

TMDB Movie Analysis

April 24, 2020

1 Project: TMDB (The Movie Database) Analysis

The primary objective of this project is to explore the 'TMDB Dataset' ([<https://www.kaggle.com/tmdb/tmdb-movie-metadata>]) by conducting suitable data-analysis using various tools and libraries like Jupyter Notebooks, IPython, Numpy, Pandas and Matplotlib. In order to make the findings more interpretable, project is further divided into the following four sections:

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1.2 Introduction

TMDD-Movie Dataset was generated by 'The Movie Database API' ([<https://www.kaggle.com/tmdb/themoviedb.org>]) which contains metadata on ~5000 movies including cast, crew, budget and revenue of each film. Using this data, we can start digging in to questions like:

- What are the highest grossing films of all time?
- Which day of week has most number of releases?
- How does popularity influence revenue of films?

```
In [102]: # Importing required libraries like Numpy, Pandas and Matplotlib
import numpy as np
```

```
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
```

Data Wrangling

1.2.1 Wrangling and munging data from its raw format into a suitable form appropriate for analysis.

```
In [103]: # Loading the data into the DataFrame using pd.read_csv() function
# There are 21 Columns and 10,866 entries in this dataset
df = pd.read_csv('tmdb-movies.csv', header=0, parse_dates=True)
df.head()
```

```
Out[103]:
```

	id	imdb_id	popularity	budget	revenue	\
0	135397	tt0369610	32.985763	150000000	1513528810	
1	76341	tt1392190	28.419936	150000000	378436354	
2	262500	tt2908446	13.112507	110000000	295238201	
3	140607	tt2488496	11.173104	200000000	2068178225	
4	168259	tt2820852	9.335014	190000000	1506249360	

	original_title	\
0	Jurassic World	
1	Mad Max: Fury Road	
2	Insurgent	
3	Star Wars: The Force Awakens	
4	Furious 7	

	cast	\
0	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi...	
1	Tom Hardy Charlize Theron Hugh Keays-Byrne Nic...	
2	Shailene Woodley Theo James Kate Winslet Ansel...	
3	Harrison Ford Mark Hamill Carrie Fisher Adam D...	
4	Vin Diesel Paul Walker Jason Statham Michelle ...	

	homepage	director	\
0	http://www.jurassicworld.com/	Colin Trevorrow	
1	http://www.madmaxmovie.com/	George Miller	
2	http://www.thedivergentseries.movie/#insurgent	Robert Schwentke	
3	http://www.starwars.com/films/star-wars-episod...	J.J. Abrams	
4	http://www.furious7.com/	James Wan	

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

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

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

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

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

[5 rows x 21 columns]

```
In [104]: # Checking the data-types of various columns using df.dtypes function
# Few datatypes are inconsistent like release_date and release_year
df.dtypes
```

```
Out[104]: id                int64
imdb_id                    object
popularity                 float64
budget                    int64
revenue                   int64
original_title             object
cast                      object
homepage                  object
director                  object
tagline                   object
keywords                  object
overview                  object
runtime                   int64
```

```

genres                object
production_companies  object
release_date          object
vote_count            int64
vote_average          float64
release_year          int64
budget_adj            float64
revenue_adj           float64
dtype: object

```

```

In [105]: # Checking the shape of the dataset using the df.shape function
          # Some columns are unnecessary for our analysis, should be excluded
          df.shape

```

```

Out[105]: (10866, 21)

```

```

In [106]: # Printing summary of entire dataset using df.info() function
          # There are missing values in few columns, must be repaired
          df.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10866 entries, 0 to 10865
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  -
0   id                    10866 non-null  int64
1   imdb_id              10856 non-null  object
2   popularity            10866 non-null  float64
3   budget               10866 non-null  int64
4   revenue              10866 non-null  int64
5   original_title        10866 non-null  object
6   cast                 10790 non-null  object
7   homepage             2936 non-null   object
8   director             10822 non-null  object
9   tagline              8042 non-null   object
10  keywords             9373 non-null   object
11  overview             10862 non-null  object
12  runtime              10866 non-null  int64
13  genres               10843 non-null  object
14  production_companies  9836 non-null   object
15  release_date         10866 non-null  object
16  vote_count           10866 non-null  int64
17  vote_average         10866 non-null  float64
18  release_year         10866 non-null  int64
19  budget_adj           10866 non-null  float64
20  revenue_adj          10866 non-null  float64
dtypes: float64(4), int64(6), object(11)
memory usage: 1.7+ MB

