

Investigate_a_Dataset

July 17, 2023

Tip: Welcome to the Investigate a Dataset project! You will find tips in quoted sections like this to help organize your approach to your investigation. Once you complete this project, remove these **Tip** sections from your report before submission. First things first, you might want to double-click this Markdown cell and change the title so that it reflects your dataset and investigation.

1 Project: Investigate a Dataset - [Database_TMDb_movie_data]

1.1 Table of Contents

Introduction

Data Wrangling

Exploratory Data Analysis

Conclusions

Introduction

1.1.1 Dataset Description

Tip: In this section of the report, provide a brief introduction to the dataset you've selected/downloaded for analysis. Read through the description available on the homepage-links present [here](#). List all column names in each table, and their significance. In case of multiple tables, describe the relationship between tables.

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

- 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.

1.1.2 Question(s) for Analysis

Tip: Clearly state one or more questions that you plan on exploring over the course of the report. You will address these questions in the **data analysis** and **conclusion** sections. Try to build your report around the analysis of at least one dependent variable

and three independent variables. If you're not sure what questions to ask, then make sure you familiarize yourself with the dataset, its variables and the dataset context for ideas of what to explore.

Tip: Once you start coding, use NumPy arrays, Pandas Series, and DataFrames where appropriate rather than Python lists and dictionaries. Also, **use good coding practices**, such as, define and use functions to avoid repetitive code. Use appropriate comments within the code cells, explanation in the mark-down cells, and meaningful variable names.

1. Explore the trend of profit and the number of movies
2. Explore the range of average vote
3. Explore the trend of profit of Universal Pictures

```
In [1]: # Use this cell to set up import statements for all of the packages that you
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
% matplotlib inline

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

Data Wrangling

Tip: In this section of the report, you will load in the data, check for cleanliness, and then trim and clean your dataset for analysis. Make sure that you **document your data cleaning steps in mark-down cells precisely and justify your cleaning decisions**.

1.1.3 General Properties

Tip: You should *not* perform too many operations in each cell. Create cells freely to explore your data. One option that you can take with this project is to do a lot of explorations in an initial notebook. These don't have to be organized, but make sure you use enough comments to understand the purpose of each code cell. Then, after you're done with your analysis, create a duplicate notebook where you will trim the excess and organize your steps so that you have a flowing, cohesive report.

```
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('Database_TMDB_movie_data/tmdb-movies.csv')
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
popularity        10866 non-null float64
budget            10866 non-null int64
revenue           10866 non-null int64
original_title    10866 non-null object
cast              10790 non-null object
homepage          2936 non-null object
director          10822 non-null object
tagline           8042 non-null object
keywords          9373 non-null object
overview          10862 non-null object
runtime           10866 non-null int64
genres            10843 non-null object
production_companies 9836 non-null object
release_date      10866 non-null object
vote_count        10866 non-null int64
vote_average      10866 non-null float64
release_year      10866 non-null int64
budget_adj        10866 non-null float64
revenue_adj       10866 non-null float64
dtypes: float64(4), int64(6), object(11)
memory usage: 1.7+ MB

```

```
In [3]: df.head()
```

```

Out[3]:
   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]

1.1.4 Data Cleaning

Tip: Make sure that you keep your reader informed on the steps that you are taking in your investigation. Follow every code cell, or every set of related code cells, with

a markdown cell to describe to the reader what was found in the preceding cell(s). Try to make it so that the reader can then understand what they will be seeing in the following cell(s).

