investigate-movie-dataset

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1 Project: Investigate TMDb Movie Data

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Introduction

This data set contains information about 10,000 movies released between 1960 and 2015. Collected from The Movie Database (TMDb), it also includes user ratings and popularities. This report aims at conducting exploratory data analysis to answer the questions below:

- 1. What are the average popularities of movies according to budget levels?
- 2. What are the profit trends of movies from year to year?
- 3. What are the average runtimes of movies over the years?
- 4. Which are the 5 cheapest and most expensive profitable movies of all time?

Key notes: In this section of the report, the following work will be done: load the data; check for cleanliness; trim and clean dataset for analysis.

1.1.1 General Properties

In [21]: # Load your data and print out a few lines. Perform operations to inspect data

```
df.head()
Out [21]:
                id
                       imdb_id
                                popularity
                                                budget
                                                           revenue
            135397
                    tt0369610
                                 32.985763
                                             150000000
                                                        1513528810
         1
             76341
                    tt1392190
                                 28.419936
                                             150000000
                                                         378436354
         2
           262500
                    tt2908446
                                 13.112507
                                             110000000
                                                         295238201
                                 11.173104
         3
           140607
                    tt2488496
                                             200000000
                                                        2068178225
           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 \
            Chris Pratt|Bryce Dallas Howard|Irrfan Khan|Vi...
            Tom Hardy | Charlize Theron | Hugh Keays-Byrne | Nic...
         1
         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
                                   http://www.madmaxmovie.com/
         1
                                                                     George Miller
         2
               http://www.thedivergentseries.movie/#insurgent
                                                                 Robert Schwentke
            http://www.starwars.com/films/star-wars-episod...
                                                                       J.J. Abrams
         3
         4
                                      http://www.furious7.com/
                                                                         James Wan
                                   tagline
         0
                         The park is open.
         1
                        What a Lovely Day.
         2
               One Choice Can Destroy You
            Every generation has a story.
         3
                                                 . . .
                       Vengeance Hits Home
                                                       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
         3 Thirty years after defeating the Galactic Empi...
                                                                     136
         4 Deckard Shaw seeks revenge against Dominic Tor...
                                                                     137
                                                 genres \
```

```
Action | Adventure | Science Fiction | Thriller
            Action | Adventure | Science Fiction | Thriller
                    Adventure | Science Fiction | Thriller
         2
         3
             Action|Adventure|Science Fiction|Fantasy
         4
                                 Action | Crime | Thriller
                                          production_companies release_date vote_count \
         O Universal Studios | Amblin Entertainment | Legenda...
                                                                       6/9/15
                                                                                     5562
           Village Roadshow Pictures | Kennedy Miller Produ...
                                                                      5/13/15
                                                                                     6185
         2 Summit Entertainment | Mandeville Films | Red Wago...
                                                                      3/18/15
                                                                                     2480
                    Lucasfilm|Truenorth Productions|Bad Robot
                                                                     12/15/15
         3
                                                                                     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
                     7.1
         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 [22]: # return a tuple of the dimensions of the dataframe
         df.shape
Out [22]: (10866, 21)
In [23]: # print the column labels in the dataframe
         for i, v in enumerate(df.columns):
             print(i, v)
0 id
1 imdb_id
2 popularity
3 budget
4 revenue
5 original_title
6 cast
7 homepage
8 director
9 tagline
10 keywords
11 overview
12 runtime
13 genres
14 production_companies
15 release_date
```

```
16 vote_count
17 vote_average
18 release_year
19 budget_adj
20 revenue_adj
In [24]: # return the datatypes of the columns
         df.dtypes
Out[24]: id
                                   int64
         imdb id
                                  object
         popularity
                                 float64
         budget
                                   int64
                                   int64
         revenue
         original_title
                                  object
                                  object
         cast
                                  object
         homepage
         director
                                  object
         tagline
                                  object
         keywords
                                  object
         overview
                                  object
                                   int64
         runtime
         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 [25]: # check for duplicates in the data
         sum(df.duplicated())
Out[25]: 1
In [26]: # check if any value is NaN in DataFrame and in how many columns
         df.isnull().any().any(), sum(df.isnull().any())
Out[26]: (True, 9)
In [27]: # displays a concise summary of the dataframe
         # including the number of non-null values in each column
         df.info()
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 10866 entries, 0 to 10865 Data columns (total 21 columns): id 10866 non-null int64 10856 non-null object imdb id popularity 10866 non-null float64 10866 non-null int64 budget 10866 non-null int64 revenue original_title 10866 non-null object 10790 non-null object cast 2936 non-null object homepage director 10822 non-null object 8042 non-null object tagline 9373 non-null object keywords 10862 non-null object overview runtime 10866 non-null int64 genres 10843 non-null object 9836 non-null object production_companies release_date 10866 non-null object 10866 non-null int64 vote_count

dtypes: float64(4), int64(6), object(11)

memory usage: 1.7+ MB

vote_average
release year

budget_adj

revenue_adj

In [28]: # Generates descriptive statistics, excluding NaN values

10866 non-null float64

10866 non-null float64

10866 non-null int64 10866 non-null float64

df.describe()

Out[28]:		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	