```

```
In [107]: # Generating descriptive statistics using df.describe() function
# Summarizes shape, central tendency and dispersion
df.describe()
```

```
Out[107]:
```

	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

```
In [108]: # Checking for unique values using df.nunique() function
# There are multiple unique values in columns like Cast, Genre
df.nunique()
```

```
Out[108]:
```

id	10865
imdb_id	10855
popularity	10814
budget	557
revenue	4702
original_title	10571
cast	10719
homepage	2896
director	5067
tagline	7997
keywords	8804
overview	10847
runtime	247
genres	2039
production_companies	7445
release_date	5909
vote_count	1289
vote_average	72
release_year	56
budget_adj	2614

```
revenue_adj          4840
dtype: int64
```

1.3 Data Cleaning

1.3.1 Detecting and cleaning data by getting rid of various inconsistencies or errors before reporting.

```
In [109]: # Dropping unnecessary columns using df.drop() function
df = df.drop(['id', 'imdb_id', 'homepage', 'tagline', 'keywords', 'overview', 'budget_
```

```
In [110]: # Dropping duplicate entries using drop_duplicates() function
df.duplicated().sum()
df.drop_duplicates()
```

```
Out[110]:
```

	popularity	budget	revenue	original_title \
0	32.985763	150000000	1513528810	Jurassic World
1	28.419936	150000000	378436354	Mad Max: Fury Road
2	13.112507	110000000	295238201	Insurgent
3	11.173104	200000000	2068178225	Star Wars: The Force Awakens
4	9.335014	190000000	1506249360	Furious 7
...
10861	0.080598	0	0	The Endless Summer
10862	0.065543	0	0	Grand Prix
10863	0.065141	0	0	Beregis Avtomobilya
10864	0.064317	0	0	What's Up, Tiger Lily?
10865	0.035919	19000	0	Manos: The Hands of Fate

	cast	director \
0	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
3	Harrison Ford Mark Hamill Carrie Fisher Adam D...	J.J. Abrams
4	Vin Diesel Paul Walker Jason Statham Michelle ...	James Wan
...
10861	Michael Hynson Robert August Lord 'Tally Ho' B...	Bruce Brown
10862	James Garner Eva Marie Saint Yves Montand Tosh...	John Frankenheimer
10863	Innokentiy Smoktunovskiy Oleg Efremov Georgi Z...	Eldar Ryazanov
10864	Tatsuya Mihashi Akiko Wakabayashi Mie Hama Joh...	Woody Allen
10865	Harold P. Warren Tom Neyman John Reynolds Dian...	Harold P. Warren

	runtime	genres \
0	124	Action Adventure Science Fiction Thriller
1	120	Action Adventure Science Fiction Thriller
2	119	Adventure Science Fiction Thriller
3	136	Action Adventure Science Fiction Fantasy
4	137	Action Crime Thriller
...
10861	95	Documentary

10862	176	Action Adventure Drama
10863	94	Mystery Comedy
10864	80	Action Comedy
10865	74	Horror

	production_companies	release_date	\
0	Universal Studios Amblin Entertainment Legenda...	6/9/15	
1	Village Roadshow Pictures Kennedy Miller Produ...	5/13/15	
2	Summit Entertainment Mandeville Films Red Wago...	3/18/15	
3	Lucasfilm Truenorth Productions Bad Robot	12/15/15	
4	Universal Pictures Original Film Media Rights ...	4/1/15	
...	
10861	Bruce Brown Films	6/15/66	
10862	Cherokee Productions Joel Productions Douglas ...	12/21/66	
10863	Mosfilm	1/1/66	
10864	Benedict Pictures Corp.	11/2/66	
10865	Norm-Iris	11/15/66	

	vote_count	vote_average	release_year
0	5562	6.5	2015
1	6185	7.1	2015
2	2480	6.3	2015
3	5292	7.5	2015
4	2947	7.3	2015
...
10861	11	7.4	1966
10862	20	5.7	1966
10863	11	6.5	1966
10864	22	5.4	1966
10865	15	1.5	1966

