 movies, "TEKKEN" is met twice. So, duplicated row will be deleted.
In [11]: df = df[~df.duplicated()]
         df.shape
Out[11]: (10865, 20)
In [12]: # Check which columns has only unique values, or don't include importnat information
         # "id" columns is a unique identifier
         print(df.id.unique().shape[0] == df.shape[0])
```

```
# imdb_id has 9 missing values and all other ids are also unique for movies
         imdb_id_null = df[df.imdb_id.duplicated() & df.imdb_id.isnull()].shape[0]
         print(df.imdb_id.unique().shape[0] + imdb_id_null == df.shape[0])
True
True
In [13]: print(df[(df.budget == 0) & (df.budget_adj != 0)].shape)
         print(df[(df.revenue == 0) & (df.revenue_adj != 0)].shape)
         df['budget_check'] = df.budget / df.budget_adj
         df['revenue_check'] = df.revenue / df.revenue_adj
         df_agg = df.groupby(df.release_date.dt.year).agg({'revenue_check': ['min', 'max'], 'bud
         df_agg.columns = map('_'.join, df_agg.columns)
         # We don't need 'budget_check' and 'revenue_check' columns anymore
         df.drop(['budget_check', 'revenue_check'], axis=1, inplace=True)
         print(int((df_agg.revenue_check_max - df_agg.revenue_check_min).sum()))
         print(int((df_agg.budget_check_max - df_agg.budget_check_min).sum()))
         df_agg.head()
(0, 20)
(0, 20)
0
Out [13]:
                       revenue_check_min revenue_check_max budget_check_min \
         release_date
         1960
                                0.135631
                                                    0.135631
                                                                      0.135631
         1961
                                0.137083
                                                    0.137083
                                                                      0.137083
         1962
                                0.138726
                                                    0.138726
                                                                      0.138726
         1963
                                0.140446
                                                    0.140446
                                                                      0.140446
         1964
                                                    0.142242
                                                                      0.142242
                                0.142242
                       budget_check_max
         release_date
         1960
                               0.135631
         1961
                               0.137083
         1962
                               0.138726
         1963
                               0.140446
         1964
                               0.142242
```

Columns "revenue" and "budget" could be deleted since corresponding adjusted columns are calculated correctly. Moreover, it's inconvenient to compare financial indicators across years and don't take into account inflation.

There're 5068 directors in this dataset. This means, on average, one person made 2 movies.

```
In [14]: df['director'].unique().shape
Out[14]: (5068,)
In [15]: directors_df = df['director'].value_counts()
         directors_df.head(15)
Out[15]: Woody Allen
                               45
         Clint Eastwood
                               34
         Martin Scorsese
                               29
         Steven Spielberg
                               29
         Ridley Scott
                               23
         Ron Howard
                               22
         Steven Soderbergh
                               22
         Joel Schumacher
                               21
         Brian De Palma
                               20
         Tim Burton
                               19
         Wes Craven
                               19
         Barry Levinson
                               19
         Rob Reiner
                               18
         John Carpenter
                               18
         David Cronenberg
                               18
         Name: director, dtype: int64
```

Top-8 directors made more than 20 movies. Whether their movies are also popular or profitable than movies of the other directos? It will be disscused in Data Analysis part.

```
In [16]: # Drop columns which won't be used in analysis
         df.drop(['id', 'imdb_id', 'budget', 'revenue', 'homepage', 'tagline', 'overview'], axis
         df.head(3)
Out[16]:
           popularity
                           original_title \
            32.985763
                            Jurassic World
            28.419936 Mad Max: Fury Road
         1
         2
            13.112507
                                 Insurgent
                                                                      director \
                                                        cast
         O Chris Pratt|Bryce Dallas Howard|Irrfan Khan|Vi...
                                                               Colin Trevorrow
         1 Tom Hardy|Charlize Theron|Hugh Keays-Byrne|Nic...
                                                                 George Miller
         2 Shailene Woodley|Theo James|Kate Winslet|Ansel... Robert Schwentke
                                                     keywords
                                                              runtime
         O monster|dna|tyrannosaurus rex|velociraptor|island
                                                                   124
            future|chase|post-apocalyptic|dystopia|australia
                                                                   120
         2 based on novel|revolution|dystopia|sequel|dyst...
                                                                  119
                                               genres \
         O Action|Adventure|Science Fiction|Thriller
```

```
Action | Adventure | Science Fiction | Thriller
2
          Adventure | Science Fiction | Thriller
                                 production_companies
                                                        vote_count \
  Universal Studios | Amblin Entertainment | Legenda...
                                                               5562
  Village Roadshow Pictures | Kennedy Miller Produ...
                                                               6185
  Summit Entertainment | Mandeville Films | Red Wago...
                                                               2480
   vote_average
                    budget_adj
                                 revenue_adj release_date
                 1.379999e+08
                                1.392446e+09
0
            6.5
                                                2015-06-09
                1.379999e+08
1
            7.1
                                3.481613e+08
                                                2015-05-13
            6.3 1.012000e+08 2.716190e+08
                                                2015-03-18
```

Replace zeros in numeric columns without missing values with NaN. The logic was explained above.

```
In [17]: col_numeric = ['runtime', 'budget_adj', 'revenue_adj']
         for col in col_numeric:
             df[col] = df[col].replace(0, np.nan)
         df.describe()
Out[17]:
                                                                             budget_adj
                  popularity
                                    runtime
                                                vote_count
                                                            vote_average
                10865.000000
         count
                               10834.000000
                                             10865.000000
                                                            10865.000000 5.169000e+03
                                 102.363855
                    0.646446
                                                217.399632
                                                                5.975012 3.688907e+07
         mean
         std
                    1.000231
                                  30.948225
                                               575.644627
                                                                0.935138
                                                                          4.196096e+07
         min
                    0.000065
                                   2.000000
                                                 10.000000
                                                                1.500000
                                                                          9.210911e-01
         25%
                                  90.000000
                                                 17.000000
                                                                5.400000 8.102293e+06
                    0.207575
         50%
                    0.383831
                                  99.000000
                                                 38.000000
                                                                6.000000 2.271505e+07
         75%
                    0.713857
                                 112.000000
                                                146.000000
                                                                6.600000 5.008384e+07
                   32.985763
                                 900.000000
                                               9767.000000
                                                                9.200000 4.250000e+08
         max
                 revenue_adj
                4.849000e+03
         count
                1.151009e+08
         mean
                1.988557e+08
         std
                2.370705e+00
         min
         25%
                1.046585e+07
         50%
                4.395666e+07
         75%
                1.316482e+08
                2.827124e+09
         max
```

1.1.3 Split columns with different values in each row

Let's look at "cast", "keywords", "genres", "production_companies" columns. Values in this columns are stored as multiple values separated by pipe (|) characters.

```
cast
```

```
In [18]: crew_df = df['cast'].str.split('|').apply(pd.Series).stack().value_counts()
         print('Shape', crew_df.shape)
         crew_df[crew_df > 40].head(15)
         #crew_df.head()
Shape (19026,)
Out[18]: Robert De Niro
                              72
         Samuel L. Jackson
                               71
         Bruce Willis
                               62
         Nicolas Cage
                               61
         Michael Caine
                               53
         Robin Williams
                               51
         John Cusack
                               50
                               49
         John Goodman
         Morgan Freeman
                               49
         Susan Sarandon
                               48
         Liam Neeson
                               48
         Julianne Moore
                               47
         Alec Baldwin
                               47
         Tom Hanks
                               46
         Johnny Depp
                               46
         dtype: int64
```

Since, there's a lot of actresses and actors in this data, columns calles "Robert_De_Niro" and "Samuel_L_Jackson" will be added, they'll indicate wheter this super enduring men were starred in the movie or not. In future, this analysis could be extend by analysing more actors from the list above.

