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# P2: Investigate a Dataset (tmdb.csv)

## Introduction

### **Data Overview**

-In the tmdb.csv contains information about 10,000 movies, in which including u ser ratings and revenue.

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### **Questions:**

```
1.Which movies had highest and lowest budgets?
```

- 2.Average runtime of all the movies
- 3. Movies with most and least earned revenue.
- 4. Movies with longest and shortest runtime values.
- 5. Year of release vs revenue earned.

Whit respect to Revenue earn by movies.

- 6. Successful genres
- 7. Most frequent cast
- 8. Average Budget
- 9. Average Duration
- 10. Year of release

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# **Data wrangling**

After observing the dataset and questions for analysis we are keeping only important data and deleting unused data so we can make our calculation easy and uderstandable.

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# Obsevations from the data set

- 1. Some missing value in budeget and revenue column are zero.
- 2. No standar unit for currency is used in this dataset So, i am assuming the currency unit is dollar.

```
In [3]: tmdb_data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 10866 entries, 0 to 10865
        Data columns (total 21 columns):
        id
                                 10866 non-null int64
        imdb id
                                 10856 non-null object
        popularity
                                 10866 non-null float64
        budget
                                 10866 non-null int64
        revenue
                                 10866 non-null int64
        original_title
                                 10866 non-null object
                                 10790 non-null object
        cast
        homepage
                                 2936 non-null object
        director
                                 10822 non-null object
                                 8042 non-null object
        tagline
                                 9373 non-null object
        keywords
        overview
                                 10862 non-null object
        runtime
                                 10866 non-null int64
        genres
                                 10843 non-null object
        production companies
                                 9836 non-null object
                                 10866 non-null object
        release_date
                                 10866 non-null int64
        vote_count
        vote_average
                                 10866 non-null float64
        release_year
                                 10866 non-null int64
                                 10866 non-null float64
        budget adj
        revenue adj
                                 10866 non-null float64
        dtypes: float64(4), int64(6), object(11)
        memory usage: 1.7+ MB
```

Out[4]: (10866, 21)

In [5]: tmdb\_data.dtypes

Out[5]: id int64 imdb\_id object popularity float64 budget int64 int64 revenue original\_title object cast object object homepage director object tagline object keywords object object overview int64 runtime object genres production\_companies object release\_date object vote\_count int64 vote\_average float64 release\_year int64 budget\_adj float64 float64 revenue\_adj dtype: object

In [6]: tmdb\_data.describe()

### Out[6]:

	id	popularity	budget	revenue	runtime	vote_count	vo
count	10866.000000	10866.000000	1.086600e+04	1.086600e+04	10866.000000	10866.000000	10
mean	66064.177434	0.646441	1.462570e+07	3.982332e+07	102.070863	217.389748	
std	92130.136561	1.000185	3.091321e+07	1.170035e+08	31.381405	575.619058	
min	5.000000	0.000065	0.000000e+00	0.000000e+00	0.000000	10.000000	
25%	10596.250000	0.207583	0.000000e+00	0.000000e+00	90.000000	17.000000	
50%	20669.000000	0.383856	0.000000e+00	0.000000e+00	99.000000	38.000000	
75%	75610.000000	0.713817	1.500000e+07	2.400000e+07	111.000000	145.750000	
max	417859.000000	32.985763	4.250000e+08	2.781506e+09	900.000000	9767.000000	
4							•

```
In [7]:
           tmdb_data.isna()
Out[7]:
                                                                    original_title
                           imdb id
                                     popularity
                                                 budget revenue
                                                                                    cast homepage
                                                                                                      director
                 0 False
                              False
                                                             False
                                                                            False
                                                                                   False
                                                                                                                  F
                                          False
                                                   False
                                                                                                False
                                                                                                          False
                    False
                              False
                                          False
                                                   False
                                                             False
                                                                            False
                                                                                   False
                                                                                                False
                                                                                                          False
                                                                                                                  F
                 2
                    False
                              False
                                          False
                                                   False
                                                             False
                                                                            False
                                                                                   False
                                                                                                False
                                                                                                         False
                                                                                                                  F
                                                                                   False
                                                                                                                  F
                 3
                    False
                              False
                                          False
                                                   False
                                                             False
                                                                            False
                                                                                                False
                                                                                                          False
                    False
                              False
                                          False
                                                   False
                                                             False
                                                                            False
                                                                                   False
                                                                                                False
                                                                                                          False
                                                                                                                  F
                                                                                                   ...
            10861
                   False
                              False
                                          False
                                                   False
                                                             False
                                                                            False
                                                                                   False
                                                                                                 True
                                                                                                         False
            10862
                   False
                              False
                                          False
                                                   False
                                                             False
                                                                            False
                                                                                   False
                                                                                                 True
                                                                                                         False
                                                                                                                  F
            10863
                   False
                              False
                                          False
                                                   False
                                                             False
                                                                            False False
                                                                                                 True
                                                                                                          False
            10864
                    False
                              False
                                          False
                                                   False
                                                             False
                                                                            False
                                                                                   False
                                                                                                 True
                                                                                                          False
                                                                                                                  F
            10865 False
                                          False
                                                             False
                                                                                   False
                                                                                                 True
                                                                                                          False
                                                                                                                  F
                              False
                                                   False
                                                                            False
           10866 rows × 21 columns
```