1. Remove unsue column

```
In [4]: # After discussing the structure of the data and any problems that need to be
        # cleaned, perform those cleaning steps in the second part of this section.
```

```
df.drop(['id', 'imdb_id', 'genres', 'original_title', 'cast', 'homepage', 'director', 'tagline']
df.to_csv('tmdb-movies-edited.csv', index=False)
df_movies = pd.read_csv('tmdb-movies-edited.csv')
df.info();
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10866 entries, 0 to 10865
Data columns (total 8 columns):
popularity                10866 non-null float64
production_companies      9836 non-null object
release_date              10866 non-null object
vote_count                10866 non-null int64
vote_average              10866 non-null float64
release_year              10866 non-null int64
budget_adj                10866 non-null float64
revenue_adj               10866 non-null float64
dtypes: float64(4), int64(2), object(2)
memory usage: 679.2+ KB
```

2. Clean the data

```
In [5]: # remove duplicate records
df_movies.drop_duplicates(inplace=True)
df_movies.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 10865 entries, 0 to 10865
Data columns (total 8 columns):
popularity                10865 non-null float64
production_companies      9835 non-null object
release_date              10865 non-null object
vote_count                10865 non-null int64
vote_average              10865 non-null float64
release_year              10865 non-null int64
budget_adj                10865 non-null float64
revenue_adj               10865 non-null float64
dtypes: float64(4), int64(2), object(2)
memory usage: 763.9+ KB
```

```
In [6]: # remove records with zero value
df_clean_invalid_values = df_movies.replace(0,np.NaN)
df_movies = df_clean_invalid_values.dropna()
df_movies.to_csv('tmdb-movies-edited.csv', index=False)
df_movies = pd.read_csv('tmdb-movies-edited.csv')
df_movies.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3808 entries, 0 to 3807
Data columns (total 8 columns):
popularity                3808 non-null float64
production_companies      3808 non-null object
release_date              3808 non-null object
vote_count                3808 non-null int64
vote_average              3808 non-null float64
release_year              3808 non-null int64
budget_adj                3808 non-null float64
revenue_adj               3808 non-null float64
dtypes: float64(4), int64(2), object(2)
memory usage: 238.1+ KB
```

3. Calculate the profit column

```
In [7]: df_movies.insert(8,'profit',df_movies['revenue_adj']-df_movies['budget_adj'])
df_movies.to_csv('tmdb-movies-edited.csv', index=False)
df_movies = pd.read_csv('tmdb-movies-edited.csv')
df_movies.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3808 entries, 0 to 3807
Data columns (total 9 columns):
popularity                3808 non-null float64
production_companies      3808 non-null object
release_date              3808 non-null object
vote_count                3808 non-null int64
vote_average              3808 non-null float64
release_year              3808 non-null int64
budget_adj                3808 non-null float64
revenue_adj               3808 non-null float64
profit                    3808 non-null float64
dtypes: float64(5), int64(2), object(2)
memory usage: 267.8+ KB
```

Exploratory Data Analysis

Tip: Now that you've trimmed and cleaned your data, you're ready to move on to exploration. **Compute statistics** and **create visualizations** with the goal of addressing

the research questions that you posed in the Introduction section. You should compute the relevant statistics throughout the analysis when an inference is made about the data. Note that at least two or more kinds of plots should be created as part of the exploration, and you must compare and show trends in the varied visualizations.

Tip: - Investigate the stated question(s) from multiple angles. It is recommended that you be systematic with your approach. Look at one variable at a time, and then follow it up by looking at relationships between variables. You should explore at least three variables in relation to the primary question. This can be an exploratory relationship between three variables of interest, or looking at how two independent variables relate to a single dependent variable of interest. Lastly, you should perform both single-variable (1d) and multiple-variable (2d) explorations.