```
In [19]: df['Robert_De_Niro'] = np.where(df['cast'].str.contains('Robert De Niro').replace(np.na
         df['Samuel_L_Jackson'] = np.where(df['cast'].str.contains('Samuel L. Jackson').replace(
         df.drop('cast', axis=1, inplace=True)
         print('Share of movies with Robert De Niro and Samuel L. Jackson')
         print(df[['Robert_De_Niro', 'Samuel_L_Jackson']].mean())
         df.head(3)
Share of movies with Robert De Niro and Samuel L. Jackson
Robert_De_Niro
                    0.006627
Samuel_L_Jackson
                    0.006535
dtype: float64
                           original_title
Out[19]:
           popularity
                                                    director \
           32.985763
                           Jurassic World Colin Trevorrow
```

```
Robert Schwentke
         2
             13.112507
                                  Insurgent
                                                       keywords
                                                               runtime \
           monster|dna|tyrannosaurus rex|velociraptor|island
                                                                   124.0
             future|chase|post-apocalyptic|dystopia|australia
                                                                   120.0
         2 based on novel|revolution|dystopia|sequel|dyst...
                                                                   119.0
                                                genres
            Action|Adventure|Science Fiction|Thriller
            Action | Adventure | Science Fiction | Thriller
         1
         2
                   Adventure | Science Fiction | Thriller
                                          production_companies vote_count \
         O Universal Studios | Amblin Entertainment | Legenda...
                                                                       5562
         1 Village Roadshow Pictures | Kennedy Miller Produ...
                                                                       6185
         2 Summit Entertainment | Mandeville Films | Red Wago...
                                                                       2480
                                                                     Robert_De_Niro
                             budget_adj
                                          revenue_adj release_date
            vote_average
         0
                     6.5 1.379999e+08 1.392446e+09
                                                        2015-06-09
                                                                                  0
                     7.1 1.379999e+08 3.481613e+08
         1
                                                        2015-05-13
                                                                                  0
         2
                     6.3 1.012000e+08 2.716190e+08
                                                        2015-03-18
                                                                                  0
            Samuel_L_Jackson
         0
                           0
         1
         2
                           0
   keywords
In [20]: keywords_df = df['keywords'].str.split('|').apply(pd.Series).stack().value_counts()
         print('Shape', keywords_df.shape)
         keywords_df.head()
Shape (7878,)
```

George Miller

1

Out[20]: woman director

sex sport

independent film

based on novel

dtype: int64

28.419936 Mad Max: Fury Road

```
There're 7878 different keywords. So, only for Top-2, columns calles "woman_director" and "independent_film" will be added, they'll indicate wheter the director of movie was woman or man, and whether the movie is independent film or not. In future, this analysis could also be extend by analysing more keywords from the list above.
```

413

396

278272

216

```
In [21]: df['woman_director'] = np.where(df['keywords'].str.contains('woman director').replace(n
         df['independent_film'] = np.where(df['keywords'].str.contains('independent film').repla
         df.drop('keywords', axis=1, inplace=True)
         print('Share of movies with woman director and independent films')
         print(df[['woman_director', 'independent_film']].mean())
         df.head(3)
Share of movies with woman director and independent films
woman_director
                    0.038012
independent_film
                    0.036539
dtype: float64
Out [21]:
            popularity
                            original_title
                                                    director runtime \
         0
             32.985763
                            Jurassic World
                                             Colin Trevorrow
                                                                 124.0
             28.419936 Mad Max: Fury Road
                                                George Miller
                                                                 120.0
             13.112507
                                 Insurgent Robert Schwentke
                                                                 119.0
                                                genres \
         O Action|Adventure|Science Fiction|Thriller
         1 Action|Adventure|Science Fiction|Thriller
         2
                   Adventure | Science Fiction | Thriller
                                         production_companies vote_count \
         O Universal Studios | Amblin Entertainment | Legenda...
                                                                      5562
         1 Village Roadshow Pictures | Kennedy Miller Produ...
                                                                      6185
         2 Summit Entertainment | Mandeville Films | Red Wago...
                                                                      2480
            vote_average
                            budget_adj
                                         revenue_adj release_date
                                                                    Robert_De_Niro \
                     6.5 1.379999e+08 1.392446e+09
         0
                                                        2015-06-09
         1
                     7.1 1.379999e+08 3.481613e+08
                                                        2015-05-13
                                                                                 0
                     6.3 1.012000e+08 2.716190e+08
                                                        2015-03-18
                                                                                 0
            Samuel_L_Jackson woman_director independent_film
         0
                           0
                                           0
                           0
                                           0
                                                              0
         1
         2
                                                              0
                           0
                                           0
  genres
In [22]: genres_df = df['genres'].str.split('|').apply(pd.Series).stack().value_counts()
         print('Shape', genres_df.shape)
         genres_df
Shape (20,)
```

```
Out[22]: Drama
                             4760
         Comedy
                             3793
         Thriller
                             2907
         Action
                             2384
         Romance
                             1712
         Horror
                             1637
         Adventure
                             1471
         Crime
                             1354
         Family
                             1231
         Science Fiction
                             1229
         Fantasy
                              916
         Mystery
                              810
         Animation
                              699
         Documentary
                              520
         Music
                              408
         History
                              334
         War
                              270
         Foreign
                              188
         TV Movie
                              167
         Western
                              165
         dtype: int64
```

For every genre will be created dummy variable which will indicate what genre the film belongs to.

```
In [23]: genre_unique = genres_df.index.tolist()
        for genre in genre_unique:
             df[genre] = np.where(df['genres'].str.contains(genre).replace(np.nan, False), 1, 0)
         df.drop('genres', axis=1, inplace=True)
         print('New shape', df.shape)
        df.head(3)
New shape (10865, 34)
Out[23]:
           popularity
                           original_title
                                                   director runtime \
                           Jurassic World
                                            Colin Trevorrow
        0 32.985763
                                                               124.0
            28.419936 Mad Max: Fury Road
                                               George Miller
                                                               120.0
         1
         2 13.112507
                                 Insurgent Robert Schwentke
                                                               119.0
                                        production_companies vote_count \
         O Universal Studios | Amblin Entertainment | Legenda...
                                                                     5562
         1 Village Roadshow Pictures | Kennedy Miller Produ...
                                                                     6185
         2 Summit Entertainment | Mandeville Films | Red Wago...
                                                                     2480
                                        revenue_adj release_date Robert_De_Niro \
           vote_average
                           budget_adj
```

```
0
            6.5 1.379999e+08 1.392446e+09
                                                2015-06-09
                                                                         0
1
            7.1 1.379999e+08 3.481613e+08
                                                2015-05-13
                                                                          0
2
            6.3 1.012000e+08 2.716190e+08
                                                2015-03-18
                                                                          0
                                      independent_film Drama
   Samuel_L_Jackson woman_director
0
                                                      0
                  0
                                   0
                                                             0
1
                                                                     0
                  0
                                                      0
                                                             0
2
                                   0
                                                                     0
   Thriller Action Romance
                              Horror
                                       Adventure Crime
                                                          Family
0
          1
                  1
                            0
                                    0
                                                1
                                                       0
                                                               0
1
          1
                  1
                            0
                                    0
                                                1
                                                       0
                                                               0
2
                            0
                                                       0
          1
                  0
                                    0
                                                1
                                                               0
   Science Fiction Fantasy Mystery
                                      Animation Documentary Music
                                                                      History \
0
                                                                              0
1
                 1
                           0
                                    0
                                                0
                                                             0
                                                                    0
                                                                              0
2
                 1
                           0
                                    0
                                                0
                                                             0
                                                                    0
                                                                              0
        Foreign
                TV Movie Western
   War
0
    0
              0
                        0
1
     0
              0
                        0
                                  0
2
     0
                                  0
```

production_companies

Shape (7879,)

Out[24]:	Universal Pictures	522
	Warner Bros.	509
	Paramount Pictures	431
	Twentieth Century Fox Film Corporation	282
	Columbia Pictures	272
	New Line Cinema	219
	Metro-Goldwyn-Mayer (MGM)	218
	Walt Disney Pictures	214
	Touchstone Pictures	178
	Columbia Pictures Corporation	160
	dtype: int64	

Top-3 production companies are used among list of 7879 different firms: * Universal Pictures

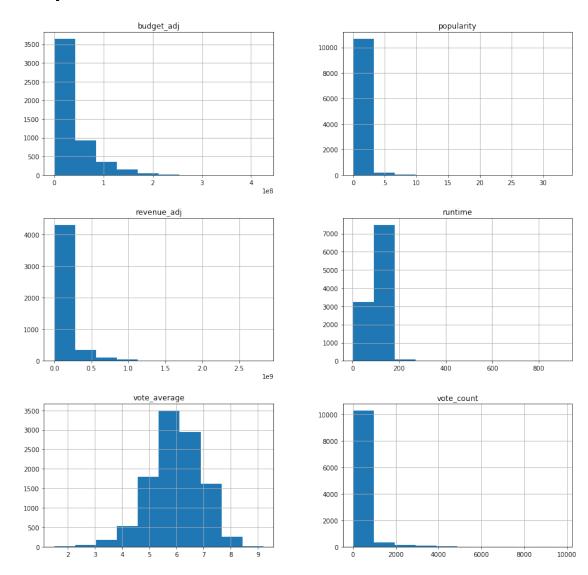
^{*} Warner Bros.

^{*} Paramount Pictures.