# **Data Cleaning**

1. Check for missing data

Based on the question, observation and the missing data i am going to remove the unued data such as id, imdb\_id, homepage, production\_companies, key words, homepage.

```
In [9]: #creating a list of columb to be deleted
    del_col=['id', 'imdb_id', 'homepage', 'keywords', 'overview', 'production_compa
    nies']

#deleting the columns
    tmdb_data= tmdb_data.drop(del_col,1)

#previewing the new dataset
    tmdb_data.head(3)
```

### Out[9]:

	popularity	budget	revenue	original_title	cast	director	tagline	runtime
0	32.985763	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	The park is open.	124
1	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	What a Lovely Day.	120
2	13.112507	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	One Choice Can Destroy You	119
4								•

# 2. Remove duplicates

```
In [10]: tmdb_data.drop_duplicates(inplace=True)
sum(tmdb_data.duplicated())
```

Out[10]: 0

```
In [11]:
            tmdb data.head(3)
Out[11]:
                popularity
                               budget
                                            revenue original_title
                                                                              cast
                                                                                       director
                                                                                                tagline runtime
                                                                             Chris
                                                                        Pratt|Bryce
                                                                                                   The
                                                           Jurassic
                                                                                         Colin
                32.985763 150000000 1513528810
                                                                                                             124
                                                                            Dallas
                                                                                                 park is
                                                                                     Trevorrow
                                                             World
                                                                      Howard|Irrfan
                                                                                                  open.
                                                                         Khan|Vi...
                                                                              Tom
                                                                     Hardy|Charlize
                                                                                                 What a
                                                         Mad Max:
                                                                                       George
                28.419936
                           150000000
                                         378436354
                                                                      Theron|Hugh
                                                                                                 Lovely
                                                                                                             120
                                                        Fury Road
                                                                                         Miller
                                                                           Keays-
                                                                                                   Day.
                                                                       Byrne|Nic...
                                                                                                   One
                                                                          Shailene
                                                                                                 Choice
                                                                                        Robert
                                                                     Woodley|Theo
                13.112507
                           110000000
                                         295238201
                                                         Insurgent
                                                                                                   Can
                                                                                                              119
                                                                       James|Kate
                                                                                    Schwentke
                                                                                                Destroy
                                                                    Winslet|Ansel...
                                                                                                    You
```

# manipulating

```
In [12]: sep_list=['budget', 'revenue']
    tmdb_data[sep_list] = tmdb_data[sep_list].replace(0, np.NAN)
    tmdb_data.dropna(subset = sep_list, inplace = True)
    rows, col = tmdb_data.shape
    print('So after removing such entries, we now have only {} no.of movies.'.form
    at(rows-1))
```

So after removing such entries, we now have only 3853 no.of movies.

# 4. Changing the release date column into standard date format

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```
In [13]: tmdb_data.release_date = pd.to_datetime(tmdb_data['release_date'])
tmdb_data.head(3)
```

Out[13]:

runtir	tagline	director	cast	original_title	revenue	budget	popularity	
1	The park is open.	Colin Trevorrow	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Jurassic World	1.513529e+09	150000000.0	32.985763	0
1	What a Lovely Day.	George Miller	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	Mad Max: Fury Road	3.784364e+08	150000000.0	28.419936	1
1	One Choice Can Destroy You	Robert Schwentke	Shailene Woodley Theo James Kate Winslet Ansel	Insurgent	2.952382e+08	110000000.0	13.112507	2
•								4

# 5. Replacing zero with nan in runtime column

```
In [14]: tmdb_data['runtime'] =tmdb_data['runtime'].replace(0, np.NAN)
```

# **Exploratory Data Analysis**

1. Which movies had highest and lowest budgets?