1.1.5 Research Question 1: Explore the trend of profit and the number of movies

```
In [8]: def movies_profit_trend(column_profit, column_year):
        df_movies = pd.read_csv('tmdb-movies-edited.csv')

        bins = [1960, 1970, 1980, 1990, 2000, 2010]
        ind = np.digitize(df_movies[column_year], bins)

        df_profit = df_movies.groupby(ind)[column_profit].sum().to_frame()

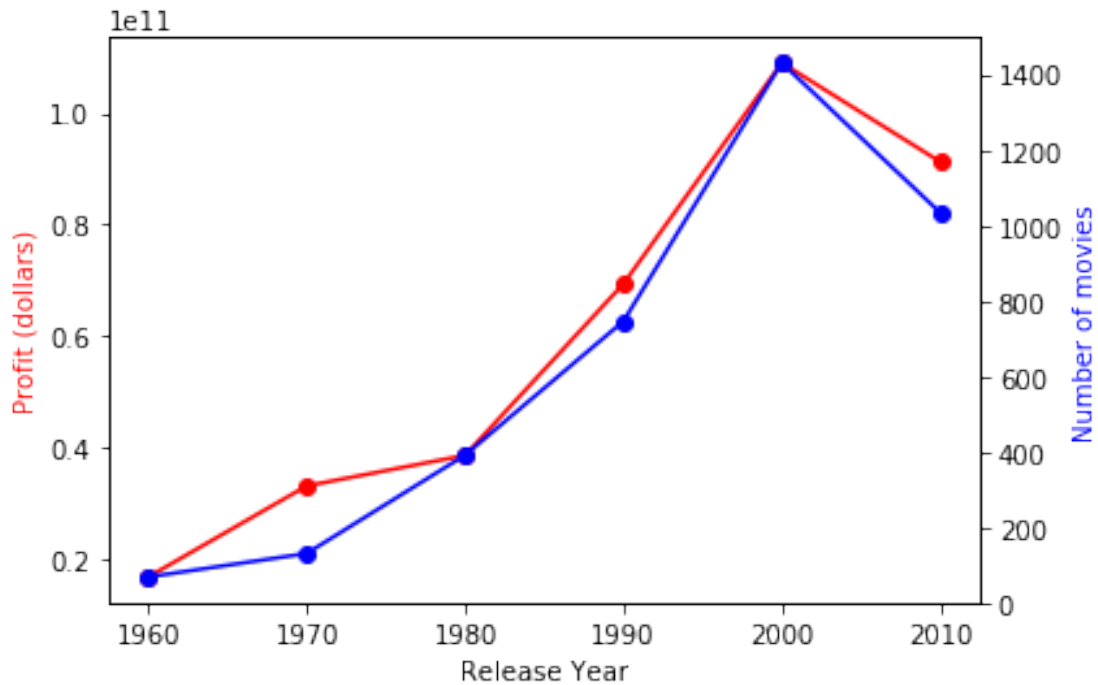
        df_movies_count = df_movies.groupby(ind)[column_year].count().to_frame()
        X_axis = np.arange(1960, 2015, 10)

        fig, ax = plt.subplots()
        ax.plot(X_axis, df_profit[column_profit], color="red", marker="o")
        ax.set_xlabel('Release Year')
        ax.set_ylabel('Profit (dollars)', color="red")

        ax2 = ax.twinx()
        ax2.plot(X_axis, df_movies_count[column_year], color="blue", marker="o")
        ax2.set_ylabel("Number of movies", color="blue")

        plt.show()

In [9]: movies_profit_trend('profit', 'release_year')
```



Conclusions

The number of movies and their profit is increasing from 1960 to 2015. The film industry is still on the rise.