```
In [25]: df['Universal_Pictures'] = (np.where(df['production_companies'].str.contains('Universal
                                               .replace(np.nan, False), 1, 0))
         df['Warner_Bros'] = np.where(df['production_companies'].str.contains('Warner Bros.').re
         df['Paramount_Pictures'] = (np.where(df['production_companies'].str.contains('Paramount
                                               .replace(np.nan, False), 1, 0))
         df.drop('production_companies', axis=1, inplace=True)
         print('Share of movies of Universal Pictures, Warner Bros., and Paramount Pictures')
         print(df[['Universal_Pictures', 'Warner_Bros', 'Paramount_Pictures']].mean() * 100)
         df.head(3)
Share of movies of Universal Pictures, Warner Bros., and Paramount Pictures
Universal_Pictures
                      4.878049
Warner_Bros
                      5.715601
Paramount_Pictures
                      3.985274
dtype: float64
Out [25]:
            popularity
                            original_title
                                                     director runtime
                                                                        vote_count \
             32.985763
                             Jurassic World
                                              Colin Trevorrow
                                                                 124.0
                                                                               5562
             28.419936 Mad Max: Fury Road
                                                George Miller
                                                                 120.0
                                                                               6185
             13.112507
                                 Insurgent Robert Schwentke
                                                                 119.0
                                                                               2480
            vote_average
                            budget_adj
                                         revenue_adj release_date Robert_De_Niro
         0
                     6.5 1.379999e+08 1.392446e+09
                                                        2015-06-09
                                                                                  0
                     7.1 1.379999e+08 3.481613e+08
                                                        2015-05-13
                                                                                  0
         1
         2
                     6.3 1.012000e+08 2.716190e+08
                                                                                  0
                                                        2015-03-18
            Samuel\_L\_Jackson woman_director
                                               independent_film Drama
                                                                        Comedy
         0
                                                                     0
                                                                              0
         1
                           0
                                            0
                                                              0
                                                                     0
                                                                              0
         2
                           0
                                            0
                                                                     0
                                                                              0
            Thriller Action Romance Horror Adventure
                                                          Crime
                                                                  Family
         0
                                    0
                                                               0
                   1
                           1
                                             0
                                                        1
                                                                       0
                   1
                                    0
                                                        1
                                                               0
                                                                       0
                           1
                                             0
         2
                                    0
                                             0
                                                               0
            Science Fiction Fantasy Mystery
                                                Animation Documentary Music
         0
                                                        0
                                                                     0
                                                                             0
                                                                                      0
                          1
                                   0
                                             0
                          1
                                   0
                                                        0
                                                                     0
                                                                             0
                                                                                      0
         1
                                             0
         2
                          1
                                   0
                                             0
                                                        0
                                                                     0
                                                                             0
                                                                                      0
                 Foreign TV Movie Western Universal_Pictures Warner_Bros
            War
         0
              0
                       0
                                 0
                                           0
                                                               0
                                                                             0
              0
                       0
                                 0
                                           0
                                                               0
                                                                             0
         1
```

2	0	0	0	0	0	0
	Paramou	ınt_Picture	es			
0			0			
1			0			
2			0			

1.1.4 Visualize results



In [27]: df_nulls = df.isnull().sum().sort_values(ascending=False)

```
print('Share of missing values')
df_nulls[df_nulls > 0] / df.shape[0]
```

Share of missing values

Missing values in "director" column isn't significant, and it's equal 0.41%. So we'll only look at non-missing data. The same could be applied to "runtime" column, and share of NaN is equal 0.29%. here we'll also look at non-missing data. More than 50% of missing values in revenues and budget will significantly distort results. Replace a half of missing values with any measure of center won't produce accurate results. So, one way is somehow impute NaN. The other way is to analyse only data with valid not-zero numbers. And some other scenarios. In the following analyses wil be used only not-zero budgets and revenues.

```
In [28]: df = df[(df.director.notnull()) & (df.runtime.notnull())]
In [29]: print('Final shape', df.shape)
         df.head()
Final shape (10792, 36)
Out[29]:
            popularity
                                        original_title
                                                                 director runtime
         0
             32.985763
                                        Jurassic World
                                                          Colin Trevorrow
                                                                              124.0
             28.419936
                                   Mad Max: Fury Road
                                                            George Miller
                                                                              120.0
         1
                                                         Robert Schwentke
         2
             13.112507
                                             Insurgent
                                                                              119.0
         3
             11.173104
                         Star Wars: The Force Awakens
                                                              J.J. Abrams
                                                                              136.0
         4
              9.335014
                                             Furious 7
                                                                James Wan
                                                                              137.0
                                                        revenue_adj release_date
            vote_count
                         vote_average
                                          budget_adj
         0
                   5562
                                   6.5
                                       1.379999e+08
                                                       1.392446e+09
                                                                       2015-06-09
         1
                   6185
                                  7.1 1.379999e+08
                                                      3.481613e+08
                                                                      2015-05-13
         2
                   2480
                                  6.3 1.012000e+08
                                                       2.716190e+08
                                                                      2015-03-18
         3
                   5292
                                  7.5 1.839999e+08 1.902723e+09
                                                                      2015-12-15
         4
                   2947
                                  7.3 1.747999e+08 1.385749e+09
                                                                      2015-04-01
            Robert_De_Niro
                             Samuel_L_Jackson
                                                woman_director
                                                                 independent_film
                                                                                    Drama
         0
                          0
                                             0
                                                              0
                                                                                         0
                          0
                                             0
                                                              0
                                                                                 0
                                                                                        0
         1
         2
                          0
                                             0
                                                              0
                                                                                 0
                                                                                        0
         3
                          0
                                             0
                                                              0
                                                                                 0
                                                                                        0
         4
                          0
                                             0
                                                              0
                                                                                 0
                                                                                         0
```

	Comedy	7 Thril	ler	Action	Romano	се	Horror A	Adventure	Crin	ne Fam:	ily	\	
0	C)	1	1		0	0	1		0	0		
1	C)	1	1		0	0	1		0	0		
2	C)	1	C	ı	0	0	1		0	0		
3	C)	0	1		0	0	1		0	0		
4	C)	1	1		0	0	C		1	0		
	Scienc	e Ficti	on	Fantasy	Myster	ΞУ	Animation	n Docume	ntary	Music	Hi	story	\
0			1	C	ı	0	()	0	0		0	
1			1	C	ı	0	()	0	0		0	
2			1	C	ı	0	()	0	0		0	
3			1	1		0	()	0	0		0	
4			0	C	l	0	()	0	0		0	
	War F	oreign	TV	Movie	Western	U	Jniversal_H	Pictures	Warne	er_Bros	\		
0	0	0		0	0			0		0			
1	0	0		0	0			0		0			
2	0	0		0	0			0		0			
3	0	0		0	0			0		0			
4	0	0		0	0			1		0			
	Paramo	ount_Pic	ture	es									
0				0									
1				0									
2				0									
3				0									
4				0									

Exploratory Data Analysis

Research questions: 1. Which genres are most popular from year to year?

- 2. How sharp is the divide between major film studios and the independents?
- 3. What kinds of properties are associated with movies that have high revenues?
- 4. Top-8 directors made more than 20 movies. Whether their movies are also popular or profitable than movies of the other directos?

1.1.5 Research Question 1: Which genres are most popular from year to year?

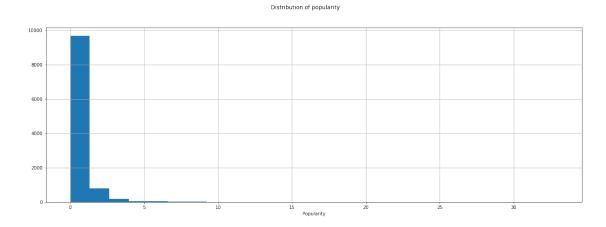
In [30]: df.popularity.describe()

Out[30]: count 10792.000000 0.649765 mean 1.002611 std 0.000188 min 25% 0.209737 50% 0.385598 75% 0.717722 max32.985763

Name: popularity, dtype: float64