[10865 rows x 13 columns]

In [111]: df.isnull().sum()

```
Out[111]: popularity      0
budget      0
revenue      0
original_title      0
cast      76
director      44
runtime      0
genres      23
production_companies      1030
release_date      0
vote_count      0
vote_average      0
release_year      0
dtype: int64
```

```
In [112]: # Checking missing values across columns using df.isnull() function
df.isnull().sum()
# Filling missing values for budget/revenue with with df.fillna() function
df['budget'].fillna(df['budget'].mean(), inplace=True)
df['revenue'].fillna(df['revenue'].mean(), inplace=True)
# Dropping null/NaN values in rows using df.dropna() function
df.dropna(axis=0, inplace=True)
```

```
In [113]: # Converting datatypes of non-date columns using pd.to_datetime()
df['release_date'] = pd.to_datetime(df['release_date'])
df['release_year'] = pd.DatetimeIndex(df['release_date']).year
df.head()
```

```
Out[113]:
```

	popularity	budget	revenue	original_title \
0	32.985763	150000000	1513528810	Jurassic World
1	28.419936	150000000	378436354	Mad Max: Fury Road
2	13.112507	110000000	295238201	Insurgent
3	11.173104	200000000	2068178225	Star Wars: The Force Awakens
4	9.335014	190000000	1506249360	Furious 7

	cast	director \
0	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
3	Harrison Ford Mark Hamill Carrie Fisher Adam D...	J.J. Abrams
4	Vin Diesel Paul Walker Jason Statham Michelle ...	James Wan

	runtime	genres \
0	124	Action Adventure Science Fiction Thriller
1	120	Action Adventure Science Fiction Thriller
2	119	Adventure Science Fiction Thriller
3	136	Action Adventure Science Fiction Fantasy
4	137	Action Crime Thriller

	production_companies	release_date	vote_count \
0	Universal Studios Amblin Entertainment Legenda...	2015-06-09	5562
1	Village Roadshow Pictures Kennedy Miller Produ...	2015-05-13	6185
2	Summit Entertainment Mandeville Films Red Wago...	2015-03-18	2480
3	Lucasfilm Truenorth Productions Bad Robot	2015-12-15	5292
4	Universal Pictures Original Film Media Rights ...	2015-04-01	2947