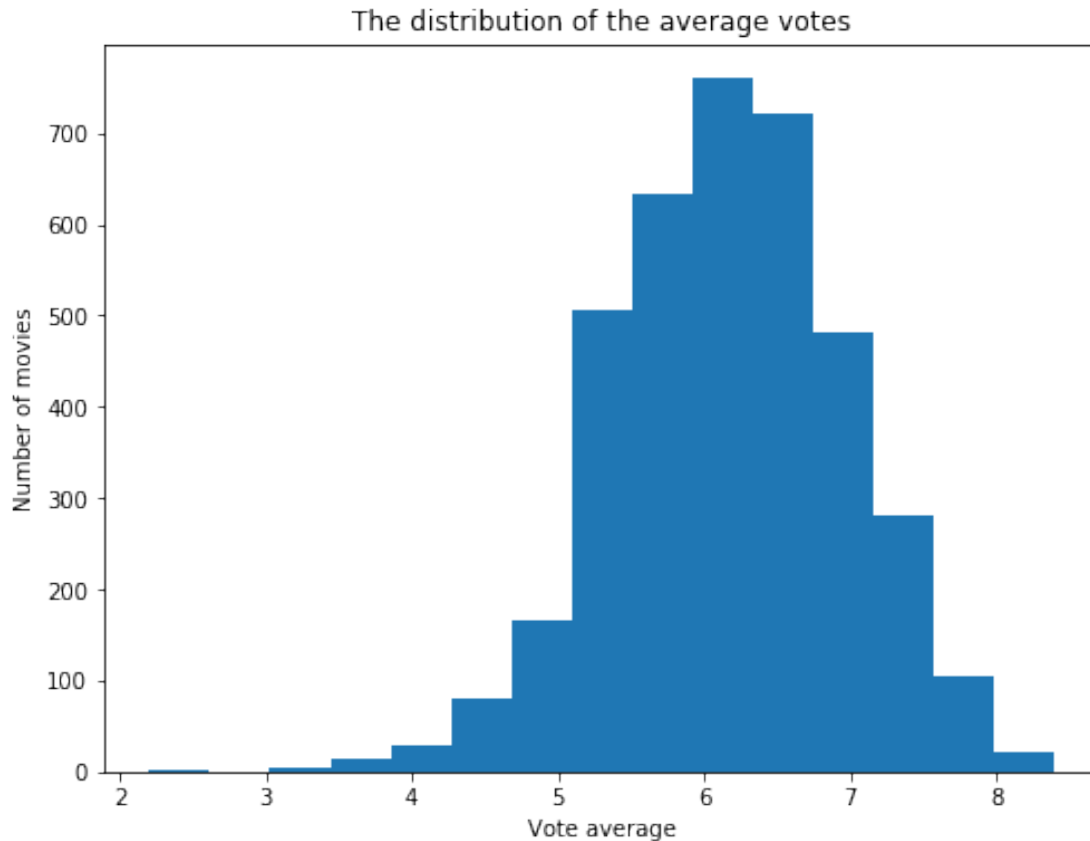
1.1.6 Research Question 2: Explore the range of average vote

```
In [10]: data = df_movies['vote_average']
        bins = np.linspace(df_movies['vote_average'].min(), df_movies['vote_average'].max(), 16)

        # Create histogram
        fig, ax = plt.subplots(figsize=(8, 6))
        ax.hist(data, bins)

        # Show plot
        plt.xlabel('Vote average')
        plt.ylabel('Number of movies')
        plt.title('The distribution of the average votes')
        plt.show()

        max = df_movies['vote_average'].max()
        min = df_movies['vote_average'].min()
        print('The highest vote: ', max, ', number of movies: ', df_movies.query('vote_average = ' + str(max)).shape[0])
        print('The lowest vote: ', min, ', number of movies: ', df_movies.query('vote_average = ' + str(min)).shape[0])
```

The highest vote: 8.4 , number of movies: 2
 The lowest vote: 2.2 , number of movies: 1

Conclusions

The average rating of movies is between 5 and 7

1.1.7 Research Question 3: Explore the trend of profit of Universal Pictures

```
In [11]: # prepare data
df_universal_pictures_movies = df_movies[df_movies['production_companies'].str.contains
df_universal_pictures_movies.to_csv('tmdb-universal-pictures-movies.csv', index=False)
df_universal_pictures_movies = pd.read_csv('tmdb-universal-pictures-movies.csv')
df_universal_pictures_movies.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 332 entries, 0 to 331
Data columns (total 9 columns):
popularity                332 non-null float64
production_companies      332 non-null object
```

```

release_date      332 non-null object
vote_count        332 non-null int64
vote_average      332 non-null float64
release_year      332 non-null int64
budget_adj        332 non-null float64
revenue_adj       332 non-null float64
profit            332 non-null float64
dtypes: float64(5), int64(2), object(2)
memory usage: 23.4+ KB