```
1.1.2 Data Cleaning
In [29]: # drop duplicates
         # confirm correction
         df.drop_duplicates(inplace=True)
         sum(df.duplicated())
Out[29]: 0
In [30]: # list of columns that are to be deleted/dropped
         col = ['id', 'imdb_id', 'budget_adj', 'revenue_adj', 'homepage', 'tagline', 'keyword
         # deleting the columns
         df.drop(col, axis = 1, inplace = True)
         #checking to see if the columns have been deleted
         df.head()
Out [30]:
            popularity
                            budget
                                        revenue
                                                                original_title \
                                                                Jurassic World
             32.985763
                        150000000
                                    1513528810
             28.419936 150000000
                                                            Mad Max: Fury Road
         1
                                      378436354
         2
             13.112507 110000000
                                      295238201
                                                                     Insurgent
                                                 Star Wars: The Force Awakens
             11.173104 200000000
         3
                                    2068178225
                                                                     Furious 7
              9.335014 190000000
                                    1506249360
                                                                           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
         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
                      Action | Adventure | Science Fiction | Thriller
                124
                      Action | Adventure | Science Fiction | Thriller
         1
                120
         2
                119
                             Adventure | Science Fiction | Thriller
         3
                       Action|Adventure|Science Fiction|Fantasy
                136
                                           Action | Crime | Thriller
         4
                137
                                           production_companies release_date
                                                                                vote count
         O 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
```

50%

75%

max

38,000000

145.750000

9767.000000

6.000000

6.600000

9.200000

2006.000000 0.000000e+00 0.000000e+00

2015.000000 4.250000e+08 2.827124e+09

3.369710e+07

2011.000000 2.085325e+07

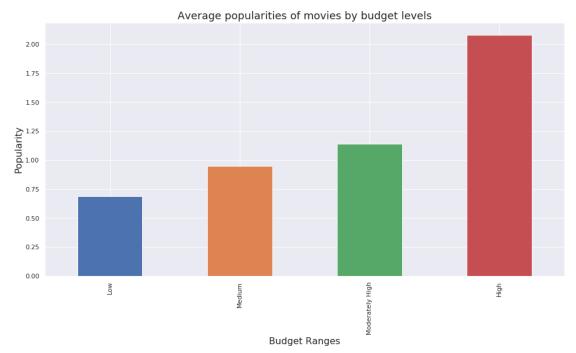
```
4 Universal Pictures | Original Film | Media Rights ...
                                                                    4/1/15
                                                                                   2947
            vote_average release_year
         0
                     6.5
                                  2015
                     7.1
         1
                                  2015
         2
                     6.3
                                  2015
         3
                     7.5
                                  2015
         4
                     7.3
                                  2015
In [31]: # Changing datatype of `release_date` column
         df['release_date'] = pd.to_datetime(df['release_date'])
In [32]: # check if the change has taken place successfully
         df.dtypes
                                        float64
Out[32]: popularity
                                          int64
         budget
         revenue
                                          int64
         original_title
                                         object
                                         object
         cast
         director
                                         object
         runtime
                                          int64
                                         object
         genres
         production_companies
                                         object
         release_date
                          datetime64[ns]
         vote_count
                                          int64
         vote_average
                                        float64
         release_year
                                          int64
         dtype: object
In [33]: # Handling O values in `budget`, `revenue` and `runtime` columns
         # Making a list of the 3 columns
         temp_col = ['budget', 'revenue', 'runtime']
         # Replacing all the O values with NaN
         df[temp_col] = df[temp_col].replace(0, np.NAN)
In [34]: # Dropping/Deleting all the NaN values
         # Subset helps to define in which columns to look for missing values
         df.dropna(subset = temp_col, inplace = True)
         rows, col = df.shape
```

1.2 Exploratory Data Analysis

Let's start with the exploration! :)

1.2.1 1. What are the average popularities of movies according to budget levels?

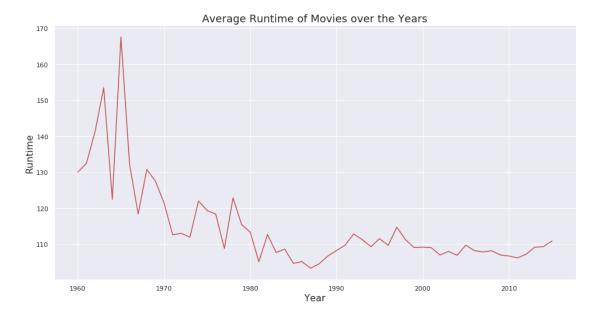
```
In [36]: # Now we find out the average popularities of each level
         df.groupby('budget_ranges')['popularity'].mean()
Out[36]: budget_ranges
         Low
                            0.686413
         Medium
                            0.951718
         Moderately High
                            1.142414
         High
                            2.080911
         Name: popularity, dtype: float64
In [37]: # Plotting the above information in a bar graph
         sns.set()
         df.groupby('budget_ranges')['popularity'].mean().plot(kind = 'bar', figsize = (16, 8)
         # Setting the title of the plot
         plt.title('Average popularities of movies by budget levels', fontsize = 18)
         # Setting the x and y axis labels
         plt.xlabel('Budget Ranges', fontsize = 16)
         plt.ylabel('Popularity', fontsize = 16);
```



1.2.2 2. What are profit trends from year to year?



1.2.3 3. What are the average runtimes of movies over the years?



1.2.4 4. Which are the 5 cheapest and most expensive profitable movies of all time?

For this we shall set a standard value of profit which has to be met, that value will be \$50,000,000