	vote_average	release_year
0	6.5	1970
1	7.1	1970
2	6.3	1970
3	7.5	1970
4	7.3	1970

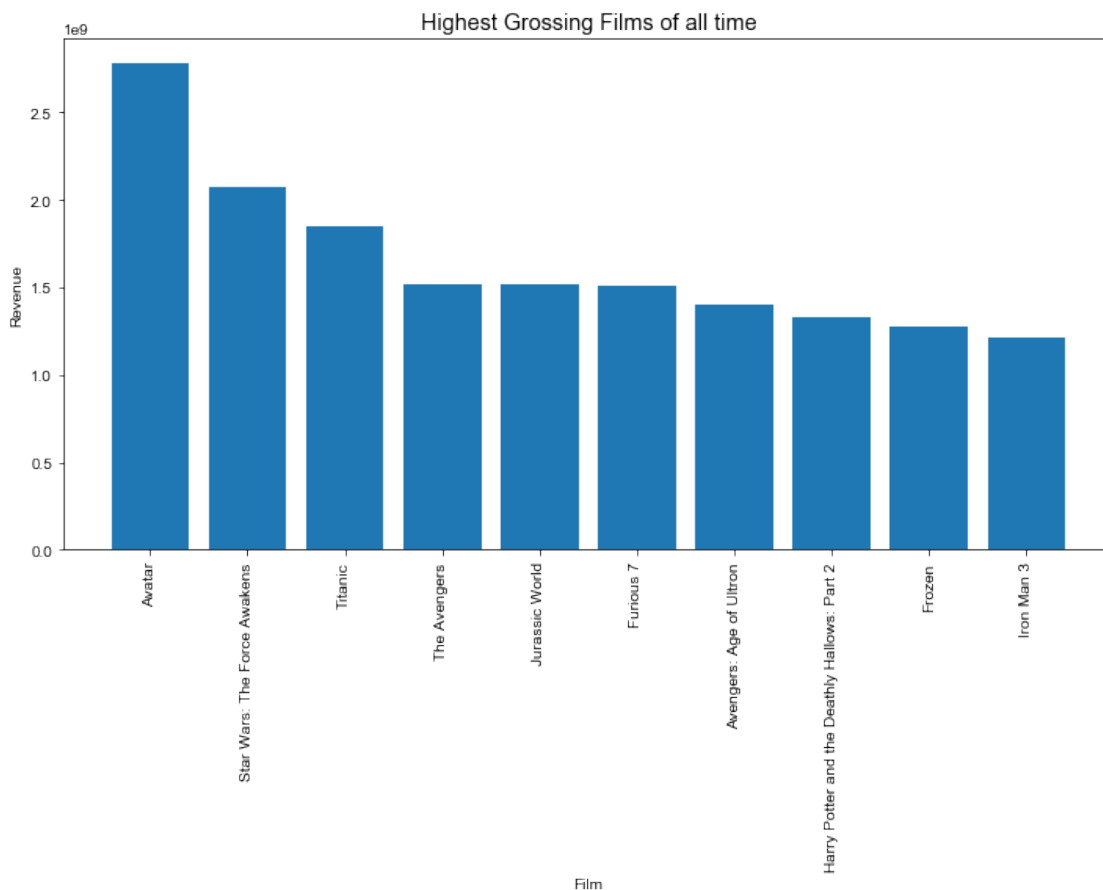
Exploratory Data Analysis

Analyzing the dataset by summarising its main characteristics. Including, choice of various of graphical techniques that will aid in bird's eye view of the data to make some sense of it. Answering critical research questions along the way to supplement our findings.

Question 1: What are the highest grossing films of all time?

Avatar(2.78B) was highest grossing film of all time, followed by Star Wars(2.06B) and Titanic(1.84B).

```
In [114]: fig, ax = plt.subplots(figsize = (12,6))
plt.rcParams['font.sans-serif'] = 'Arial'
plt.rcParams['font.family'] = 'sans-serif'
plt.rcParams['font.size']=12
high_gross = df.sort_values(by='revenue', ascending=False)[:10]
plt.bar(high_gross['original_title'], high_gross['revenue'])
plt.xlabel('Film')
plt.ylabel('Revenue')
plt.title('Highest Grossing Films of all time')
plt.xticks(rotation=90)
plt.show()
```