```
In [15]: #defining the function
    def calculate(column):
        #for highest earned profit
        high= tmdb_data[column].idxmax()
        high_details=pd.DataFrame(tmdb_data.loc[high])

    #for lowest earned profit
    low= tmdb_data[column].idxmin()
    low_details=pd.DataFrame(tmdb_data.loc[low])

    #collectin data in one place
    info=pd.concat([high_details, low_details], axis=1)

    return info
    calculate('budget')
```

### Out[15]:

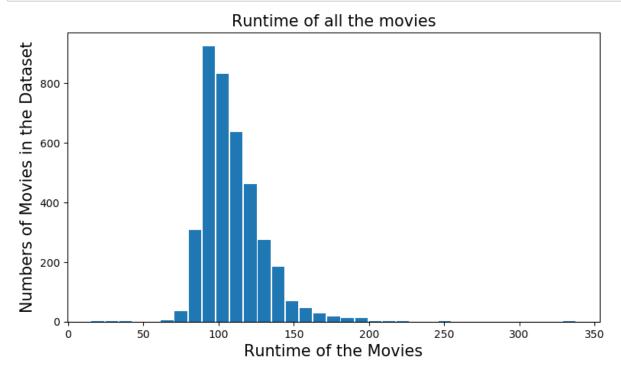
	2244	2618
popularity	0.25054	0.090186
budget	4.25e+08	1
revenue	1.10876e+07	100
original_title	The Warrior's Way	Lost & Found
cast	Kate Bosworth Jang Dong-gun Geoffrey Rush Dann	David Spade Sophie Marceau Ever Carradine Step
director	Sngmoo Lee	Jeff Pollack
tagline	Assassin. Hero. Legend.	A comedy about a guy who would do anything to
runtime	100	95
genres	Adventure Fantasy Action Western Thriller	Comedy Romance
release_date	2010-12-02 00:00:00	1999-04-23 00:00:00
vote_count	74	14
vote_average	6.4	4.8
release_year	2010	1999
budget_adj	4.25e+08	1.30905
revenue_adj	1.10876e+07	130.905

### 1. Average runtime of all the movies

```
In [16]: def avg_fun(column):
    return tmdb_data[column].mean()
avg_fun('runtime')
```

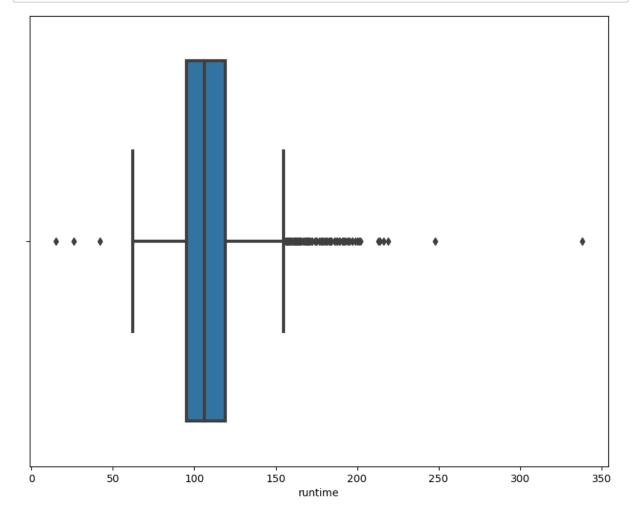
Out[16]: 109.22029060716139

```
In [17]: #plotting a histogram of runtime of movies
   plt.figure(figsize=(9,5), dpi = 100)
   plt.xlabel('Runtime of the Movies', fontsize = 15)
   plt.ylabel('Numbers of Movies in the Dataset', fontsize=15)
   plt.title('Runtime of all the movies', fontsize=15)
   plt.hist(tmdb_data['runtime'], rwidth = 0.9, bins =35)
   plt.show()
```

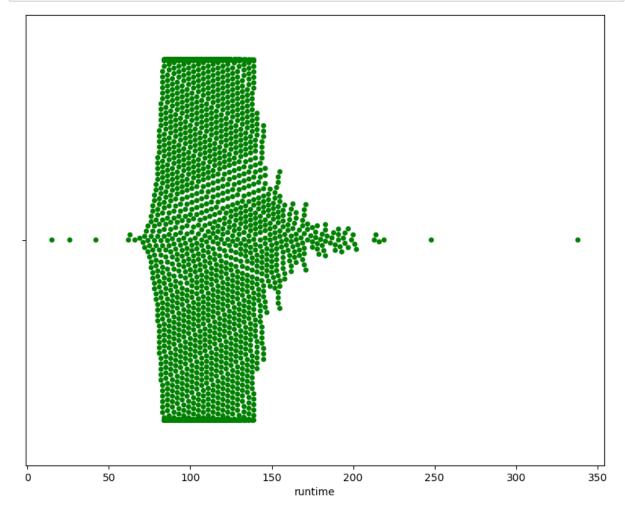


In this histogram is showing positively right skewed. Most of the movies runtime are between 90 to 140. Almost 6,000 and more no.of movies fall in this criteria.