```
In [31]: df[df.popularity == df.popularity.max()]
Out [31]:
            popularity original_title
                                               director runtime
                                                                  vote_count
             32.985763
                        Jurassic World Colin Trevorrow
                                                            124.0
                                                                         5562
            vote_average
                            budget_adj
                                         revenue_adj release_date
                                                                    Robert_De_Niro
                                                       2015-06-09
         0
                     6.5
                          1.379999e+08 1.392446e+09
            Samuel_L_Jackson woman_director
                                              independent_film Drama
         0
                                                                     0
            Thriller
                     Action
                             Romance
                                       Horror
                                               Adventure
                                                          Crime
                                                                 Family \
         0
                   1
                           1
                                    0
                                            0
                                                        1
                                                               0
                                                                       0
            Science Fiction Fantasy Mystery
                                                          Documentary Music
                                               Animation
         0
                          1
                                   0
                                            0
                                                                     0
                                                                            0
                                                                                     0
                         TV Movie Western Universal_Pictures Warner_Bros
         0
              0
                       0
                                 0
                                          0
            Paramount_Pictures
         0
```

Most of the popularity ratings are less than 1. The distribution of popularity is right skewed with maximum rating 32.99 of movie "Jurassic World" (2015).



From 1960 to 2015 the number of movies increased with the corresponding increase of average opularity rate. However, some movies became popular since 20 century. For example, "The Godfather" (1972) has 5.74 popularity rate.

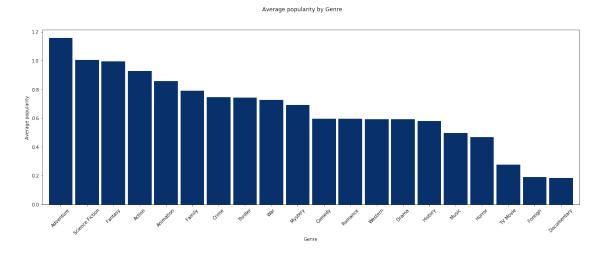
```
In [33]: df[(df.release_date >= '1972-01-01') & (df.release_date <= '1974-01-01')].sort_values('
                                                                         ascending=False).iloc[0, :9
Out[33]: popularity
                                          5.73803
         original_title
                                   The Godfather
         director
                            Francis Ford Coppola
         runtime
                                              175
                                             3970
         vote_count
                                              8.3
         vote_average
         budget_adj
                                     3.12874e+07
         revenue_adj
                                      1.27791e+09
         release_date
                             1972-03-15 00:00:00
         Name: 7269, dtype: object
In [34]: df.plot(x='release_date', y='popularity', rot=0, legend=False, kind='line', figsize=(21
         plt.xlabel('Release date')
         plt.ylabel('Popularity')
         plt.suptitle('Popularity across time')
         plt.show()
                                       Popularity across time
```

For the next figure and following barplots create a function which will plot bar charts.

```
In [35]: xlabel = 'Release Date'
    ylabel = 'Average popularity'

def plot_bar_chart(df, title, cmap, rot=0, legend=True, legend_lst=None, loc='best', xl
    df.plot(kind='bar', rot=rot, figsize=(21,7), cmap=cmap, legend=legend, width=.9)
    if legend and legend_lst:
        plt.legend(legend_lst, loc=loc)
    elif legend:
        plt.legend(loc=loc)
    plt.xlabel(xlabel)
    plt.ylabel(ylabel)
    plt.suptitle(title)
    plt.show()
```

Among all 10 thound movies, Adventure is the most popular genre with average popularity about 1.2.



Now, that we have average popularity separately by release date and by genres, let's look at the most popular genres over time.

In [37]: # Reshape dataset such that genre became one row and create popularity_by_genre dataset

popularity_by_genre = df[df[genre_unique[0]] == 1].groupby(df.release_date.dt.year)['popularity_by_genre['genre'] = genre_unique[0]
 years = df.release_date.dt.year.tolist()
 for col in genre_unique[1:]:
 temp = df[df[col] == 1].groupby(df.release_date.dt.year)['popularity'].mean().reset
 temp['genre'] = col
 popularity_by_genre = popularity_by_genre.append(temp)

print('Genres\n', popularity_by_genre.genre.unique())
 print('Average popularity\n', popularity_by_genre.popularity.mean())
 popularity_by_genre.head()
Genres

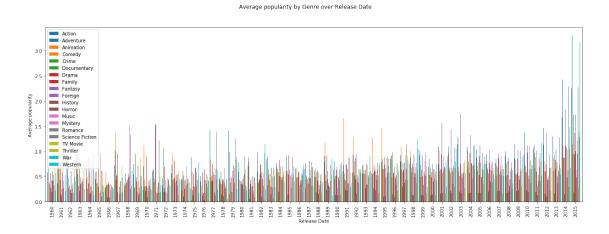
['Drama' 'Comedy' 'Thriller' 'Action' 'Romance' 'Horror' 'Adventure'

```
Average popularity
0.586307521846
           release_date popularity genre
Out [37]:
        0
                   1960
                           0.566305 Drama
         1
                   1961
                           0.432233
                                     Drama
         2
                   1962
                           0.392000 Drama
        3
                   1963
                           0.322448 Drama
         4
                   1964
                           0.364818 Drama
In [38]: # Create dateset with popularity means to make visuals
         df_plot = popularity_by_genre.groupby(['release_date', 'genre']).popularity.mean()
         df_plot = df_plot.sort_index(level='release_date', ascending=True).unstack()
         df_plot.head()
Out[38]: genre
                                                                          Documentary \
                        Action Adventure Animation
                                                         Comedy
                                                                   Crime
        release_date
         1960
                      0.590724
                                 0.700981
                                                 NaN 0.396000 0.346480
                                                                                   NaN
         1961
                      0.365913
                                 0.767323
                                             2.631987 0.679838
                                                                0.806519
                                                                                  NaN
         1962
                       0.708945
                                 0.942513
                                                 NaN 0.273183 0.472816
                                                                                  NaN
                       0.951729
         1963
                                 0.728487
                                             2.180410 0.444481
                                                                0.552367
                                                                                  NaN
                                 0.736431
         1964
                       0.813087
                                            0.445746 0.491229 0.261690
                                                                                  NaN
                          Drama
                                  Family
                                           Fantasy
                                                     Foreign
                                                               History
                                                                          Horror
        genre
         release_date
         1960
                      0.566305 0.278064 0.428247 0.194948 0.412364 0.591450
         1961
                      0.432233 0.755722 0.154073 0.113651
                                                              0.334463 0.254578
         1962
                      0.392000 0.323463 0.235542 0.235542
                                                              0.600060 0.179368
         1963
                       0.322448 0.852632 0.214071 0.351647
                                                              0.481618 0.376901
         1964
                       0.364818  0.878298  0.599791  0.152209
                                                              0.316377 0.189650
                         Music
                                 Mystery
                                            Romance Science Fiction TV Movie
         genre
         release_date
         1960
                       0.423531
                                     {\tt NaN}
                                         0.541227
                                                           0.327905
                                                                          NaN
         1961
                      0.564749 0.712793 0.426036
                                                           0.274103
                                                                          NaN
         1962
                       0.323933 0.259220
                                          0.365621
                                                           0.179212
                                                                          NaN
         1963
                           NaN 0.421033
                                         0.399573
                                                           0.467836
                                                                          NaN
         1964
                      0.753799 0.308091
                                         0.221586
                                                           0.226636
                                                                          NaN
                      Thriller
         genre
                                     War
                                           Western
         release_date
         1960
                      0.811910 0.225863 0.567759
                            NaN 0.531184
         1961
                                          0.210021
         1962
                      0.760582 0.506162 0.414180
```

'Crime' 'Family' 'Science Fiction' 'Fantasy' 'Mystery' 'Animation' 'Documentary' 'Music' 'History' 'War' 'Foreign' 'TV Movie' 'Western']

```
1963 0.719929 0.664326 0.246957
1964 0.622580 0.930959 0.127679
```

It's not a good idea to use all the genres for the last 55 years. Let's fix this in the following figure.



Get only top-5 genres from each year, so we could distinguish their popularity rates.