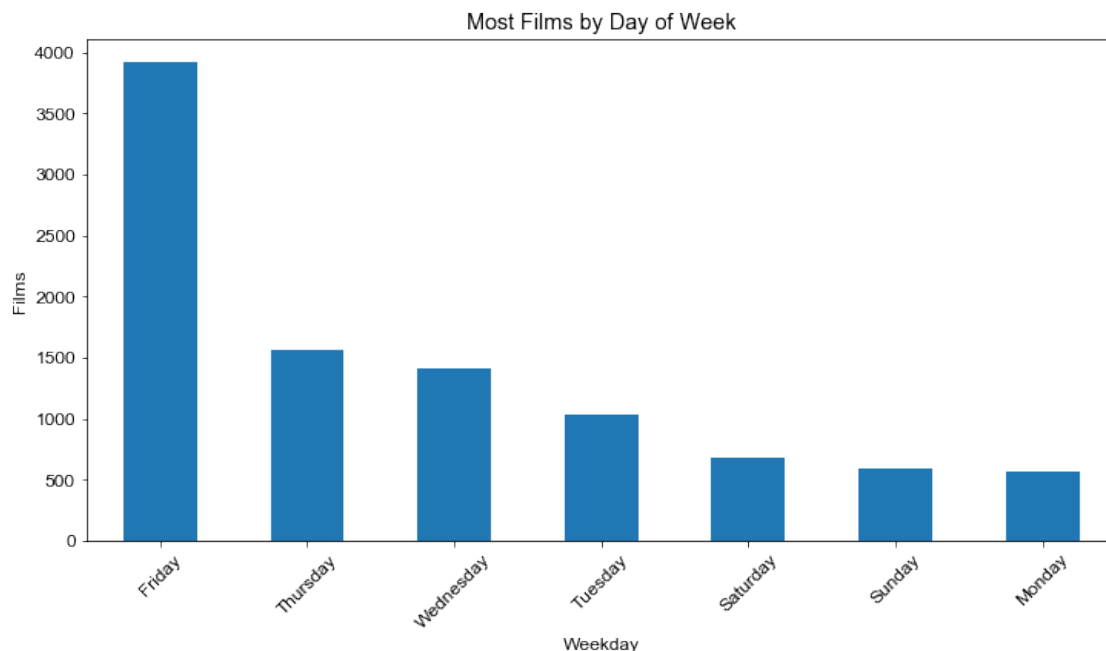


Question 2: Which day of the week has most number of releases?

Friday seems to have the most number of releases, followed by Thursday and Wednesday.

```
In [115]: fig, ax = plt.subplots(figsize = (12,6))
plt.rcParams['font.sans-serif'] = 'Arial'
plt.rcParams['font.family'] = 'sans-serif'
plt.rcParams['font.size']=12
day_week = df['release_date'].dt.day_name()
day_week.value_counts().plot.bar()
plt.xticks(rotation=45)
plt.xlabel('Weekday')
plt.ylabel('Films')
plt.title('Most Films by Day of Week')
```

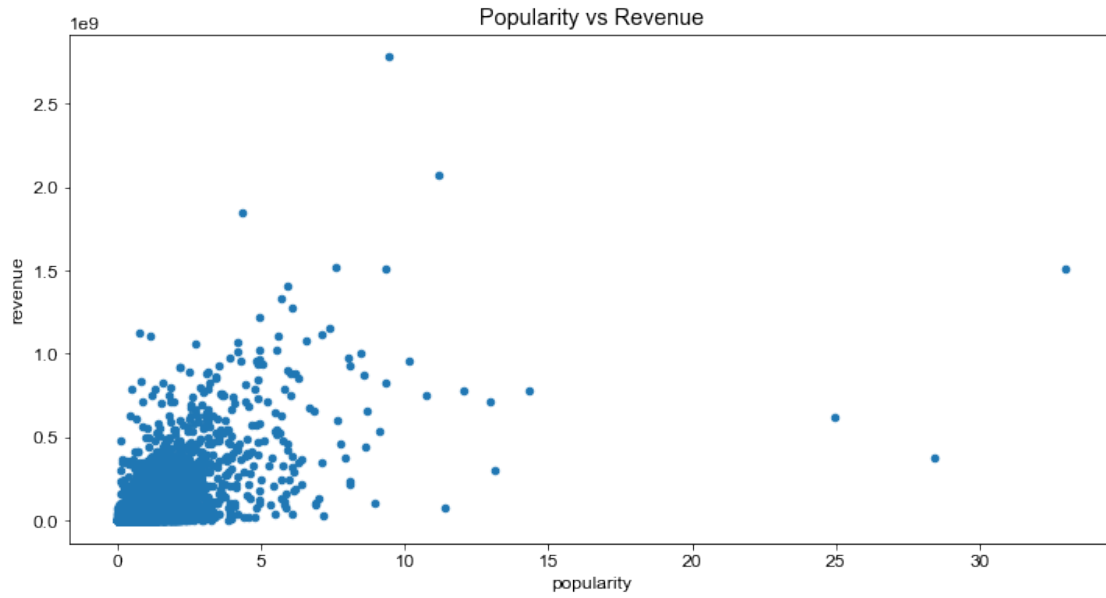
```
Out[115]: Text(0.5, 1.0, 'Most Films by Day of Week')
```



Question 3: How does popularity influence revenue of a film?