```

```

In [12]: def universal_trend(column_profit, column_year):
          df_movies = pd.read_csv('tmdb-universal-pictures-movies.csv')

          plt.figure(figsize=(8,3))

          plt.plot(df_movies.groupby(column_profit)[column_year].sum())

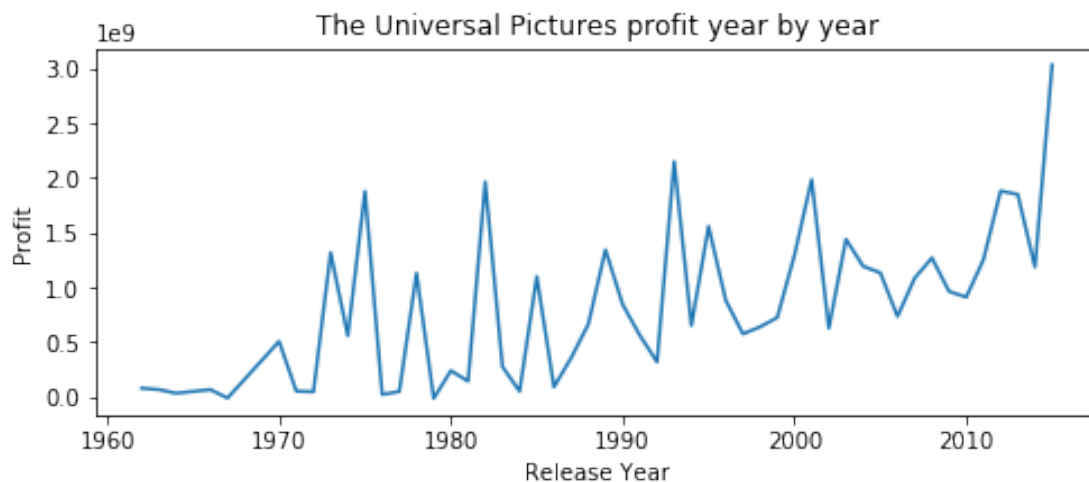
          plt.xlabel('Release Year')
          plt.ylabel('Profit')
          plt.title('The Universal Pictures profit year by year')
          plt.show()

```

```

In [13]: universal_trend('release_year', 'profit')

```



Conclutions

Universal Pictures' profits are still on an upward trend from 1960 to 2015

Conclusions

In this investigation, we can find that the number of movies and their profit is increasing from year to year. This also applies to Universal Pictures as its profit is still on an upward trend from 1960 to 2015. We also find out that the average rating of movies is between 5 and 7, the lowest average vote 2.2 is and the highest average vote is 8.4.

However, the above conclusion may be not accurate because of the lacking of data, since we removed many records with empty data.

1.2 Submitting your Project

Tip: Before you submit your project, you need to create a .html or .pdf version of this notebook in the workspace here. To do that, run the code cell below. If it worked correctly, you should get a return code of 0, and you should see the generated .html file in the workspace directory (click on the orange Jupyter icon in the upper left).

Tip: Alternatively, you can download this report as .html via the **File > Download as** submenu, and then manually upload it into the workspace directory by clicking on the orange Jupyter icon in the upper left, then using the Upload button.

Tip: Once you've done this, you can submit your project by clicking on the "Submit Project" button in the lower right here. This will create and submit a zip file with this .ipynb doc and the .html or .pdf version you created. Congratulations!

```
In [14]: from subprocess import call
         call(['python', '-m', 'nbconvert', 'Investigate_a_Dataset.ipynb'])
```

```
Out[14]: 0
```

```
In [ ]:
```