```
In [40]: df_plot = popularity_by_genre.groupby(['release_date', 'genre']).popularity.mean()
         df_plot_top5 = df_plot.reset_index().sort_values(by=['release_date', 'genre'],
                                                           ascending=[True, False]).groupby('rele
         df_plot_top5.set_index(['release_date', 'genre'], inplace=True)
         df_plot_top5.head(10)
Out [40]:
                                        popularity
         release_date genre
         1960
                      Western
                                          0.567759
                      War
                                          0.225863
                      Thriller
                                          0.811910
                      Science Fiction
                                          0.327905
                                          0.541227
                      Romance
         1961
                      Western
                                          0.210021
                      War
                                          0.531184
                      Science Fiction
                                          0.274103
                      Romance
                                          0.426036
                      Mystery
                                          0.712793
In [41]: df_plot_top5.groupby('genre').popularity.mean().sort_values()
```

0.285852

Out[41]: genre

TV Movie

```
Romance 0.464334

Mystery 0.501919

Western 0.537787

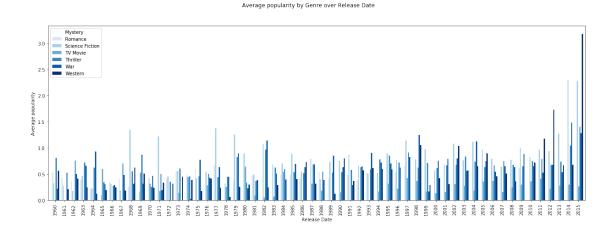
War 0.626728

Thriller 0.645515

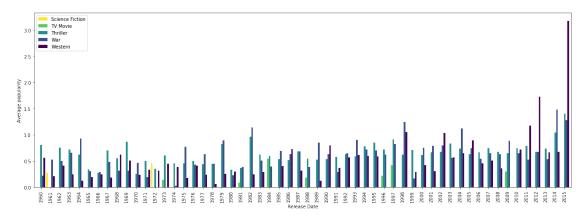
Science Fiction 0.763493

Name: popularity, dtype: float64
```

The most popular genres by popularity rate during the whole period is Science Fiction, Thriller, and Wars. Moreover, during the 20 century Romance and TV movies were also as popular as Science Fiction genre. In the 21st century Western genre has reached leading positions.



If we look only on top-3 genres of each year, Thriller and war genres bagin to compete for first place with each other from year to year.



1.1.6 Research Question 2: How sharp is the divide between major film studios and the independents?

Top-2 production companies are:

- 1. Universal Pictures (produced 522 movies),
- 2. Warner Bros. (produced 509 movies).

Let's look at difference between major film studios and the independents using this production companies data.

```
In [44]: df.groupby(['independent_film', 'Universal_Pictures']).mean()[['popularity', 'runtime',
                                                                         'budget_adj', 'revenue_ad
Out[44]:
                                               popularity
                                                               runtime
                                                                        vote_count
         independent_film Universal_Pictures
         0
                          0
                                                 0.646776
                                                           102.220736
                                                                        215.945171
                          1
                                                 0.962337
                                                           108.499055
                                                                        407.565217
                          0
         1
                                                 0.306643
                                                            98.840506
                                                                         36.240506
                          1
                                                 0.327996 102.000000
                                                                         23.000000
                                                                budget_adj
                                                                             revenue_adj
                                               vote_average
         independent_film Universal_Pictures
                          0
                                                   5.968410 3.657895e+07
                                                                            1.138762e+08
                          1
                                                   6.073535 5.267046e+07
                                                                            1.660731e+08
                          0
         1
                                                   5.951646
                                                             9.054178e+06
                                                                            5.819831e+06
                          1
                                                   5.800000 4.577416e+06
                                                                                     NaN
In [45]: pd.crosstab(df.independent_film, df.Universal_Pictures)
Out[45]: Universal_Pictures
                                 0
                                      1
         independent_film
         0
                             9867
                                    529
         1
                              395
                                      1
```

```
In [46]: df.groupby(['independent_film', 'Warner_Bros']).mean()[['popularity', 'runtime', 'vote_
                                                                        'budget_adj', 'revenue_ad
Out[46]:
                                       popularity
                                                      runtime vote_count
         independent_film Warner_Bros
                          0
                                         0.638903 102.101156 210.486550
                          1
                                         1.040808 109.474960 465.922456
                          0
                                         0.305501
                                                   98.829949
         1
                                                                36.269036
                                         0.542260 102.500000
                          1
                                                                24.000000
                                                       budget_adj
                                       vote_average
                                                                    revenue_adj
         independent_film Warner_Bros
                          0
                                           5.960949 3.546243e+07 1.125606e+08
                          1
                                           6.176090 6.425075e+07 1.769029e+08
         1
                          0
                                           5.951269 8.617312e+06 5.754943e+06
                          1
                                           5.950000 6.792302e+07 1.432024e+07
In [47]: pd.crosstab(df.independent_film, df.Warner_Bros)
Out[47]: Warner_Bros
                                   1
         independent_film
         0
                           9777
                                 619
         1
                            394
```

Universal Pictures made only 1 independent film (it's about 0.19% of all movies made by the company), and Warner Bros. made 2 independent films (it's about 0.32% of all movies made by the company). In comparison, 393 movies of the other companies were independent (it's about 3.67% of all movies).

So, why major film studios don't made independent?

Popularity rates of independent movies are twice lower (0.65 vs. 0.31, or 0.96 vs. 0.33) than not-independent ones.

If we don't take into account missing data in revenue and budget columns, the result for independent movies is more than twice worse. Despite average votes for independent and not-independent movies are approximately the same (6 points), number of votes for independent movies is more then 5 times smaller (200-500 vs. 20-40).

So, producing companies don't make independent movies because revenues from them are lower, or revenues from independent movies are lower because producing comanies don't make them? It's really interesting question for the further investigation and analysis.

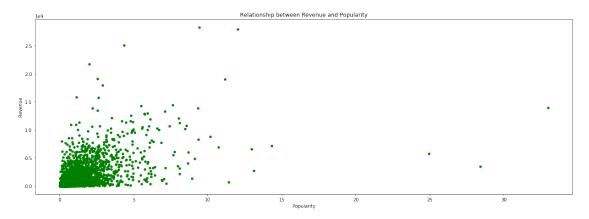
1.1.7 Research Question 3: What kinds of properties are associated with movies that have high revenues?

To answer this question, only numeric columns were considered.