General positive correlation among popularity and revenue with few outliers.

```
In [116]: plt.rcParams['font.sans-serif'] = 'Arial'
plt.rcParams['font.family'] = 'sans-serif'
plt.rcParams['font.size']=12
df.plot(x='popularity', y='revenue', kind='scatter', figsize=(12,6))
plt.title('Popularity vs Revenue')
plt.show()
```

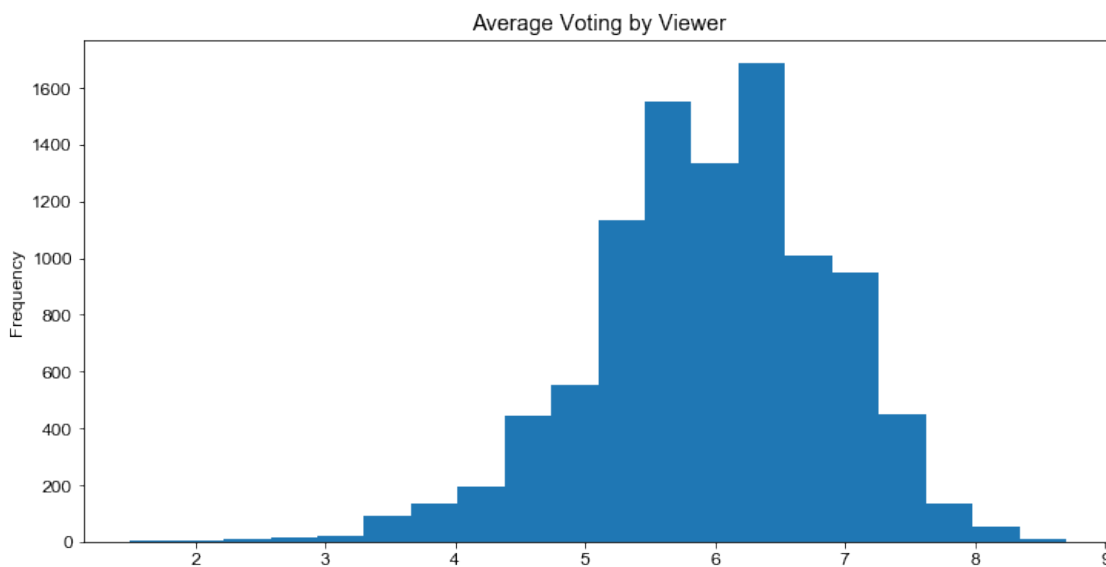


Question 4: What does average voting among films look like?

In general viewers have rated the films in following range, with very few ratings above 8.

```
In [117]: plt.rcParams['font.sans-serif'] = 'Arial'
plt.rcParams['font.family'] = 'sans-serif'
plt.rcParams['font.size']=12
df.vote_average.plot(kind='hist', bins = 20, figsize = (12,6), title='Average Voting b
```

```
Out[117]: <matplotlib.axes._subplots.AxesSubplot at 0x11ec2f890>
```



Conclusions

Some of the meaningful insights extracted from the above analysis of data include:

- Avatar is highest grossing film of all time, followed by Star Wars: Force Awakens and Titanic.
- Friday saw the most number of film releases, followed by Thursday based on the day of week.
- Popularity has a positive affect on the success of a given film or its commercial revenue.
- Average vote among various films is around 6.0, with very few films having vote higher than 8.0.

Limitations

- Extracted data may be incomplete, as there are several missing values. This affects its usability. (For example, missing values in columns like budget/revenue are replaced with their means).
- Data collected above varies in quality and format. Many unique values appear in individual columns. (For example, cast/genres columns have lots of unique values seperated by '|' operator).
- Such data needs lot of preprocessing before its ready for analysis. There may be incorrect data types. (For example, release_date/release_year had to be transformed before using them in analysis).
- Such inconsistencies, must be treated with caution until making important decisions as it may produce biases.

References

1. Python Plotting Basics [<https://towardsdatascience.com/python-plotting-basics-simple-charts-with-matplotlib-seaborn-and-plotly-e36346952a3a>]
2. Quick dive into Pandas [<https://towardsdatascience.com/quick-dive-into-pandas-for-data-science-cc1c1a80d9c4>]
3. Data Wrangling with Pandas [<https://towardsdatascience.com/data-wrangling-with-pandas-5b0be151df4e>]