```
In [41]: # creating a list of columns that will be viewed
         col = ['original_title', 'cast', 'director', 'budget', 'revenue', 'profit_loss']
         # Using query function to show records of movies which have a profit of more than $50
         # Also using sort_values function to make sure it is sorted according to the budget c
         df.query('profit_loss>50000000')[col].sort_values('budget', ascending = False).head(5
Out [41]:
                                             original_title \
              Pirates of the Caribbean: On Stranger Tides
         3375
         7387
                  Pirates of the Caribbean: At World's End
         14
                                   Avengers: Age of Ultron
         6570
                                          Superman Returns
         1929
                                                    Tangled
                                                             cast \
         3375
               Johnny Depp|PenÃilope Cruz|Geoffrey Rush|Ian M...
         7387
               Johnny Depp|Orlando Bloom|Keira Knightley|Geof...
         14
               Robert Downey Jr. | Chris Hemsworth | Mark Ruffalo...
               Brandon Routh | Kevin Spacey | Kate Bosworth | James...
         6570
               Zachary Levi|Mandy Moore|Donna Murphy|Ron Perl...
         1929
                                                                       profit_loss
                                director
                                               budget
                                                             revenue
         3375
                            Rob Marshall 380000000.0 1.021683e+09 6.416830e+08
         7387
                          Gore Verbinski 300000000.0 9.610000e+08 6.610000e+08
         14
                             Joss Whedon 280000000.0 1.405036e+09 1.125036e+09
         6570
                            Bryan Singer
                                          270000000.0 3.910812e+08 1.210812e+08
              Nathan Greno Byron Howard 260000000.0 5.917949e+08 3.317949e+08
         1929
In [42]: df.query('profit_loss>50000000')[col].sort_values('budget', ascending = True).head(5)
Out [42]:
                         original_title \
                The Karate Kid, Part II
         10495
         7447
                    Paranormal Activity
                The Blair Witch Project
         2449
         7057
                             Open Water
         10759
                              Halloween
                                                              cast \
         10495 Ralph Macchio | Pat Morita | Martin Kove | Charlie T...
                Katie Featherston|Micah Sloat|Mark Fredrichs|A...
         7447
```

```
2449 Heather Donahue|Michael C. Williams|Joshua Leo...
7057 Blanchard Ryan|Daniel Travis|Saul Stein|Michae...
10759 Donald Pleasence|Jamie Lee Curtis|P.J. Soles|N...
```

	director	budget	revenue	profit_loss
10495	John G. Avildsen	113.0	115103979.0	115103866.0
7447	Oren Peli	15000.0	193355800.0	193340800.0
2449	Daniel Myrick Eduardo SÃanchez	25000.0	248000000.0	247975000.0
7057	Chris Kentis	130000.0	54667954.0	54537954.0
10759	John Carpenter	300000.0	70000000.0	69700000.0

Conclusions

Question 1: It can be observed that movies with a higher budget range tend to be more popular with the audience.

Question 2: Profits have increased exponentially with each passing year especially after the beginning of the 21st century.

Question 3: The runtime of movies has decreased with each passing year. It experienced a hike during the 60s but has then steadily decreased over the years. The lowest was around 100-110 minutes. Presently, movies tend to last around the 110 minute mark.