```
1329
        12.037933
                        Star Wars
                                         George Lucas
                                                          121.0
                                                                        4428
5231
         4.355219
                           Titanic
                                        James Cameron
                                                          194.0
                                                                        4654
10594
         2.010733
                     The Exorcist
                                   William Friedkin
                                                          122.0
                                                                        1113
9806
                                   Steven Spielberg
         2.563191
                              Jaws
                                                          124.0
                                                                        1415
       vote_average
                        budget_adj
                                       revenue_adj release_date
                                                                   Robert_De_Niro
1386
                 7.1
                     2.408869e+08 2.827124e+09
                                                      2009-12-10
1329
                 7.9 3.957559e+07
                                      2.789712e+09
                                                      1977-03-20
                                                                                 0
5231
                 7.3 2.716921e+08 2.506406e+09
                                                                                 0
                                                      1997-11-18
10594
                 7.2 3.928928e+07
                                      2.167325e+09
                                                      1973-12-26
                                                                                 0
9806
                 7.3 2.836275e+07 1.907006e+09
                                                                                 0
                                                      1975-06-18
       Samuel_L_Jackson
                           woman_director
                                            independent_film Drama
                                                                       Comedy
                                                                    0
1386
                       0
                                         0
                                                            0
                                                                             0
                       0
                                         0
                                                            0
                                                                    0
1329
                                                                             0
5231
                       0
                                         0
                                                            0
                                                                    1
                                                                             0
10594
                       0
                                         0
                                                            0
                                                                    1
                                                                             0
                                                                    0
                                                                             0
9806
                       0
                                         0
                                                            0
       Thriller
                 Action
                                    Horror
                                             Adventure
                                                         Crime
                                                                Family
                           Romance
               0
1386
                       1
                                 0
                                          0
                                                      1
                                                             0
               0
                                 0
                                          0
                                                      1
                                                             0
                                                                      0
1329
                       1
5231
               1
                       0
                                 1
                                          0
                                                      0
                                                             0
                                                                      0
10594
               1
                       0
                                 0
                                          1
                                                      0
                                                             0
                                                                      0
9806
               1
                       0
                                 0
                                          1
                                                      1
                                                             0
                                                                      0
       Science Fiction Fantasy Mystery
                                             Animation
                                                        Documentary
                                                                       Music
1386
                      1
                                1
                                          0
                                                      0
                                                                    0
                                                                           0
                                0
1329
                                          0
                                                      0
                                                                    0
                                                                           0
5231
                      0
                                0
                                          0
                                                      0
                                                                    0
                                                                           0
10594
                      0
                                0
                                          0
                                                      0
                                                                    0
                                                                           0
9806
                      0
                                0
                                          0
                                                      0
                                                                           0
                               TV Movie
                                          Western Universal_Pictures
       History
                 War
                      Foreign
                                        0
1386
              0
                   0
                             0
                                                 0
                                                                       0
                                        0
                                                 0
                                                                       0
1329
              0
                   0
                             0
                             0
                                        0
                                                 0
                                                                       0
5231
                   0
10594
              0
                   0
                             0
                                        0
                                                 0
                                                                       0
9806
              0
                   0
                             0
       Warner_Bros
                     Paramount_Pictures
1386
                  0
                                        0
1329
                  0
                                        0
                  0
                                        1
5231
                                        0
10594
                  1
9806
                                        0
```

In [49]: revenue_df[['popularity', 'revenue_adj']].plot(kind='scatter', x='popularity', y='revenue_adj']]

```
color='green')
plt.xlabel('Popularity')
plt.ylabel('Revenue')
plt.title('Relationship between Revenue and Popularity')
plt.show()
```



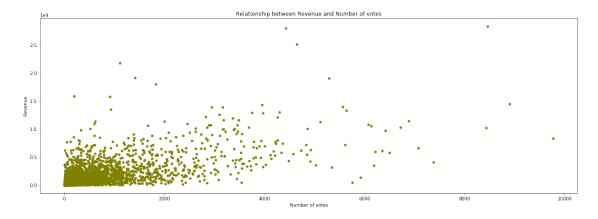
Average Revenue 771099276.712

```
Out[50]:
              popularity
                               original_title
                                                          director runtime
                                                                              vote count
         0
               32.985763
                               Jurassic World
                                                  Colin Trevorrow
                                                                      124.0
                                                                                    5562
         629
               24.949134
                                 Interstellar Christopher Nolan
                                                                      169.0
                                                                                    6498
         1
               28.419936 Mad Max: Fury Road
                                                    George Miller
                                                                      120.0
                                                                                    6185
                               budget_adj
                                             revenue_adj release_date Robert_De_Niro
              vote_average
         0
                        6.5
                            1.379999e+08
                                            1.392446e+09
                                                            2015-06-09
                                                                                      0
         629
                        8.0 1.519800e+08
                                            5.726906e+08
                                                            2014-11-05
                                                                                      0
         1
                             1.379999e+08 3.481613e+08
                                                            2015-05-13
                                                                                      0
              Samuel_L_Jackson woman_director independent_film
                                                                     Drama
                                                                             Comedy
         0
                              0
                                               0
                                                                  0
                                                                          0
                                                                                  0
         629
                              0
                                               0
                                                                  0
                                                                                  0
                                                                          1
         1
                              0
                                               0
                                                                  0
                                                                          0
                                                                                  0
                         Action
              Thriller
                                 Romance
                                           Horror
                                                   Adventure
                                                               Crime
                                                                      Family
         0
                      1
                                        0
                                                0
                                                            1
                                                                   0
                                                                            0
                              0
         629
                      0
                                        0
                                                0
                                                            1
                                                                   0
                                                                            0
                              1
                                                0
         1
                                        0
                                                                   0
                                                                            0
```

Science Fiction Fantasy Mystery Animation Documentary Music \

```
0
                     1
                               0
                                                                   0
                                                                           0
629
                     1
                               0
                                         0
                                                                   0
                                                                           0
                     1
                               0
                                         0
                                                                           0
1
                     Foreign
                              TV Movie Western Universal_Pictures
     History
0
            0
                           0
            0
                                       0
                                                 0
                                                                       0
629
                 0
                           0
                           0
                                                 0
1
                                                                       0
     Warner_Bros Paramount_Pictures
0
                0
629
                1
                                       1
1
                0
                                       0
```

```
plt.xlabel('Number of votes')
plt.ylabel('Revenue')
plt.title('Relationship between Revenue and Number of votes')
plt.show()
```



Average Revenue 1527637054.52

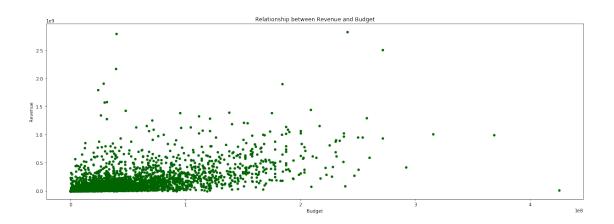
Out[52]:		popularity	$original_title$	director	runtime	vote_count	\
	1386	9.432768	Avatar	James Cameron	162.0	8458	
	4361	7.637767	The Avengers	Joss Whedon	143.0	8903	
	2875	8.466668	The Dark Knight	Christopher Nolan	152.0	8432	
	1919	9.363643	Inception	Christopher Nolan	148.0	9767	