```
In [18]: import seaborn as sns
    plt.figure(figsize=(10,8), dpi = 100)
    #using seaborn to generate the boxplot
    sns.boxplot(tmdb_data['runtime'], linewidth = 3)
    plt.show()
```



```
In [33]: plt.figure(figsize=(10,8), dpi = 100)
    sns.swarmplot(tmdb_data['runtime'], color = 'green')
    plt.show()
```



```
In [19]: | tmdb_data['runtime'].describe()
Out[19]: count
                   3854.000000
         mean
                    109.220291
                     19.922820
          std
                     15.000000
         min
          25%
                     95.000000
          50%
                    106.000000
          75%
                    119.000000
                    338.000000
          max
         Name: runtime, dtype: float64
```

By looking the plot and calcilations, I can conclude that 1. Average runtime for 25% of movies is 95 minutes 2. Average runtime for 50% of movies is 106 minutes 3. Average runtime for 75% of movies is 119 minutes

1. Movies with most and least earned revenue.

```
In [20]: #defining the function
def calculate(column):
    #for highest earned profit
    high= tmdb_data[column].idxmax()
    high_details=pd.DataFrame(tmdb_data.loc[high])

#for Lowest earned profit
    low= tmdb_data[column].idxmin()
    low_details=pd.DataFrame(tmdb_data.loc[low])

#collectin data in one place
    info=pd.concat([high_details, low_details], axis=1)

    return info
calculate('revenue')
```

### Out[20]:

	1386	5067
popularity	9.43277	0.462609
budget	2.37e+08	6e+06
revenue	2.78151e+09	2
original_title	Avatar	Shattered Glass
cast	Sam Worthington Zoe Saldana Sigourney Weaver S	Hayden Christensen Peter Sarsgaard Chloë Sevi
director	James Cameron	Billy Ray
tagline	Enter the World of Pandora.	NaN
runtime	162	94
genres	Action Adventure Fantasy Science Fiction	Drama History
release_date	2009-12-10 00:00:00	2003-11-14 00:00:00
vote_count	8458	46
vote_average	7.1	6.4
release_year	2009	2003
budget_adj	2.40887e+08	7.11212e+06
revenue_adj	2.82712e+09	2.37071

Column with 1386 shows the most revenue earned. That is 278151e + 09 And the column with 5067 shows least earned. That is 2.

1. Movies with longest and shortest runtime values.

In [21]: calculate('runtime')

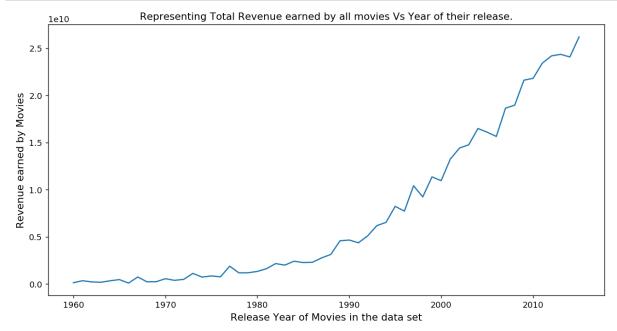
Out[21]:

	2107	5162
popularity	0.534192	0.208637
budget	1.8e+07	10
revenue	871279	5
original_title	Carlos	Kid's Story
cast	Edgar RamÃrez Alexander Scheer Fadi Abi Samra	Clayton Watson Keanu Reeves Carrie-Anne Moss K
director	Olivier Assayas	Shinichiro Watanabe
tagline	The man who hijacked the world	NaN
runtime	338	15
genres	Crime Drama Thriller History	Science Fiction Animation
release_date	2010-05-19 00:00:00	2003-06-02 00:00:00
vote_count	35	16
vote_average	6.2	6.8
release_year	2010	2003
budget_adj	1.8e+07	11.8535
revenue_adj	871279	5.92676

Longest runtime show in column 2107 that is 338 minutes. And shortest runtime in column 5162 that is 15 minutes.

1. Year of release vs revenue earned.