```
1386
                         7.1 2.408869e+08 2.827124e+09
                                                             2009-12-10
         4361
                         7.3 2.089437e+08 1.443191e+09
                                                             2012-04-25
                                                                                       0
         2875
                         8.1 1.873655e+08 1.014733e+09
                                                             2008-07-16
                                                                                       0
         1919
                         7.9 1.600000e+08 8.255000e+08
                                                             2010-07-14
                                                                                       0
               Samuel_L_Jackson
                                  woman_director
                                                   independent_film
                                                                      Drama
         1386
                                                                           0
                                                                                   0
         4361
                               0
                                                0
                                                                   0
                                                                           0
                                                                                   0
         2875
                               0
                                                0
                                                                   0
                                                                           1
                                                                                   0
         1919
                               0
                                                0
                                                                   0
                                                                           0
                                                                                   0
               Thriller
                          Action
                                  Romance
                                            Horror
                                                    Adventure
                                                                Crime
                                                                        Family
         1386
                       0
                               1
                                         0
                                                 0
                                                             1
                                                                    0
         4361
                       0
                                         0
                                                 0
                                                                    0
                                                                             0
                                         0
         2875
                       1
                               1
                                                 0
                                                             0
                                                                    1
                                                                             0
         1919
                       1
                               1
                                         0
                                                 0
                                                             1
                                                                    0
                                                                             0
               Science Fiction Fantasy Mystery Animation Documentary
         1386
                              1
                                        1
                                                 0
                                                             0
                                                                           0
                                                                                  0
                              1
         4361
                                        0
                                                 0
                                                             0
                                                                           0
                                                                                  0
         2875
                              0
                                        0
                                                 0
                                                             0
                                                                           0
                                                                                  0
                              1
                                        0
                                                 1
                                                                           0
         1919
               History War Foreign TV Movie Western Universal_Pictures
         1386
                      0
                           0
                                               0
                                                         0
                                                                              0
                                     0
         4361
                      0
                           0
                                     0
                                               0
                                                         0
                                                                              0
         2875
                           0
                                     0
                                               0
                                                         0
                                                                              0
                      0
                           0
                                               0
                                                         0
                                                                              0
         1919
                      0
               Warner_Bros Paramount_Pictures
         1386
                          0
         4361
                          0
                                               0
         2875
                          1
                                               0
         1919
                          1
                                               0
In [53]: revenue_df[['budget_adj', 'revenue_adj']].plot(kind='scatter', x='budget_adj', y='revenue_adj')
                                                          color='darkgreen')
         plt.xlabel('Budget')
         plt.ylabel('Revenue')
         plt.title('Relationship between Revenue and Budget')
         plt.show()
```

revenue_adj release_date Robert_De_Niro

vote_average

budget_adj



In [54]: print('Average Revenue', revenue_df[revenue_df.budget_adj > 300000000].revenue_adj.mear
revenue_df[revenue_df.budget_adj > 300000000]

Average Revenue 670719525.753

Out[54]:		popularity				0	riginal_t	itle	direc	tor \
	7387	4.965391	Pira	ates of t	the Car	ibbean: A	t World's	End Gor	e Verbin	ski
	3375	4.955130	Pirate	s of the	Caribb	ean: On S	tranger T	ides F	Rob Marsh	all
	2244	0.250540				The	Warrior's	Way	${\tt Sngmoo}$	Lee
		runtime vo	ote_coun	t vote_a	average	budge	t_adj r	evenue_ad	lj \	
	7387	169.0	262	6	6.8	3.15500	6e+08 1.	010654e+0	9	
	3375	136.0	3180)	6.3	3.68371	3e+08 9.	904175e+0	8(
	2244	100.0	7	4	6.4	4.25000	0e+08 1.	108757e+0)7	
									_	
		release_date		t_De_Nir	o Samu	el_L_Jack	son woma	n_directo	or \	
	7387	2007-05-19		()		0		0	
	3375	2011-05-11		()		0		0	
	2244	2010-12-02	2	()		0		0	
			L # 2 7 1 1	D 0	a	Trl. : 1 7	٨ ٠٠	D	TT	`
	7007	independent			•					\
	7387		0	0	0	0	1	0	0	
	3375		0	0	0	0	1	0	0	
	2244		0	0	0	1	1	0	0	
		Adventure	Crime 1	Family S	Science	Fiction	Fantasy	Mvsterv	Animati	on \
	7387	1	0	0		0	1	0		0
	3375	1	0	0		0	1	0		0
	2244	1	0	0		0	1	0		0
		Documentary	y Music	History	y War	Foreign	TV Movie	Westerr	n \	
	7387	(0 0	(0 0	0	0	C)	

3375	0 0	0 0	0 0	0
2244	0 0	0 0	0 0	1
	Universal_Pictures	Warner_Bros	Paramount_Pictures	
7387	0	0	0	
3375	0	0	0	
2244	0	0	0	

The relationship between revenue and popularity, revenue and number of votes, revenue and budget is positive.

Some outliers are present in the dataset, so:

- * 3 movies with the highest popularity made on average \$771099276.71 (Jurassic World (2015), Interstellar (2014), Mad Max: Fury Road (2015)),
- * 4 movies with the number of votes higher than 8000 gathered on average, \$1527637054.52 (Avatar (2009), The Avengers (2012), The Dark Knight (2008), Inception (2010)),
- * 3 movies with the highest budget made on average \$670719525.753 (Pirates of the Caribbean: At World's End (2001), Pirates of the Caribbean: On Stranger Tides (2011), The Warrior's Way (2010)).

1.1.8 Research Question 4: Top-8 directors made more than 20 movies.

top8_directors = df[df.director.isin(directors)]

In [55]: directors = directors_df.iloc[:8].index.tolist()

Whether their movies are also popular or profitable than movies of the other directos?

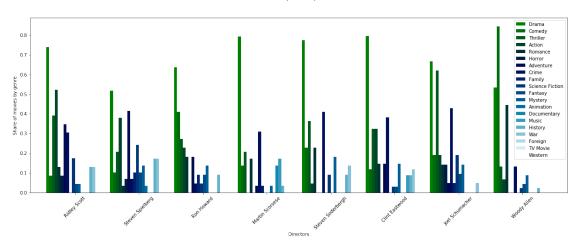
```
print('Shape', top8_directors.shape)
         top8_directors.head()
Shape (225, 36)
Out [55]:
                                                            director runtime \
              popularity
                                   original_title
                                                        Ridley Scott
         7
                7.667400
                                      The Martian
                                                                        141.0
         33
                                  Bridge of Spies Steven Spielberg
                                                                        141.0
                3.648210
         66
                2.345821 In the Heart of the Sea
                                                          Ron Howard
                                                                        122.0
                1.007054
                                   Irrational Man
                                                         Woody Allen
                                                                         95.0
         155
                                     The Audition
                                                    Martin Scorsese
                                                                         16.0
         572
                0.082569
                                          budget_adj
                                                        revenue_adj release_date \
              vote_count vote_average
         7
                                   7.6 9.935996e+07
                                                      5.477497e+08
                                                                      2015-09-30
                    4572
                                   7.1 3.679998e+07
         33
                    1638
                                                      1.496016e+08
                                                                      2015-10-15
                                   6.4 9.199996e+07
         66
                     805
                                                      8.631506e+07
                                                                      2015-11-20
         155
                     319
                                   6.1 1.012000e+07
                                                      2.519979e+07
                                                                      2015-07-17
         572
                      10
                                   6.1
                                                 NaN
                                                                      2015-10-27
                                                                NaN
              Robert_De_Niro Samuel_L_Jackson woman_director independent_film
         7
                           0
                                             0
                                                                                0
         33
                           0
                                             0
                                                              0
                                                                                0
         66
                           0
                                             0
                                                              0
                                                                                0
         155
                           0
                                             0
                                                              0
                                                                                0
```

```
Comedy
                              Thriller Action Romance
                                                           Horror
                                                                    Adventure
              Drama
         7
                   1
                           0
                                      0
                                               0
                                                        0
                                                                             1
         33
                   1
                           0
                                      1
                                               0
                                                        0
                                                                 0
                                                                             0
                                                                                    0
                   1
                           0
                                      1
                                               1
                                                        0
                                                                 0
                                                                             1
                                                                                    0
         66
         155
                   1
                           0
                                      0
                                               0
                                                        0
                                                                 0
                                                                             0
                                                                                    0
         572
                           1
                                      0
                                               0
                                                        0
                                                                 0
                                                                             0
                                                                                    0
                                                            Animation Documentary
              Family
                       Science Fiction Fantasy
                                                   Mystery
                                                                                      Music
         7
                    0
                                                0
                                                         0
                                                                     0
                                                                                   0
                                      1
                                                                                           0
         33
                    0
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              History
                        War
                             Foreign
                                      TV Movie
                                                 Western Universal_Pictures
         7
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              Warner_Bros Paramount_Pictures
         7
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                         0
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         572
                                               0
In [56]: means_ = top8_directors.groupby('director')[['popularity', 'vote_average', 'budget_adj'
                                                        + genre_unique].mean().sort_values('popular
         means_
Out [56]:
                             popularity vote_average
                                                            budget_adj
                                                                          revenue_adj
         director
         Ridley Scott
                                                         7.909806e+07
                                2.009294
                                               6.478261
                                                                        1.917985e+08
         Steven Spielberg
                                1.875138
                                               6.824138
                                                         7.419896e+07
                                                                        5.630536e+08
         Ron Howard
                                1.295499
                                               6.377273
                                                         8.682048e+07
                                                                        2.515053e+08
                                                                        1.348079e+08
         Martin Scorsese
                                               6.958621
                                1.128548
                                                         5.473410e+07
         Steven Soderbergh
                                0.945138
                                               6.172727
                                                         4.068058e+07
                                                                        1.456043e+08
         Clint Eastwood
                                0.830037
                                               6.508824
                                                         4.264728e+07
                                                                        1.407573e+08
         Joel Schumacher
                                0.768102
                                               5.971429
                                                         5.768043e+07
                                                                        1.309330e+08
         Woody Allen
                                               6.44444
                                                        1.848060e+07
                                                                        5.882820e+07
                                0.563424
                                 Drama
                                          Comedy Thriller
                                                                Action
                                                                          Romance
                                                                                     Horror \
         director
         Ridley Scott
                             0.739130  0.086957  0.391304  0.521739  0.130435  0.086957
```

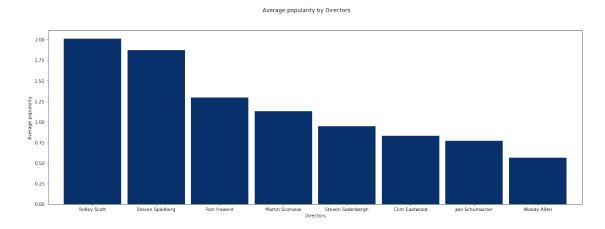
Steven Spielberg Ron Howard Martin Scorsese Steven Soderbergh Clint Eastwood Joel Schumacher Woody Allen	0.517241 0.636364 0.793103 0.772727 0.794118 0.666667 0.533333	0.103448 0.409091 0.137931 0.227273 0.117647 0.190476 0.844444	0.272727 0 0.206897 0 0.363636 0 0.323529 0 0.619048 0	0.227273 0.1 0.000000 0.1 0.045455 0.2 0.323529 0.1 0.190476 0.1	0.068966 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
	Adventure	Crime	Family	Science Fict	ion Fantasy \
director			J		, ,
Ridley Scott	0.347826	0.304348	0.000000	0.173	3913 0.043478
Steven Spielberg	0.413793	0.068966	0.103448	0.241	
Ron Howard	0.181818	0.045455	0.090909	0.045	5455 0.090909
Martin Scorsese	0.034483	0.310345	0.034483	0.000	0.000000
Steven Soderbergh	0.000000	0.409091	0.000000	0.090	909 0.000000
Clint Eastwood	0.147059	0.382353	0.000000	0.029	9412 0.029412
Joel Schumacher	0.047619	0.428571	0.047619	0.190	0.095238
Woody Allen	0.000000	0.133333	0.000000	0.022	2222 0.044444
	Mystery	Animation	Documentar	ry Music	History \
director					
Ridley Scott	0.043478	0.000000	0.00000	0.000000	0.130435
Steven Spielberg	0.137931	0.034483	0.00000	0.000000	0.172414
Ron Howard	0.136364	0.000000	0.00000	0.000000	0.090909
Martin Scorsese	0.034483	0.00000	0.13793	31 0.172414	0.034483
Steven Soderbergh	0.181818	0.000000	0.00000	0.000000	0.090909
Clint Eastwood	0.147059	0.000000	0.00000	0.088235	0.088235
Joel Schumacher	0.142857	0.000000	0.00000	0.000000	0.00000
Woody Allen	0.088889	0.000000	0.00000	0.000000	0.022222
	War	Foreign :	TV Movie V	Western	
director					
Ridley Scott	0.130435	0.0	0.0 0.	.000000	
Steven Spielberg	0.172414	0.0	0.0 0.	.000000	
Ron Howard	0.000000	0.0	0.0 0.	.090909	
Martin Scorsese	0.000000	0.0	0.0 0.	.000000	
Steven Soderbergh	0.136364	0.0	0.0 0.	.000000	
Clint Eastwood	0.117647	0.0	0.0 0.	. 176471	
Joel Schumacher	0.047619	0.0	0.0 0.	.000000	
Woody Allen	0.000000	0.0	0.0 0.	.000000	

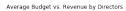
The most popular genres are Drama, Action, and Crime.

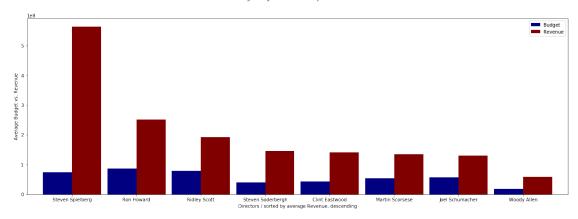
Share of movies by Genre by Directors



Movies of Ridley Scott have the highest average popularity rate (2.01), compared to the next Steven Spielberg's movies with average popularity rate equals 1.88.







Average ROI 246.324639484

Out[60]: director

Steven Spielberg	658.842923
Steven Soderbergh	257.920973
Clint Eastwood	230.049812
Woody Allen	218.324059
Ron Howard	189.684298
Martin Scorsese	146.295994
Ridley Scott	142.481863
Joel Schumacher	126.997194
dtype: float64	

Finally, average ROI of movies of directors from these group is 246%. Moreover, ROI of movies directed by Steven Spielberg equals 658%.

Conclusions

In this project TMDb Movie Data was analyzed. This dataset contains information about 10,000 movies, including user ratings and revenue.

There're 2 main parts in this Jupyter notebook:

- 1. Data Wrangling,
- 2. Exploratory Data Analysis.

First part consists of the following blocks:

- * General Properties,
- * Data Cleaning,
- * Data visualization.

In the second part the following research questions were considered:

1. Which genres are most popular from year to year? Most of the popularity ratings are less than

1. The distribution of popularity is right skewed with maximum rating 32.99 of movie "Jurassic World" (2015).

From 1960 to 2015 the number of movies increased with the corresponding increase of average opularity rate. However, some movies became popular since 20 century. For example, "The Godfather" (1972) has 5.74 popularity rate.

Among all 10 thound movies, Adventure is the most popular genre with average popularity about 1.2.

The most popular genres by popularity rate during the whole period is Science Fiction, Thriller, and Wars. Moreover, during the 20 century Romance and TV movies were also as popular as Science Fiction genre. In the 21st century Western genre has reached leading positions.

If we look only on top-3 genres of each year, Thriller and war genres bagin to compete for first place with each other from year to year.

- 2. How sharp is the divide between major film studios and the independents? Top-2 production companies are:
- -> Universal Pictures (produced 522 movies),
- -> Warner Bros. (produced 509 movies).

Universal Pictures made only 1 independent film (it's about 0.19% of all movies made by the company), and Warner Bros. made 2 independent films (it's about 0.32% of all movies made by the company). In comparison, 393 movies of the other companies were independent (it's about 3.67% of all movies).

So, why major film studios don't made independent?

Popularity rates of independent movies are twice lower (0.65 vs. 0.31, or 0.96 vs. 0.33) than not-independent ones.

If we don't take into account missing data in revenue and budget columns, the result for independent movies is more than twice worse. Despite average votes for independent and not-independent movies are approximately the same (6 points), number of votes for independent movies is more then 5 times smaller (200-500 vs. 20-40).

So, producing companies don't make independent movies because revenues from them are lower, or revenues from independent movies are lower because producing comanies don't make them? It's really interesting question for the further investigation and analysis.

3. What kinds of properties are associated with movies that have high revenues?

To answer this question, only numeric columns were considered.

The relationship between revenue and popularity, revenue and number of votes, revenue and budget is positive.

Some outliers are present in the dataset, so:

- ~ 3 movies with the highest popularity made on average \\$771099276.71 (Jurassic World (2015), Interstellar (2014), Mad Max: Fury Road (2015)),
- ~ 4 movies with the number of votes higher than 8000 gathered on average, \\$1527637054.52 (Avatar (2009), The Avengers (2012), The Dark Knight (2008), Inception (2010)),
- \sim 3 movies with the highest budget made on average \\$670719525.753 (Pirates of the Caribbean: At World's End (2001), Pirates of the Caribbean: On Stranger Tides (2011), The Warrior's Way (2010)).

Movies of Ridley Scott have the highest average popularity rate (2.01), compared to the next

^{4.} Top-8 directors made more than 20 movies. Whether their movies are also popular or profitable than movies of the other directos?

Steven Spielberg's movies with average popularity rate equals 1.88.

The most popular genres are Drama, Action, and Crime.

Finally, average ROI of movies of directors from these group is 246%. Moreover, ROI of movies directed by Steven Spielberg equals 658%.

It's worth to add, this analysis has limitations due to constraints on data that impact the findings of the conclusion.

First of all, there's a lot of missing data in revenue and budget columns. Therefore, we can't say exactly that whether independent movies are less profitable than movies mase by major film studios. The same is true for comparing film and making conclusions about the most expensive movies (based on budgets column) and their revenues. Lastly, we can't campare film revenues, budgets, and ROI of directors who made more than 20 movies due to mising data.

One way is to keep only movies with not-zero budget and revenue values. The other way is to find missing data from the other sources in order not to lose important information.