```
In [22]: revenue_year = tmdb_data.groupby('release_year')['revenue'].sum()
    plt.figure(figsize=(12,6), dpi = 130)
    plt.xlabel('Release Year of Movies in the data set', fontsize = 12)
    plt.ylabel('Revenue earned by Movies', fontsize = 12)
    plt.title('Representing Total Revenue earned by all movies Vs Year of their re lease.')
    plt.plot(revenue_year)
    plt.show()
```



```
In [23]: revenue_year.idxmax()
```

Out[23]: 2015

From 1960 to 2015 there was a significant increase in revenue year wise. And in year 2015 earn highest revenue.

# What when we fix the revenue earn by movies is \$ 50M

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```
In [24]: revenue_data = tmdb_data[tmdb_data['revenue'] >= 50000000]
         revenue_data.index = range(len(revenue_data))
         revenue_data.index = revenue_data.index + 1
         revenue_data.head(3)
```

#### Out[24]:

		popularity	budget	revenue	original_title	cast	director	tagline	runtir
	1	32.985763	150000000.0	1.513529e+09	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	The park is open.	1
	2	28.419936	150000000.0	3.784364e+08	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	What a Lovely Day.	1
	3	13.112507	110000000.0	2.952382e+08	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	One Choice Can Destroy You	1
	4								•
In [25]:	ler	n(revenue_	_data)# no d	of rows					
Out[25]:	183	36							

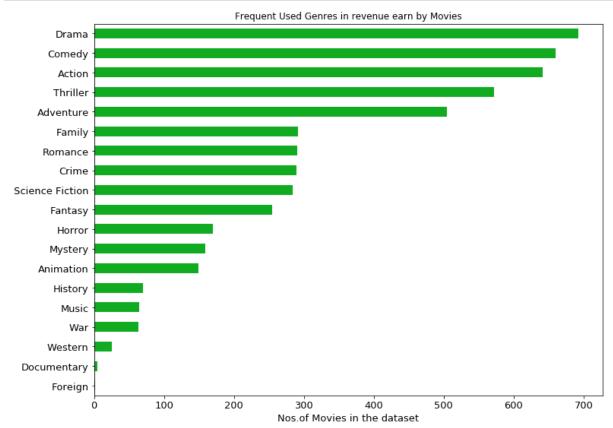
So dataset is reduced to 1836 from 3853.

# Part 2

1. Succesful Genres

```
In [26]: def data(column):
              data = revenue_data[column].str.cat(sep = '|')
              data = pd.Series(data.split('|'))
              count = data.value_counts(ascending = False)
              return count
          count = data('genres')
          count.head()
Out[26]: Drama
                       693
         Comedy
                       660
         Action
                       642
                       572
         Thriller
         Adventure
                       505
         dtype: int64
```

```
In [40]: count.sort_values(ascending = True, inplace = True)
    lt = count.plot.barh(color = '#10ab21', fontsize = 13)
    lt.set(title = 'Frequent Used Genres in revenue earn by Movies')
    lt.set_xlabel('Nos.of Movies in the dataset', color = 'black', fontsize = '13')
    lt.figure.set_size_inches(12, 9)
    plt.show()
```



### 1. Most Frequent Cast

Bruce willis is on the top with 32 cast and Tom Cruise on second with 30 cast.

#### 1. Average budget

```
In [28]: def revenue_avg(column):
    return revenue_data[column].mean()
    revenue_avg('budget')

Out[28]: 58559101.38344227
```

When movies average budget is 59million than moviese arn revenue 50 million more.

1. Average duration

```
In [29]: revenue_avg('runtime')
Out[29]: 113.12581699346406
```

when average duration movies is 113 minutes than movies earn \$50 million more.

1. Movies popularity

```
In [30]: def revenue_avg(column):
    return revenue_data[column].mean()
    revenue_avg('release_year')

Out[30]: 2002.6252723311547
```

Average revenue earn by year by release is \$2002 million

# **Conclusions**

In this data analysis i can able to find the various factors which involed to earn revenue from the movies. Some finding are below

- 1. Average budget must be around \$60 millon.
- 2. Average duration should be 113 minutes.
- 3. Genre should be Drama, Comedy, Action, Thriller, Adventure.
- Most Frequent Cast are Bruce Willis, Tom Cruise, Robert De Niro, Tom Hanks, Brad Pitt.

These are some factor through which movies earn more revenue and average revenue \$2002 million dollar by release years.

## Limitations:

The data analysis process was done which had significant amount of revenue earn is around \$50 million on various runtime, budget, genres. But In this dataset i observe that the profit column is missing so we can not analysis on the various fector which lead to profiable movies. And Currency unit is not there so it might be some possibility to miss match in the results.

In [ ]: